

***Northwest Forest Plan—The First Twenty Years (1994-2013):  
Status and Trend of Marbled Murrelet Populations and  
Nesting Habitat***

**Technical Coordinators**

Gary A. Falxa

Martin G. Raphael

## PREFACE

In the 1980s, public controversy intensified in the Pacific Northwest over timber harvest in old-growth forests, declining species populations [such as northern spotted owl (*Strix occidentalis caurina*), marbled murrelet (*Brachyramphus marmoratus*), and Pacific salmon], and the role of federal forests in regional and local economies. This ultimately led to the adoption of the Northwest Forest Plan (NWFP), which amended existing management plans for 19 national forests and 7 Bureau of Land Management districts in California, Oregon, and Washington (24 million acres of federal land within the 57-million-acre range of the northern spotted owl). The NWFP provides a framework for an ecosystem approach to the management of those 24 million acres of federal lands. It established the overarching conservation goals of (1) protecting and enhancing habitat for species associated with late-successional and old-growth forests, (2) restoring and maintaining the ecological integrity of watersheds and aquatic ecosystems, and (3) providing a predictable level of timber sales and other services, as well as maintaining the stability of rural communities and economies.

The NWFP relies on monitoring to detect changes in ecological and social systems relevant to its success in meeting conservation objectives, and on adaptive management processes that evaluate and use monitoring information to adjust conservation and management practices (Mulder et al. 1999). An interagency effectiveness monitoring framework was implemented to meet requirements for tracking status and trend for watershed condition, late-successional and old-growth forests, social and economic conditions, tribal relationships, and population and habitat for marbled murrelets and northern spotted owls. This report is one of a set of status and trend monitoring reports on these topics, which addresses questions about the effectiveness of the NWFP in meeting its objectives through its first 20 years. Monitoring results for the first 10 years and first 15 years are documented in a series of reports posted at: <http://www.reo.gov/monitoring/reports/index.shtml>.

This is the third in a series of monitoring reports from the Marbled Murrelet Effectiveness Monitoring module under the NWFP, and focuses on monitoring results on the status and trends for marbled murrelet populations and nesting habitat through the first 20 years of the NWFP (1994--2013), following the design described in the “Marbled Murrelet Effectiveness Monitoring Plan for the Northwest Forest Plan” (Madsen et al. 1999). This report is composed of three chapters. The first chapter discusses the status and trend of the portion of the murrelet population associated with the NWFP area. The second chapter presents the status and trend of murrelet nesting habitat. The third chapter presents results from an evaluation of the relationships between murrelet distribution at sea off the NWFP area, nesting habitat distribution and other terrestrial factors, and marine factors. This chapter is a first step toward meeting the long-term monitoring goal of the murrelet monitoring strategy, as described in Madsen et al. (1999), of developing a predictive model that relates forest habitat conditions to the demographic health of the murrelet population. In addition, the third chapter provides a brief synthesis of the results of all three chapters in this report, and a discussion of management implications of those results.

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## REPORT ABSTRACT

The Northwest Forest Plan (NWFP) is an ecosystem management plan for federal forest lands in the Pacific Northwest of the United States. It was implemented, in part, to conserve and restore old-growth and late-successional forests that would contribute to the conservation and recovery of threatened species including the marbled murrelet (*Brachyramphus marmoratus*). Monitoring of murrelet populations and nesting habitat helps inform land managers of the effectiveness of the NWFP in meeting its goals and objectives.

A specific conservation goal of the NWFP is to stabilize and increase murrelet populations by maintaining and increasing nesting habitat. We monitored marbled murrelet populations annually from 2000 to 2013 in near-shore marine waters associated with the NWFP area, using boat-based transects and distance estimation methods, in coastal waters off Washington, Oregon, and northern California. We divided this area of coastal waters into five geographic subareas corresponding to conservation zones established in the U.S. Fish and Wildlife Service's recovery plan for the marbled murrelet, and estimated population size and trend for each conservation zone, and for all zones combined. At the conservation zone scale, the most recent (2013) population estimates ranged from about 71 murrelets in Conservation Zone 5 (San Francisco Bay north to Shelter Cove, California) to 7,880 murrelets in Conservation Zone 3 (from Coos Bay north to the Columbia River, Oregon). Estimated density of murrelets on the surveyed waters ranged from approximately 0.1 murrelets per sq. km in Conservation Zone 5 to 5.2 murrelets per sq. km in Conservation Zone 4 (from Shelter Cove, California north to Coos Bay, Oregon). Annual population estimates for the entire NWFP area ranged from about 16,600 to 22,800 murrelets during the 14-year period, with a 2013 estimate of 19,700 (95 percent confidence interval: 15,400 to 23,900). We computed linear trends of the annual population estimates through 2013 at multiple scales. At the conservation-zone scale, there was strong evidence of a linear decline in the two conservation zones in Washington: Conservation Zone 1 (3.9 percent decline per year), which includes the Strait of Juan de Fuca, San Juan Islands, and Puget Sound and Conservation Zone 2 (6.7 percent decline per year), which includes the outer coast of Washington. We found no evidence of a linear trend in Conservation Zone 3 or Conservation Zone 5 (confidence intervals broadly overlap zero). In Conservation Zone 4, the trend estimate was positive, but the evidence for a trend was not conclusive because the estimate's 95 percent confidence overlapped zero (1.5 percent per year; 95 percent confidence interval: -0.9 to 4.0). At the state scale, which combines conservation zones and portions of conservation zones, we found strong evidence for a declining linear trend in Washington (4.6 percent decline per year) and no evidence of a trend in Oregon. For California, as for Zone 4, no trend was detected; while the trend estimate was positive, the evidence for a trend was not conclusive. For the entire NWFP area the trend estimate for the 2001 to 2013 period

was negative, but here also the confidence interval for the estimate overlapped zero and the evidence for a trend was inconclusive. This result differs from the decline previously reported at the NWFP-scale for the 2001 to 2010 period. This difference was the result of high population estimates for 2011 through 2013 compared to the previous several years, which reduced the slope of the trend and increased variability. Continued monitoring should help to better understand population trends and to assess underlying factors that might explain trends and variability in annual estimates. The population monitoring results to date indicate that the NWFP goal of stabilizing and increasing marbled murrelet populations has not yet been achieved throughout the NWFP area.

Another objective of the effectiveness monitoring plan for the marbled murrelet includes mapping baseline nesting habitat (at the start of the NWFP) and estimating changes in that habitat over time. Using maximum entropy (Maxent) models, we modeled nesting habitat suitability over lands in the murrelet's range in Washington, Oregon, and California. The models used vegetation and physiographic attributes, and a sample of 368 murrelet nest sites (184 confirmed murrelet nest sites and 184 occupied sites) for model training, and provided estimates of suitable nesting habitat for a baseline year (1993) and 20 years later (2012). We estimated there were about 2.5 million acres of potential nesting habitat over all lands in the murrelet's range in Washington, Oregon, and California at the start of the plan (1993). Of this, 0.46 million acres were identified as highest suitability, matching or exceeding the average conditions for the training sites. Most (90 percent) of potential nesting habitat in 1993 on federally-administered lands occurred within federal reserved-land-use allocations. A substantial amount (41 percent) of baseline habitat occurred on non-federal lands, including 44 percent of the highest suitability habitat. We found a net loss of about 2 percent of potential nesting habitat from 1993 to 2012 on federal lands, compared to a net loss of about 27 percent on nonfederal lands. For federal and nonfederal lands combined, the net loss was about 12 percent. Fire was the major cause of nesting habitat loss on federal lands since the NWFP was implemented, but timber harvest, and insect damage or disease also caused losses; timber harvest was the primary cause of loss on non-federal lands. The large amount of younger forest of lower suitability located in reserves has the potential to offset habitat losses over time, but this merits further investigation using spatially-explicit forest development models.

While the NWFP can provide nesting habitat, the murrelet depends upon the marine environment to meet its foraging and roosting requirements, in addition to its use of terrestrial forest to meet its nesting requirements. To assess the relative contributions of terrestrial and marine factors on murrelet abundance, distribution, and trends, we synthesized data on the status and trend of murrelet populations, status and trend of inland nesting habitat, and status and trend of marine factors. Specifically, we initially examined the spatial and temporal correlations of marine and terrestrial factors with the spatial distribution and trend of murrelets. We then conducted a multivariate analysis by using a boosted regression tree method to concurrently investigate the contributions of a suite of marine and terrestrial factors to at-sea murrelet abundance and trends. In both analyses, we found that numbers of murrelets are positively correlated with amounts and pattern (large contiguous patches) of suitable nesting habitat, and that population trend is

most strongly correlated with trend in nesting habitat although marine factors also contribute to this trend. Model results suggest that conservation of suitable nesting habitat is key to murrelet conservation, but that marine factors, especially factors that contribute to murrelet prey abundance, play a role in murrelet distribution and trend. Conservation of habitat within reserves, as well as management actions that are designed to minimize loss of suitable habitat or improve quality of nesting habitat on all lands, should contribute to murrelet conservation and recovery.

## Chapter 1. Status and Trend of Marbled Murrelet Populations in the Northwest Forest Plan Area

Gary A. Falxa, Martin G. Raphael, Craig Strong, Jim Baldwin, Monique Lance, Deanna Lynch, Scott F. Pearson, Richard D. Young

*Gary A. Falxa is a Fish and Wildlife Biologist, U.S. Fish and Wildlife Service, Arcata Fish and Wildlife Office, 1655 Heindon Rd., Arcata, CA 95521; Martin G. Raphael is a Research Wildlife Biologist, USDA Forest Service, Pacific Northwest Research Station, 3625 93<sup>rd</sup> Ave. SW, Olympia WA 98512; Craig Strong is a Consultant Researcher, Crescent Coastal Research, 7700 Bailey Road, Crescent City, CA 95531; Jim Baldwin is a Statistician Unit Leader, USDA Forest Service, Pacific Southwest Research Station, 800 Buchanan Street, West Annex Building, Albany, CA 94710; Monique Lance is a Research Biologist, Washington Department of Fish and Wildlife, 7801 Phillips Road SW, Lakewood, WA. 98498; Deanna Lynch is a Fish and Wildlife Biologist, U.S. Fish and Wildlife Service, Washington Fish and Wildlife Office, Branch of Listing, 510 Desmond Dr., Suite 102, Lacey, WA 98503; Scott Pearson is a Senior Research Scientist, Washington Department of Fish and Wildlife, 1111 Washington Street SE, Olympia, WA 98501; and Richard D. Young is a GIS Analyst, U.S. Fish and Wildlife Service, Pacific Regional Office, Ecological Services, 911 NE 11<sup>th</sup> Avenue, Portland, OR 97232.*

### ABSTRACT

The Northwest Forest Plan (NWFP) is an ecosystem management plan for federal forest lands in the Pacific Northwest of the United States. It incorporates a program to monitor the effectiveness of the NWFP in meeting various objectives, including supporting populations of species associated with late-successional and old-growth forests. To evaluate the NWFP's effectiveness in conserving species associated with older forests, we monitored marbled murrelet (*Brachyramphus marmoratus*) populations annually from 2000 to 2013 in near-shore marine waters associated with the NWFP Area. We counted murrelets along transect lines using boats in coastal waters off Washington, Oregon, and northern California (north of San Francisco Bay) and used distance estimation methods to account for detectability. We divided this area of coastal waters into five geographic subareas corresponding to conservation zones established in the U.S. Fish and Wildlife Service's recovery plan for the marbled murrelet, and estimated population size and trend for each conservation zone, and for all zones combined. At the conservation zone scale, the most recent (2013) population estimates ranged from about 71 murrelets in Conservation Zone 5 (San Francisco Bay north to Shelter Cove, California) to 7,880 murrelets in Conservation Zone 3 (from Coos Bay north to the Columbia River, Oregon). The density estimates ranged from 0.1 murrelets per km<sup>2</sup> in Conservation Zone 5 to 5.2 murrelets per km<sup>2</sup> in Conservation Zone 4 (from Shelter Cove, California, north to Coos Bay, Oregon). Annual population estimates for the entire NWFP area ranged from about 16,600 to 22,800 murrelets during the 14-year period, with a 2013 estimate of 19,700 (95 percent confidence interval: 15,400 to 23,900). We assessed for potential linear trends of the annual population estimates through 2013 at the NWFP-wide (all 5 conservation zones), single-

zone, and state scales. At the scale of the individual conservation zone, there was strong evidence of a linear decline in the two conservation zones in Washington: Conservation Zone 1 (3.9 percent decline per year; 95 percent confidence interval: -7.6 to 0.0), which includes the Strait of Juan de Fuca, San Juan Islands, and Puget Sound and Conservation Zone 2 (6.7 percent decline per year; 95 percent confidence interval: -11.4 to -1.8), which includes the outer coast of Washington. In contrast, we found no evidence of a linear trend in Conservation Zone 3 or Conservation Zone 5 (confidence intervals broadly overlap zero). In Conservation Zone 4, the trend estimate was positive, but the evidence for a trend was not conclusive because the estimate's 95 percent confidence overlapped zero (1.5 percent per year; 95 percent confidence interval: -0.9 to 4.0). At the state scale, which combines conservation zones and portions of conservation zones, we found strong evidence for a declining linear trend in Washington (4.6 percent decline per year; 95 percent confidence interval: -7.5 to -1.5 percent) and no evidence of a trend in Oregon. For California, as for Zone 4, no trend was detected; while the trend estimate was positive, the evidence for a trend was not conclusive (+2.5 percent per year; 95 percent confidence interval: -1.1 to 6.2). No trend was detected for the overall NWFP area; while the trend estimate was negative, the evidence was not conclusive (-1.2 percent per year; 95 percent confidence interval: -2.9 to 0.5) over the 2001 to 2013 period. The NWFP-area trend for this period differs from the decline previously observed for the 2001 to 2010 period. This difference was the result of high population estimates for 2011 through 2013 compared to the previous several years, which reduced the slope of the trend and increased variability. Contributing to the recent high NWFP-area estimates were higher estimates in Conservation Zone 1 in 2011 and 2012, and in Conservation Zones 3 and 4. Continued monitoring should help to better understand population trends and to assess underlying factors that might explain trends and variability in annual estimates. The population monitoring results indicate that the NWFP goal of stabilizing and increasing marbled murrelet populations has not yet been achieved; potential causes for this are discussed in a companion chapter.

Keywords: *Brachyramphus marmoratus*, marbled murrelet, Northwest Forest Plan, population monitoring, population trends, effectiveness monitoring, seabird, old-growth forest.

## INTRODUCTION

Established in 1994, the Northwest Forest Plan (hereafter, NWFP) represented a major change in how federal forest lands were managed in western Washington and Oregon and northwest California. It was developed in response to public controversy during the late 1980's and early 1990's over the harvest of old forests on federal lands. While public concerns included the loss of old-growth forest ecosystems as a whole, the controversy was fueled and focused in part by concerns over the impacts of harvest on the northern spotted owl (*Strix occidentalis caurina*), which was listed in 1990 as threatened under the federal Endangered Species Act (USFWS 1990). In 1992, the marbled murrelet (*Brachyramphus marmoratus*) (murrelet), a seabird dependent on old-growth forests for nesting was also listed as threatened in Washington, Oregon, and California (USFWS 1992). For both species, loss and degradation of habitat from timber

harvesting, exacerbated by catastrophic events including fire and wind storms, were the primary factors contributing to the listings (USFWS 1990, 1992).

The NWFP provides a framework for an ecosystem approach to the management of about 10 million hectares (24.5 million acres) of federal lands within the range of the northern spotted owl (USDA USDI 1994). It established the overarching conservation goals of (1) protecting and enhancing habitat for species associated with late-successional and old-growth forests, (2) restoring and maintaining the ecological integrity of watersheds and aquatic ecosystems, and (3) providing a predictable level of timber sales and other services, as well as maintaining the stability of rural communities and economies. A more specific conservation goal of the NWFP is to stabilize and increase marbled murrelet populations by maintaining and increasing nesting habitat (Madsen et al. 1999). The NWFP (USDA USDI 1994) identified the following as a primary question for evaluating the plan's effectiveness in achieving this goal: *Is the marbled murrelet population stable or increasing?* The objective of this chapter is to address this question based on data collected during the NWFP's first 20 years.

Ecological monitoring programs were established to evaluate the effectiveness of the NWFP in meeting conservation objectives, and to inform management decisions (Mulder et al. 1999). Specifically, monitoring programs were established to assess the status and trends of: (1) late-successional and old-growth forests, (2) northern spotted owl habitat and populations, (3) marbled murrelet habitat and populations, (4) federal agency relationships with Indian tribes, (5) watershed conditions, and (6) socioeconomic conditions.

Although the marbled murrelet is a seabird that spends most of its time living and foraging in coastal marine waters, it was selected for monitoring because it is strongly associated with late-successional and old-growth forests for nesting (Madsen et al. 1999). It nests mostly on large branches or other suitable platforms in large coniferous trees (Ralph et al. 1995, Nelson 1997). Nesting habitat is key to marbled murrelet conservation (Ralph et al. 1995; USFWS 1997, 2009; Raphael 2006; Piatt et al. 2007). Due mostly to timber harvesting, only a small percentage (5 to 20 percent, depending on region) of original old-growth forest remains (Morrison 1988, Norheim 1996, 1997), mostly in relatively small, fragmented patches or in forest parks and reserves. The NWFP identified goals for marbled murrelet nesting habitat including providing substantially more suitable habitat for marbled murrelets than existed at the start of the plan, providing large contiguous blocks of murrelet nesting habitat, and improving or maintaining the distribution of populations and habitat (Madsen et al. 1999). Monitoring murrelet population trends provides a key indicator of whether the NWFP is successfully providing nesting habitat to support a stable and well-distributed murrelet population (Madsen et al. 1999); chapter 2 of this volume provides results from the monitoring of nesting habitat.

Marbled murrelet monitoring for the NWFP has both habitat and population components (Madsen et al. 1999). For habitat monitoring, the approach is to establish a baseline level of nesting habitat by first modeling habitat relationships, and then comparing habitat changes to the baseline (Huff et al. 2006; Raphael et al. 2006; Raphael et al. 2011). Population size and trends are monitored using a unified sampling design and standardized survey methods (Miller et al. 2006; Raphael et al. 2007, Miller et al. 2012).

Thus, trends in both murrelet nesting habitat and populations are tracked over time. The ultimate goal is to relate population trends to the amount and distribution of nesting habitat (Madsen et al. 1999).

### **WHAT IS NEW SINCE THE 15-YEAR REPORT?**

In this report, the status and trend analyses incorporate several more years sampling data, through 2013. While methods have remained consistent for murrelet population monitoring, we also conducted a number of new analyses including:

- An extensive review of all data 2000 to 2013 for consistency and archival purposes, followed by a re-analysis of the density, size and trends of murrelet populations associated with the NWFP area;
- A new analysis of the effect of Beaufort sea state on the detectability of murrelets, and thus on estimated murrelet densities and trends throughout the analysis area, at multiple spatial scales;
- An evaluation of whether murrelet distribution with respect to distance from shore (inshore versus offshore subunit) changed over the 2000 to 2013 period;
- An evaluation of State-level population status and trends, for use by state managers and others (e.g., evaluating State-level recovery); this is in addition to the ongoing analysis of status and trends at the conservation zone and NWFP area scales.
- An updated power analysis using sampling data through 2013 to forecast the program's ability to detect trends in future surveys under a reduced monitoring effort.

## **METHODS**

### **Sampling Design**

The objectives of our murrelet population monitoring are to estimate population size and trend in coastal waters adjacent to the NWFP area), which extends from the United States border with British Columbia south to the Golden Gate of San Francisco Bay (Figure 1-1). The NWFP area encompasses five of the six marbled murrelet conservation zones (sampling strata) designated by the Marbled Murrelet Recovery Plan (USFWS 1997) and lies offshore of the NWFP area. The target population is also defined by the area of navigable waters within 3-8 km of shore (distance varies by conservation zone), and temporally from mid-May through the end of July, when breeding murrelets at sea are likely to be associated with inland nesting habitat. The total area of coastal waters within this area and containing the target population was about 3,392 mi<sup>2</sup> (8,785 km<sup>2</sup>). Within each conservation zone (Figure 1-1), two or three geographic strata were designated based on patterns of murrelet density (Miller et al. 2006, Raphael et al. 2007). The distance from shore of the offshore boundary for the target population varied among conservation zones and strata, and was selected in

each area to capture at least 95 percent of the murrelets on the water (Bentivoglio et al. 2002; Miller et al. 2006, 2012; Raphael et al. 2007). Sampling was designed to allocate more effort to strata with higher murrelet densities (Raphael et al. 2007).

To assess murrelet density and population size within each conservation zone and stratum, we established Primary Sampling Units (PSU) that are roughly rectangular areas of approximately 20 km of coastline and are contiguous over the entire sampling area. PSU and strata boundaries remained constant over the sampling period. Each conservation zone includes from 14 to 22 unique PSUs, except for Conservation Zone 1, where the complex shoreline of the Puget Sound area resulted in 98 PSUs. Although the NWFP was implemented in 1994, it took several years to develop an effectiveness monitoring plan and sampling design. Following completion of the effectiveness monitoring plan for murrelets (Madsen et al. 1999), population monitoring began in 2000 for all conservation zones. Our target sample size in Conservation Zones 2 through 5 was 30 PSU surveys per conservation zone per year; most or all unique PSUs in these conservation zones were sampled each year and in a random sampling order. In Conservation Zone 1, an initial sample of 30 PSUs was randomly selected out of the 98 available PSUs, and each selected PSU was sampled twice each year (Raphael et al. 2007). This same random Conservation Zone 1 subsample was sampled each year to minimize between-year variance. In Conservation Zone 5, the target sample was reduced to 15 PSUs in 2004 to balance logistics, cost and precision in this area of very few murrelets. Conservation Zone 5 was not sampled in 2006, 2009, 2010, or 2012 due to funding limitations. We discuss below, under *Treatment of Years with No Surveys for Conservation Zone 5*, how we dealt with missing data from Conservation Zone 5 in population size and trend estimates.

We divided PSUs into inshore and offshore subunits (Figure 1-2), which allows more sampling effort in nearshore subunits with higher murrelet density (Bentivoglio et al. 2002). However, PSUs in stratum 3 of Conservation Zone 1 were not divided into subunits, as murrelet density was low throughout the stratum. The inshore unit extended to either 1500 or 2000 meters from shore, except in stratum 2 of Conservation Zone 1, where narrow inlets and passages between opposite shorelines limited the inshore subunit to within 500 meters of shore. As discussed below, for Conservation Zone 5 we changed the division between inshore and offshore PSU subunits in 2005 from 2000 meters offshore to 1200 meters. Inshore PSU subunits generally have higher murrelet densities, so they were sampled with more effort using transects placed parallel to shore. Offshore PSU subunit transects are oriented diagonally with the shoreline, often in a zigzag configuration (Figure 1-2) to sample across the gradient of murrelet density that, generally, declines with distance from shore (Ralph and Miller 1995). PSU sampling details for each conservation zone and stratum are summarized in Raphael et al. (2007).

We use two observers for each survey, one on each side of the boat's centerline, surveying a 90° arc to the left or right of the bow, but emphasizing the area in front of the boat. We estimated murrelet density using line transect methods (Buckland et al. 2001, Thomas et al. 2004), where the perpendicular distance to each detected murrelet or group of murrelets was estimated. Accuracy of distance estimates is key to density estimates using line transect methods. Distance training and calibration occurred throughout the season to maintain consistency in distance estimates between observers

and across years. Because surface waves can obscure murrelets on the water, observers noted sea state using the Beaufort scale. The Beaufort scale is an empirical measure that relates wind speed to observed conditions at sea, ranging from a value of zero (calm, flat sea conditions) to 12 (hurricane force winds). Surveys were generally conducted under sea conditions of Beaufort 2 or less, although occasionally surveys continued after conditions increased to Beaufort 3. Description of the complete survey protocol is provided in Raphael et al. (2007), and in Miller et al. (2006). Minor adjustments to the survey protocol are described below under *Protocol Clarifications and Refinements*. In addition to recording all marbled murrelet detections, observers also recorded other seabirds and marine mammals detected during sampling.

Using this protocol, we conducted population monitoring surveys in the five conservation zones beginning in 2000, and sampled all conservation zones except Conservation Zone 5 in each year between 2000 and 2013. In any year, we conducted 150 to 200 PSU surveys across all conservation zones combined, and recorded approximately 4,000 to 6,000 marbled murrelet observations along roughly 5,500 to 6,500 km of transect (Table 1-1). Because some PSUs are sampled more than once in a year, the number of unique PSUs sampled annually is about 90 to 95 PSUs across the five Conservation Zones (Raphael et al. 2007).

## Analysis

### ***Density and Population Estimates***

We conducted surveys from 2000 through 2013. Departures from the protocol in Conservation Zones 1 and 2 in 2000 may have affected density estimates for those Conservation Zones. Therefore we used data from only 2001 through 2013 for all estimates and analyses involving these conservation zones, namely those for Conservation Zone 1, Conservation Zone 2, Washington State, and “All-Zones” (the 5 conservation zones combined). Conservation Zone 5 was not sampled in four of the years (2006, 2009, 2010, and 2012), and we interpolated Conservation Zone 5 densities for those years based on data from adjacent years and methods described below (*Treatment of Years with No Surveys in Conservation Zone 5*).

For each year of survey, we estimated average marbled murrelet densities (murrelets per km<sup>2</sup>), with an associated estimate of precision for each conservation zone, for the entire target population, and for the three states within the area sampled. We used distance sampling methods (Buckland et al. 2001, 2004) and the software program DISTANCE version 6.2 (Thomas et al. 2010) to estimate the probability for detecting a murrelet that is present at distance zero [ $f(0)$ ] and the mean number of murrelets per group [or cluster size;  $E(s)$ ] for each year and conservation zone from inshore and offshore subunit surveys. We truncated the distance data prior to analysis by discarding the five percent of observations with the greatest distances for each conservation zone, which can improve modeling of detection functions, as recommended by Buckland et al. (2001). We set DISTANCE to use the mean observed cluster size as the estimate for  $E(s)$  unless an internal test found evidence that detection is a function of cluster size, in which case DISTANCE applied a correction (Buckland et al. 2001). For each year, the data from Conservation Zones 4 and 5 were combined for estimating the detection

function,  $E(s)$ ,  $f(0)$ , and truncation distance. We did this because the low number of murrelet detections in Conservation Zone 5 were insufficient for estimating these parameters. DISTANCE also provided the number of groups of murrelets observed per km ( $ER$ =encounter rate) for each PSU subunit survey. We then estimated density (murrelets/km<sup>2</sup>) for each PSU subunit survey (Raphael et al. 2007) using the estimates and encounter rate from DISTANCE with the following formula:

$$\hat{d} = 1000 \cdot \hat{f}(0) \cdot \hat{E}(s) \cdot ER/2$$

The “hats” over the letters designate estimates. Strata, conservation zone, and All-Zones density estimates were constructed from average densities weighted by the area of the respective geographic scale.

Target population estimates for each conservation zone and for the five-conservation zones combined were produced using standard methods for stratified sampling (Cochran 1977, Sokal and Rohlf 1981). We used the total area within each stratum to expand the density estimates from DISTANCE, and associated estimates of precision, to calculate the average total numbers of murrelets by conservation zone, state, and for all conservation zones combined for the target period. Estimates of precision were produced using bootstrap resampling methods with consideration of PSU samples that might be clustered in time or space (Miller et al. 2006; Raphael et al. 2007). Density and population estimates were equivalent for purposes of trend analysis because the total area (area sampled) was constant over the study for all conservation zones, and because population is simply a multiple of density. Details on methods used to calculate population estimates and confidence intervals are provided in Raphael et al. (2007).

To portray variation in at-sea density at a finer scale, we obtained a mean density at the PSU scale by first averaging the annual density for each PSU at two scales: the entire PSU, and for the separate inshore and offshore subunits. We then calculated the mean density for each PSU and its subunits by averaging the annual values throughout the sample period.

### ***Estimating Trends***

We assessed for linear trends in murrelet density in the NWFP area from 2000 through 2013, excluding the year 2000 from analyses which involved Conservation Zones 1 and 2, as previously noted. We estimated trends for each conservation zone, for All-Zones, and for each state. For Conservation Zone 5, the single-conservation zone trend analysis used data from all years with surveys from 2000 through 2013; for the All-Zones analyses, we used the interpolated Conservation Zone 5 densities for the years not sampled (2006, 2009, 2010, and 2012). Because Conservation Zone 5 supports less than one percent of the target population, missing data had very little effect on population estimates and no measurable influence on trend magnitude or significance; this was confirmed empirically by analyzing trends for the NWFP area with and without Conservation Zone 5 included.

We fit a linear regression to the natural logarithm of annual density estimates to test for declining trends in individual Conservation Zones 1 through 5 and in All-Zones. For our analysis, the natural logarithm best fits and tests existing demographic models (USFWS 1997; McShane et al. 2004) that predicted a constant declining murrelet population. We tested the null hypothesis that the slope equals zero or greater (no change or increase in murrelet numbers) against the alternative hypothesis of the slope being less than zero (i.e. a one-tailed test for decreasing murrelet densities). In a model where the percent change ‘ $r$ ’ is constant from year to year, and  $d$  represents the murrelet density estimate in a given year:

$$d_{Year} = d_{2000} \times \left(1 + \frac{r}{100}\right)^{Year-2000} \times e^{error}$$

and when we take the natural logarithm of both sides we end up with a standard linear model:

$$\begin{aligned} \log(d_{Year}) &= \log(d_{2000}) + \log\left(1 + \frac{r}{100}\right) \times (Year - 2000) + error \\ &= a + b \times (Year - 2000) + error \end{aligned}$$

where  $a$  and  $b$  are constants to be estimated, and  $error \sim N(0, \sigma^2)$ . Under such a model the percent change from year to year is constant and is equal to  $r = 100(e^b - 1)$ .

For the purposes of evaluating the evidence for a linear trend, we considered: (1) the magnitude of the annual trend estimate, particularly in relation to zero, where zero represents a stable population, and (2) the width and location of the 95 percent confidence intervals surrounding that trend estimate, also in relation to zero. The evidence for a population trend, versus a stable population, is stronger when the trend estimate and its 95 percent confidence interval do not overlap zero, and when the trend estimate is farther from zero. When the confidence interval of a trend estimate is tight around zero, then we would conclude that there is no evidence of a trend. Finally, when the confidence interval of a trend estimate broadly overlaps zero and the trend estimate is not close to zero, this indicates evidence that is not conclusive for or against a non-zero trend. Confidence intervals which are mainly above or below zero, but slightly overlap zero, can provide some evidence of a trend.

In order to illustrate the cumulative, multi-year effect on population size of the annual population trend estimates from our analyses, we calculated for each trend estimate the cumulative population change over a ten-year period during the period of sampling. For this calculation, we defined a ten-year period as one encompassing 10 increments of change at the annual rate, such as the time period 2003 to 2013. Our calculation used the following formula, where “ $R$ ” is the estimated annual percentage rate of change:

$$\text{Cumulative change (\%)} = 100 \left( \left(1 + \frac{R}{100}\right)^{10} - 1 \right)$$

The cumulative change value assumes a constant rate of change at the estimated trend rate, based on an exponential model of population change.

### ***Effects of Sea Conditions on Density and Trend Estimates***

We evaluated the influence of sea conditions (Beaufort sea state) during surveys on detection functions and ultimately on density and trend estimates.

We treated the Beaufort values for each observation (typically 0, 1, or 2, occasionally 3, and rarely 4) as a categorical rather than continuous variable, because we do not know if the change in detectability at any distance changes in the same amount going from Beaufort 0 to Beaufort 1 as it would be going from Beaufort 2 to Beaufort 3. Using the Distance methods previously described, we obtained separate detection curves for each Beaufort category, which along with the encounter rate and mean group size were used in Distance to estimate murrelet density. We then compared the densities of the default no-covariate model with those from the model with the sea condition covariate. In some years and conservation zones there were too few detections within a particular Beaufort class to meet the Distance method recommendation (Buckland et al. 2001) of an average of at least 3 detections per distance class for modeling detection curves. In this situation, we pooled values in adjacent Beaufort classes. This resulted in the merging of Beaufort class 3 into class 2 observations (12 instances), merging Beaufort class 4 into class 3 (5 instances) or class 2 (2 instances), merging Beaufort class 0 observations into class 1 (6 instances), and merging Beaufort class 2 into class 1 (1 instance). As in other analyses, we pooled Conservation Zone 5 data with Conservation Zone 4 data, because of too few murrelet detections in Conservation Zone 5.

We used AIC methods (Johnson and Omland 2004), to identify the best model for each year, conservation zone, and Beaufort category combination, and also compared the variance in density estimates for competing models. We evaluated the effect of the sea condition covariate on the population-trend estimates by using the regression methods described above to compare the trend estimated from the density estimates provided by the no-covariate and Beaufort covariate models for each conservation zone.

Our at-sea observations are expected to have a detection probability of 1 at zero distance from the transect line and then to decline (with no subsequent increases) with increasing distance from the line—i.e., they are assumed to be monotonic, and to be bounded by zero and one. When including the sea state covariate in our model, the program Distance would not always use a monotonic detection function, due to inclusion of the cosine adjustment, and some estimates of probabilities exceeded 1 at some distances from the line. Cosine adjustments are intended to allow more flexibility in fitting the detection function to the observed data. However, where detection probabilities exceed one, the resulting estimate of density can be erroneous and vastly different from estimates not using Beaufort as a covariate. To address this issue, we examined Beaufort models with and without cosine adjustments for each zone-year combination. Between the 2 models, we selected the Beaufort model with the smallest AICc value, except when the model with the cosine adjustment produced an estimated

detection probability greater than 1, in which case we used the “no cosine adjustment” Beaufort model results.

In three cases (2001 for Conservation Zone 1 and 2001 and 2002 for Conservation Zone 2), sea- condition data were not available, so these conservation zone-year combinations were excluded from the analysis.

We previously evaluated and reported on potential observer (crew) effects with a subset of data (Miller et al. 2012) and did not repeat that evaluation for this report.

### ***Power Analysis (Appendix 1-A)***

We conducted a new power analysis, based on population monitoring data through 2013. The goal of this analysis was to examine the power to detect trends under the reduced sampling effort which we initiated in 2014. The methods and results from this new power analysis are fully reported in Appendix 1-A.

### ***Temporal and Spatial Variation in Marbled Murrelet Distribution as a Function of Distance from Shore***

During the planning phase of the monitoring program, researchers subdivided each PSU into inshore and offshore subunits, to allow allocation of greater sampling effort to inshore areas, where densities of murrelets tend to be greater (Raphael et al. 2007). The allocation of effort was based on data collected prior to 2000, and subject to future adjustment based on new data (Bentivoglio et al. 2002, Raphael et al. 2007). We calculated and inspected the ratios of inshore to offshore density for each year-conservation zone combination to evaluate whether those ratios support the protocol's current allocation of greater sampling effort nearshore. Ratio values >1.0 indicate a greater density of murrelets in the inshore subunits relative to the offshore.

We also evaluated whether the ratio of inshore density to offshore density changed in a consistent manner over time during the years of sampling through 2013. If such changes were observed, then that would trigger a reconsideration of the current allocation of total survey length between the inshore and offshore subunits within PSUs. We conducted this analysis at the stratum scale, the minimum scale at which the survey design allows adjustment of survey effort allocation between subunits. Using all years of survey data through 2013, we calculated the average annual density in the inshore and offshore subunits analyses at the scale of the 2 or 3 strata within each conservation zone. This provided sample sizes of about 30 to 55 PSU samples for each subunit per year. Conservation Zone 5 was excluded from this analysis because the data include many density estimates of zero. Stratum 3 of Conservation Zone 1 was also excluded from the analysis, as PSUs within this stratum do not have an offshore subunit. For each PSU stratum, we visually looked for patterns suggesting a systematic change between 2000 and 2013 in murrelet distribution as a function of distance from shore.

A change in distribution might have implications for any trend patterns observed. In particular, if a shift in murrelet distribution resulted in a smaller proportion of the

population occurring within our sample area (and thus being sampled) in the latter years of this study, this might lead to underestimates of population size in those years and an erroneous decline signal. Our analysis of inshore to offshore density does not provide a rigorous test for such a shift. However, if we were to observe a higher proportion of murrelets offshore in the later years of this study, this could be consistent with such a shift in distribution. Similarly, should we observe no change in the nearshore/offshore ratio over time, this would lend some support to such a shift not occurring.

## **Protocol Clarifications and Refinements**

The field and analytical methods used in the marbled murrelet population monitoring have been presented in detail elsewhere (Raphael et al. 2007). In this section, we document several clarifications and refinements of the methods and protocol described in that publication.

### Estimates of Population Size and Trend at State Scale

In this report, we include for the first time estimates of marbled murrelet population size and trend at the state scale, because this scale is relevant for evaluating conservation actions and regulations at that scale. We used the same analytic approach as described above, except that we calculated average annual murrelet densities for each of the three states within the sample area: Washington, Oregon, and California. We calculated average densities by weighting the murrelet density for each conservation zone, or portion thereof, within a state, by the area of coastal waters sampled within that conservation zone or portion of conservation zone. For Washington, this involved the weighted average density for Conservation Zones 1 and 2. The Oregon estimate averaged the density for Conservation Zone 3, and for the portion of Conservation Zone 4 within Oregon (PSUs 1 through 9); PSU 9 spans the Oregon-California border, but is predominately in Oregon. The California estimate averaged the density for the California portion of Conservation Zone 4 (PSUs 10 through 22) and all of Conservation Zone 5. Our California estimate does not include murrelets occurring in Conservation Zone 6 (south of the Golden Gate of San Francisco Bay), because Conservation Zone 6 is outside of the NFWP area, and thus is not sampled by this program.

### Treatment of Years with No Surveys in Conservation Zone 5

Conservation Zone 5 was not surveyed in 4 years: 2006, 2009, 2010, and 2012. We instituted measures to formalize treatment of missing Conservation Zone 5 data in our analyses, which have been applied to the entire data set. For regressions used to estimate trend for Conservation Zone 5, we use only data from years with surveys. For All-Zones population and density estimates and trend analyses, we used interpolation methods. When Conservation Zone 5 has been sampled both before and after the year without surveys (as is the case for all years in this report), we use mean of the prior and following year densities to estimate the missing year's density. If Conservation Zone 5 is not surveyed for 2 consecutive years, as occurred in 2009 -2010, we interpolate using the prior and following years with surveys. For example, for 2009 and 2010, we

estimated density ( $\hat{d}$ ) using 2008 and 2011 Conservation Zone 5 data and the following formula:

$$\hat{d}_{2009 \text{ (or 2010)}} = \hat{d}_{2008} + \frac{\hat{d}_{2011} - \hat{d}_{2008}}{3}$$

When Conservation Zone 5 was not surveyed in the last year of analysis period, we use data from the most recent prior year with Conservation Zone 5 surveys to extrapolate density for the missing data.

We also used the interpolated values for Conservation Zone 5 in our “All Zones” trend estimate. We estimated the “All-Zones” density and standard error of density using the following formulas, where  $a_z$  is the area of Conservation Zone  $Z$ :

$$\hat{d}_{All} = \frac{\sum_{z=1}^5 \hat{d}_z a_z}{\sum_{z=1}^5 a_z}$$

$$\hat{\sigma}_{All} = \frac{\sqrt{\sum_{z=1}^5 a_z^2 \hat{\sigma}_z^2}}{\sum_{z=1}^5 a_z}$$

#### Adjusted Boundary Separating Inshore and Offshore Subunits in Conservation Zone 5

In early 2005 we used the 2000-2004 data to review murrelet distribution as a function of distance from shore. This review indicated that most murrelets were observed within 1300 m of shore. As a result, we adjusted the location of the boundary separating the inshore and offshore subunits from 2000 m offshore to 1200 m offshore. By reducing the area of the inshore subunit while maintaining the same survey effort in that subunit, we increased survey effort to that area of higher density. Concurrently, the length of the offshore effort increased from about 6 km to about 9 km per PSU sample. The adjusted length of the offshore transect was calculated using the following formula (details in Raphael et al. 2007):

$$r = \frac{a_1}{a_2} \times \sqrt{\frac{d_1}{d_2}}$$

where the ratio ( $r$ ) of the optimal inshore to offshore transect length (which minimizes the variance of the PSU density estimator) is based on the mean densities in the two subunits ( $d_1$  and  $d_2$ ) and the area of the subunits ( $a_1$  and  $a_2$ ) when a Poisson distribution is assumed for the observed counts. Because the length of the inshore transects is fixed as the length of the PSU measured parallel to shore (about 20 km), the optimal ratio is determined by adjusting the length of the offshore transect.

These changes took effect with the 2005 surveys and were continued; the protocol allows such data-informed adjustments (Madsen et al. 1999 and Raphael et al. 2007).

This reallocation of sampling effort does not affect estimated densities and population sizes, but should reduce the confidence intervals associated with those estimates.

### Bootstrap Method Used to Construct Confidence Intervals

We have previously described the bootstrap method that we use for constructing 95 percent confidence intervals for density and population estimates at the conservation zone and strata scales (Raphael et al. 2007). Here, we provide additional details of the methods used, in particular we explain how surveys are grouped into “clusters”, and how those clusters of surveys are then sampled in the bootstrap process.

For a given conservation zone and year, the different PSU samples typically show some grouping in space and time. This results from the practical limitations and efficiencies of conducting surveys from of a limited number of coastal ports where survey vessels can be launched, compounded by bad weather limiting the days when surveys can be conducted. For example, PSU’s 3 and 4 in Conservation Zone 3, Stratum 1 might be surveyed on the same day. We need to account for the spatial and temporal dependence of these surveys when estimating confidence intervals. The estimates of  $E(s)$ ,  $f(0)$ , truncation distance, and density presented in this report and used in all other analyses are based on the original data as described in Raphael et al. (2007), and not on bootstrap estimates. While the bootstrap process results in estimates of parameters  $E(s)$ ,  $f(0)$ , truncation distance, and density, we used those estimates only to estimate confidence intervals.

These are the bootstrap analysis steps used to estimate the standard errors and confidence intervals, for each year and conservation zone:

1. Within each stratum of the Conservation Zone, we assign labels (“clusters”) to groups of surveys close in time and space for that year. “Close” is defined as being both within 3 PSU’s of each other spatially and surveyed within 4 or fewer days of each other temporally. This produces a set of  $n$  clusters for that stratum and year.
2. We then randomly select  $n$  clusters with replacement from that set of clusters. Sampling with replacement means that any cluster might be chosen more than once or not at all for a single bootstrap selection.
3. Suppose there are  $k$  surveys within a selected cluster. We then randomly select with replacement  $k$  surveys within the cluster.
4. All of the observations from the selected surveys in all strata are placed in one bootstrap-created data set which is then used to provide estimates of density,  $f(0)$ ,  $E(s)$ , and the truncation for the Conservation Zone.
5. This process is repeated 1,000 times for each Conservation Zone for a given year.
6. The standard errors of the estimates of density for each stratum and conservation zone, and for  $f(0)$ ,  $E(s)$ , and the truncation distance for each conservation zone are estimated using the standard deviations of the 1,000 bootstrap estimates. As noted above, the original data are used to estimate density,  $f(0)$ ,  $E(s)$ , and truncation distance, and the bootstrap process provides only the estimates of precision for those parameter estimates.

### Treatment of Abbreviated PSU Surveys

The target survey effort for a PSU was occasionally not achieved due to deteriorating weather conditions, resulting in an incomplete survey. In 2004, we clarified the treatment of incomplete PSU surveys, allowing for limited use of data from such surveys. For a given conservation zone in a single year, one but not both of the following cases of incomplete survey data would be allowed for each conservation zone:

- Data from up to 3 incomplete PSU samples could be used, providing that no more than 25 percent of the total transect length was missing from any PSU sample, and that no PSU would have more than one incomplete survey;
- or
- Data can be used from one PSU sample with up to 50 percent of either the total inshore or offshore segment length missing.

For any incomplete survey used, the survey length is adjusted in the analyses to match the actual transect length. Surveys not meeting the above criteria were discarded from all analyses.

In addition, effective in 2004, data for a single PSU sample must be collected within a single day. Prior to 2004, sampling effort for a single PSU sample was occasionally conducted over two days, with the inshore subunit sampled one day, and the offshore subunit sampled on a second day.

### Minimum Visibility Conditions for Conducting Surveys

Effective since 2011, we adopted a rule that surveys would be conducted only in conditions when surveyors can see a murrelet at 150 m. Murrelets beyond this distance have little effect on density or population estimates, in part due to the truncation which occurs in program Distance. Previously, the minimal visibility distance was not standardized, and varied from 100 to 200 meters, depending on conservation zone.

### Comprehensive Review of Data

In 2014, we developed and implemented a new, automated procedure to screen all data from 2000 through 2013, as an improved data quality assurance process. This improved our ability to detect potential data inconsistencies, such as might have occurred during data entry or transcription by the different field crews and data managers, and employs cross-referencing between and within database fields, as well as screening for values which are outside of the range of values normally observed for a given data field. Each problematic data line identified by this process was manually reviewed by the individual(s) responsible for data maintenance for each conservation zone, and original field data forms and records consulted as needed. We corrected any errors found, and created a new database which was the basis for all population density and trend analyses presented in this report. While the corrections represent a very small percentage of data records, it did affect several years, and some density and

trend estimates presented here differ slightly from previous versions, including those in the program's 2013 annual data summary (Falxa et al. 2014).

### Field Audit form

As part of the field observer training, the methods (pg. 12 *in* Raphael et al. 2007) call for one of the crew supervisors for a given zone to accompany survey crews three times during the survey season as an audit to observe their overall performance and ability to detect murrelets. To assist in conducting audits of crews, we developed a field audit form (Appendix 1-B). The survey leader for each Conservation Zone conducted audits of crews in their zone each season, and the monitoring program coordinator (Falxa) audited crews from the different zones periodically to evaluate for consistency in protocol implementation across crews and conservation zones. In addition to helping maintain consistency with the protocol and among crews, audits led to clarifications, including the minimum visibility rule discussed above.

### Changes in Conservation Zone Leads for Population Surveys

In addition to the above refinements and clarifications, the responsibility for data collection has changed for some conservation zones since our last report. In 2013, the Washington Department of Fish and Wildlife assumed lead for conducting population surveys in Conservation Zone 1; until that year, researchers with the US Forest Service's Pacific Northwest Research Station conducted the Conservation Zone 1 sampling. In Conservation Zones 4 and 5, Crescent Coastal Research assumed responsibility for all surveys in 2010. Previously, researchers from the US Forest Service's Pacific Southwest Research Station had led surveys in the California portion of Conservation Zone 4, as well as contributing to data collection in Conservation Zone 5. Currently, the Washington Department of Fish and Wildlife conducts all surveys in Conservation Zones 1 and 2, and Crescent Coastal Research conducts all surveys in Conservation Zones 3, 4 and 5.

Finally, effective 2014, a decision was made by agency managers to implement a "contingency plan" due to budget restrictions, which reduced sampling effort to once every 2 years rather than annually. Conservation Zones 1 and 3 would be sampled in even-numbered years, and Conservation Zones 2 and 4 in odd-numbered years. Conservation Zone 5 would be sampled every four years, during years when Conservation Zone 4 is sampled. This plan was partially implemented in 2014, when Conservation Zone 4 was not sampled, and Conservation Zone 2, instead of being skipped, was sampled because funding was available.

## **RESULTS**

### **Population Estimates**

Estimates of density and population size by conservation zone and for all conservation zones are presented by year in Table 1-2. Among conservation zones, murrelet density

varied greatly, from less than 0.1 murrelets per km<sup>2</sup> in Conservation Zone 5 to greater than 5 murrelets per km<sup>2</sup> in Conservation Zone 4 (Table 1-2). Based on these densities, our most recent (2013) population size estimates at the conservation zone scale ranged from about 71 murrelets in Conservation Zone 5 to 7,880 murrelets in Conservation Zone 3 (Table 1-2). Conservation Zones 1 and 3 had the two highest population estimates in all years except 2008 and 2013, when Conservation Zones 3 and 4 had the highest estimates (Table 1-2). Conservation Zone 5 supported far fewer murrelets than any other conservation zone, with population estimates never exceeding 300 murrelets.

Because population estimates are the product of both density and area of coastal waters sampled, density patterns at the conservation zone scale did not closely track population estimates across conservation zones. For example, although Conservation Zone 3 had the largest single-conservation zone estimate of population size in 2013, murrelet density was slightly higher in Conservation Zone 4, which has a sample area that is 30 percent smaller than Conservation Zone 3. Sample area size also contributes to the high murrelet numbers in Conservation Zone 1, which encompasses an area of sampled coastal waters (about 3,500 km<sup>2</sup>) more than double that of the next largest Conservation Zone (about 1,650 km<sup>2</sup>, Conservation Zone 2).

At the All-Zones scale, the mean murrelet density ranged from 1.89 per km<sup>2</sup> (2010) to 2.56 murrelets per km<sup>2</sup> (2002; Table 1-2). Population estimates at this scale varied from 16,600 in 2010 to 22,800 in 2003 (Table 1-2, Figure 1-3). From 2011 through 2013, the All-Zones population estimates were higher than observed since 2005. These higher estimates reflect higher population estimates in Conservation Zone 1 in 2011-2012, in addition to high Conservation Zones 3 and 4 estimates in 2011 and 2013 (Table 1-2).

At the scale of individual states, average density was markedly higher off the coast of Oregon, where density was about four murrelets per km<sup>2</sup> in most years, compared to densities about half this in Washington and California (Table 1-3). California supported fewer than half the number of murrelets estimated for the other two states (Table 1-3); this does not include the small, isolated population in central California (Henry et al. 2012), which is outside of the area monitored under the NWFP. Population sizes for both Oregon and Washington were fairly similar (Table 1-3), but were more variable among years in Washington (Oregon mean: 7,874 murrelets, coefficient of variation: 13.8 percent; Washington mean: 8,798, coefficient of variation: 24.9 percent).

At a finer scale, the average density over the years of this study varied among PSUs. Some of the observed variation mirrored general density patterns between Conservation Zones, such that all 15 PSUs in Conservation Zone 5 had low average density. Elsewhere, average density among PSUs within a given stratum or conservation zone displayed variation by as much as 10 times or even more in some cases (Figure 1-4).

Note that estimates at the stratum scale, with the distance estimation parameters  $f(0)$ ,  $E(s)$ , and truncation distance are presented in Appendix 1-C.

## Trend analyses

### *Population Trends*

Estimated rates of linear annual change at the scales of the individual conservation zones and State-scale are presented in Figures 1-5 and 1-6, where the estimated rate of linear annual change is shown relative to zero (no change) and the overlap of the 95 percent confidence intervals with zero indicates where the evidence is stronger (no or minimal overlap of 0) or not (extensive overlap with 0). In Conservation Zone 1, the data indicate a linear declining trend of 3.9 percent per year (95 percent confidence interval: -7.6 to 0.0). The data also provide strong evidence for a linear decline in Conservation Zone 2 (6.7 percent decline per year; 95 percent confidence interval: -11.4 to -1.8) (Table 1-4, Figures 1-5 and 1-7a). Assuming a constant rate of decline, these rates would translate to cumulative population declines over a ten-year period of about 33 percent in Conservation Zone 1 and 50 percent in Conservation Zone 2, based on an exponential model of population change (Table 1-4). We found no evidence of linear trends in Conservation Zones 3 and 5. In Conservation Zone 4, no trend was detected but the evidence was not conclusive; the trend estimate was above zero and the confidence interval for the estimate overlapped zero (1.5 percent per year, 95 percent confidence interval: -0.9 to 4.0) (Figure 1-5, Table 1-4).

No trend was detected for the combined “all-zones” 5-conservation zone area for the 2001-2013 period. While the trend estimate was below zero, the evidence was not conclusive because the estimate’s 95 percent confidence interval overlapped zero (-1.2 percent per year; 95 percent confidence interval: -2.9 to 0.5) (Figure 1-5, Table 1-4).

At the scale of the three states (Figure 1-6, Table 1-4), the murrelet population in Washington experienced an estimated average annual rate of decline of -4.6 percent (95 percent confidence interval: -7.5 to -1.5), with the data providing strong evidence for a declining linear trend for the 2001 to 2013 period (Figure 1-6). There was no evidence for a population decline or increase in Oregon (+0.3 percent per year; 95 percent confidence interval: -1.8 to 2.5). For California, no trend was detected; while the trend estimate was positive, the evidence for a trend was not conclusive because the 95 percent confidence interval overlapped zero (+2.5 percent per year; 95 percent confidence interval: -1.1 to 6.2).

### **Effects of Sea Conditions on Density and Trend Estimates**

The no-covariate model had the smallest AICc value for 61 percent of the 51 conservation zone-year combinations provided by our analysis. The covariate model (with Beaufort sea state) had the smallest AICc value in the remaining 39 percent (n=20 cases). Of these covariate models with lower AICc values, the difference in AICc values between the Beaufort and no-covariate model was less than 4.0 in 9 of the 20 cases, suggesting that both models were competitive in these cases.

While the AICc values do not support a consistent best single approach to be applied to all conservation zones and years, the resulting Beaufort and no covariate density

estimates do not show large differences relative to the standard errors for the density estimates (Figure 1-8).

Based on the results of this analysis, we are continuing the method of using “no-covariate” models to estimate murrelet population density and trends throughout this report, but will evaluate alternatives for future analyses. One such alternative would be to make a separate decision on which model to use for every year and conservation zone combination, based on AICc values.

### **Power to Detect Trends**

One measure for assessing the effectiveness of the monitoring design is its power to detect linear trends in the density (and the correlated population size) of murrelets over time. Detailed results are provided in Appendix 1-A, where Tables 1-A-1 and 1-A-2 present the estimated calendar year when sampling will have been sufficient to detect a population trend for two levels of power. These power numbers measure the ability of the sampling design to detect a trend of a specified magnitude or greater, and are presented for a range of rates of annual decrease. This power analysis is based on annual density estimates through 2013, and forecasts the program’s ability to detect trends under a reduced sampling effort beginning in 2014.

For a given level of power, fewer years of sampling are required to detect a larger decline than a smaller decline. For example, the power analysis indicates that sampling through 2013 has already been sufficient to test for a 5 percent decline with 80 percent power in three of five conservation zones. In comparison, for every zone, and with a sampling frequency of every-other year, sampling would need to continue through 2020 (Conservation Zone 3) or later to test for a two percent decline with 80 percent power. The power to detect a given trend varies among zones, and one factor influencing power appears to be murrelet density. The number of years required to detect a decline was inversely related to average density for a conservation zone (Figure 1-9; linear regression  $R^2 = 0.79$ ,  $P=0.04$ , for detecting a 2 percent decline with 80 percent power,  $n=5$ ). This is consistent with the general pattern of greater variability for estimates at smaller spatial scales (standard error for single conservation zone versus for all conservation zones combined), and for conservation zones with lower density (Table 1-2). For example, Conservation Zone 5 supports the lowest murrelet density of the five conservation zones, and requires the largest number of years to achieve a given level of statistical power (Tables 1-A-1 and 1-A-2).

### **Temporal and Spatial Variation in Marbled Murrelet Distribution as a Function of Distance from Shore**

To minimize the variance in our overall murrelet density estimate, we devoted more sampling effort in the nearshore region where, based on preliminary data, murrelet density was higher (Bentivoglio et al. 2002; Raphael et al. 2007). Comparing density ratios in Figure 1-10, we see that our assumption of greater nearshore density is supported in nearly all year/strata/ zone combinations. The ratios shown represent 108

unique zone-stratum-year combinations, of which only 4 are clearly below 1.0 and two are approximately 1; the mean ratio at the stratum scale (averaged over all strata and years) is 8.3. In other words, in only 4 percent of the cases did we observe a higher density in the offshore subunit. Even in Stratum 2 of Conservation Zone 2, where most of these low ratios are observed, the ratio is above 1.0 or  $\approx 1.0$  for the majority (10 of 13) of years. In addition, this particular strata has very low densities of murrelets and consequently has little influence on the precision of the range-wide population estimates and trend.

In Figure 1-10, we present the inshore to offshore murrelet density ratio for each year-conservation zone-stratum combination. When examining these ratios for all eight conservation zone-strata combinations in Figure 1-10, we see no pattern to these ratios over time in most conservation zones. In Conservation Zone 4 there is some evidence for a declining ratio between 2003 and 2013 for both strata, but this pattern breaks down for the full time series (linear regression, Stratum 1:  $R^2 = 0.16$ ,  $P=0.15$ ; Stratum 2:  $R^2 = 0.09$ ,  $P=0.29$ ).

## **DISCUSSION**

This report provides the third evaluation of murrelet population status and trends, following previous reports associated with 10 years (Miller et al. 2006) and 15 years (Miller et al. 2012; Davis et al. 2011) of NWFP implementation. The new analyses reported here indicate that marbled murrelet population numbers vary over space and time throughout the NWFP area, and evaluation at a finer-scale than the NWFP area is informative for conservation purposes. Such variation is not surprising given that the factors affecting murrelet density and trend are expected to vary across the NWFP area, which encompasses about 11 degrees of latitude. Previously, the number of years in our sample size limited our ability (statistical power) to test for population trends at the conservation zone scale, thus limiting our interpretations to the NWFP-wide scale.

At the conservation zone scale, as observed in previous reports, murrelet density and abundance varied widely among conservation zones, with the most recent (2013) population estimates ranging from about 71 murrelets in Conservation Zone 5 to 7,880 murrelets in Conservation Zone 3. Differences among conservation zone population estimates are a result of variation of both murrelet density and the area of marine coastal waters being sampled. The density estimates ranged from to 0.1 murrelets per  $\text{km}^2$  in Conservation Zone 5 to 5.2 murrelets per  $\text{km}^2$  in Conservation Zone 4. The target population for Conservation Zone 1 inhabits an area of marine waters double that of any other zone, due to the large area of marine waters associated with the complex shorelines of the Puget Sound and the San Juan Islands. Annual population estimates for the entire NWFP area ranged from about 16,600 to 22,700 murrelets during the 14-year period, and included the lowest population estimate to date, for 2010.

At the 10-year interval, which followed the first four years of monitoring for the Northwest Forest Plan, we did not detect a decline in murrelet densities (Miller et al. 2006). At the 15-year interval, which included monitoring results from 2001 through 2010, we estimated a decline of about 3.7 percent per year at the spatial scale of the

five conservation zones combined (Miller et al. 2012). With the addition of estimates through 2013, the evidence for a trend is no longer conclusive at this scale.

With more years of data, the murrelet monitoring program had greater power to evaluate for trends at the scale of the five conservation zones. At that scale, we now observe strong evidence of linear population declines in Conservation Zones 1 and 2, and no trend in Conservation Zones 3 and 5. In Conservation Zone 4, which includes southern Oregon and northern California, the trend estimate was positive but with confidence intervals for the estimate overlapping zero, the evidence for a trend was not conclusive, and we concluded that no trend was detected for the 2000 to 2013 period.

For the first time, we evaluated for trends in the annual population estimates at the state scale (Table 1-3, Figure 1-6). We found strong evidence of an annual 4 to 5 percent linear decline in Washington ( $P < 0.01$ ). No evidence of a linear trend was found for the murrelet in Oregon. In California, as for Conservation Zone 4, no trend was detected; while the trend estimates were positive, the evidence for a trend was not conclusive based on the magnitude of the annual rate estimate and the overlap of the confidence intervals with zero (representing no change). It is worth noting that for both Conservation Zone 4 and California, if the pattern of population estimates continues into the future, an increasing trend may be detectable.

With more years of data, we have found that population estimates can vary markedly between years, particularly at the conservation zone scale, with annual estimates being above or below the average trend line. This variability, combined with reduced future sampling effort, contributes to the relatively large number of years of sampling required to confidently test for lower rates of change at the conservation zone scale.

## **Trend Pattern**

In Washington, there is strong evidence for a declining trend in both Conservation Zones 1 and 2. In northern and central Oregon, as well as at the Oregon state-level, we observed no evidence for a trend. In California (Conservation Zone 5 plus the California portion of Conservation Zone 4), the evidence for a trend was not conclusive, but the trend estimate was positive. For the entire Conservation Zone 4, which spans northern California and southern Oregon, the trend analysis indicated a non-significant positive slope. These trend results are suggestive of north-south trend pattern, with clear declines to the north, relatively stable populations in the middle (Oregon) and stable to the south in California, but with trend estimate values greater than zero. Results of an analysis of factors contributing to variability in murrelet distribution and trends are presented and discussed in chapter 3 (Raphael et al., this volume). Further analysis of the factors contributing to variability in murrelet distribution and population trends is merited, but the results to date suggest that habitat loss is likely contributing to the trend pattern.

One analytic approach for distinguishing factors that contribute to population trends, and which we plan to pursue in future trend analyses, is the use of state-space models (Humbert et al. 2009; Kery and Schaub 2012). These models have the potential to better separate year to year variation due to sampling (observation) error from variation

due to biological processes, such as a population trend or environmental variability. State-space models generally require more years of data than the linear regression method we have used to date, but the monitoring program's sample size should soon be sufficient to merit use of these models (P. Lukacs, and E. Ward, personal communications).

The magnitude and strength of evidence for a NWFP-wide decline have decreased relative to our previous assessments. This difference may be driven by a variety of factors, most notable being the higher population estimates for 2011 and 2012, specifically in Conservation Zone 1, and especially in Stratum 1 of that conservation zone (Strait of Juan de Fuca). In 2011, estimates of murrelet population size increased in all conservation zones except Conservation Zone 2, compared to estimates from previous years. For Conservation Zone 1, magnitude of increase in 2011 was such that the 95 percent confidence intervals for the 2010 and 2011 estimates do not overlap each other. In 2012, the All-Zones estimate remained higher, as did the Zone 1 estimate (Figure 1-7a; Table 1-2, Appendix 1-C). In 2013, the All-Zones population estimate was lower than in 2011 or 2012 and the population estimate in Conservation Zone 1 declined from the 2011-2012 levels, while the Conservation Zone 4 estimate was the highest in 14 years, and the Conservation Zone 3 estimate was the second highest (Table 1-2). The reasons for the pattern observed in 2011 to 2013 in Conservation Zones 1 and 4 are unknown, but we discuss some potential causes below.

### **Sampling and Interpretation Challenges**

The challenges of accurately sampling such a mobile and patchily distributed species, and associated uncertainty in density estimates, could have contributed to the increased estimates in recent years, as could other factors. Results of murrelet population monitoring in 2014 and beyond will help clarify population status and trend, as will data explorations underway. For the latter, we have identified several topics to explore:

*Has the distribution of marbled murrelets relative to distance from shore changed?*

Specifically, did murrelet distribution shift closer to shore in 2011-2013, such that murrelets previously too far offshore to be within our sampling areas moved closer in those years, to put them within the sampled area? We would expect such a shift to be reflected in higher ratios of the inshore to offshore density. Our evaluation of this ratio (Figure 1-10) did not find values outside of ratios observed in prior years. The only exception was Conservation Zone 4 Stratum 2 in 2012; however, this is an area with relatively few murrelets which contributes little to the observed increases for all conservation zones combined.

*Do any of the parameters used to estimate density differ in 2011-2012 from previous years?* Parameters of interest include the probability density function of detection distances [ $f(0)$  in DISTANCE], the mean number of murrelets per murrelet group detected [ $E(s)$  in DISTANCE], and the encounter rate of murrelets during surveys. Data inspection indicate that  $f(0)$  and  $E(s)$  did not differ markedly in 2011-2013 from prior years in any conservation zone (Appendix 1-C), but encounter rates did increase in 2011-2013. Higher encounter rates are consistent with higher murrelet densities, while

average group size and the density function for detection distances remained similar to other years.

*Could the distribution of marbled murrelets within Conservation Zone 1 have shifted from unsampled PSUs to sampled PSUs?* Conservation Zone 1 differs from other conservation zones in having only a subset of PSUs sampled each year, with the same PSUs sampled based on an initial random sample in 2000 (Raphael et al. 2007). Changes in density estimates in Stratum 1 of Conservation Zone 1 contributed heavily to the 2011-2012 increases in Conservation Zone 1 estimates. Because not all PSUs are surveyed in this stratum, a movement of murrelets between PSUs could contribute to an increase in estimates. Lacking data from the unsampled areas, this question is difficult to evaluate.

The removal of two large dams on the Elwha River between September 2011 and August 2014 generated large sediment plumes within Stratum 1 of Conservation Zone 1 during the survey season, and might also have influenced the distribution and abundance of murrelets in this area. Removal of Elwha Dam began in September 2011 and the final segment of Glines Canyon Dam was removed in August 2014. Conceivably, the sediment input from dam removal could have created a foraging opportunity that attracted murrelets from other PSUs or strata. However, we first observed an increase in murrelet density in Stratum 1 of Conservation Zone 1 in the 2011 survey, months prior to the September 2011 Elwha Dam removal (Appendix 1-C).

*Could the distribution of marbled murrelets have shifted from areas outside of Conservation Zone 1 into Conservation Zone 1 during the 2011 to 2013 period?* With respect to movements from south of Conservation Zone 1, our data do not support a systematic shift that would account for the increase during this period (Table 1-2, Figure 1-7a). In 2011, the first year of the observed increase in Conservation Zone 1, estimates for other conservation zones were fairly stable (Conservation Zones 2 and 3) or increased (Conservation Zone 4) compared to 2010 (Table 1-2), not providing any evidence for movement from the south between 2010 and 2011. In Conservation Zone 2, which is adjacent to Conservation Zone 1, murrelet densities were very similar from 2009 through 2013. Our murrelet density estimates from Conservation Zone 3 increased or were stable in 2010, 2011 and 2013 compared to the several years prior to 2010 (Figure 1-7a), which is not consistent with movement of murrelets from this zone into Conservation Zone 1. Similarly, in Conservation Zone 4, density estimates in 2011 through 2013 were equal or greater than densities in 2009-2010. Numbers of murrelets decreased in 2012 (when Conservation Zone 1 numbers increased) in Conservation Zones 3 and 4, compared to 2011 and 2013. While between-zone movements could have contributed to annual variations observed within this 3-year period, they do not explain the overall increase observed in Conservation Zone 1 during the 2011-2012 period, compared to previous years (Table 1-2, Figure 1-7a).

Murrelets could have moved from the north into our sample area, such as across the Strait of Juan de Fuca from British Columbia to northern Washington. Comparable regional results are not available from British Columbia. However, available data for evaluating this possibility are limited to a single long-term at-sea sampling effort from about 100 km of transects on the southwest coast of Vancouver Island during May to July. Results from this effort suggest a marked increase in murrelet numbers during the

2001 to 2013 period, especially during the years 2010, 2011, and 2013 compared to previous years (Y. Zharikov, pers. comm.). The data from this small area, which is in part on the Strait of Juan de Fuca, are not consistent with a marked emigration of murrelets out of their study area. It remains possible that the influx of murrelets in Conservation Zone 1 came from areas even further north in British Columbia or Alaska.

*Were fewer murrelets breeding in 2011 and 2012, thus more adult murrelets on the water versus at nest sites?* During the approximate 29-day incubation period and first 1-2 days after hatching, one of the murrelet adults is present on the nest (Nelson 1997), and thus is not available to be counted at-sea. The incubation period (Nelson 1997) closely overlaps our sampling period, thus, if fewer murrelets attempted nesting, we would expect to detect more murrelets. The ratio of hatch-year to after-hatch-year murrelets observed on surveys during the fledging period is one measure of reproductive success (Peery et al. 2007). This ratio, as well as numbers of hatch-year murrelets counted around the San Juan Islands in 2011-2012 were comparable to or greater than numbers in other years (M. Raphael and T. Bloxton, unpublished data), which is not consistent with fewer murrelets nesting for the murrelet population associated with the waters around the San Juan Islands, in Conservation Zone 1. Also, anecdotal observations from the California Current System, which includes western Washington, and peripherally the western portion of Conservation Zone 1, indicate good years for marine productivity and forage fish during 2011-2013 (Peterson et al. 2013), which would tend to lead to more murrelets nesting (Peery et al. 2004). Indicative of this, in Conservation Zone 3, hatch-year murrelets occurred at near-average densities in 2011 and 2012, and at anomalously high densities in 2013 (Strong 2014). This would not support a lower proportion of murrelets breeding, at least within the California Current System.

*Potential effects of high rates of nesting success and recruitment, and/or early fledging.* High rates of nesting success, particularly if combined with earlier fledging than normal, could potentially result in higher densities of hatch-year murrelets. Murrelet fledgling numbers at sea typically peaks in late July to August (Nelson 1997), which only slightly overlaps our sampling, which extends from 15 May to 31 July. As noted above, marine productivity was good in 2011 to 2013, at least in the California Current System, and good ocean productivity and prey quality tend to be associated with greater nesting success and recruitment in marbled murrelets (Becker and Beissinger 2006; Becker et al. 2007; Norris et al. 2007).

*Could habitat change have caused the observed pattern?* Murrelet nesting habitat takes many decades to several centuries to develop (USFWS 1997), thus is a process too slow to account for the rapid increase in density estimates observed in a period of less than 10 years.

### **Potential Uncertainties in Sampling**

We reviewed several sources of potential bias that could affect our observed trends. For example, we anticipated a seasonal increase in murrelet density during the sampling period as chicks hatched and incubating murrelets returned to the water to forage, making only short flights inland to feed chicks (Peery et al. 2007). However, examination of the data early in the monitoring program (Miller et al. 2006) did not find a

temporal trend within a season. This lack of a seasonal trend may be due to a variety of factors, including a small proportion of murrelets breeding in any one year or the early return to the ocean of breeders whose nests failed during incubation. Our objectives in this study were to estimate the average number of murrelets in our target area between 15 May and 31 July and to be able to detect trends in those estimates. Our sampling design, which distributed sampling effort consistently through this period, allows us to meet our objectives even if the number of murrelets on the water during the sampling period was not constant.

More sampling effort was devoted to the inshore subunit, where data from previous work had shown densities to be highest. Based on average densities since 2000, densities in the inshore subunit were typically higher, and on average about eight times higher than in the offshore subunit. Densities were higher in the offshore subunit in only 4 of 108 stratum-year combinations sampled. Three of these year-stratum combinations were clustered in the southern half of Conservation Zone 2, near the Columbia River estuary and plume, as well as being in an area where the continental shelf is broad. Perhaps murrelet distribution in this area extends in some years further offshore than assumed. Murrelets can occur farther offshore where shallow waters and islets extend farther offshore (Ralph and Miller 1995, Speich and Wahl 1995, Strong et al. 1995, Raphael et al. 1999). If so, it should not influence our trend results for the target population, but could mean that in this stratum the area of coastal waters sampled represents less than the design target of 95 percent of the local population. The effect on our population estimates should be relatively small, as this area has very low densities of murrelets (Figure 1-4). A long-term (multi-year), systematic shift in murrelet distribution toward further offshore could affect our ability to assess population trends, if it resulted in a substantial change in the proportion of murrelets occurring beyond the waters sampled. We evaluated the annual density estimates for each conservation zone for evidence of a trend since 2000-2001 of murrelet distribution shifting further offshore, and found no evidence of such a pattern to date. In the future, we will evaluate whether the data collected by the monitoring program can be used to explicitly evaluate the protocol's assumption that coastal waters sampled encompass at least 95 percent of the local population during the sample period.

Other studies have found year, observer, and sea-state effects on detectability and at-sea density estimates for murrelets (Ronconi and Burger 2009). These factors, if not accounted for, can potentially increase error in our estimates, and thus reduce the power to detect trends. Our trend analyses explicitly accounted for year effects. We assessed observer (crew) effects in an earlier analysis of Conservation Zone 1 data (Miller et al. 2012) and found no observer effect on density or trend estimates, which may be reflective of our training efforts and low crew turnover. Our analysis of sea-state demonstrated a relatively slight effect on density and trend estimates, and no effect that would change our conclusion about the direction or magnitude of any trend. Sea-state effects are in part reduced by our protocol which precludes surveys during poor sea conditions. . While we did not include a sea state covariate in the analyses in this report, we will continue to evaluate the influence of sea state on murrelet detection and ultimately on trend estimates.

Given the goals of the NWFP and the monitoring program, ideally, any population trends we observe through monitoring should reflect changes in nesting habitat

conditions within the NWFP area. However, biological systems are rarely closed, particularly when defined by political boundaries. It is likely that there is some movement of murrelets between the northern portion of our sampling area and Canadian waters to the north. Suitable nesting habitat continues north from Washington into British Columbia, both on the mainland and Vancouver Island, and such movements have been observed. In a telemetry study, Raphael and colleagues (Bloxtton and Raphael 2009; unpublished data) recorded movements between U.S. waters and nesting sites on nearby Vancouver Island but no long distance movements consistent with individuals shifting their distribution from Washington to areas north of our study area, or vice-versa. Similarly, a telemetry study in northern California (Hebert and Golightly 2008) found that murrelets traveled less than 50km away from the mouth of the watershed where most nesting occurred (Hebert and Golightly 2008). However, these studies occurred during the breeding season, when nesting murrelets would be unlikely to make long shifts in location.

A northward shift of the murrelet's distribution from Washington into Canada could mimic the decline observed in Conservation Zone 1 (Puget Sound and Strait of Juan de Fuca) and could also affect trends in coastal Washington, Conservation Zone 2. However, we know of no evidence or causal mechanism for such a shift from 2001 to 2013 and the available data indicate that such a shift is unlikely. The murrelet's distribution at sea during the breeding season generally coincides with the distribution of potential nesting habitat directly inland (Burger 2002; Miller et al. 2002; Meyer et al. 2002; Raphael 2006; Raphael et al. 2002, 2015), suggesting that most murrelets observed on the water represent local breeding populations. A large northward population shift would suggest that breeding individuals are shifting nest locations, which is not supported by the limited information on nest-site fidelity. Nest-site fidelity is common in other alcids (Divoky and Horton 1995), and individual marbled murrelets have been observed re-nesting in the same stands and trees in successive years, suggesting some fidelity to nest areas (Hebert et al. 2003, Piatt et al. 2007). Also, population-trend data from British Columbia from the 1990s to 2006 do not support a shift from Washington waters to British Columbia, where there is some evidence for a decline during this period (Piatt et al. 2007). When examining the previously-mentioned yearly monitoring by Zharikov on the southwest coast of Vancouver Island in British Columbia, there is no evidence that murrelets are shifting between Conservation Zone 1 and southwest Vancouver Island during the monitoring period. A recent analysis of British Columbia murrelet population trends during 1996 through 2013, based on a radar-based monitoring program, found negative annual trends for two of the three sampling regions adjacent to Washington (East Vancouver Island and South Mainland Coast), and no trend in the third region (West and North Vancouver Island) (Bertram et al. in revision). Finally, Piatt et al. (2007) reported a substantial and continuing loss of likely murrelet nesting habitat on Vancouver Island and Haida Gwaii since the 1970s.

### *Potential Causes for Decline*

The NWFP population monitoring reported here does not address the causes of population trends. However, the other chapters in this volume report on companion analyses which provide some insight into potential causes. Chapter 2 (Raphael et al. this volume), documents the loss of murrelet nesting habitat within the area of the

NWFP, with the most acres of loss occurring in Washington. Chapter 3 (Raphael et al., this volume) uses the findings from murrelet population and nesting habitat monitoring to explore the relationships between quantity and quality of inland forest, prey availability, ocean conditions, and murrelet densities at sea in the NWFP area, building on a previous analysis (Raphael et al. 2015). As detailed in chapter 3, analysis of those relationships suggests that the amount and spatial pattern of nesting habitat, and changes therein, were the strongest predictors of murrelet numbers and trend in nearby marine waters.

In conclusion, this monitoring program provides population information, available nowhere else, on the status of marbled murrelets in the Northwest Forest Plan area, as well as being the only population information available to inform the species' recovery and ultimately delisting. A conservation goal of the NWFP is to stabilize and increase murrelet populations by maintaining and increasing nesting habitat. In this report, we address a primary question for evaluating the plan's effectiveness in achieving this goal during the first 20 years of NWFP implementation: *Is the marbled murrelet population stable or increasing?* Our findings indicate that the answer to this question is "no," the murrelet population associated with the NWFP area is not stable or increasing, at least not in Washington. We believe that the magnitude of the decline observed for Washington State and its two conservation zones, based on the 2001 to 2013 period, is sufficient to cause concern, and may merit a review of potential management implications and responses.

Management implications of results to date from the marbled murrelet effectiveness monitoring program are provided in detail in chapter 3 (Raphael et al. this volume). The trend pattern to date is of concern, particularly for Washington, where the murrelet population has not stabilized. Both the NWFP (FEMAT 1993) and the species' recovery plan (USFWS 1997) anticipated a challenge in maintaining murrelet populations for 50 to 200 years, until new nesting habitat develops. In light of observed population trends, our findings underscore the importance of the short-term goal to maintain existing nesting habitat. Long-term monitoring of murrelet populations and their environment, including nesting habitat, should reveal whether the NWFP meets its conservation goal of stabilizing and ultimately increasing marbled murrelet populations by maintaining and increasing nesting habitat. With long-term monitoring, we may also better understand the mechanisms underlying population change, and the degree to which population changes are due to nesting habitat conditions on the lands managed under the Northwest Forest Plan.

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**Table 1-1.** The number of marbled murrelet population monitoring primary sampling unit (PSU) surveys completed for the Northwest Forest Plan, and the total kilometers of survey transect sampled from 2000 to 2013.

Year	Zone	Number of PSU Surveys	Survey Effort (km)	Year	Zone	Number of PSU Surveys	Survey Effort (km)
2000	1	N/A	N/A	2007	1	60	2213
	2	N/A	N/A		2	31	1429
	3	24	1002		3	30	1151
	4	57	1493		4	29	750
	5	29	792		5	14	423
2001	1	60	2158	2008	1	60	2235
	2	22	1039		2	31	1441
	3	27	1067		3	30	1122
	4	54	1421		4	31	802
	5	22	602		5	13	385
2002	1	60	2228	2009	1	60	2230
	2	22	983		2	31	1380
	3	31	1239		3	31	1111
	4	56	1397		4	35	912
	5	26	705		5	No surveys	
2003	1	60	2210	2010	1	60	2246
	2	30	1359		2	30	1342
	3	30	1132		3	30	1169
	4	55	1418		4	26	676
	5	19	508		5	No surveys	
2004	1	57	2133	2011	1	60	2222
	2	30	1375		2	30	1356
	3	30	1188		3	31	1201
	4	32	836		4	32	813
	5	16	412		5	16	469
2005	1	60	2234	2012	1	60	2231
	2	26	1136		2	34	1567
	3	28	1108		3	29	1168
	4	31	812		4	27	702
	5	15	432		5	No surveys	
2006	1	60	2230	2013	1	60	2246
	2	29	1300		2	30	1361
	3	31	1185		3	29	1159
	4	30	776		4	31	808
	5	No surveys			5	15	454

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**Table 1-2.** Marbled murrelet population estimates, 2000 to 2013. Based on at-sea surveys conducted in Conservation Zones 1 through 5.

Year	Zone	Density	% Std. Err.	Murrelets	Murrelets		Area (km <sup>2</sup> )
		(birds per km <sup>2</sup> )			95% CL Lower	95% CL Upper	
2000	1&2	N/A	N/A	N/A	N/A	N/A	N/A
2000	3	4.13	18.6%	6,587	3,987	8,756	1,595
2000	4	4.22	30.9%	4,887	3,417	9,398	1,159
2000	5	0.09	80.6%	79	0	260	883
2000	All	N/A	N/A	N/A	N/A	N/A	N/A
2001	1	2.55	18.0%	8,936	5,740	11,896	3,501
2001	2	0.90	41.9%	1,518	524	2,942	1,688
2001	3	4.64	13.2%	7,396	5,230	9,075	1,595
2001	4	3.28	24.0%	3,807	2,983	6,425	1,159
2001	5	0.12	52.5%	106	27	244	883
2001	All	2.47	10.1%	21,763	17,472	26,053	8,826
2002	1	2.79	21.5%	9,758	5,954	14,149	3,501
2002	2	1.23	29.2%	2,031	800	3,132	1,650
2002	3	3.58	24.1%	5,716	3,674	9,563	1,595
2002	4	4.11	15.1%	4,766	3,272	6,106	1,159
2002	5	0.28	42.3%	249	27	400	883
2002	All	2.56	11.9%	22,521	17,264	27,777	8,788
2003	1	2.43	16.6%	8,495	5,795	11,211	3,498
2003	2	2.41	28.8%	3,972	2,384	6,589	1,650
2003	3	3.69	16.1%	5,881	3,992	7,542	1,595
2003	4	3.81	17.3%	4,412	3,488	6,495	1,159
2003	5	0.05	61.1%	48	0	85	883
2003	All	2.60	9.6%	22,808	18,525	27,091	8,786
2004	1	1.56	22.0%	5,465	2,921	7,527	3,498
2004	2	1.82	27.0%	3,009	1,669	4,634	1,650
2004	3	5.05	13.7%	8,058	5,369	9,819	1,595
2004	4	4.27	26.9%	4,952	3,791	9,021	1,159
2004	5	0.10	60.5%	88	18	214	883
2004	All	2.46	10.5%	21,572	17,144	26,000	8,786
2005	1	2.28	20.5%	7,956	4,900	11,288	3,497
2005	2	1.56	20.4%	2,576	1,675	3,729	1,650
2005	3	3.67	16.9%	5,854	3,580	7,447	1,595
2005	4	3.17	23.6%	3,673	2,740	6,095	1,159
2005	5	0.17	31.8%	149	69	251	883
2005	All	2.30	10.7%	20,209	15,976	24,442	8,785
2006	1	1.69	18.1%	5,899	4,211	8,242	3,497
2006	2	1.44	18.0%	2,381	1,702	3,433	1,650
2006	3	3.73	12.7%	5,953	4,546	7,617	1,595
2006	4	3.41	14.9%	3,953	3,164	5,525	1,159
2006	5	0.10	32.8%	89	35	150	883
2006	All	2.08	8.2%	18,275	15,336	21,214	8,785

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**Table 1-2** (continued). Marbled murrelet population estimates.

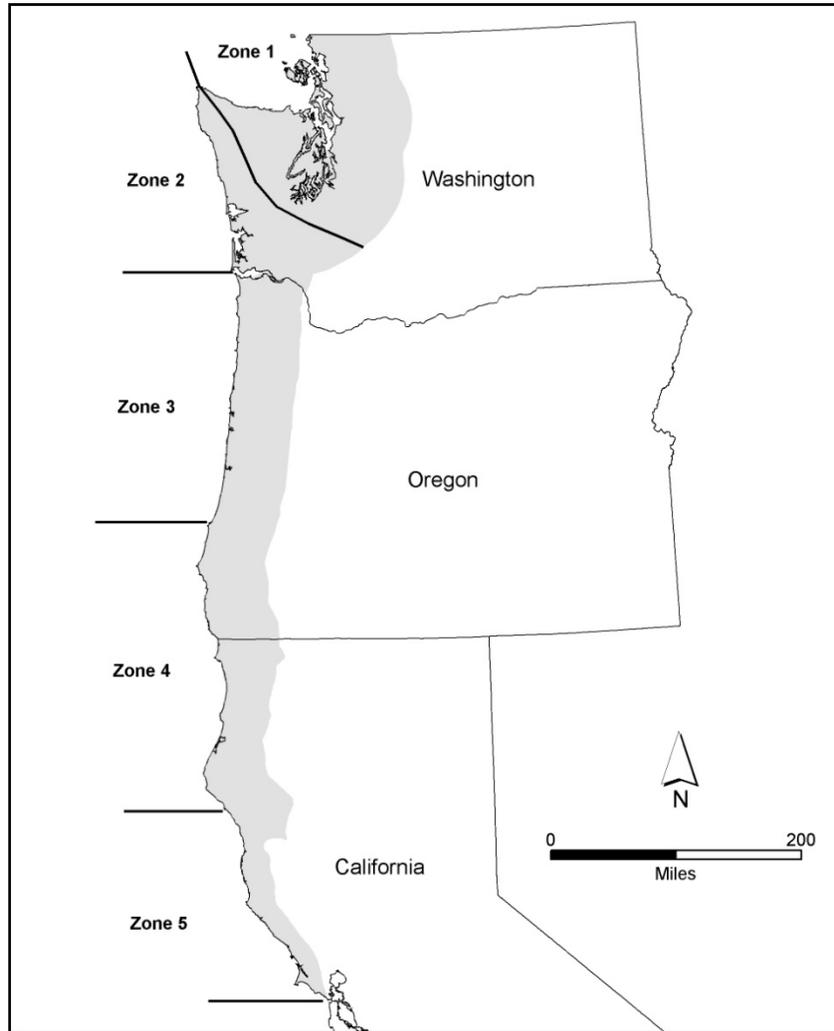
Year	Zone	Density	% Std. Err.	Murrelets	Murrelets 95% CL		Area (km <sup>2</sup> )
					Lower	Upper	
2007	1	2.00	24.2%	6,985	4,148	10,639	3,497
2007	2	1.54	26.7%	2,535	1,318	3,867	1,650
2007	3	2.52	19.8%	4,018	2,730	5,782	1,595
2007	4	3.23	34.8%	3,749	2,659	7,400	1,159
2007	5	0.03	37.7%	30	0	49	883
2007	All	1.97	13.7%	17,317	12,654	21,980	8,785
2008	1	1.34	17.6%	4,699	3,000	6,314	3,497
2008	2	1.17	22.1%	1,929	1,164	2,868	1,650
2008	3	3.86	14.7%	6,153	4,485	8,066	1,595
2008	4	4.56	17.9%	5,285	3,809	7,503	1,159
2008	5	0.08	48.1%	67	9	132	883
2008	All	2.06	8.9%	18,134	14,983	21,284	8,785
2009	1	1.61	21.2%	5,623	3,786	8,497	3,497
2009	2	0.77	21.9%	1,263	776	1,874	1,650
2009	3	3.70	17.7%	5,896	3,898	7,794	1,595
2009	4	3.79	19.9%	4,388	3,599	6,952	1,159
2009	5	0.10	50.6%	90	11	186	883
2009	All	1.96	10.6%	17,260	13,670	20,851	8,785
2010	1	1.26	20.0%	4,393	2,719	6,207	3,497
2010	2	0.78	25.5%	1,286	688	1,961	1,650
2010	3	4.50	16.7%	7,184	4,453	9,425	1,595
2010	4	3.16	28.5%	3,665	2,248	6,309	1,159
2010	5	0.13	52.1%	114	13	241	883
2010	All	1.89	11.1%	16,641	13,015	20,268	8,785
2011	1	2.06	17.4%	7,187	4,807	9,595	3,497
2011	2	0.72	33.4%	1,189	571	2,106	1,650
2011	3	4.66	16.3%	7,436	5,067	9,746	1,595
2011	4	5.20	34.9%	6,023	2,782	10,263	1,159
2011	5	0.16	53.0%	137	16	295	883
2011	All	2.50	12.6%	21,972	16,566	27,378	8,785
2012	1	2.41	20.7%	8,442	5,090	12,006	3,497
2012	2	0.72	33.5%	1,186	564	2,360	1,650
2012	3	3.99	15.5%	6,359	4,136	8,058	1,595
2012	4	4.28	24.9%	4,960	3,414	8,011	1,159
2012	5	0.12	50.4%	104	10	206	883
2012	All	2.40	11.4%	21,052	16,369	25,736	8,785
2013	1	1.26	27.9%	4,395	2,298	6,954	3,497
2013	2	0.77	18.5%	1,271	950	1,858	1,650
2013	3	4.94	16.3%	7,880	5,450	10,361	1,595
2013	4	5.22	20.5%	6,046	4,531	9,282	1,159
2013	5	0.08	45.4%	71	5	118	883
2013	All	2.24	11.1%	19,662	15,398	23,927	8,785

**Table 1-3.** Summary of 2000 to 2013 marbled murrelet density and population size estimates at the State scale.

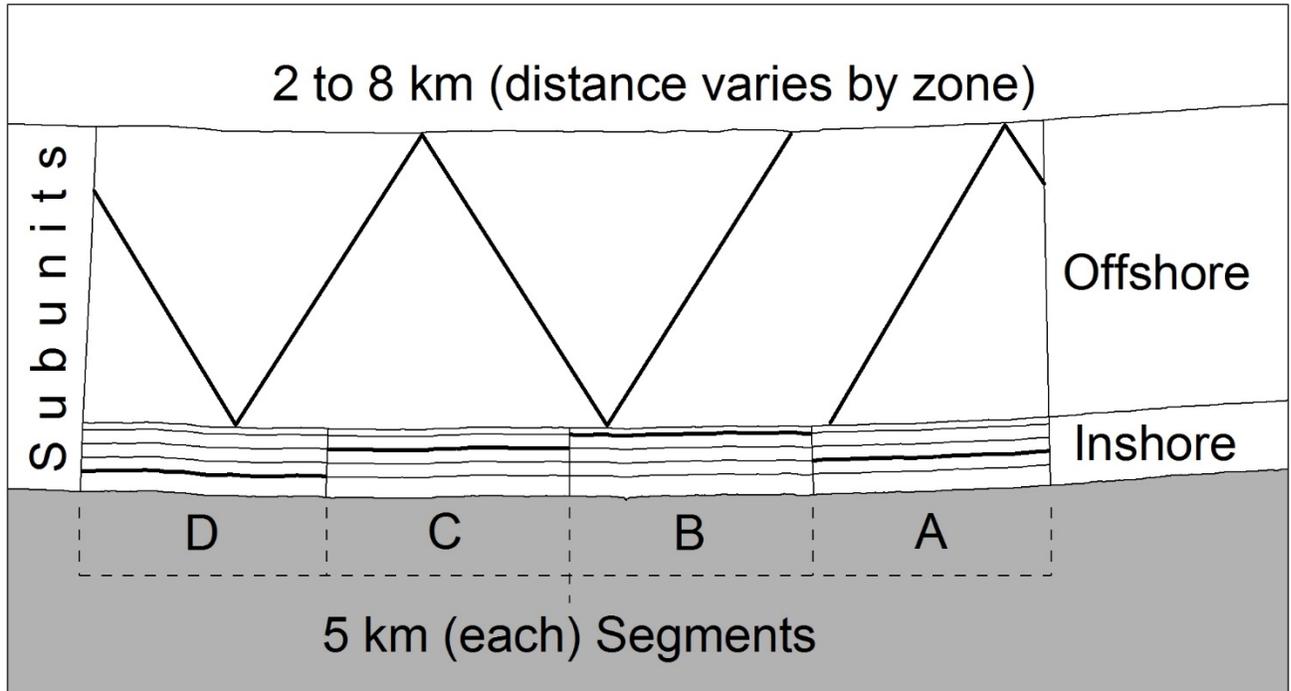
Year	State	Density (murrelets per km <sup>2</sup> )	Murrelets	Murrelets 95% CL Lower	Murrelets 95% CL Upper	Area (km <sup>2</sup> )
2001	WA	2.01	10,453	7,057	13,849	5,188
2002	WA	2.29	11,789	7,507	16,071	5,151
2003	WA	2.42	12,467	8,906	16,028	5,149
2004	WA	1.65	8,474	5,625	11,322	5,149
2005	WA	2.05	10,533	7,179	13,887	5,148
2006	WA	1.61	8,280	6,024	10,536	5,148
2007	WA	1.85	9,520	5,946	13,095	5,148
2008	WA	1.29	6,628	4,808	8,448	5,148
2009	WA	1.34	6,886	4,486	9,285	5,148
2010	WA	1.10	5,679	3,840	7,518	5,148
2011	WA	1.63	8,376	5,802	10,950	5,148
2012	WA	1.87	9,629	6,116	13,142	5,148
2013	WA	1.10	5,665	3,217	8,114	5,148
2000	OR	3.85	7,983	4,095	11,870	2,071
2001	OR	4.43	9,168	5,935	12,402	2,071
2002	OR	3.64	7,530	4,473	10,586	2,071
2003	OR	3.56	7,380	4,547	10,213	2,075
2004	OR	4.40	9,112	5,532	12,692	2,071
2005	OR	3.36	6,966	4,589	9,344	2,071
2006	OR	3.68	7,617	5,779	9,455	2,071
2007	OR	2.59	5,357	3,009	7,704	2,071
2008	OR	3.64	7,541	4,893	10,189	2,071
2009	OR	3.58	7,423	4,454	10,393	2,071
2010	OR	3.95	8,182	4,678	11,686	2,071
2011	OR	4.05	8,379	2,209	14,550	2,071
2012	OR	3.76	7,780	4,183	11,377	2,071
2013	OR	4.74	9,819	6,158	13,480	2,071
2000	CA	2.28	3,571	2,556	4,585	1,566
2001	CA	1.31	2,051	1,030	3,073	1,566
2002	CA	2.04	3,202	2,425	3,980	1,566
2003	CA	1.90	2,985	2,392	3,579	1,569
2004	CA	2.55	3,986	3,009	4,964	1,566
2005	CA	1.73	2,710	2,106	3,313	1,566
2006	CA	1.52	2,378	1,781	2,976	1,566
2007	CA	1.56	2,440	1,709	3,170	1,566
2008	CA	2.53	3,964	3,414	4,515	1,566
2009	CA	1.88	2,952	2,148	3,755	1,566
2010	CA	1.72	2,691	1,959	3,424	1,566
2011	CA	3.33	5,217	4,155	6,279	1,566
2012	CA	2.22	3,481	2,795	4,167	1,566
2013	CA	2.67	4,178	3,561	4,795	1,566

**Table 1-4.** Estimates of average annual rate of population change and cumulative population change for each conservation zone, for ‘All Zones’ combined, and for each state. Note that we used data from 2001-2013 for the Washington, and Conservation Zones 1 and 2 estimates, and for all others we used data from 2000 to 2013. All trends assume a constant (linear) annual rate of change; see text for details.

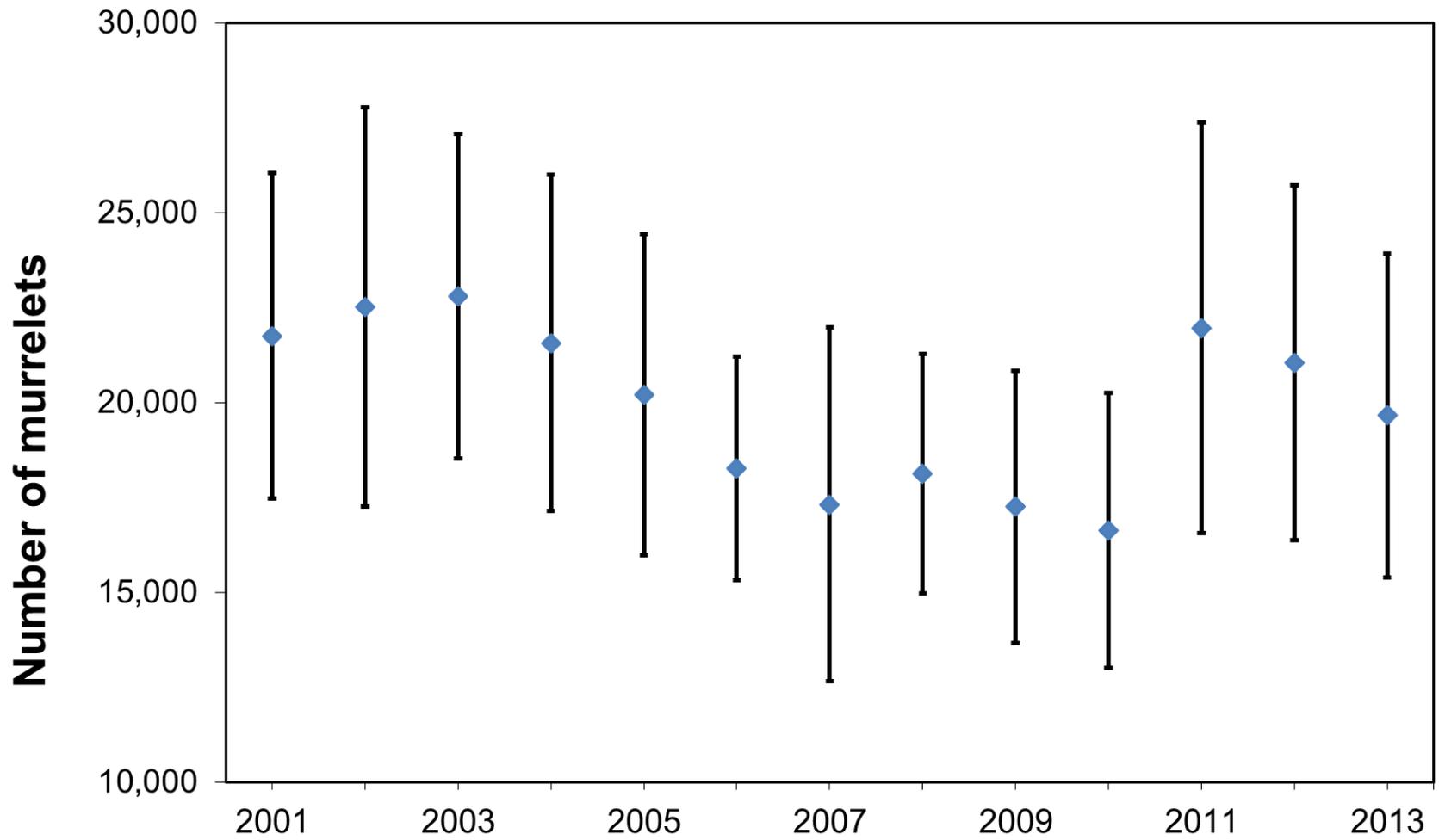
Conservation Zone	Annual Rate of Change (%)	95% Conf. Limits		Cumulative change over 10 years (%)	Adjusted $R^2$	P-value
		Lower	Upper			
All Zones	-1.2	-2.9	0.5	-11.3	0.099	0.156
1	-3.9	-7.6	0.0	-32.8	0.244	0.050
2	-6.7	-11.4	-1.8	-50.0	0.396	0.013
3	0.6	-2.1	3.3	+6.2	0.000	0.643
4	1.5	-0.9	4.0	+16.1	0.064	0.195
5	-1.0	-8.3	6.9	-9.6	0.000	0.785
Washington	-4.6	-7.5	-1.5	-37.6	0.449	0.007
Oregon	0.3	-1.8	2.5	+3.0	0.000	0.756
California	2.5	-1.1	6.2	+28.0	0.092	0.154



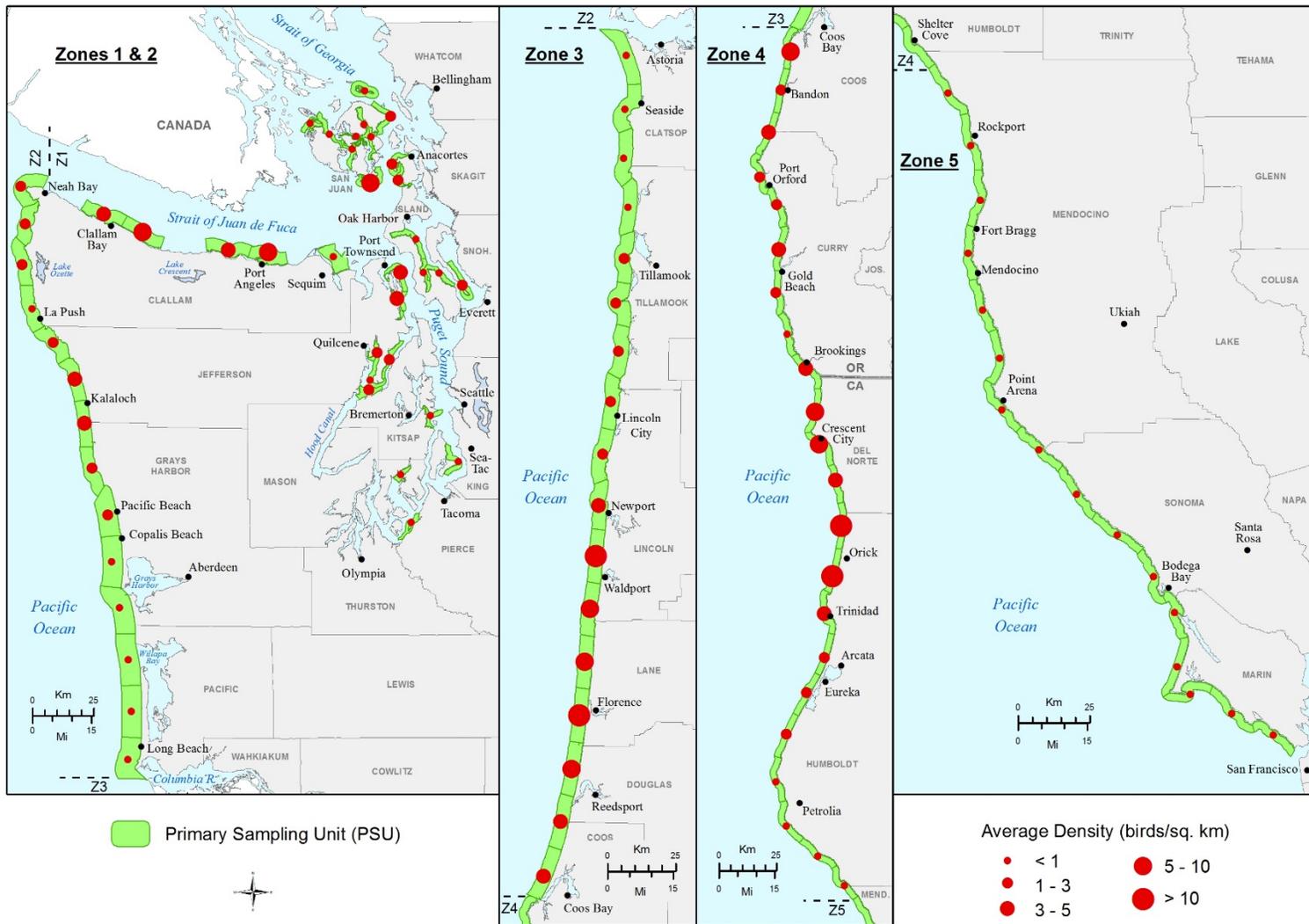
**Figure 1-1.** The five at-sea marbled murrelet survey (conservation) zones adjacent to the NWFP area. The shaded area corresponds to the overlap between the NWFP area and the breeding distribution of the murrelet. See Figure 1-4 for the offshore boundaries of the marine waters sampled (adapted from U.S. Fish and Wildlife Service 1997).



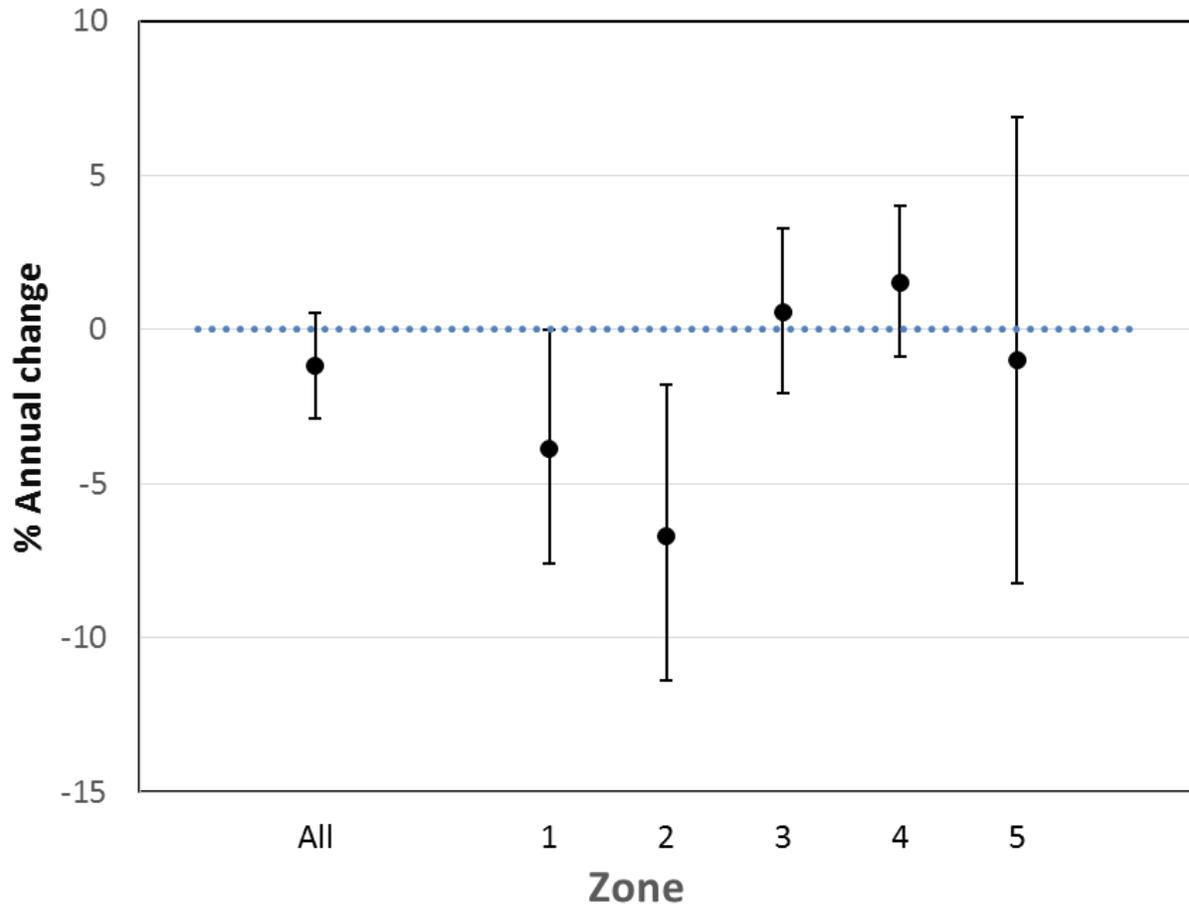
**Figure 1-2.** Marbled murrelet primary sampling unit with inshore and offshore subunits showing parallel and zigzag transects. The inshore subunit is divided into four equal-length segments (approximately 5 km each) and four equal-width bins (bands parallel to and at increasing distances from shore). One bin is selected without replacement (depicted by heavier line) for each segment of transect.



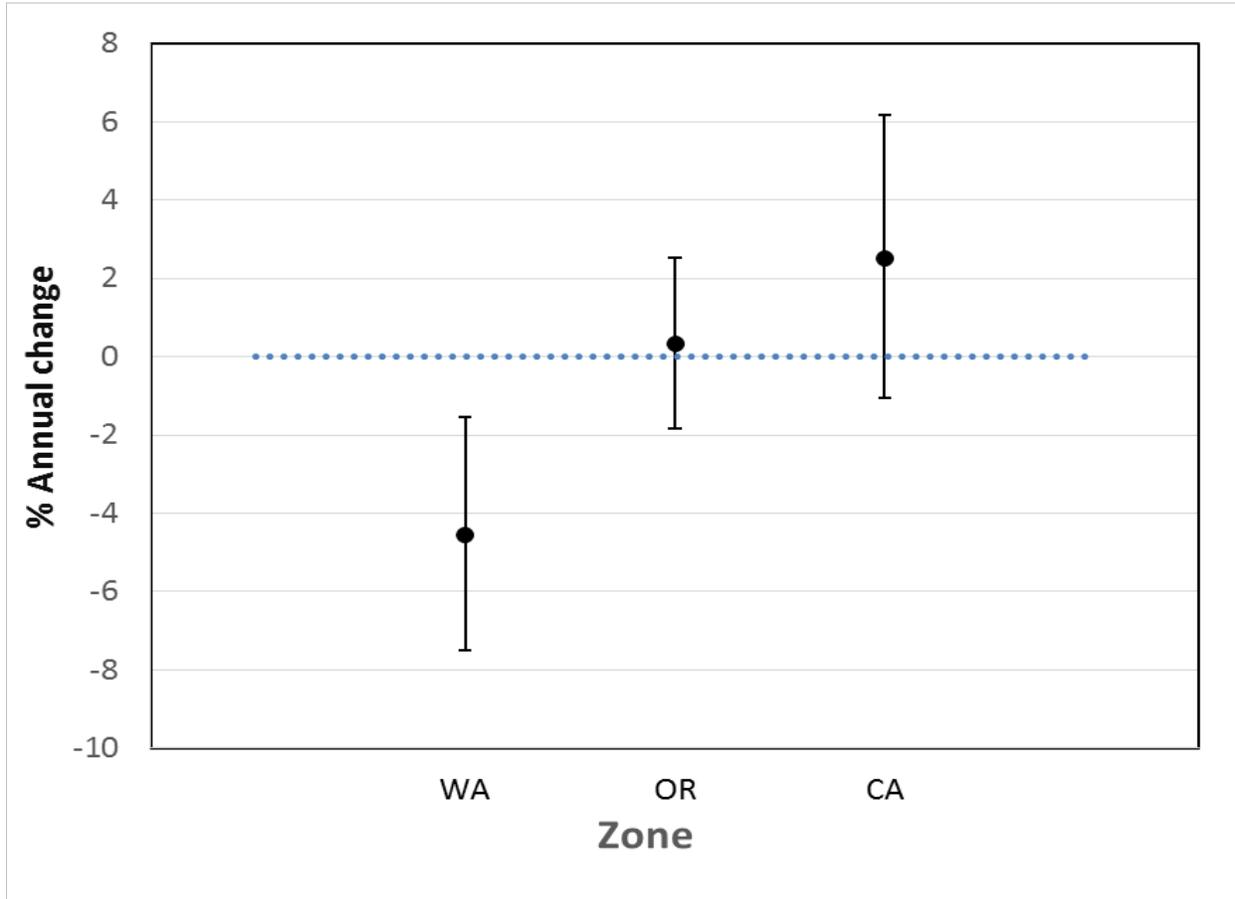
**Figure 1-3.** Annual marbled murrelet population estimates and 95 percent confidence intervals, for Conservation Zones 1 through 5 combined.



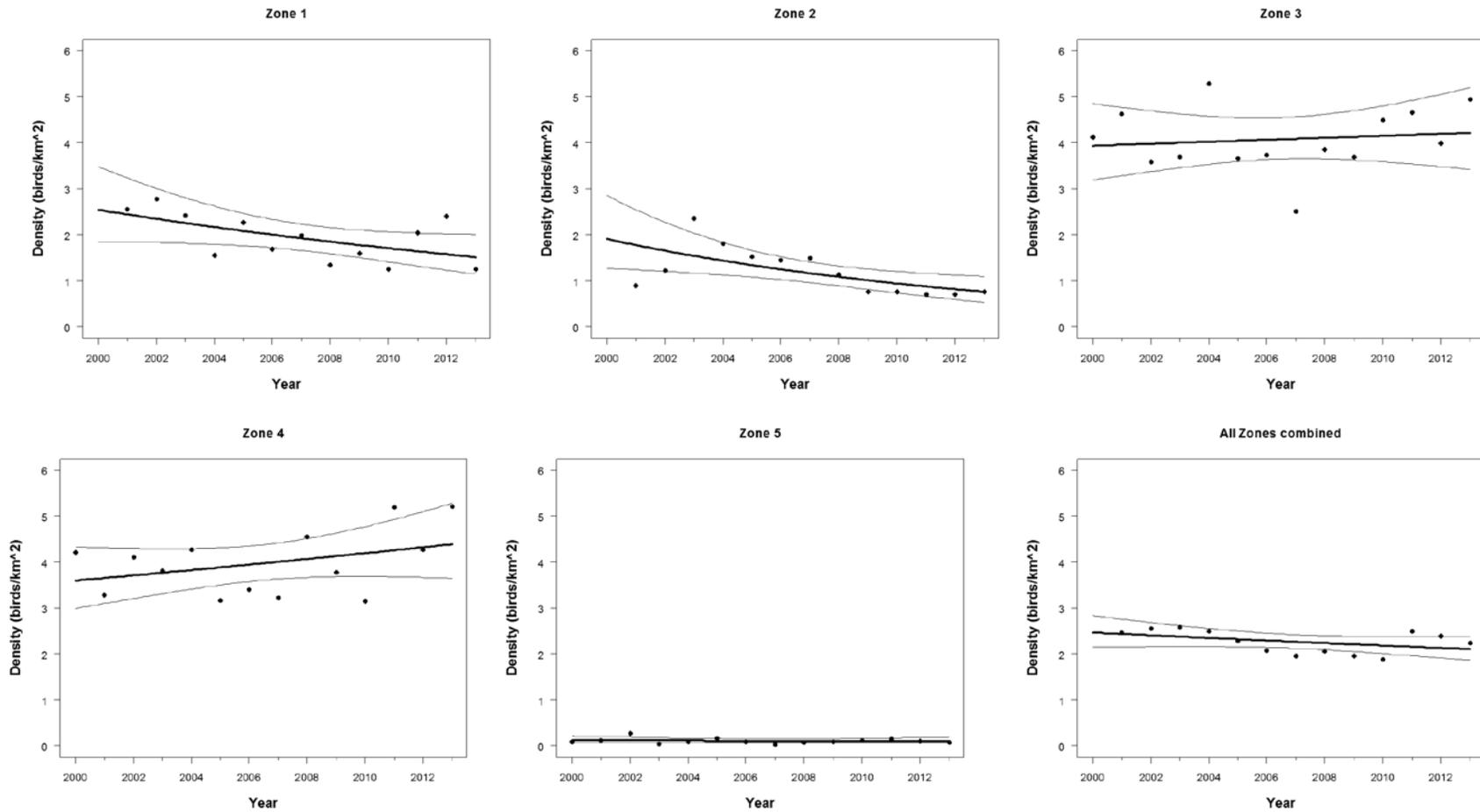
**Figure 1-4.** Average marbled murrelet densities at sea by primary sampling unit for each Conservation Zone. Based on mean densities from 2000 to 2013 monitoring data (2001 to 2013 data for Conservation Zones 1 and 2).



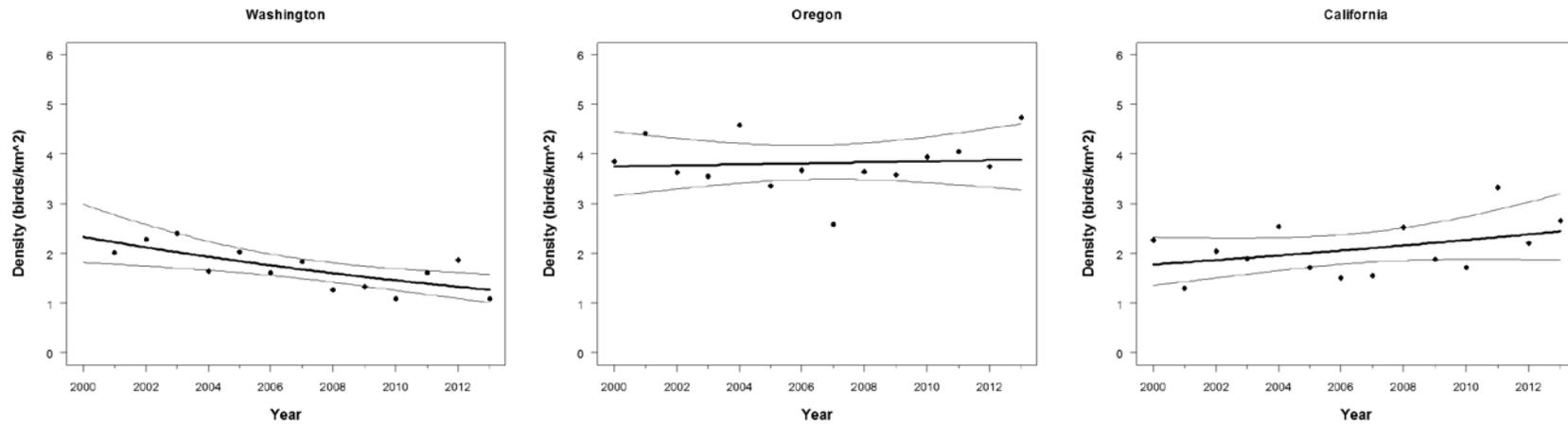
**Figure 1-5.** Trend results: average rate of annual change by conservation zone and for all conservation zones combined, 2000 to 2013, with 95 percent confidence intervals. The All-Zones, Conservation Zone 1, and Conservation Zone 2 trends are based on 2001 to 2013 data.



**Figure 1-6.** Trend results: average rate of annual change by state, 2000 to 2013, with 95 percent confidence intervals. Washington trend is based on 2001 to 2013 data.



**Figure 1-7a.** Results of trend analyses at All-Zones and individual Conservation Zone scales. Graphs show fitted regression lines through the annual population estimates for the period of analysis, with 95 percent confidence limits.



**Figure 1-7b.** Results of trend analyses at the State scale. Graphs show fitted regression lines through the annual population estimates for the period of analysis, with 95 percent confidence limits.

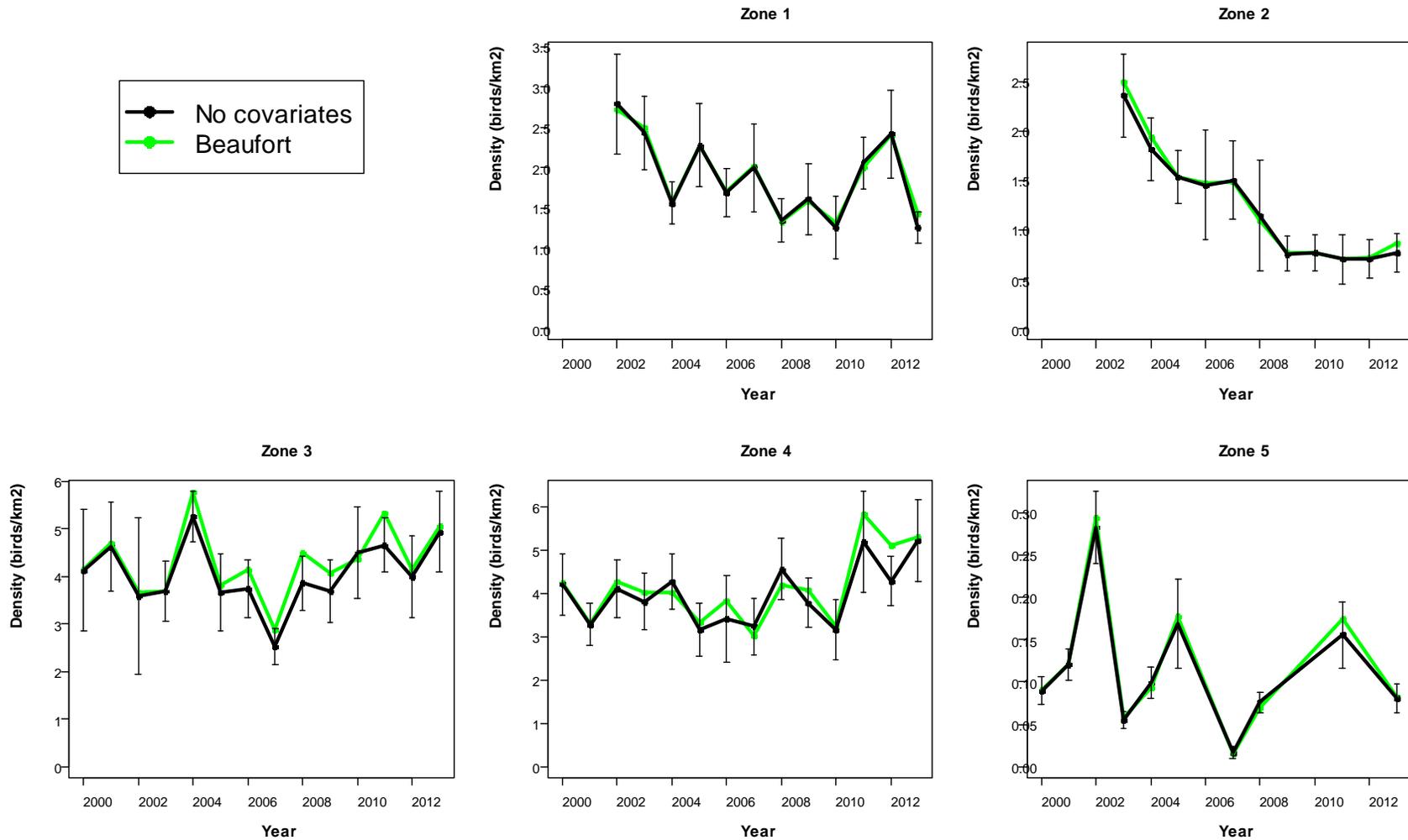
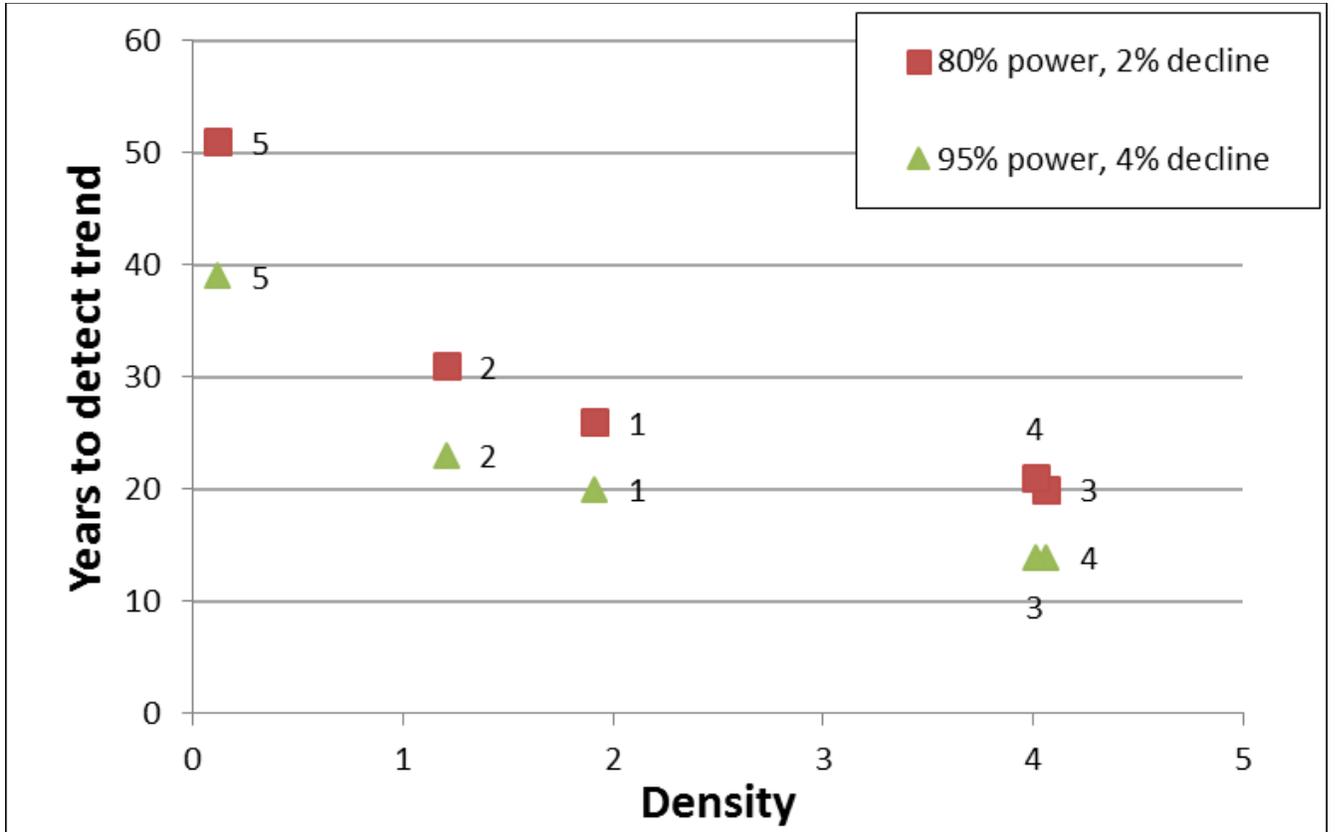
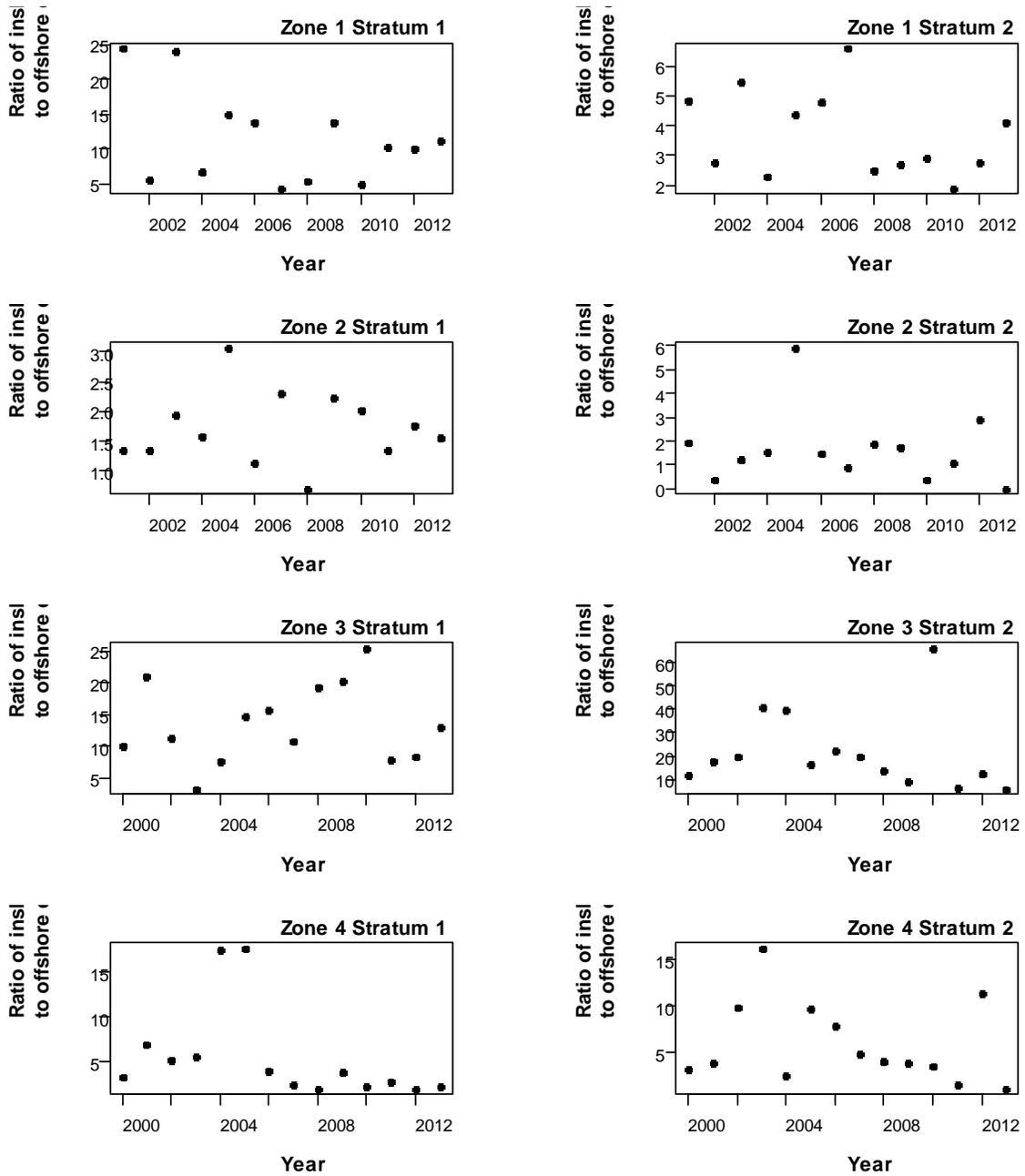


Figure 1-8. Effect of sea condition (Beaufort) covariate on annual density estimates at the conservation zone scale.



**Figure 1-9.** The number of years of sampling (starting in 2000 or 2001) needed to detect a declining trend at the conservation zone scale as a function of mean conservation zone density, averaged over 2001 to 2013. Each point represents a conservation zone, with points labeled with corresponding conservation zone.



**Figure 1-10.** Temporal pattern of marbled murrelet density at stratum scale. See text for details.

## Appendix 1-A: Power Analysis

### *Introduction*

Because one objective of our monitoring is to detect a change over time, it is desirable to have the statistical power to detect a decline in a time frame that allows managers to respond by altering land management strategies. We conducted a new power analysis for this report, using the larger set of available data to determine the sampling design's power to detect a decline for each conservation zone and for all conservation zones combined. This replaces previous power analyses by Miller et al. (2006) and Falxa et al. (2011), which were based on 2000-2003 and 2001-2009 data, respectively. The reason for conducting this analysis is to assess power associated with a reduced sampling effort from 2014 forward. This assumption reflects a decision among agency managers, discussed below, to reduce sampling frequency to every other year effective in 2014.

### *Methods*

Our method for calculating power combines those described by Hogg and Craig (1995) and Draper and Smith (1998), and used monitoring data from 2001 through 2013.

As noted above, for this power analysis we assumed that conservation zones will be surveyed every other year such that surveys will now be made in even years for two of the conservation zones (Conservation Zones 1 and 3), in odd years for Conservation Zones 2 and 4, and every 4 years in Conservation Zone 5. To approximate the power to detect linear trends (in the log of the density) we assumed the full complement of existing surveys up to 2013 for Conservation Zones 2, 4, and 5 and up to 2014 for Conservation Zones 1 and 3. Then we assumed surveys only in every other year except for Conservation Zone 5 for which surveys in every 4 years were considered.

We estimated the power to detect annual rates of population change of 1, 2, 3, 4, and 5 percent. We did this by simulating data from a regression of the surveyed years using the slope associated with 1, 2, 3, 4, and 5 percent annual declines and the estimated root mean square error from the data collected so far, through 2013. There were 1,000 simulations per conservation zone, annual change, and year combination. The estimated power curves were smoothed in an attempt to obtain more accurate power results by regressing the inverse normal distribution function of the power estimates and the year. The predicted values were used to display the power curves (Figure 1-A-1).

## RESULTS

### *Power to Detect Trends*

Our measure for assessing the effectiveness of the monitoring design is its power to detect changes in the mean density and the resulting mean total population of marbled murrelets over time. Tables 1-A-1 and 1-A-2 present the estimated years when sampling will be sufficient to detect a trend in a population declining at various rates of annual population decrease for two levels of power: 0.80 and 0.95. These power numbers measure the ability of the sampling design to detect a significant trend, if one exists. For example, if we use a power level of 0.80 and for a rate of decline of 3 percent per year, the power analysis results would estimate the number of years of sampling required to detect a real 3 percent decline, with an 80 percent probability of detecting that trend (equals the probability of correctly rejecting the null hypothesis).

For the population of the 5 conservation zones combined, the power analysis estimated that by 2013 enough years had been sampled to detect an annual decrease of 3 percent with greater than 95 percent power, and that an annual decrease of 2 percent could be detected with about 5 more years of sampling (through 2018) at 95 percent power (Appendix 1-A Table 1-A-2). Sampling through 2024 would be required to detect an annual decrease of 1 percent at the scale of the combined conservation zones with 80 percent power; sampling through 2028 would increase the power to 95 percent. At the single-conservation zone scale, the decrease we could detect with monitoring to date differs among the five conservation zones. Power is fairly similar for Conservation Zones 3 and 4 where sampling to date is sufficient to detect annual decreases of 3 percent or greater with at least 80 percent power, and a decrease of 4 percent or greater with about 95 percent power. In Conservation Zones 1 and 2, where interannual variability has been slightly greater, statistical power is slightly higher for sampling to date, being sufficient by 2017 to detect annual declines since 2001 of 5 percent with 80 percent power (Table 1-A-1); we note that a significant decline has already been detected for Conservation Zone 2, at an estimate annual rate of about 7 percent per year. The number of years required to detect a trend is highest for Conservation Zone 5.

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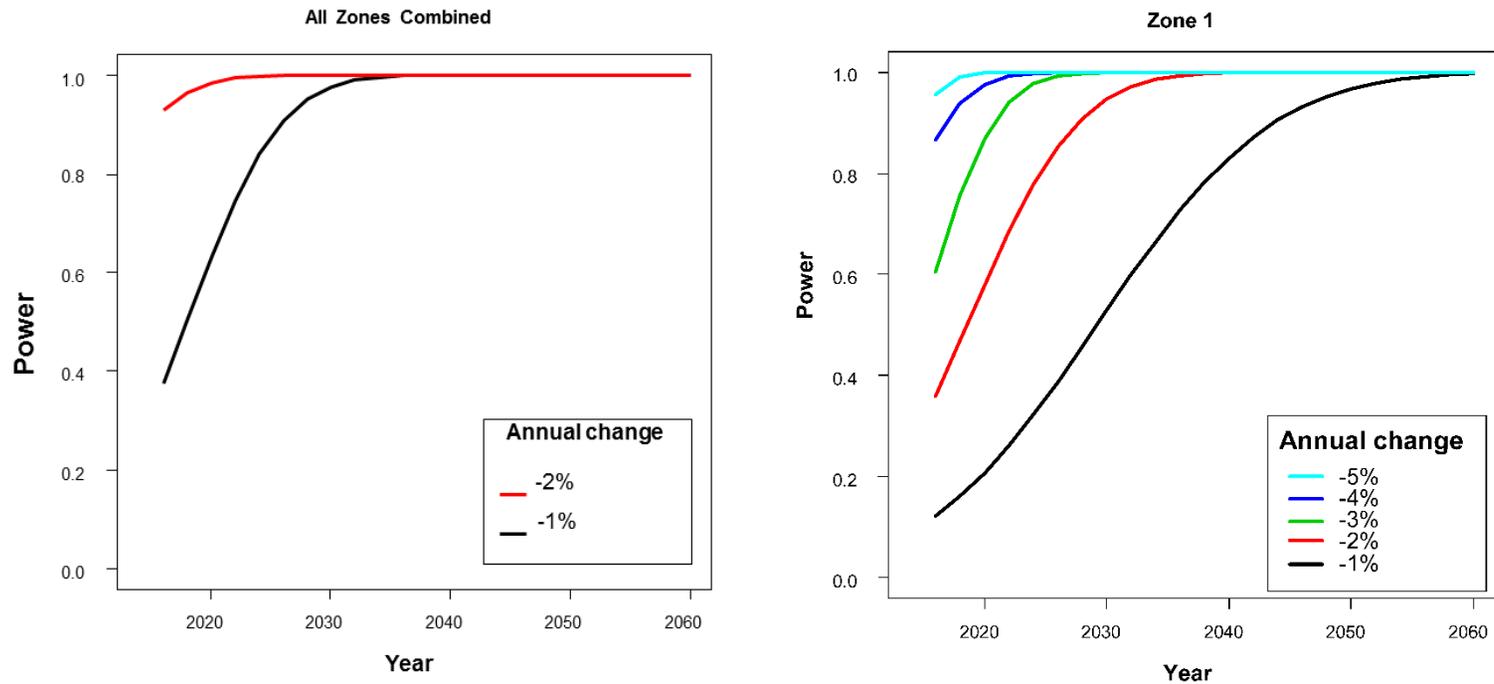
**Table 1-A-1.** Estimates of the year when 80 percent or greater power will be achieved to detect various rates of annual decrease in the NWFP marbled murrelet population, based on data from surveys starting in 2000/2001. As described in text, the power analysis assumed that starting in 2014, Zones 1 through 4 are sampled every other year, and Zone 5 is sampled every 4 years. Reported for all Conservation Zones combined and by individual conservation zone. An “A” in a cell indicates that this level of power has already been achieved. A decline can have been detected already, even if 80 percent power has not been reached, such as for Conservation Zone 2. See text for details.

Annual Rate of Decrease (%)	Conservation Zone					
	All	1	2	3	4	5
	Year When 80% Power Achieved					
1	2024	2040	2049	2034	2031	>2059
2	A	2026	2031	2020	2021	2051
3	A	2020	2023	A	A	2039
4	A	2016	2019	A	A	2031
5	A	A	2017	A	A	2027

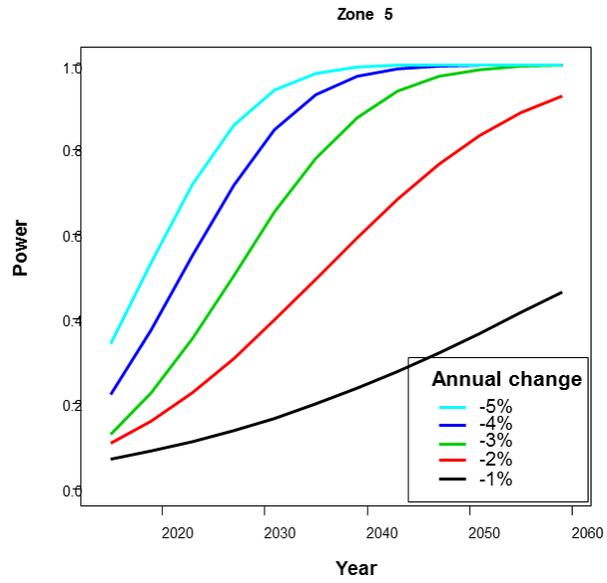
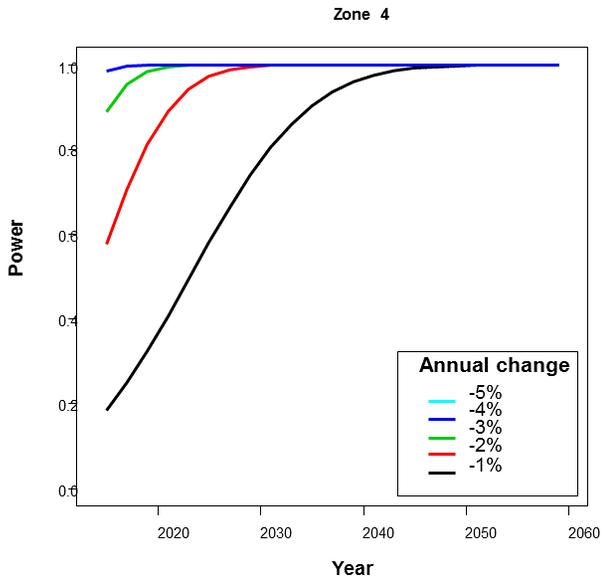
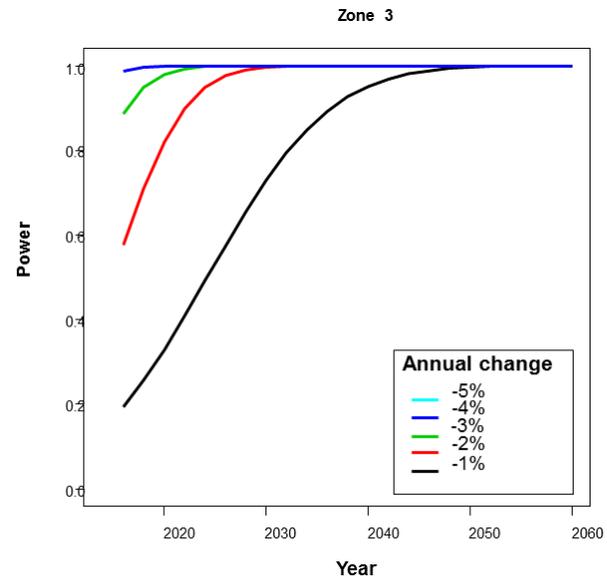
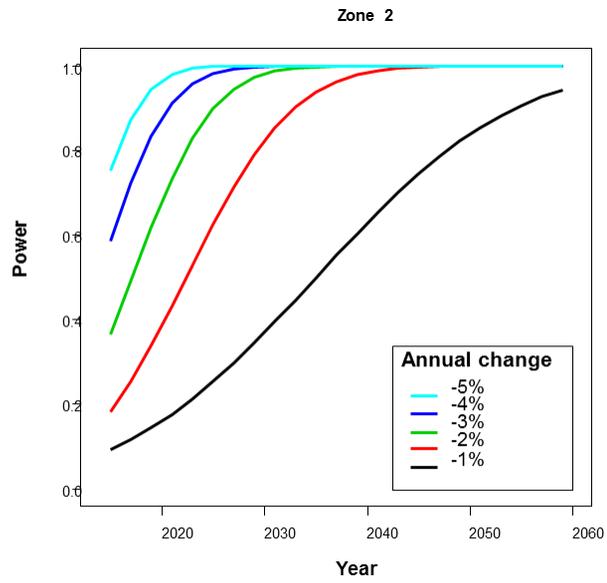
**Table 1-A-2.** Similar to Table 1-A-1, but reporting the year when 95 percent or greater power will be achieved to detect various percentages of annual decrease in the NWFP murrelet population, for all conservation zones combined and by individual zone.

Annual Rate of Decrease (%)	Conservation Zone					
	All	1	2	3	4	5
	Year When 95% Power Achieved					
1	2028	2048	2059	2042	2039	>2059
2	2018	2030	2037	2024	2023	>2059
3	A	2024	2029	2018	2017	2047
4	A	2020	2023	A	A	2039
5	A	2016	2021	A	A	2035

**Figure 1-A-1.** The following charts represent the power to detect population declines at annual rates of change ranging from 1 to 5 percent. Power estimates assumed the use of data from annual sampling to date (since 2000-2001), and assumes sampling in every-other year from 2014 on. The exception is Conservation Zone 5, which is based on only those years actually sampled through 2013, and on sampling every 4 years from 2014 on.



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Appendix 1-B: Field Audit Form

Checklist for Field Audit  
Marbled Murrelet Long-term Monitoring Surveys

Name of Auditor \_\_\_\_\_ Date \_\_\_\_/ \_\_\_\_/ \_\_\_\_

Field Crew (name or location) \_\_\_\_\_ Survey Vessel \_\_\_\_\_

Time Begin \_\_\_\_\_ End \_\_\_\_\_

Was each crewmember observed as a: Survey observer \_\_\_\_\_ Boat operator \_\_\_\_\_

If not, was the observer/navigator configuration different from the previous audit? \_\_\_\_\_

Is the track line of the transect:

\_\_\_\_\_ being recorded by GPS and downloaded each day?

\_\_\_\_\_ paused during forays off the transect line and between segment start and end points?

Did observers:

\_\_\_\_\_ Scan with greater effort close to the transect line?

\_\_\_\_\_ Record distances at first detection?

\_\_\_\_\_ Record all murrelet groups detected, regardless of distance from the line?

\_\_\_\_\_ Record flying murrelets?

\_\_\_\_\_ Communicate with each other on groups close to the line?

\_\_\_\_\_ Alert each other to murrelet groups to minimize missed detections?

\_\_\_\_\_ Define group size consistent with each other and with protocol definition?

Are distance trials:

\_\_\_\_\_ Conducted in sets of 5, with all 5 estimates within 15%?

\_\_\_\_\_ Up to date (completed every 3 days)?

## Appendix 1-C

### Population Estimates at Stratum Scale, with Distance Parameters

Year	Zone	Stratum	Density	CV	Birds	Lower 95% CI	Upper 95% CI	Area	f(0)	E(s)	Truncation Distance (m)
2000	3	All	4.129	18.6%	6,587	3,987	8,756	1,595	0.0165	1.623	100
2000	3	1	1.336	32.2%	883	357	1,350	661			
2000	3	2	6.104	19.6%	5,704	3,296	7,608	935			
2000	4	All	4.216	30.9%	4,887	3,417	9,398	1,159	0.0097	1.730	180
2000	4	1	6.024	34.0%	4,420	2,931	8,784	734			
2000	4	2	1.097	32.1%	467	297	881	425			
2000	5	All	0.090	80.6%	79	-	260	883	0.0097	1.730	180
2000	5	1	0.179	80.6%	79	-	260	441			
2000	5	2	0.000		-	-	-	441			
2001	All	All	2.466	10.1%	21,763	17,472	26,053	8,826			
2001	1	All	2.553	18.0%	8,936	5,740	11,896	3,501	0.0133	1.594	142
2001	1	1	4.506	23.1%	3,809	2,432	5,689	845			
2001	1	2	1.764	21.4%	2,111	948	2,816	1,196			
2001	1	3	2.067	37.2%	3,016	404	5,003	1,459			
2001	2	All	0.899	41.9%	1,518	524	2,942	1,688	0.0125	1.444	80
2001	2	1	1.430	55.7%	1,040	91	2,364	727			
2001	2	2	0.497	72.5%	478	106	1,317	961			
2001	3	All	4.636	13.2%	7,396	5,230	9,075	1,595	0.0166	1.735	140
2001	3	1	1.724	23.0%	1,140	657	1,700	661			
2001	3	2	6.695	14.1%	6,257	4,241	7,814	935			
2001	4	All	3.284	24.0%	3,807	2,983	6,425	1,159	0.0101	1.749	170
2001	4	1	4.567	27.2%	3,351	2,436	5,880	734			
2001	4	2	1.072	30.1%	456	313	854	425			
2001	5	All	0.121	52.5%	106	27	244	883	0.0101	1.749	170
2001	5	1	0.198	39.1%	87	-	138	441			
2001	5	2	0.043	231.6%	19	-	129	441			
2002	All	All	2.563	11.9%	22,521	17,264	27,777	8,788			
2002	1	All	2.788	21.5%	9,758	5,954	14,149	3,501	0.0103	1.761	194
2002	1	1	7.207	32.8%	6,092	2,716	9,782	845			
2002	1	2	1.879	26.9%	2,248	909	3,309	1,196			
2002	1	3	0.972	34.7%	1,419	580	2,515	1,459			
2002	2	All	1.233	29.2%	2,031	800	3,132	1,650	0.0195	1.400	70
2002	2	1	2.448	32.1%	1,774	559	2,840	724			
2002	2	2	0.278	41.2%	258	-	417	926			
2002	3	All	3.583	24.1%	5,716	3,674	9,563	1,595	0.0118	1.892	150
2002	3	1	0.696	34.1%	460	258	886	661			
2002	3	2	5.624	24.7%	5,256	3,301	8,732	935			
2002	4	All	4.112	15.1%	4,766	3,272	6,106	1,159	0.0108	1.724	175
2002	4	1	5.186	15.9%	3,805	2,501	4,892	734			
2002	4	2	2.260	33.1%	961	437	1,665	425			
2002	5	All	0.282	42.3%	249	27	400	883	0.0108	1.724	175
2002	5	1	0.510	46.1%	225	8	371	441			
2002	5	2	0.054	71.1%	24	-	54	441			
2003	All	All	2.596	9.6%	22,808	18,525	27,091	8,786			
2003	1	All	2.428	16.6%	8,495	5,795	11,211	3,498	0.0087	1.817	300
2003	1	1	6.644	22.1%	5,617	3,372	7,795	845			
2003	1	2	1.441	32.9%	1,721	911	2,794	1,195			
2003	1	3	0.793	32.8%	1,156	252	1,912	1,458			
2003	2	All	2.407	28.8%	3,972	2,384	6,589	1,650	0.0171	1.399	80
2003	2	1	2.639	26.0%	1,912	1,132	3,048	724			
2003	2	2	2.225	48.4%	2,061	1,019	4,229	926			
2003	3	All	3.686	16.1%	5,881	3,992	7,542	1,595	0.0132	1.664	130
2003	3	1	1.192	23.8%	788	499	1,212	661			
2003	3	2	5.450	17.8%	5,093	3,244	6,680	935			
2003	4	All	3.806	17.3%	4,412	3,488	6,495	1,159	0.0086	1.704	180
2003	4	1	4.960	19.7%	3,640	2,622	5,392	734			
2003	4	2	1.816	27.2%	773	557	1,424	425			
2003	5	All	0.055	61.1%	48	-	85	883	0.0086	1.704	180
2003	5	1	0.109	61.1%	48	-	85	441			
2003	5	2	0.000		-	-	-	441			
2004	All	All	2.455	10.5%	21,572	17,144	26,000	8,786			
2004	1	All	1.562	22.0%	5,465	2,921	7,527	3,498	0.0108	1.789	280
2004	1	1	3.833	30.0%	3,241	1,365	4,845	845			
2004	1	2	1.513	25.4%	1,807	1,042	2,777	1,195			
2004	1	3	0.286	60.0%	417	-	727	1,458			
2004	2	All	1.823	27.0%	3,009	1,669	4,634	1,650	0.0116	1.411	115
2004	2	1	3.373	33.4%	2,444	1,217	4,093	724			
2004	2	2	0.611	25.0%	565	314	841	926			
2004	3	All	5.051	13.7%	8,058	5,369	9,819	1,595	0.0143	1.6979	110
2004	3	1	1.721	20.7%	1,137	707	1,732	661			
2004	3	2	7.405	15.1%	6,921	4,278	8,564	935			
2004	4	All	4.272	26.9%	4,952	3,791	9,021	1,159	0.0093	1.700	200
2004	4	1	5.331	32.2%	3,911	2,729	7,732	734			
2004	4	2	2.447	43.5%	1,041	608	2,421	425			
2004	5	All	0.099	60.5%	88	18	214	883	0.0093	1.700	200
2004	5	1	0.091	64.5%	40	-	104	441			
2004	5	2	0.107	93.6%	47	-	137	441			

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Year	Zone	Stratum	Density	CV	Birds	Lower 95% CI	Upper 95% CI	Area	f(0)	E(s)	Truncation Distance (m)
2005	All	All	2.300	10.7%	20,209	15,976	24,442	8,785			
2005	1	All	2.275	20.5%	7,956	4,900	11,288	3,497	0.0156	1.758	150
2005	1	1	2.501	37.7%	2,114	698	3,661	845			
2005	1	2	2.426	25.4%	2,895	1,186	4,210	1,194			
2005	1	3	2.021	30.1%	2,947	1,198	5,019	1,458			
2005	2	All	1.561	20.4%	2,576	1,675	3,729	1,650	0.0136	1.4184	130
2005	2	1	2.785	19.1%	2,018	1,233	2,764	724			
2005	2	2	0.603	56.7%	558	166	1,461	926			
2005	3	All	3.669	16.9%	5,854	3,580	7,447	1,595	0.0127	1.841	150
2005	3	1	0.808	32.2%	534	269	962	661			
2005	3	2	5.693	17.8%	5,320	3,156	6,760	935			
2005	4	All	3.169	23.6%	3,673	2,740	6,095	1,159	0.0108	1.518	170
2005	4	1	4.487	25.5%	3,292	2,329	5,562	734			
2005	4	2	0.895	42.1%	381	243	901	425			
2005	5	All	0.169	31.8%	149	69	251	883	0.0108	1.518	170
2005	5	1	0.141	48.1%	62	8	121	441			
2005	5	2	0.197	39.7%	87	36	156	441			
2006	All	All	2.080	8.2%	18,275	15,336	21,214	8,785			
2006	1	All	1.687	18.1%	5,899	4,211	8,242	3,497	0.0138	1.765	139
2006	1	1	2.760	16.3%	2,333	1,628	3,182	845			
2006	1	2	1.418	24.9%	1,693	777	2,551	1,194			
2006	1	3	1.284	40.4%	1,873	595	3,440	1,458			
2006	2	All	1.455	18.0%	2,381	1,702	3,433	1,650	0.0130	1.5678	107
2006	2	1	2.261	19.9%	1,638	1,038	2,372	724			
2006	2	2	0.802	34.0%	743	380	1,344	926			
2006	3	All	3.731	12.7%	5,953	4,546	7,617	1,595	0.0114	1.814	145
2006	3	1	1.034	29.6%	684	352	1,070	661			
2006	3	2	5.638	14.1%	5,269	3,886	6,827	935			
2006	4	All	3.410	14.9%	3,953	3,164	5,525	1,159	0.0106	1.622	150
2006	4	1	4.821	15.5%	3,538	2,698	4,894	734			
2006	4	2	0.977	47.8%	416	209	981	425			
2006	5	All	<i>Interpolated</i>		89	35	150	883	0.0106	1.622	150
2006	5	1	<i>Interpolated</i>		69	4	85	441			
2006	5	2	<i>Interpolated</i>		65	18	103	441			
2007	All	All	1.971	13.7%	17,317	12,654	21,980	8,785			
2007	1	All	1.997	24.2%	6,985	4,148	10,639	3,497	0.0117	1.642	378
2007	1	1	3.445	27.6%	2,912	1,025	4,392	845			
2007	1	2	1.218	21.9%	1,453	708	1,993	1,194			
2007	1	3	1.796	51.3%	2,620	206	5,629	1,458			
2007	2	All	1.536	26.7%	2,535	1,318	3,867	1,650	0.0135	1.496	126
2007	2	1	2.851	32.0%	2,065	964	3,336	724			
2007	2	2	0.508	25.5%	470	234	666	926			
2007	3	All	2.518	19.8%	4,018	2,730	5,782	1,595	0.0106	1.653	150
2007	3	1	0.526	58.5%	348	26	744	661			
2007	3	2	3.927	20.4%	3,670	2,525	5,378	935			
2007	4	All	3.234	34.8%	3,749	2,659	7,400	1,159	0.0106	1.607	180
2007	4	1	4.730	37.5%	3,470	2,329	7,025	734			
2007	4	2	0.655	36.9%	279	146	549	425			
2007	5	All	0.033	37.7%	30	-	49	883	0.0106	1.607	180
2007	5	1	0.067	37.7%	30	-	49	441			
2007	5	2	0.000		-	-	-	441			
2008	All	All	2.064	8.9%	18,134	14,983	21,284	8,785			
2008	1	All	1.344	17.6%	4,699	3,000	6,314	3,497	0.0109	1.739	206
2008	1	1	3.572	25.1%	3,019	1,439	4,472	845			
2008	1	2	0.899	27.6%	1,073	580	1,640	1,194			
2008	1	3	0.416	30.8%	607	288	970	1,458			
2008	2	All	1.169	22.1%	1,929	1,164	2,868	1,650	0.0112	1.535	187
2008	2	1	2.584	22.4%	1,872	1,132	2,801	724			
2008	2	2	0.062	49.1%	57	-	116	926			
2008	3	All	3.857	14.7%	6,153	4,485	8,066	1,595	0.0113	1.750	130
2008	3	1	0.337	28.4%	223	107	353	661			
2008	3	2	6.345	15.3%	5,930	4,233	7,816	935			
2008	4	All	4.560	17.9%	5,285	3,809	7,503	1,159	0.0100	1.705	200
2008	4	1	6.386	19.5%	4,685	3,167	6,687	734			
2008	4	2	1.410	39.0%	600	302	1,195	425			
2008	5	All	0.076	48.1%	67	9	132	883	0.0100	1.705	200
2008	5	1	0.065	60.1%	29	-	81	441			
2008	5	2	0.087	70.3%	38	-	68	441			
2009	All	All	1.965	10.6%	17,260	13,670	20,851	8,785			
2009	1	All	1.608	21.2%	5,623	3,786	8,497	3,497	0.0094	1.694	254
2009	1	1	3.811	27.7%	3,221	1,777	5,107	845			
2009	1	2	0.689	26.3%	822	489	1,302	1,194			
2009	1	3	1.083	42.9%	1,580	410	3,299	1,458			
2009	2	All	0.765	21.9%	1,263	776	1,874	1,650	0.0092	1.475	191
2009	2	1	1.609	23.3%	1,166	693	1,766	724			
2009	2	2	0.105	61.0%	97	-	209	926			
2009	3	All	3.696	17.7%	5,896	3,898	7,794	1,595	0.0131	1.696	120

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Year	Zone	Stratum	Density	CV	Birds	Lower 95% CI	Upper 95% CI	Area	f(0)	E(s)	Truncation Distance (m)
2009	3	1	0.650	42.5%	430	187	893	661			
2009	3	2	5.849	19.0%	5,467	3,339	7,250	935			
2009	4	All	3.786	19.9%	4,388	3,599	6,952	1,159	0.0100	1.661	150
2009	4	1	5.304	20.9%	3,892	3,031	6,170	734			
2009	4	2	1.167	67.3%	497	244	1,390	425			
2009	5	All	<i>Interpolated</i>		90	11	186	883	0.0100	1.661	150
2009	5	1	<i>Interpolated</i>		55	2	140	441			
2009	5	2	<i>Interpolated</i>		36	-	67	441			
2010	All	All	1.894	11.1%	16,641	13,015	20,268	8,785			
2010	1	All	1.256	20.0%	4,393	2,719	6,207	3,497	0.0100	1.717	200
2010	1	1	2.004	26.8%	1,694	957	2,712	845			
2010	1	2	1.783	23.6%	2,128	1,021	3,052	1,194			
2010	1	3	0.391	43.1%	571	62	1,142	1,458			
2010	2	All	0.779	25.5%	1,286	688	1,961	1,650	0.0114	1.582	145
2010	2	1	1.336	23.8%	968	552	1,439	724			
2010	2	2	0.343	71.9%	318	-	784	926			
2010	3	All	4.503	16.7%	7,184	4,453	9,425	1,595	0.0138	1.770	160
2010	3	1	1.071	50.1%	708	239	1,354	661			
2010	3	2	6.930	17.7%	6,476	3,691	8,468	935			
2010	4	All	3.162	28.5%	3,665	2,248	6,309	1,159	0.0120	1.624	165
2010	4	1	3.774	34.3%	2,769	1,463	5,087	734			
2010	4	2	2.106	36.3%	896	431	1,700	425			
2010	5	All	<i>Interpolated</i>		114	13	241	883	0.0120	1.624	165
2010	5	1	<i>Interpolated</i>		81	3	200	441			
2010	5	2	<i>Interpolated</i>		33	-	66	441			
2011	All	All	2.501	12.6%	21,972	16,566	27,378	8,785			
2011	1	All	2.055	17.4%	7,187	4,807	9,595	3,497	0.0089	1.666	289
2011	1	1	5.580	20.3%	4,717	2,621	6,399	845			
2011	1	2	1.243	23.7%	1,484	790	2,147	1,194			
2011	1	3	0.676	65.8%	986	206	2,384	1,458			
2011	2	All	0.721	33.4%	1,189	571	2,106	1,650	0.0110	1.4967	161
2011	2	1	1.314	30.8%	952	400	1,572	724			
2011	2	2	0.256	102.0%	237	38	772	926			
2011	3	All	4.661	16.3%	7,436	5,067	9,746	1,595	0.0126	1.678	120
2011	3	1	0.980	38.6%	648	343	1,455	661			
2011	3	2	7.264	17.4%	6,788	4,304	9,054	935			
2011	4	All	5.196	34.9%	6,023	2,782	10,263	1,159	0.0122	1.644	145
2011	4	1	6.724	42.2%	4,933	1,643	8,767	734			
2011	4	2	2.561	47.3%	1,090	592	2,472	425			
2011	5	All	0.155	53.0%	137	16	295	883	0.0122	1.644	145
2011	5	1	0.243	64.8%	107	5	259	441			
2011	5	2	0.068	78.8%	30	-	66	441			
2012	All	All	2.396	11.4%	21,052	16,369	25,736	8,785			
2012	1	All	2.414	20.7%	8,442	5,090	12,006	3,497	0.0109	1.847	164
2012	1	1	7.166	24.4%	6,056	3,289	8,823	845			
2012	1	2	1.507	30.4%	1,799	812	2,892	1,194			
2012	1	3	0.402	48.1%	587	168	1,227	1,458			
2012	2	All	0.719	33.5%	1,186	564	2,360	1,650	0.0132	1.485	106
2012	2	1	1.178	29.2%	853	325	1,289	724			
2012	2	2	0.360	89.9%	333	-	1,459	926			
2012	3	All	3.986	15.5%	6,359	4,136	8,058	1,595	0.0112	1.765	186
2012	3	1	0.895	34.9%	591	227	1,042	661			
2012	3	2	6.172	15.9%	5,768	3,775	7,330	935			
2012	4	All	4.279	24.9%	4,960	3,414	8,011	1,159	0.0107	1.652	140
2012	4	1	6.050	27.6%	4,439	2,916	7,497	734			
2012	4	2	1.225	39.6%	521	166	940	425			
2012	5	All	<i>Interpolated</i>		104	10	206	883	0.0107	1.652	140
2012	5	1	<i>Interpolated</i>		89	5	189	441			
2012	5	2	<i>Interpolated</i>		15	-	33	441			
2013	All	All	2.238	11.1%	19,662	15,398	23,927	8,785			
2013	1	All	1.257	27.9%	4,395	2,298	6,954	3,497	0.0109	1.695	137
2013	1	1	2.379	31.4%	2,010	861	3,253	845			
2013	1	2	0.657	20.1%	784	508	1,124	1,194			
2013	1	3	1.097	64.4%	1,600	381	3,717	1,458			
2013	2	All	0.770	18.5%	1,271	950	1,858	1,650	0.0117	1.569	132
2013	2	1	1.605	19.0%	1,163	854	1,722	724			
2013	2	2	0.117	59.3%	108	-	274	926			
2013	3	All	4.939	16.3%	7,880	5,450	10,361	1,595	0.0112	1.637	160
2013	3	1	0.991	43.8%	655	151	1,226	661			
2013	3	2	7.731	17.8%	7,225	4,707	9,667	935			
2013	4	All	5.216	20.5%	6,046	4,531	9,282	1,159	0.0128	1.607	146
2013	4	1	7.384	21.8%	5,418	3,939	8,516	734			
2013	4	2	1.477	36.7%	629	279	1,184	425			
2013	5	All	0.080	45.4%	71	5	118	883	0.0128	1.607	146
2013	5	1	0.160	45.4%	71	5	118	441			
2013	5	2	0.000		-	-	-	441			

## Chapter 2. Status and trend of nesting habitat for the Marbled Murrelet under the Northwest Forest Plan

Martin G. Raphael, Gary A. Falxa, Deanna Lynch, S. Kim Nelson, Scott F. Pearson, Andrew J. Shirk, Richard D. Young

*Martin G. Raphael is a Research Wildlife Biologist, USDA Forest Service, Pacific Northwest Research Station, 3625 93<sup>rd</sup> Ave. SW, Olympia WA 98512; Gary A. Falxa is a Fish and Wildlife Biologist, U.S. Fish and Wildlife Service, Arcata Fish and Wildlife Office, 1655 Heindon Rd., Arcata, CA 95521; Deanna Lynch is a Fish and Wildlife Biologist, U.S. Fish and Wildlife Service, Washington Fish and Wildlife Office, Branch of Listing, 510 Desmond Dr., Suite 102, Lacey, WA 98503; S. Kim Nelson is a Research Wildlife Biologist, Oregon Cooperative Fish and Wildlife Research Unit, Oregon State University, Department of Fisheries and Wildlife, 104 Nash Hall, Corvallis, OR 97331; Scott F. Pearson is a senior scientist, Wildlife Science Division, Washington Department of Fish and Wildlife, 1111 Washington St. SE, 5th Floor, Olympia, WA 98501; Andrew J. Shirk is an ecologist, Climate Impacts Group, College of the Environment, University of Washington, Box 355674, Seattle, WA 98195 and Richard D. Young is a GIS Analyst, U.S. Fish and Wildlife Service, Pacific Regional Office, Ecological Services, 911 NE 11<sup>th</sup> Avenue, Portland, OR 97232.*

### Abstract

The primary objectives of the effectiveness monitoring plan for the marbled murrelet (*Brachyramphus marmoratus*) include mapping baseline nesting habitat (at the start of the Northwest Forest Plan) and estimating changes in that habitat over time. Using maximum entropy (Maxent) models, we modeled nesting habitat suitability over all lands in the murrelet's range in Washington, Oregon and California. The models used vegetation and physiographic attributes, and a sample of 368 murrelet nest sites (184 confirmed murrelet nest sites and 184 occupied sites) for model training. We estimated there were 2.53 million acres of potential nesting habitat over all lands in the murrelet's range in Washington, Oregon, and California at the start of the plan (1993). Of this, 0.46 million acres were identified as highest suitability, matching or exceeding the average conditions for the training sites. Most (90 percent) of potential nesting habitat in 1993 on federally-administered lands occurred within reserved-land allocations. A substantial amount (41 percent) of baseline habitat occurred on nonfederal lands, including 44 percent of the highest-suitability habitat. We found a net loss of about 2 percent of potential nesting habitat from 1993 to 2012 on federal lands, compared to a net loss of about 27 percent on nonfederal lands. For federal and nonfederal lands combined, the net loss was about 12 percent. Fire was the major cause of nesting habitat loss on federal lands since the Northwest Forest Plan was implemented; timber harvest was the primary cause of loss on nonfederal lands. The large amount of younger forest of lower suitability located in reserves has the potential to offset habitat losses over time, but this merits further investigation using spatially-explicit forest development models. As evidenced by the high proportion of currently suitable nesting habitat that occurs within reserved land use designations, the Northwest Forest Plan has been successful in conserving murrelet habitat on federal lands. Losses of habitat on federal lands will continue due to fires and other disturbance events, but we expect

those losses to be exceeded by recovery of currently unsuitable habitat within reserves as forests mature. Incentives are needed, however, to curb losses of suitable habitat on nonfederal lands.

## **Introduction**

Although the marbled murrelet (*Brachyramphus marmoratus*) (murrelet) is a seabird that spends most of its time foraging on small fish and invertebrates in coastal waters, it was selected for monitoring the effectiveness of the Northwest Forest Plan (NWFP) in conserving old-forest species because it is associated with late-successional and old-growth forests for nesting. It nests mostly on large branches or other suitable platforms in large trees (Ralph et al. 1995, Nelson 1997). Conservation of the bird's nesting habitat is central to murrelet recovery (USFWS 1997). Due mostly to timber harvest, only a small percentage (5 to 20 percent) of original old-growth forests remain in Washington, Oregon, and California (Morrison 1988; Norheim 1996, 1997; USFWS 1997), and mostly in relatively small, fragmented patches or in forest parks and reserves.

Marbled murrelet effectiveness monitoring (Madsen et al. 1999) assesses status and trends in marbled murrelet populations and nesting habitat to answer the questions: Are the marbled murrelet populations associated with the NWFP area stable, increasing, or decreasing? Is the NWFP maintaining and restoring marbled murrelet nesting habitat? To address these questions, NWFP marbled murrelet monitoring has two components: population and habitat (Madsen et al. 1999). For habitat monitoring, the approach is to establish a baseline level of nesting habitat by first modeling habitat relationships, and then comparing habitat changes to the baseline (Huff et al. 2006a, Raphael et al. 2011). An underlying assumption is that murrelets are responding to the habitat characteristics used in the models as predictors of nesting habitat conditions. Population size and trends are monitored at sea using a unified sampling design and standardized transect-based survey methods (Miller et al. 2006, 2012; Raphael et al. 2007; Chapter 1, this volume). Thus, trends in both murrelet nesting habitat and populations are tracked over time. The ultimate goal is to relate population trends to nesting habitat conditions (Madsen et al. 1999; Chapter 3, this volume). A specific conservation goal of the NWFP is to stabilize and increase murrelet populations by maintaining and increasing nesting habitat (Madsen et al. 1999). The objective of this chapter is to evaluate the effectiveness of the NWFP in maintaining and increasing murrelet nesting habitat during the plans' first 20 years, by developing new baseline estimates for 1993 and comparing those with 2012 conditions to assess status and trend in nesting habitat.

The previous NWFP monitoring report for murrelets (15-yr report; Raphael et al. 2011, [http://www.fs.fed.us/pnw/pubs/pnw\\_gtr848.pdf](http://www.fs.fed.us/pnw/pubs/pnw_gtr848.pdf)) presented results from monitoring of murrelet habitat during the first 15 years of the NWFP by using species distribution modeling to develop baseline estimates of the amount and distribution of marbled murrelet potential nesting habitat. This publication builds upon and updates the 15-year report. As in the 15-year report, we used a habitat suitability modeling approach to

estimate the amount, spatial distribution and trends of potential nesting habitat. Model inputs included location data for nest sites and occupied stands (stands with observations of murrelet behaviors considered evidence of nesting), and spatial data on a suite of habitat characteristics hypothesized to affect the suitability of forest as marbled murrelet nesting habitat. What is new in this analysis are updated spatial data on habitat attributes at the start (1993) and end (2012) of the period (using gradient nearest neighbor methods [GNN]), updated spatial data on vegetation disturbances and causes during the period (using Landsat-based detection of Trends in Disturbance and Recovery methods [LandTrendr]), and a slightly expanded set of murrelet nest and occupied sites in Oregon and California. Also new for this analysis, the starting (1993) and ending (2012) years for the trend analyses are standardized throughout the NWFP area. The baseline (1993) level for marbled murrelet potential nesting habitat that is established in this report, using these improved data and technologies, replaces the baseline estimates in the 15-year report (Raphael et al. 2011). We then use these new baseline estimates to compare with those that we derive here for 2012 to assess changes in nesting habitat.

## **Methods**

### **Analytical Methods**

To assess the status and trend of nesting habitat for marbled murrelets, we used species distribution models (Guisan and Zimmermann 2000) to model the distribution and relative suitability of forests within the NWFP area as marbled murrelet nesting habitat. Specifically, we used the habitat suitability modeling software Maxent (version 3.3.3k, Phillips et al. 2006, Phillips and Dudík 2008) following methods used for the 15-year report and documented by Raphael et al. (2011). Maxent uses a machine learning process to estimate the most uniform probability of occurrence (maximum entropy) at unobserved (background) locations given known constraints (observations of presence data). In other words, it estimates the relative probability of occurrence at unobserved locations throughout the study area by comparing environmental conditions (covariate values) at locations where murrelets nest (presence sites) to conditions at the unobserved locations, assigning a higher probability of occurrence to locations with environmental conditions more similar to presence sites (Baldwin 2009). It uses presence-only data (in our study, known murrelet nesting locations) and does not use locations where the species is known to be absent (to not nest), as data is very scarce on sites where absence has been reliably documented. Maxent is similar to Biomapper software (Hirzel et al. 2002), used to develop the habitat maps in the 10-year monitoring report (Raphael et al. 2006).

When compared to other habitat modelling approaches, Maxent performs as well or better (Elith et al. 2006, Hernandez et al. 2006, Phillips et al. 2006, Merow and Silander 2014). The Maxent approach has been criticized (e.g., Royle et al. 2012, Yackulic et al. 2012; see also the response by Phillips and Elith 2013) because some authors find that presence-only models do not perform as well as presence-absence models. Others find that there are problems with those models as well, primarily due to issues with false

absences (Hirzel et al. 2002). Using a set of murrelet nest locations, Raphael et al. (2011) compared the performance of Maxent with other modeling platforms for predicting nesting habitat suitability and concluded that Maxent performed better; we found no compelling reasons to adopt another modeling platform for the current analysis. In addition, the available data on locations of murrelet absence were more limited in quantity and spatial distribution than the available data on locations of murrelet presence, which favored a presence-only model for our purposes.

Maxent is now the most widely used software for conducting presence-only species distribution modeling (Merow and Silander 2014) and a recent survey of over 300 scientists found Maxent software to be one of the most useful methods currently available for species distribution modeling (Ahmed et al. 2015).

## **Study Area**

Our target area was all habitat-capable land, including both federally administered and nonfederal lands, within the range of the murrelet in Washington, Oregon, and California, except for the portion of the murrelet range south of San Francisco, which is outside the NWFP area. “Habitat-capable” lands were defined as lands capable of supporting forest, and delineated for all of our map-based analyses by a 30-m resolution raster map that represents areas within the NWFP boundary that are capable of developing into forests. This map was created for the 15-yr monitoring reports (Davis et al. 2011; Raphael et al. 2011) and was not updated for this report. It was largely based on the U.S. Geological Society (USGS) Gap Analysis Program (GAP) and the “impervious layer” from National Land Cover Database (Vogelmann et al. 2001; Herold et al. 2003). It excluded urbanized areas, major roads, agricultural areas, water, lands above tree line, snow, rock, and other nonforested features. We used this map to “mask out” nonforested areas for each time period map. Therefore, estimates of habitat area and other analyses in this report only applied to habitat-capable areas.

Our analysis covers only lands within the NWFP and marbled murrelet range. The terrestrial (nesting) portion of the marbled murrelet range was defined during NWFP development and consists of NWFP marbled murrelet inland zones 1 and 2 (FEMAT 1993). Inland zone 1 is where the majority of murrelet nests and detections are located; inland zone 2 is further from the coast and includes areas where detection data indicated only a small fraction of the murrelet population nests (FEMAT 1993). NWFP inland zone 1 extends from the coastline to 40 miles inland in Washington, 35 miles in Oregon, and up to 25 miles inland in California (Figures 2-1 and 2-2). In California, inland zone 1 is narrower toward the southern end of the NWFP area, and inland zone 2 drops out (Figure 2-2); this reflects a narrower distribution of forested potential nesting habitat in that area. As described later in this report, our habitat modeling excluded NWFP inland zone 2 in Oregon and California because of the scarcity or lack of known murrelet nest and occupied sites from those areas with which to train the habitat suitability models.

## Land Use Allocations

The NWFP assigned the federal lands in the plan area to different land uses by creating a number of land use allocation (LUA) classes, where management would vary according to their designated use. These classes were broadly categorized as reserved and nonreserved lands (Huff et al. 2006a). Based on these classes, we summarized murrelet habitat data within the NWFP area using three categories: federal reserved LUAs, federal nonreserved LUAs, and nonfederal lands. In reserved lands, commercial timber harvest is generally not permitted and younger stands, if managed, are managed to attain tree size and stand structure resembling old growth (Thomas et al. 2006). Reserved lands include such areas as national park lands and designated wilderness areas, as well as national forest and Bureau of Land Management (BLM) lands designated as late-successional reserves. In most cases, on nonreserved federal lands, commercial timber harvest is permitted.

Updates to the original 1994 LUA GIS map were produced in 2002, 2009, and 2013 for the 10-year, the 15-year, and the 20-year monitoring reports, respectively. Each successive update improved the accuracy of some of the mapped allocation boundaries based on subsequent work by individual federal entities, as well as corrected some mapping anomalies and inconsistencies (i.e., “gaps” and “slivers”) that had been inadvertently introduced during earlier mapping efforts. More importantly, these updates incorporated major allocation changes that had occurred since the previous mapping effort. Examples of these types of changes include the designation of new wilderness areas and land swaps between federal and nonfederal entities.

Each update represents a significant amount of time and effort on the part of monitoring team personnel, who made every effort to procure and incorporate the best available data at that time. The current (2013) version of the LUA map (Figure 2-1) represents the cumulative result of these three updates. Even so, some issues and limitations remain. These include the inability to map NWFP riparian reserves (which can cover significant amounts of land where stream densities are high) and inconsistencies in how administratively withdrawn areas (e.g., withdrawn from the acres available for timber harvest at the discretion of individual National Forests) were mapped (Davis and Lint 2005, Huff et al. 2006a). The lack of mapped NWFP riparian is because, as Moeur et al. (2005) noted, “...at the Plan scale, they cannot be reliably distinguished from [the adjacent nonreserved or matrix [lands] because of a lack of consistency in defining intermittent stream corridors and varying definitions for riparian buffers”. As those authors note, this affects only NWFP riparian reserves that are not within another NWFP reserve type (such as late-successional reserve) This limitation has no effect on the Maxent model, or on the suitability class assigned to any area, but would affect whether habitat in a riparian area on federal lands is classified as “reserved” or “nonreserved”. This would result in our estimates for reserved federal lands being biased low and estimates for federal nonreserved lands being biased higher by the same amount, than if riparian reserves were mapped. The NWFP initially estimated the amount of riparian reserve within nonreserved LUAs to represent about 32 percent of the nonreserved LUA area of federal lands (USDA and USDI 1994a). Our analyses

(below) assigned about 9 percent of the higher-suitability murrelet nesting habitat on NWFP federal lands to ‘nonreserved’ LUAs, for the baseline year. Applying the 32 percent estimate to the 9 percent of habitat on nonreserved federal lands suggests that about 3% of higher-suitability habitat in federal riparian reserves would be incorrectly classified as nonreserved; this provides a rough estimate of the potential error resulting from the lack of mapped riparian reserves. Another minor issue involves a small amount of federally-owned lands that are awaiting official land use allocation designation. These areas, which represent about 0.1 percent of the total area modeled, are identified as “not designated” in the 2013 map and are reported in the nonreserved category in this report.

Land use allocations within the NWFP area will continue to change. We plan to update the LUA map with the intent of improving it for each successive monitoring effort. Previous versions of the LUA map have been archived, and for monitoring purposes, we always report vegetation and habitat changes within the reference frame of the most up-to-date version.

## **Data Sources for Covariates**

### **GNN Covariates**

Many of the covariates used in our habitat-suitability models were based on gradient nearest neighbor (GNN) maps of forest composition and structure (Ohmann and Gregory 2002). GNN maps were developed specifically for landscape and regional scale analysis and monitoring in forest ecosystems (Ohmann and Gregory 2002; Moeur et al. 2005, 2011; Spies et al. 2007). As part of the NWFP Effectiveness Monitoring program, scientists mapped detailed attributes of forest composition and structure for all forested land in the NWFP area using GNN imputation; the gradient nearest neighbor method integrates vegetation measurements from regional grids of field plots, mapped environmental data, and Landsat imagery to ascribe detailed ground attributes of vegetation to each pixel in a digital landscape map (Ohmann and Gregory 2002; Ohmann et al. 2010; Ohmann et al. 2014). GNN also provides a suite of diagnostics detailing model reliability and map accuracy (see appendix 5 in Davis et al. 2015 for a summary). The GNN analyses created attribute maps for 2 time periods: a baseline year (1993), and a year representing the end of the analysis period (2012).

The resulting GNN vegetation attribute data provided the core source of covariates used for our habitat modeling and mapping, and covered the entire breadth of the species’ nesting range from Washington to northern California for the bookend years of 1993 and 2012. We called these two time periods “bookends” because the changes in habitat that we analyzed and report on occurred between these two endpoints. The satellite imagery from which GNN was created was from 1993 and 2012. The on-the-ground plot data used by GNN to create the vegetation maps covers the period from 1991-2000 for the baseline period, and 2001-2008 for time period two. The resolution of the GNN products we used was 98-feet (30-meter). GNN covariates used in our models (Table 2-1) included MOD\_OGSI\_NWFP (an old-growth forest structure index,

described below), CANCOV\_CON (conifer canopy cover), CANCOV\_HDW (hardwood canopy cover), DDI (diameter diversity index), MNDBHBA\_CON (basal-area weighted mean diameter of conifers), QMDC\_DOM (quadratic mean diameter of dominant conifer trees), TPHC\_GE\_100 (density of conifer trees  $\geq 100$  cm dbh), STNDHGT (stand height), and AGE\_DOM\_BA\_NO\_REM (basal-area weighted stand age based on dominant and co-dominant trees, and excluding remnant trees). GNN attributes also contributed to other covariates, as described below.

For both the 15- and 20-year reports, the GNN covariate maps were developed using Landsat time-series data that were temporally normalized using the LandTrendr algorithm as described below. The GNN covariate data used for the 20-year report included several incremental improvements over the 15-year data and methods as summarized below, some of which were also documented in Ohmann et al. (2014):

**New field plots added:** The 15-year maps were based on plots measured through 2007. The 20-year GNN products included four more years of field plot data, measured through 2011.

**Screening for field plot outliers:** This process benefited immensely from having yearly time-series mosaics from 1984 to 2012, as well as LandTrendr disturbance maps, to aid in determination of timing of disturbances relative to plot measurement. In screening, potential outlier plots were flagged using various algorithms that compared observed plot attributes to the predicted map attributes to identify mismatches. These plots were viewed in the LandTrendr imagery and digital aerial photography to identify and exclude plots that straddled contrasting forest conditions (e.g., older forest and clearcut), or that had been disturbed between plot measurement and imagery dates.

**New spatial predictor:** Spatial predictor variables normally used in the GNN process include maps of abiotic variables such as climate, topography, latitude, and longitude, as well as Landsat imagery of tasseled cap brightness, greenness, and wetness. For the 20-yr report, analysts added the normalized burn ratio (NBR), which is a vegetation index used as a form of change detection. By comparing the NBR before and after a fire or other disturbance event, one can identify the change brought about by that event.

**Matching of plots to imagery:** For the 15-year report, we had only two LandTrendr imagery dates to work with, and plots were matched either to imagery time 1 or time 2. This resulted in as much as a six-year difference between plot measurement date and imagery date. Many plots were excluded from modeling because of disturbance between plot and imagery dates. For this 20-year report, the GNN analysis implemented yearly matching of field plots to LandTrendr data, with plots matched to the same imagery year as plot measurement. This was made possible by having yearly LandTrendr mosaics available from 1984 to 2012. This resulted in many fewer plots excluded because of disturbance, and effectively eliminated differences between plot and imagery dates associated with growth.

While GNN was designed for regional-scale analyses, the maps portrayed forest structure and composition at finer spatial scales. In a recent assessment of the MOD\_OGSI\_NWFP covariate, GNN performed well at the scale of a 30-km hexagon (distance from the center of one hexagon to the next), which covered slightly over 190,000 ac (Ohmann et al. 2014). More information on the GNN mapping for the NWFP, and map products, are available at:

<http://lemma.forestry.oregonstate.edu/projects/nwfp>.

### Platform Covariate

Marbled murrelets in the NWFP area most often nest on larger limbs of coniferous trees. This type of nest location is termed a “platform” and the presence and abundance of platforms are very often good predictors of suitable murrelet nesting habitat (Nelson 1997, Burger 2002, Burger et al. 2010). The PLATFORM covariate was computed from the GNN data using data from previous studies (Raphael, unpublished data) in which numbers of platforms were counted on a very large sample of trees from plots scattered throughout the murrelet range, and then summarized by tree species and diameter class (Raphael et al. 2011). We computed mean numbers of platforms by tree species and DBH (diameter-at-breast-height) class, and then applied these means to tree counts from the GNN data. The mean number of platforms for each species and DBH group (Table 2-1 in Raphael et al. 2011) was multiplied by the associated GNN attribute data on conifer trees per hectare (e.g., TPH\_PSME\_50\_75 for density of Douglas-fir trees in the 50-75 cm DBH class) to estimate total number of platforms per acre. This latter number was used as the covariate value. While the abundance of platforms is likely important for nest habitat quality and nest site selection, other ecological and environmental factors such as vegetative cover, stand characteristics, and local climate, may also be important in nest site selection (Nelson 1997). For this reason we included covariates that represented other factors that could also be important to habitat suitability and nest site selection by murrelets.

### Old Growth Structure Index Covariate (MOD\_OGSI\_NWFP)

The “old growth structure index” (OGSI) was conceptually developed by Spies and Franklin (1988) and further refined by Franklin and Spies (1991). Specifically, it was designed to reflect the continuous nature of ecological succession as opposed to identifying one point along a continuum to separate old growth from younger forests (Franklin and Spies 1991). The OGSI consists of measurable forest structure elements, in our case: (1) density of large live trees, (2) diversity of live tree size classes, (3) density of large snags, and (4) percent cover of down woody material. These are elements commonly considered as key ecological and structural attributes of old growth forests within the NWFP area. Low index values represented younger or less structurally complex forests and high index values represented older or more structurally complex forests. The OGSI covariate we used was based on attribute data provided by GNN, and calculated separately for different forest vegetation types or zones, to account for structural differences among forest types, such as what diameter

constitutes a “large” tree. Davis et al. (2015) developed the OGSi covariate we used, and their publication provides additional details. Whereas Davis et al. (2015) conducted further analyses using the OGSi variable to identify older, late successional forests at different stages of stand development, we used the variable more simply as continuous input covariate in our habitat models.

### Landscape Covariates

Previous studies (Raphael et al. 1995, Raphael et al. 2011, Meyer and Miller 2002) found that murrelets select larger patches of contiguous forest for nesting. To address patch characteristics, we created two covariates, MULTISTORY\_50 and PCTMATURE\_50. These covariates were derived from GNN IMAP\_LAYERS and VEGCLASS covariates, respectively. For each pixel, we evaluated forest condition on a 124-acre (50-hectare) circular neighborhood centered on the pixel, assigning the percentage of the circle in mature-forest condition to the pixel (see Table 2-1).

### Fog Covariate

We included FOG as a covariate because cool summer fog can greatly moderate summer temperatures and humidity near the coast, especially in northern California, where fog plays an important ecological role in the distribution of coast redwood (*Sequoia sempervirens*) forest (Sawyer et al. 2000) with which murrelets are closely associated (Meyer et al. 2002). Fog may also create suitable conditions for the development of epiphytes on branches, contributing to potential murrelet nesting platforms. The fog spatial layer was developed by Henderson, et al. (2011) and represents the average value of effective precipitation added by fog drip and low clouds. One unit of “fog effect” equals 20 inches (508 mm) of effective precipitation. We included FOG in the Oregon and Washington model regions, in part to maintain consistency in covariate sets across the NWFP area.

### Climate Covariates

We obtained monthly 30 year normal (1981-2010) raster climate data (800m resolution) from PRISM models (PRISM Climate Group 2012), including July maximum temperature (JULY\_MAXT) and mean precipitation for the months from May through September. We averaged these five monthly mean precipitation models to produce a mean summer precipitation model (SMR\_PRECIP).

### Other Data Sources

We used 2009 versions of the physiographic province layer (which also defines the NWFP area; FEMAT 1993) and the marbled murrelet range layer to define the extent of our analysis, and report outcomes based on these areas. Revisions to the original FEMAT (1993) physiographic province layer involved correction of state boundaries using 1:24,000 scale digital topographic maps and inclusion of a more detailed, higher resolution coastline which included several islands that were previously omitted. Revisions to the 2004 version of the murrelet range layer were confined to inclusion of the higher resolution coastline. The murrelet’s range south of Canada was divided into the six marbled murrelet “conservation zones” identified in the species’ recovery plan

(USFWS 1997), and the conservation zones were further broken into strata for purposes of population monitoring (Miller et al. 2006, Raphael et al. 2007). Five of these conservation zones (1 through 5) overlap the NWFP area. We extended inland the breaks between conservation zones and strata, primarily by following watershed lines, so that we could summarize nesting habitat data and examine relationships between inland nesting habitat and at-sea murrelet populations at the scale of conservation zones (Figure 2-2).

As described elsewhere in this report, our habitat modeling excluded NWFP inland zone 2 in Oregon and California because of the scarcity or lack of known murrelet nest and occupied sites from those areas. Additionally, other data indicate that murrelets rarely nest in NWFP inland zone 2 in Oregon and California (Alegria et al. 2002, Hunter et al. 1998).

### **Covariate Selection and Screening**

From the literature (including Raphael et al. 2011) and our experience, we selected a candidate set of environmental covariates. In contrast to the 15-year analysis, we did not eliminate correlated variables because our intent was to use the available data to produce the strongest predictive accuracy of habitat suitability and we were less interested in describing the environmental drivers of murrelet distribution and habitat (Jim Baldwin, personal communication; Merow et al. 2013). For this same reason, we used the same full set of covariates in each modeling region. Our final covariate list is summarized in Table 2-1.

#### **Accuracy Assessment**

When screening potential GNN covariates, we considered accuracy assessment data provided by the GNN/IMAP project. The assessments used a form of ground-truthing, by comparing observed values for a grid of field inventory plots with the GNN-predicted (modeled) values for those same plots. This provided accuracy data for eight GNN attributes used directly in our models, as well as for OGSi\_NWFP, upon which the MOD\_OGSi\_NWFP covariate is based (Table 2-2). Accuracy assessments were not available for non-GNN covariates or for derived GNN covariates, but were available for some GNN attributes which contributed to the PCTMATURE\_50 covariate. GNN accuracy assessments are available at:

<http://lemma.forestry.oregonstate.edu/data/structure-maps>

Table 2-2 summarizes, by GNN modeling region, the accuracy assessment results for GNN attributes that contributed directly or indirectly to model input covariates. Accuracy, as measured by correlation “r” values, ranged from 0.37 to 0.81 among modeling regions for the eight attributes used directly as covariates, plus the three that were the basis for the PCTMATURE\_50. When averaged across the 4 modeling regions, covariate accuracy ranged from r values of 0.56 to 0.75 for covariates, and averaged lowest in the California Coast region, where the sample size of field inventory plots was smaller.

The PLATFORM covariate was based in part on GNN attribute data on tree density by species and DBH class. Creating these required the tree density data to be subdivided into 30 categories for each modeling region (5 DBH classes for each of 6 species groups). Accuracy assessments were not available for these categories, however trees equal to or greater than 100 cm DBH contribute the most to platform numbers, and when all conifers of this size class were pooled (TPHC\_GE\_100), accuracy data is available, with correlations ranging from 0.48 to 0.67 (Table 2-2). Thus, accuracy was moderately high for the pooled set of large conifer trees which provide the greatest number of platforms.

### **Data Preparation**

All covariates were processed as ArcGIS (ESRI 1999-2013) rasters at 30-meter resolution, the native resolution of the GNN data. A smoothing function (focal mean) was applied to all covariate rasters, except MULTISTORY\_50 and PCTMATURE\_50, to assign the mean value of the 3 x 3-pixel neighborhood to the cell. We used this smoothing function to reflect the spatial uncertainty in our murrelet location data, but still maintained a spatial resolution < 1 ha. All covariate rasters were converted to ASCII files for input into Maxent and Maxent ASCII output back to rasters using ArcGIS.

### **Murrelet Locations**

We used agency records to identify two types of murrelet nest locations to serve as species presence sites for training the Maxent models: known nest locations, and stand locations where murrelet occupancy behavior was observed during audio-visual surveys of potential habitat (Evans Mack et al. 2003), using all available records through 2013. In both cases, we used only records where inspection of digital aerial photographs confirmed that undisturbed forest was present at the location in 1993, our baseline modeling year. As described in a previous report (Raphael et al. 2011), we initially focused on known nest locations, but this yielded relatively small sample sizes, and did not always provide representative spatial distribution across potential murrelet habitat. Therefore, we added a random sample of “occupied” sites equal in number to the sample of nest sites for each state. We used an equal number of occupied and nest sites to minimize any potential bias in one data set or the other, as neither data set was collected via random sampling and may have biases. For example, many of the occupied sites were surveyed prior to timber harvest, so site selection for these surveys was guided by timber considerations. We assumed that when pooled, the nest and occupied location data used in our habitat modeling represented the breadth of possible murrelet habitat types (McShane et al. 2004; Nelson et al. 2006, Raphael et al. 2006). These known nest and occupied location data (Figure 2-2) were used to train the Maxent habitat suitability model.

Location data for known nest and occupied sites were collected from a variety of sources. In Washington, the source was a database maintained by the Washington

Department of Fish and Wildlife. California sources included a database maintained by the California Department of Fish and Wildlife, supplemented by records assembled by the US Fish and Wildlife Service, and also included a number of nests located by a radiotelemetry study (Hebert and Golightly 2008). For Oregon, data sources were from a database currently maintained by Oregon State University and populated with records from the US Forest Service, the US Bureau of Land Management, the Oregon Department of Forestry, Oregon Parks and Recreation Department, and Oregon State University (S.K. Nelson, unpublished data).

For known nest locations, we included: (1) known nest trees located by visual observations and by radiotelemetry of nesting murrelets ( $n = 134$ ); (2) sites where downy, flightless murrelets had been found on the ground ( $n = 7$ ); and (3) sites where murrelet eggshells had been found on the ground, typically at the base of a suitable nest tree ( $n = 43$ ). Numbers of locations of downy young plus eggshells were 9, 4, and 33 in WA, OR, and CA, respectively. For the occupied sites, behaviors are considered evidence of nesting at or near the location of the behavior (Evans Mack et al. 2003). For our analyses, these behaviors included one or more of the following: murrelets circling at or below the forest canopy; circling above the canopy by no more than 1.0 canopy height; flying through in a straight flight path below the canopy; landing in, perching, or departing from a tree; or birds emitting  $\geq 3$  calls from a fixed point in a tree within 100 meters (328 feet) of an observer (Evans Mack et al. 2003).

We manually screened the data on known nest site locations with the aid of aerial photography, base GNN vegetation mapping data, and communications with original data sources to confirm and correct locations and remove duplicate records. We further screened the data so that all presence locations included in the final dataset were greater than 30 meters apart (the resolution of our modeling).

The selection of occupied sites used to train the Maxent model involved a series of filters and screenings. Filters were used to eliminate duplicate sites and those that fell within 164 feet (50 meters) of a known nest site. Additionally, the Washington and Oregon databases were so robust (4,900+ and 4,300+ records respectively), that a filter was applied to randomly eliminate sites within 5,774 feet (1,760 meter) of each other. This was done in an effort to maximize the distribution of the points among different habitat stands, as well as to reduce the number of records in the databases to a more manageable size for the manual screening process. The subset of occupied sites produced by the filtering process was then screened by manual inspection of each site location using digital aerial photography. Sites were eliminated if forest conditions at the site were clearly non-habitat (e.g., clear-cut, young forest, roadway, open water) in the baseline year. In total, about 20 percent of potential occupied sites were eliminated in this last process. Finally, a stratified random selection was made from the remaining sites equal to the number of known nest sites within a state, and stratified by physiographic provinces within states proportional to the amount of habitat-capable lands in each province. In Oregon and California, we limited our habitat analysis to the NWFP inland zone 1 (Figure 2-2) because of the scant evidence for murrelet use of inland areas (Hunter et al. 1998; Alegria et al. 2002). In California, we found no records of murrelet use of NWFP inland zone 2. In Oregon, there are 9 known occupied sites in

NWFP inland zone 2 that are all clustered near the boundary with NWFP inland zone 1 in a small area near Roseburg. Elsewhere in Oregon, data are lacking (no surveys conducted or data not available in database); an exception is the Siskiyou Mountains in the southern end of the state, where surveys found negligible use in NWFP inland zone 2 (no detections in 3,300 survey visits, Alegria et al. 2002). In Washington, the occupied sites from NWFP inland zone 2 were much more evenly distributed spatially and with respect to distance from the coast, and we decided to include NWFP inland zone 2 in our analysis area.

## **Habitat Change**

We used two methods to assess change in the amount and distribution of habitat from the 1993 baseline to current conditions (2012). For the first method (the “bookend approach”), we compared amounts of habitat estimated by the Maxent models for two time periods: (1) the baseline year, and (2) 2012, which we obtained by projecting the Maxent model from the baseline period onto a map of the covariate values for 2012. Projecting the model in this manner could result in a projected model with validity issues if covariate values in the 2012 data were outside the range of covariate values in the baseline study area used to build the Maxent model (Phillips et al. 2006). However, all covariate values were within the range of baseline values. Using this approach we estimated net change during the period, which represented the difference between area of nesting habitat gains (change in Maxent score from below to above our threshold) and losses (change in Maxent score from above the threshold to below that threshold). Our second approach, “LandTrendr-verified”, used both the bookend model loss results and LandTrendr data to estimate habitat loss from the baseline condition, and to identify causes of observed losses.

For the second approach, we calculated losses as follows: first, we used the Maxent bookend model results from each state to identify areas that had changed from ‘suitable’ to ‘unsuitable’ habitat during the analysis period. We then examined these bookend losses spatially by using a reclassification of the LandTrendr change maps (see below) from the same time period, which identified four disturbance types: wildfire, timber harvest (primarily harvest, but can include short-term disturbances other than fire and harvest), insect and disease (and other long-term disturbance agents) and other natural disturbance. Lands within our study area that were not classified by LandTrendr were assumed to have ‘no disturbance’. As described below, the ‘other natural disturbance’ class was used only in Congressionally Reserved or Administratively Withdrawn (CRAW) lands, which primarily comprise national parks, wilderness areas and national wildlife refuges, plus other federal lands identified for uses which do not include timber harvest. We considered bookend losses that overlapped one of the four disturbance classes as ‘verified’ by LandTrendr. If both the bookend analyses indicated a loss of suitable habitat for that pixel and the LandTrendr data also indicated a disturbance it was assigned a particular disturbance type.

Differences between these 2 methods of estimating habitat change are (1) for the “bookend approach” we used the net change in habitat as a result of gains and losses, while the “LandTrendr-verified approach” estimated only losses, and (2) the latter method used information from two sources (the Maxent models and LandTrendr) to

estimate losses, and provided data on cause of habitat loss. The strengths and weaknesses of the two approaches are addressed in the Discussion.

#### LandTrendr Change Detection

Davis et al. (2015) provide a detailed description of this topic, which is summarized here. The annual time-series of LandTrendr-generated maps identified where, when, how much, and how long disturbances had occurred between 1993 and 2012. They also showed areas where the forest vegetation had been stable or was recovering. These time series maps of forest vegetation disturbance and recovery were similar to what was used in the 15-yr monitoring reports. They were developed following methods in Kennedy et al. (2010, 2012) and verified for accuracy using the TimeSync method (Cohen et al. 2010). The LandTrendr maps represented three aspects of vegetation change: (1) year of disturbance, (2) magnitude of disturbance, and (3) duration of disturbance. Davis et al. (2015) further classified disturbances identified by these maps to produce a map of where timber harvesting, wildfire, insect and disease, and other natural disturbances (e.g., blowdown, floods, landslides, etc.) occurred between 1993 and 2012 (see Appendix 4 in Davis et al. 2015 for details). Where this map overlapped losses of suitable murrelet nesting habitat, it helped to explain the causes for habitat loss since the NWFP's implementation.

Landsat imagery for the NWFP area was acquired from the United States Geological Service (USGS) Glovis website for the summer period (usually July and August) from 1984 to 2012. Images were atmospherically corrected (Masek et al. 2008) and, to minimize cloud coverage, multiple image dates within a given season were used to produce a clear-pixel composite image for that year.

The composite imagery was then processed using the LandTrendr segmentation algorithm (Kennedy et al. 2010), which computes the NBR spectral index for each pixel in the time-series. The NBR (van Wagtenonk et al. 2004; Key and Benson 2006) is a vegetation index for which change in NBR can be used to detect vegetation change. By comparing the NBR before and after a fire or other disturbance event, one can identify the magnitude of change brought about by that event. The algorithm identified year-dates (vertices) where changes in NBR had occurred (normally associated with changes in vegetation). Between vertices, temporal segments were established and each segment was labeled as disturbance, recovery, or stable based on spectral direction. These data were used by Davis et al. (2015) to identify the year and duration of disturbance. For each segment Davis et al. (2015) estimated the percentage of vegetation cover for the beginning and ending vertices using a statistical model that relates NBR to vegetation cover (Cohen et al. 2010). The difference between vertex predictions represented the magnitude of the disturbance or recovery for that segment in terms of percent vegetation cover. Finally, the absolute magnitude was scaled to the starting value to compensate for varying pre-disturbance forest cover values, such that all magnitudes were expressed as a proportion of the starting condition. The final step was to spatially filter the pixels to a minimum mapping unit of 11 pixels, or about 1 hectare. For a given area, the change detection analysis used the highest magnitude disturbance that occurred throughout the time series. Kennedy et al. (2012) found

overlapping disturbances to be rare in the NWFP area, covering less than 5 percent of the area of primary disturbances.

The next step in identifying causes of disturbance was to separate short-duration events, such as timber harvest, from slower, longer-duration causes such as insect damage. An analysis of wildfire and timber harvest unit polygons showed an average disturbance duration signal of about  $1.5 \pm 1.0$  years (mean  $\pm 1$ SD) for regeneration (e.g., clear-cut) harvests,  $1.7 \pm 1.7$  years for wildfires, and  $2.2 \pm 2.2$  years for thinning harvests (Davis et al. 2015). Based on this information, the duration map was classified into a binary map of fast (1-4 years duration) and slow ( $>4$  years) disturbance. Fast disturbance represented abrupt events such as a wildfire, timber harvest, wind blowdown, or debris flow. Slow disturbances represented insects or disease, or post-fire mortality. To help identify wildfire disturbances, Davis et al. (2011; 2015) used maps of the perimeters of wildfires that occurred between 1993 and 2012. The maps of wildfire disturbance are fairly inclusive of all major wildfires exceeding a few acres, but the many wildfires smaller than this were not mapped. Insect damage detection data (USDA 2008) helped classify insect-caused disturbance. Finally, LUAs helped identify areas (CRAW lands) of fast disturbance where timber harvest was unlikely. Agency “forest activity” GIS layers were used where available to identify harvest areas, but similar GIS layers were not available for timber harvest on nonfederal lands. Davis et al. (2015; appendix 4) provide a complete classification rule set used to produce the map of cause of disturbance and classification accuracy assessment.

The disturbance maps did not capture all disturbances that occurred during the final year of our analysis (2012), particularly for those disturbances that occurred in 2012 after the satellite image acquisition date, or that were obscured by smoke from wildfires. Those changes will be captured in subsequent monitoring efforts. Our disturbance classification map provides a general sense of the amount and change agents behind losses of suitable murrelet nesting habitat between 1993 and 2012.

Specific criteria used to identify causes of change were:

*Timber Harvest* – Represents timber harvesting; including thinning and regeneration harvests. Classified as fast (duration  $\leq 4$  years) disturbances outside of CRAW lands and also outside of wildfire perimeters; or if within a wildfire perimeter, then predating the fire year by more than two years. Some fast disturbances meeting these criteria occurred within areas identified as having insect damage. Visual inspection with high resolution aerial imagery indicated that most of these disturbances were from timber harvesting, some likely related to insect damage salvage (Davis et al. 2015). Due to a lack of spatial data on landslides, floods and blowdown, the ‘timber harvest’ category includes some fast disturbances due those other causes, when located outside of CRAW lands.

*Wildfire* – Fast or slow (duration  $>4$  years) disturbances within a mapped wildfire perimeter, but only when no other disturbance preceded the fire year. Slow disturbance within wildfire perimeters likely represent post-fire mortality.

*Insect and Disease* – Slow duration disturbances that occurred outside of wildfire perimeters, or if within a wildfire perimeter, preceded the fire year. Also includes fast disturbances that occurred within areas identified as having insect damage, where mapped for two or more consecutive years.

*Other Natural Disturbance* – Fast disturbances that occurred within CRAW lands and outside of wildfire perimeters, or if within fire, then preceded fire year. Includes blowdown, floods, and landslides. Due to the lack of data that would allow us to distinguish timber harvest from other fast natural disturbances (other than fire), this cause was only used within CRAW lands, where timber harvest was very unlikely.

### **Model refinements**

Once we selected our final set of covariates, we conducted a series of Maxent model runs to evaluate model performance. To evaluate model performance, we used training and test model gain, and area under the curve statistics (AUC; Fielding and Bell 1997, Boyce et al. 2002). Gain is closely related to deviance, a measure of goodness of fit used in generalized additive and generalized linear models and is available as part of the model output in Maxent (Phillips et al. 2006). The lowest value of gain is 0 and gain usually increases towards an asymptote as the fit between the model and the training data improves. During a run, Maxent is generating a probability distribution over pixels in the grid, starting from a uniform distribution and repeatedly improving the fit to the data. The gain is defined as the average log probability of the presence samples, minus a constant that makes the uniform distribution have zero gain. At the end of a run, the gain indicates how closely the model is concentrated around the presence samples; for example, if the gain is 2, it means that the average likelihood of the presence samples is  $\exp(2) \approx 7.4$  times higher than that of a random background pixel (Phillips, unpublished tutorial, available at <http://www.cs.princeton.edu/~schapire/maxent/>). For a given model run, separate gain statistics were generated for the training (75 percent) and test (25 percent) portions of the available presence sites.

The other measure of model performance, AUC, is the area under a ROC (receiver operator characteristic) curve (Boyce et al. 2002, Hirzel et al. 2006). AUC is a measure of model performance that illustrates how well one can distinguish presence sites from the available background sites (some of which are likely to be occupied by and/or suitable for murrelets). Values range from 0 to 1.0 and location data that cannot be distinguished from the background with any greater probability than a random coin toss would yield an AUC score of 0.5. We present AUC values generated using test data, which is data held back during model development and then used to test model fit and accuracy. Test AUC provides a measure of model performance in classifying an independent set of presence points.

Maxent also provides a choice of covariate relationships to include in a model, called “features.” Feature types include Linear, Quadratic, Threshold, Hinge, and Product. These features set the possible shapes of the relationship between a covariate and the response (i.e., the Maxent probability distribution) or allow for covariate interactions

(product features). A user can select any combination of these feature types. A model with linear features requires the fewest parameters, as only 2 parameters (slope and intercept) are estimated for each covariate. Quadratic relationships require both slope and intercept as well as exponent parameters for each covariate. Hinge features create a piece-wise approximation to any distribution. The number of parameters for any one covariate increases for each “hinge” in the modeled distribution which can result in a complex distribution and many parameters. The Product feature allows for interactions among all pairs of covariates. The total number of parameters for any model depends, therefore, on the types of features selected and the complexity of the response curves between the covariates and the probability scores. In addition, Maxent has a “regularization” constant that can be specified. Increasing the regularization value above the default has the effect of smoothing the response curve, thereby reducing the number of parameters in the model. Regularization is a common approach in model selection to balance model fit and complexity, allowing both accurate prediction and generality (Elith et al. 2011). Maxent uses a default regularization setting of 1.0, which is derived for a given set of training sites and designed to achieve this balance (see Phillips and Dudik 2008 and Elith et al. 2011 for a thorough examination of the regularization settings). A regularization setting less than 1.0 produces an output distribution that is a closer fit to the training sites, but which can result in overfitting, and values greater than 1.0 will provide a more spread out, less localized prediction (Phillips, unpublished tutorial, available at <http://www.cs.princeton.edu/~schapire/maxent/>). Based on our initial model evaluations, we used Maxent’s default regularization value of 1.0 in the models reported here.

For the previous (15-year) report, we ran a number of Maxent models, each time varying the set of features we selected and the setting for regularization. We then plotted AUC and gain for each model against the total number of parameters required by the model. One would expect greater gain and greater AUC in models with larger numbers of parameters, just as a regression model with more covariates will generally explain more variance in a dataset than a model with fewer covariates. The penalty for large numbers of parameters can be overfitting the data. If the model is overfitted to training data, then it will perform badly when applied to new data (i.e., test data that were not used to create the model). We used this method to refine models for each of the three states, and in each case the Linear plus Quadratic plus Product features performed best relative to numbers of parameters required (Raphael et al. 2011). Based on the results of the model selection process reported by Raphael et al. (2011), all models reported here used those three features.

### **Summarizing Maxent Output**

For each state, we ran the Maxent model by using the combined sets of nest sites and occupied sites as training data, and using 1993 covariate values to build the final habitat suitability models, with 25 replicated model runs, each of which produced a model as well as a map of 1993 habitat suitability. This approach differs slightly from those of the analysis of the 15-year report (Raphael et al. 2011), which used 10 replicated model

runs. We used a larger number of replicates to better represent the central tendency and variation for a given model. For each set of these 25 model runs we set Maxent to partition the presence sites into 75 percent to be used to train the model, and withholding 25 percent for testing the performance of the resulting model. We retained this approach for the final model runs for each modeling region because the replicated model iterations with randomly-partitioned presence sites provided data to assess the average behavior of the models; this also allowed for statistical testing of performance (see below). Because the presence sites were re-partitioned for each of the 25 replicate model runs, the resulting models and maps differed among the replicates. To estimate 2012 conditions, each of the 1993 model replicates for a state was then projected onto 2012 conditions (covariate values). Thus, for each state (modeling region) the result was 25 maps of habitat conditions for 1993 and for 2012. The Maxent modeling platform also produced maps with the average habitat suitability scores for each state for 1993 and 2012, based on the 25 replicate maps, and maps of the standard deviation of the 25 scores; Maxent computed the average and standard deviation scores at the pixel scale.

Once we selected our final model structure, we used *k*-fold crossvalidation to build our models in each modeling region (i.e., for each state) and computed an area-adjusted frequency index (AAF, Boyce et al. 2002, Hirzel et al. 2006) from the set of 25 replicated model runs for each state and version (Figure 2-3). The primary output from the Maxent model is a logistic probability for each pixel in the model region. The logistic probability can be interpreted as the relative likelihood that the conditions at a given pixel are suitable habitat for nesting murrelets. The AAF method as applied here used the model's logistic probability values for test sites (the 25 percent of presence sites randomly chosen by the model as test sites) to evaluate whether, for a given range of model logistic scores (or score class), the presence site values occur in that score class more or less than expected by chance. Within a given score class, the AAF method calculates "P/E" ratio (predicted ÷ expected), with the numerator being the model's predicted frequency of test sites and the denominator the expected frequency of test sites within the class, if test sites were randomly distributed across the modeling region.

To compute AAF indices (P/E ratios), we subdivided the range of a Maxent models' logistic probability output values, which scale continuously from zero to one, into a set of overlapping probability value classes, each representing a subset of the full range. We used a moving window of 0.30 width and a resolution of 0.05 to perform these calculations (Boyce et al. 2002, Hirzel et al. 2006); thus each overlapping class was 0.30 units wide and centered on points 0.05 units apart. These settings differ from those used for the 15-year report (Raphael et al. 2011); preliminary exploration of settings indicated that these settings produced smoother P/E curves (Hirzel et al. 2006) for the current set of model output. We calculated the AAF index for each of those classes as the relationship between the proportion of all murrelet locations (training sites) with estimated logistic probability values in that class divided by the proportion of the available landscape that is estimated to have probability values in that same class. Values less than 1.0 indicate that the proportion of murrelet locations in those probability classes were less than the proportion in the landscape, whereas AAF values > 1 indicate the proportion of murrelet locations in those probability classes were greater

than the proportion of the landscape in those same probability classes. For example, if 1% of the landscape was estimated to fall within a logistic probability value class centered on 0.8, but 10% of the murrelet locations were estimated to have logistic probability values for that same class, the AAF value would be 10. This indicates that murrelets were much more likely to occur (nest) in the 0.8 probability class than expected by chance (that is, if responding at random to the environmental conditions in our covariate set).

To evaluate the status and trend of murrelet nesting habitat, we divided the logistic probability scores among 4 categories, based on the results of the 1993 models. Using categories was necessary to convert the continuous scores into a form that allowed computation of acres of habitat, and to describe how those categories changed over time. To accomplish this categorization, we first computed the AAF values from the test data for each of the 25 runs, then computed a mean and SD for the AAF values across all Maxent logistic probabilities (Figure 2-3). We subsequently used the point where the P/E ratio equals 1.0 (that is, where the predicted frequency of test sites equals the expected frequency of test sites) as a threshold to separate “higher-suitability” habitat from “lower-suitability” habitat (the dashed line dividing Class 3 from Class 2 in Figure 2-3). We performed a further separation of Maxent scores below the P/E threshold into two classes by calculating the mean score of all pixels with logistic scores below the P/E threshold, and using the mean to create two lower classes of suitability. For pixels above the P/E threshold, we computed the mean logistic probability score for all nest and occupied locations used for modeling in that region (state) and used that mean to separate the two higher classes of habitat suitability (Figure 2-3). Thus, we created four classes of habitat suitability:

**Class 1 (lowest suitability):** includes all pixels with logistic probability values between (a) zero and (b) the mean logistic value for all pixels below the P/E=1 threshold.

**Class 2 (marginal suitability):** includes all pixels with logistic probability values between (a) the mean logistic value for all pixels below the P/E=1 threshold, and (b) the P/E=1 threshold.

**Class 3 (moderate suitability):** includes all pixels with logistic probability values between (a) the P/E=1 threshold and (b) the mean logistic value for the mean logistic score for all nest and occupied locations used for modeling in that state. One can also think of this category as including those pixels for which the relative likelihood of murrelet presence exceeds that expected by chance, given the set of environmental conditions (covariate values) at that pixel, while excluding those in the highest-suitability Class 4.

**Class 4 (highest suitability):** includes all pixels with logistic probability values between the mean logistic score for all nest and occupied locations used for modeling in that state, and the maximum score (generally close to 1.0). Thus, this class included pixels with probability scores equal or exceeding the average score of the presence (training) sites used in that state’s model. One can think of this class as approximating locations where environmental conditions equal to or exceeding those of the average nest/occupied location in our data set.

After calculating the values used to classify the 1993 model output (logistic scores) for each state into these four classes, we applied those same values to the 2012 maps of habitat suitability (logistic scores) which had been created by projecting the 1993 models onto the 2012 covariate conditions. Therefore, for each state, the same models and thresholds were used to model habitat suitability and then to classify the resulting habitat suitability maps for both the 1993 baseline and the 2012 bookend.

To portray variability in our estimates of amounts of suitable nesting habitat, we computed the standard deviation and mean acres of higher-suitability habitat (Classes 3 and 4 combined). To do this calculation, we obtained the 25 sets of logistic probability maps from the 25 replicated Maxent runs, calculated acres above the threshold from each map, and then computed the mean and standard deviation of those acre values from the 25 replicate maps for each study region (state).

We tested for a change in nesting habitat area from 1993 to 2012 by running matched-pair t-tests on the estimates of higher-suitability habitat at the scales of state, physiographic province, and main land use allocation class (nonfederal, federal reserved, federal nonreserved). We used a matched-pair approach because we wanted to control for variability among the 25 replicated model runs conducted for each state model. Each replicate was based on a model built on the 1993 environmental data and that same model was projected to the new 2012 environmental data, thus providing a matched pair consisting of 1993 and 2012 habitat estimates for each replicate. We used those matched estimates for each replicate to compute the difference in acres of higher-suitability habitat from 1993 to 2012 for each replicate, and then computed the mean difference across all replicates to test the likelihood of observing those data under a null hypothesis of no difference in mean acres between time periods.

The plot of P/E against the mean habitat suitability of each class (Figure 2-3) provides a test for model performance, as a good model is expected to show a monotonically increasing curve (Boyce et al. 2002), for which we tested using the Spearman rank correlation coefficient (Boyce et al. 2002; Hirzel et al. 2006).

### **Landscape Habitat Pattern – Edge versus Core**

Marbled murrelet nest success is reduced along forested edges, due to higher rates of nest depredation near edges (Manley and Nelson 1999, Raphael et al. 2002, Malt and Lank 2007). For that reason, we investigated the configuration of potential habitat by computing how much of that habitat occurred along edges versus within forest interior conditions. We used the morphometric spatial pattern analysis (MSPA) tool in the GUIDOS (Vogt 2013) toolbox to characterize murrelet nesting habitat configuration. The GUIDOS toolbox is a software program with functions for processing and visualizing spatial data. MSPA is an algorithm that classifies pixels in a binary raster image based on their geometry and connectivity to other pixels. Contiguous clusters of habitat pixels are considered a patch. Each patch has a zone of 'edge' habitat along its periphery, defined by the 'edge depth' parameter. In the middle of the patch, inside of the edge zone, pixels are classified as 'core' habitat. MSPA may optionally further classify certain

edge pixels based on their configuration in areas of transition between core and edge. These transitional classes include 'islets' (edge pixels in patches too small to contain core habitat), 'bridges' (edge pixels that link separate patches of core habitat), 'loops' (edge pixels that connect lobes within a core habitat patch), 'perforations' (the edge of a non-habitat patch contained within a habitat patch), and 'branches' (narrow protrusions of edge habitat extending outwards from patches of core habitat). In addition, MSPA allows edge classes to be further classified as combinations of the above subtypes (e.g. 'loop in edge', or 'bridge in perforation').

We performed MSPA based on the maps of higher-suitability nesting habitat (classes 3 plus 4) identified for each state, for both 1993 and 2012. MSPA parameters included an edge depth of 90m, an 8-cell connection rule (i.e., diagonal connections were allowed when assigning pixels to habitat patches), transitions were enabled (i.e., transitional classes were allowed), and 'intext' was enabled (i.e., combinations of edge classes were allowed). Classification using these settings produced 11 MSPA classes, which we aggregated into three classes, including:

1. Core (only includes 'core' MSPA class). This class represents core higher-suitability habitat that is further than 90m from an edge.
2. Core-edge (including MSPA classes 'islet', 'loop', 'bridge', and 'branch'). This class represents edges of higher-suitability habitat within 90m of core habitat.
3. Edge (including MSPA classes 'perforation', 'edge', 'loop in edge', 'loop in perforation', 'bridge in edge', and 'bridge in perforation'). This class includes narrow ribbons of edge higher-suitability habitat which occur more than 90m beyond core pixels, or isolated patches too small or narrow to contain core pixels.

## **Human Disturbance**

Marbled murrelet nest site selection is thought to be sensitive to human modification of the landscape. We quantified human landscape modification based on a 'human footprint' model (Leu et al. 2008) of anthropogenic impacts in the western United States. This model considers human habitation, roads, railroads, irrigation canals, power lines, linear feature densities, agricultural land, campgrounds, highway rest stops, landfills, oil and gas development, and human induced fires. These impacts were summarized spatially in a raster model with values corresponding to 10 human footprint ranks, with a value of 1 being the least modification and 10 the greatest. To assess the amount of potential nesting habitat potentially degraded by these human impacts, we calculated the acres of nesting habitat within each of the 10 ranks, summarized by state and bookend year. In addition, to determine if the degree of human modification of the landscape differed between suitable core and edge habitats, we calculated the mean human footprint rank of habitat pixels in each landscape pattern class (core, core-edge, and edge).

## Results

### Covariates

Mean values of each covariate differed between training sites and study region (Table 2-3). Maxent output includes estimates of the relative contribution of each covariate to the final model (Table 2-4). These values are estimated by Maxent during the iterative model optimization process, and are based on the increase in training gain associated with each covariate. The contribution values should be interpreted with caution for covariates that are highly correlated. This is because there is an element of chance in how the percent contribution is divided among highly correlated covariates; one of a pair of such covariates may be assigned a high contribution and the other a low contribution when in fact both may be important to the species ((Phillips, unpublished tutorial, available at <http://www.cs.princeton.edu/~schapire/maxent/>). As described earlier, we included correlated variables in models because our intent was to use the available data to produce the strongest predictive accuracy of habitat suitability.

Overall contributions of the covariates show that MOD\_OGSI\_NWFP, FOG, and PLATFORMS made the greatest contributions to the Washington model; SMRPRECIP, MOD\_OGSI\_NWFP, and STANDHT were the strongest in Oregon, and CANCOV\_CON, PLATFORMS, FOG, and SMRPRECIP were the strongest in California (Table 2-4). Combining the results from Tables 2-3 and 2-4, and focusing on the covariates with the greatest model contributions, we see that training sites had, on average, much higher values for covariates indicative of older forest (old-growth structure index, stand height, and platform density), compared to available (all habitat capable lands). Training sites also had greater mean conifer canopy cover and fog index values. The pattern for summer precipitation differed among states, being higher for training sites compared to available in Washington and Oregon, but lower for training sites in California (Table 2-3). For the landscape-scale covariates, in all modeling regions the nest and occupied sites had on average more than twice the amount of forest in older-forest vegetation classes within the 50-hectare circles centered on the training site (PCTMATURE\_50), compared to the random pixel. At this same scale, the number of canopy layers (MULTISTORY\_50) tended to be greater at training sites, but not as much as for PCTMATURE\_50 (Table 2-3). Plots of Maxent scores against each of the covariates (Figure 2-4) show that all covariates except CANCOV\_HDW have either positive linear relationships (increasing values of covariate have increasing Maxent scores) or quadratic relationships (as in AGE\_DOM\_BA\_NO\_REM, PLATFORMS, and TPH\_GT\_100). For example, PLATFORMS has a quadratic relationship in Oregon and Washington (Figures 2-4a and 2-4b), indicating that the habitat suitability score (Maxent logistic output) is greatest for intermediate densities of platforms. We might hypothesize that this relationship is due to platforms being less dense, but of higher quality (larger on average, for example) in old-growth forest dominated by larger but less dense trees, compared to a younger forest which supports a higher density of trees large enough to provide platforms, but the average platform is smaller, and perhaps less likely to be selected as a nest site by a murrelet. In California, this relationship is asymptotic rather than quadratic (Figure 2-4c), perhaps because forests available to nesting murrelets in California differ from forests to the

north by being dominated by coast redwood trees, which may have different platform characteristics than the tree species used to the north.

Another way to evaluate contributions is to compare training gain of each covariate modeled alone against the gain from the global model (when all covariates are included) and to compare the effect on global gain when that covariate is removed and all other covariates are retained (Figure 2-5). Covariate contributions evaluated in this way differ somewhat from the previous comparisons. Evaluated in this way, the strongest covariates in Washington are MNDBHBA\_CON, PLATFORMS, QMDC\_CON, and DDI. In Oregon, the strongest contributors are PLATFORMS, STNDHGT, QMDC\_CON, MOD\_OGSI\_NWFP, and MNDBHBA\_CON. In California, CANCOV\_C was strongest, followed by TPHC\_GT\_100, PLATFORMS, MNDBHBA\_CON, and STNDHGT.

### Model Performance

We summarized gain for each state in Figure 2-6 and contrasted test gain and training gain. Training gain was estimated from the data used to build the model. Test gain was estimated from the 25 percent of murrelet locations withheld in each Maxent model iteration. If a model were overfit (that is, had an overabundance of parameters) then we would expect training gain to be much larger than test gain. As shown in Figure 2-6, test gain was close to or larger than training gain in all three model regions. Gain also indicates how markedly the model distinguishes the presence samples (nest plus occupied sites) from the background, using the equation  $e^{\text{gain}}$  [also written as 'exp(gain)'], where  $e \approx 2.718$ . For example, if the gain is 2, it means that the average likelihood of all the presence samples is  $\exp(2)$ , or about 7.4 times higher than that of a random background pixel. As measured by test gain, model performance was strongest in California [gain = 3.065, and  $\exp(3.065) = 21.4$ ], indicating a much stronger distinction between murrelet sites and the background area in that state compared with the other states. Test gains were lower in Oregon (gain = 1.634,  $\exp(1.634) = 5.1$ ) and Washington (gain = 2.671,  $\exp(2.671) = 14.5$ ). In all states, test gains were higher than in the NWFP 15-year analysis (Washington: 1.092, Oregon 1.041, California 2.976, Raphael et al. 2011).

Test AUC values were ranked among the model regions in the same pattern as gain: AUC was greatest in the California model (AUC = 0.960) and lower in the models for Oregon (AUC = 0.892) and Washington (AUC = 0.914) (Figure 2-6). For all three models, the plot of P/E values against the mean habitat suitability of each class showed a monotonically increasing curve with high correlation value ( $R_s > 0.99$ ;  $P < 0.001$ ), indicative of strong model performance (Figure 2-3).

### Habitat Suitability

Our models estimated the suitability of conditions for murrelet occurrence at two points in time: the start of the NWFP (1993) and current (2012) (Tables 2-5 to 2-8; Figures 2-7 to 2-13). As summarized in Figure 2-12, most land is classified in the lower suitability

classes 1 and 2, with successively fewer acres in the higher classes above our suitability threshold. The proportion of habitat-capable land that was above the threshold (in Classes 3 plus 4) at the start of the NWFP varied among model regions (states) and land ownerships. In Washington 14.3 percent of all habitat-capable land was classified above the threshold in 1993; in Oregon and California 12.9 and 4.1 percent, respectively were above the threshold in 1993 (Table 2-8). The proportion in suitable classes was greater on federal lands than nonfederal in Washington (25.9 percent versus 8.8 percent) and Oregon (23.4 percent versus 6.7 percent), but lower in California (2.7 percent versus 4.7 percent), where relatively little habitat-capable area occurs on federal land (Tables 2-5 to 2-8). Over all lands, we estimated a total of 2.53 million acres of higher-suitability habitat in 1993 (12.2 percent of habitat capable land, Table 2-8). Most of this nesting habitat (54 percent) was on federally reserved lands but a substantial amount (41 percent) was on nonfederal land. On federal lands over the 3-state region, 91 percent of above-threshold habitat fell within reserves, while 82 percent of all habitat-capable federal lands were within reserves. Among physiographic provinces, the largest amounts of above-threshold habitat on federal lands occurred in the Olympic Peninsula, Western Cascades of Washington, and Oregon Coast (Tables 2-5 and 2-6, Figures 2-7c, and 2-9 to 2-11).

The highest-suitability habitat (Class 4 only) was relatively scarce on the study area compared to Class 3 suitable habitat, and formed a greater percent of habitat-capable lands on federal lands in all states (Tables 2-5 and 2-6; Figures 2-9 to 2-11). Class 4 lands represented 4.4 and 1.9 percent of federal and nonfederal habitat-capable lands in Washington, respectively, 3.8 and 1.1 percent in Oregon, and 0.8 and 0.7 percent in California (based on data in Tables 2-5 to 2-7).

## **Habitat Change**

As discussed above, we used a bookend approach to assess net change in the amount and distribution of habitat from the baseline (1993) to “current” conditions represented by 2012 data. We observed both losses (that is, lands that were classified above our habitat suitability threshold in 1993 but fell below the threshold in 2012) and gains (lands that were below the threshold in 1993 but above the threshold in 2012) (Tables 2-9 and 2-10). Under the bookend approach, which considers net change after accounting for both gains and losses, we estimated that the net amount of above-threshold, higher-suitability habitat declined over all lands from 2.53 million acres to 2.23 million acres (12.1 percent decline). Amount of habitat above the threshold on all lands declined in all three states, by 13.3, 9.2, and 17.8 percent in Washington, Oregon, and California, respectively (Tables 2-5 to 2-8, Figures 2-7 and 2-13). As illustrated in Figure 2-8a, total acres of higher-suitability habitat declined by the greatest amount in Washington (215 thousand acres), and by lesser amounts in Oregon (88 thousand acres) and California (22 thousand acres). Declines in all three states were statistically significant (matched-pair t-test,  $P < 0.01$ ). At the scale of land use allocation, we estimated a decline of 293 thousand acres for nonfederal lands, ( $P = 0.00$ ), a decline of 34 thousand acres for federal reserved lands ( $P = 0.00$ ), and a small increase of 3 thousand acres ( $P = 0.01$ ) on federal nonreserved lands (match-pair t-tests, Figures 2-

7b and 2-8b). At the physiographic province scale (Figures 2-7c and 2-8c), statistically significant declines ( $P < 0.01$ ) were observed in all but the Washington Eastern Cascades and California Klamath provinces; in those two provinces we observed a slight but significant increase in amount of habitat. We observed net declines in amounts of both Class 3 and Class 4 habitat in all three states, and increases in Class 1 (lowest suitability) in Washington and Oregon (Figure 2-13). Proportions of habitat loss (relative to baseline amount of suitable habitat) were roughly similar among physiographic provinces (Table 2-10, Figure 2-7c) but were somewhat greater in the Washington Western Lowlands and somewhat less in the Washington Western Cascades provinces. We also summarized habitat change at the conservation zone scale, which is a primary scale for murrelet population estimation (Chapter 1, this volume). At this scale, the proportionate loss of higher-suitability habitat was greatest in Conservation Zone 2 (-16.1 percent of baseline); losses in Zones 1, 3, and 4 were 10.9, 7.9, and 17.0, percent, respectively (Table 2-11; Figure 2-7d). We observed a gain of 17.2 percent habitat in Zone 5 (Figure 2-7d; Table 2-11); because little habitat exists in this zone, the gain in terms of acres is relatively small.

Loss of higher-suitability habitat was greatest on nonfederal lands (losses were 29.8, 21.1, and 21.8 percent of baseline in Washington, Oregon and California, respectively; Tables 2-9 and 2-10). On nonfederal lands, almost all loss (98 percent) was due to harvest (Tables 2-12 and 2-13). Losses were lower from federally reserved lands, totaling 1.7, 3.8, and 1.1 percent from the three states (Tables 2-9 and 2-10). The cause of loss varied by land ownership, based on the LandTrendr-verified losses. On federal lands, most of this loss of higher-suitability habitat (62 percent) was due to fire and about 23 percent due to harvest (Table 2-12). On federally reserved lands, wildfire accounted for 66 percent of losses (Table 2-12). Most of these losses (62 percent of all losses in reserves) occurred in the Oregon Klamath physiographic province, and from a single fire, the 2002 Biscuit Fire, which was Oregon's largest contiguous, single-year fire on record (Azuma et al. 2004). Outside of the Oregon Klamath province, fire was less dominant as a cause for losses on federal reserved lands, accounting for 12 percent of habitat losses, followed by timber harvest (38 percent) and natural causes (35 percent) (Table 2-12). Across all federal lands, insect and disease accounted for about 5 percent of habitat loss, but 15 percent of losses in Washington; natural disturbances were also mostly a cause of loss in Washington. While timber harvest is generally not allowed in federal reserved lands, some harvest did occur in reserves after NWFP implementation where timber sales had been approved prior to 1994. In addition, as described above under *LandTrendr Change Detection*, the 'timber harvest' category likely includes rapid habitat losses due to blowdown, landslides, and floods. The exception is for the CRAW subset of reserved lands, where no harvest is allowed and thus where fire did not cause rapid habitat loss, other rapid losses could be reliably assigned to the natural disturbance category, which includes landslides, blowdown and floods (Table 2-12).

While at broader scales the amount of higher-suitability habitat declined, some gains in habitat were observed at finer scales, notably for the Oregon Coast Range province, where net gains were estimated on federal lands (reserved and nonreserved allocations) for both Class 3 and 4 combined (higher-suitability habitat), and for Class 4

(highest suitability) alone (Table 2-9). Most notable was the net gain of about 5,500 acres of Class 4 on federal reserved lands in that province. Also at a finer scale, in terms of habitat suitability, the loss rate (as percent of baseline) of the highest-suitability habitat (Class 4) was generally less than (Oregon and California) or comparable to (Washington) the loss rate of higher-suitability habitat (Class 3 plus 4; Table 2-10; Figure 2-8d). For all lands combined, Class 4 habitat losses were greatest in Washington (15.5 percent of baseline), slightly lower in California (12.8 percent) and least in Oregon (0.4 percent; Table 2-10).

### **Habitat Pattern**

The spatial configuration of higher-suitability habitat varied by state and land allocation. We used the ratio of edge habitat (represented as the sum of edge and core-edge) to total habitat (that is, the proportion of higher-suitability habitat that occurs within 90 meters of an edge) to assess habitat configuration patterns. Higher-suitability habitat on nonfederal lands occurred mostly within edges, especially in Oregon and Washington where habitat in edges was about 80 to 90 percent of total habitat (Table 2-14, Figure 2-143); habitat in reserves on federal lands had the lowest proportion of edge habitat in all three states, but that proportion still exceeded 50 percent in all states (Figure 2-14).

In Washington, in both bookend years, about half of all higher-suitability habitat was present in small patches classified as edge by the landscape pattern analysis (Table 2-14). Of the habitat distributed in larger patches, slightly over half was classified as core-edge and slightly less than half was classified as core. The loss of habitat occurring between the bookend years in Washington was approximately equally divided among the three landscape pattern classes. In Oregon the distribution of habitat among the three landscape pattern classes was similar to Washington, except that a greater proportion of habitat was associated with core areas (core or core-edge classes) compared to edge. Habitat losses in Oregon were proportionately greatest from core habitat areas for all land uses, with small gains in the two edge classes on federal lands (Table 2-14). In California, the limited amount of nesting habitat was distributed about equally among the three classes, and the loss between bookend years was mainly in core and core-edge habitat, with limited reductions in edge habitat.

### **Human Disturbance**

Modeled nesting habitat was strongly correlated with areas of low human footprint. In each of the three states and in both bookend years, over 95% of nesting habitat was in five lowest human footprint ranks, over 80% was in three lowest human footprint ranks, and over 50% was in the two lowest human footprint ranks. The most common human footprint rank of nesting habitat in all states was 2 (on a scale of 1 to 10, where 10 is the highest human modification). The mean human footprint rank (Table 2-15) within the three landscape pattern classes was approximately equal in Oregon and California in both bookend years, however, in Washington, the human footprint rank increased

somewhat from core habitat to edge (Table 2-15). Overall, the mean human footprint rank in all landscape pattern classes was lowest in Washington, highest in California, and intermediate in Oregon.

## Discussion

### Sources of Uncertainty

This work represents a third update of a range-wide map of potential murrelet nesting habitat from consistent baseline vegetation information. We believe the effort has resulted in a more robust understanding of the current amount and distribution of nesting habitat (based on a model using satellite data) compared to the information available at the time of our earlier reports (Raphael et al. 2006, Raphael et al. 2011). As in our previous efforts, there are a number of sources of uncertainty that should be recognized.

#### Vegetation Mapping

First, there is uncertainty and error in the underlying GNN vegetation classification. We have previously discussed accuracy assessment information for the vegetation data (see *Methods* above). Error rates in the original vegetation attributes such as tree diameter and canopy cover varied among modeling regions, but on average the accuracy assessments of the GNN covariates indicated moderate to moderately high accuracy in predicting those attributes, as indicated by correlation coefficients. However, at the scale of GNN modeling region, correlations between GNN predictions and ground-based measurements for pixels province-scale sometimes fell below 0.5, indicating lower accuracy by the GNN model due to GNN model error in predicting that vegetation attribute within a GNN modeling region. Some of our covariates were derived from combinations of GNN covariates (such as PLATFORMS) and we do not have a measure of accuracy of these derived covariates. In general, we can assume that finer-scale covariates (such as the count of stems in diameter classes) will be less accurate than more broadly defined covariates. Another derived covariate is PCTMATURE\_50 and while we have an accuracy assessment for some of the GNN attributes, such as QMDA\_DOM and CANCOV, used by GNN to classify pixels as large conifer, we do not know the accuracy of our estimate of the percentage of a 50-ha circle that is classified as large conifer.

Resolution is also a source of uncertainty. In general, finer resolution data, such as the 30-meter resolution GNN data, will show more variation and detail than coarser resolution data. Engler et al. (2004) found that models using higher resolution (finer-scale) habitat predictors performed better than models using coarser resolution data (82-ft versus 1,640-ft resolution raster data). The lower model performances they observed at the 1,640-ft resolution (roughly 62 acres pixel size) were probably caused by a loss of information that is inevitable when aggregating environmental maps. This aggregation may, in some cases, hide important combinations of habitat predictors which would be expressed with finer resolution data. Our method of computing a 9-

pixel average for all covariates helped reduce the effect of errors at the single-pixel scale resulting from imagery noise, while retaining much of the fine-scale richness of the GNN data.

Errors in GNN attribute data also resulted in some model covariate values that did not match the actual vegetation on the ground. For example, we checked GNN attribute data against aerial imagery for murrelet nest and occupied sites used to train the model, and in some cases observed mismatches, where aerial photos showed old forest with large trees, but the GNN attributes for the site indicated forest with primarily small trees. This kind of error in the vegetation characteristic data could introduce error into the Maxent models, by training the models on a broader range of ecological conditions at murrelet location sites than actually occurs.

An underlying source of error is noise in the LandSat imagery used by GNN and LandTrendr. Individual LandSat images can have ‘impulse noise’, which is a general term for single-pixel spots that are not authentic imagery (USGS [http://landsat.usgs.gov/science\\_an\\_impulsenoise.php](http://landsat.usgs.gov/science_an_impulsenoise.php)). These can result in ‘salt and pepper’ patterns in which individual pixels are misclassified. The LandTrendr procedure of using annual time-series to temporally normalize imagery reduces this error, but does not eliminate it.

### Murrelet Locations

We recognize three primary sources of uncertainty in our marbled murrelet database. First, for the occupied detection sample we assumed there were no false positives, that is, we assumed murrelets were correctly identified during surveys and that their behavior was correctly observed so that sites with occupied detections were not recorded in error. Occupied detections were those that were believed to be associated with nesting (Evans Mack et al. 2003), but we cannot know if murrelets were actually nesting at all such detection sites. To the extent that our training sites included occupied sites which were in fact unsuitable sites, our models could be less accurate by including attribute data from sites that were not actually used by nesting murrelets. Also, our sample of nest sites includes locations where downy young were observed on the ground, and/or egg fragments were located, and it is possible these signs were not correctly attributed to the actual nest tree or its close proximity (especially in the case of egg fragments that could be carried off by predators like ravens). Thus, these sites may have less spatial accuracy than our sample of confirmed nest trees. Third, there is variation in forest attributes among the pixels that we delineated at murrelet locations. Some pixels within areas treated as species sites may not have been the exact locations used by the birds. To the extent that some pixels within 3 x 3-pixel neighborhood that contributed to averaged covariate values for presence locations may have included unsuitable habitat, our description of mean vegetation conditions at the site may have greater variance than a more homogenous site of truly suitable habitat.

Forest changes could have occurred between the year when nesting or occupied behaviors were observed and 1993, the year of the GNN vegetation covariates used in the Maxent models. We reduced this potential source of error by using aerial imagery to confirm the presence of older forest at all training sites in 1993. However, subtler

forest changes might have occurred, resulting in a difference between forest conditions at the time murrelets selected a site for nesting and 1993 forest conditions. A related temporal mismatch could occur if marbled murrelet nest selection behavior changed over time. Given the short time period involved in the present analysis, we believe an evolutionary change in nest selection behavior unlikely in a long-lived species. Behavioral changes or forest changes are a consideration for future analysis, particularly in the event of significant changes in forest or environmental attributes associated with climate change.

The allocation of murrelet location survey effort was not random with respect to the vegetation and physiographic covariates. Murrelet surveys were not conducted according to any planned survey design but, rather, some of the surveys in our database were done in advance of timber sales in forest that was judged likely to be murrelet habitat. As a result, there are likely biases in the distribution of survey effort and hence in the distribution of occupied sites in our data set (Daw et al. 1998, Scott et al. 2002, Edwards et al. 2006). This could result in a bias that less survey effort was expended in younger forest types with scattered older trees, where the dominant forest is too young to harvest. However, radiotelemetry studies are not subject to this bias and have not found this forest type to be selected often by nesting murrelets (Manley et al., 2001, Baker et al. 2006, Zharikov et al. 2006, Raphael unpublished data). In addition, the standard survey protocol (Evans Mack et al. 2003) recommends surveys for any stand where potential nest platforms occur, so surveys do occur in younger stands. The filtering, screening and stratification we conducted on the initial set of occupied sites reduced but did not eliminate the potential for biases in the spatial distribution of training sites within the modeling regions.

#### Model uncertainty

Projecting model results from one set of environmental data to another set can create uncertainties. We found that the range of values in each of our covariates from the current period fell within the ranges of those covariates in the baseline period, which helps justify our method of projection. However, projecting data in this way assumes that murrelets were selecting habitat conditions in the same way for each time period. If murrelets change habitat preferences in relation to changing environments, then our projections could be inaccurate. We have no evidence that habitat selection has changed.

As noted, our bookend method provides data on both habitat losses and habitat gains. Some of these gains may be due to the different sources of error and uncertainty we have discussed, just as some of the bookend losses may be due to error. Remote sensing approaches have demonstrated their ability to detect both losses and gains in forest cover (Staus et al. 2002, Kennedy et al. 2007, Hais et al. 2009, Coops et al. 2010), but the ecological characteristics of good murrelet nesting habitat are more complex than simple forest cover. The satellite imagery used to develop GNN covariates, as well as our analytic methods, may be less effective at distinguishing real but gradual increases in habitat quality from false gains due to background random noise, compared to its ability to detect habitat losses which are frequently due to substantial and usually abrupt loss or reduction in forest canopy. For this reason, in a

similar analysis for northern spotted owls (Davis 2011), the author questioned whether gains identified by bookend models were as reliable as losses of the owl's complex nesting and roosting habitat, over a short period of analysis. While additional error may occur for projecting a model to a new data set, versus the error associated with the original model, we used the exact same habitat models and model input sources for both 1993 and 2012, and losses and gains were determined by consistent criteria. However, it is possible that for short analysis periods there could be more error associated with detecting gains across any suitability threshold value, versus detecting losses across that same threshold, because losses tend to have a stronger signal (greater average loss in suitability) than gains. If this were true, our methods would tend to overestimate gains, and as a consequence underestimate losses when using the 'bookend' (net loss) method. We have assumed that model errors are not biased toward losses or gains, but this may be an area for future research. If classification errors occur as a consistent percent of the pixels in a suitability class, then we might expect a bias toward false gains (more pixels erroneously classified as changing to higher-suitability in 2012), versus errors in classifying changes from higher to lower suitability because most (88 percent over the entire NWFP area) of baseline habitat was classified as lower suitability (Class 1 or 2).

Because we performed 25 replicated model runs for each model region, we are able to portray some measure of uncertainty in our prediction of habitat suitability, and in our estimates of habitat change (see Figures 2-7 and 2-8). Doing so represents a major advance in the representation of habitat suitability. The magnitude of variation among model runs, represented by the 95 percent confidence interval around estimates, provides a useful way to judge model performance and helps interpret estimates of habitat suitability.

We used presence-only species distribution modeling methods because of the nature of the marbled murrelet nest location data that were available range wide, notably the lack of adequate and well-distributed samples of absence locations. Given the newness of presence-only methods such as Maxent, caution has been advised in their use (Royle et al. 2012, Yackulic et al. 2013, Ahmed et al. 2015). We exercised caution through the development of our modeling and calibration procedures, during consideration and critical examination of data sources and of modeled relationships between species occurrence and environmental covariates, and took steps to minimize potential sampling bias, within the time and resources available for monitoring murrelet nesting habitat. In addition, we have provided information including response curves, model performance metrics, and other information that readers may use to critically evaluate our results.

Notwithstanding these potential errors, our models all had very good (if not excellent) classification skill as measured by the AUC and gain values) and also were well calibrated as evidenced by the P/E (AAF) plots and associated Spearman test results. The sources of uncertainty we mention should predispose the models to perform worse – not better. Nonetheless, even with the “deck stacked against” good models, good models were generated.

## **Interpretation of Model Output**

We have presented maps depicting relative suitability of nesting habitat for the murrelet at a resolution of 30 meters. Predicted suitability at a single pixel can be far less reliable than predicted suitability at a larger scale, where small-scale errors are smoothed out by using average suitability over the larger area. Such smoothing can also reduce the accuracy of some single pixels, but predictions at this scale more reliably match the larger-scale patterns on the ground. Further, the GNN metadata specifically advises users that the most appropriate use of that data is across landscapes, counties, large watersheds, or ecoregions (areas much larger than stands or patches). For these reasons, we strongly caution users that estimates of amount of suitable habitat should be based on larger areas, such as for USGS hydrologic units (HUC; Seaber et al. 1987) of size 6 or larger (that is, HUC codes of 6 or smaller), and not for individual sites or stands. In addition, using our maps to locate specific areas of suitable murrelet habitat on a specific ownership is inappropriate at any scale, unless combined with ground-truthing or other form of verification.

## **Comparison with Previous Estimates**

Results presented in this report differ from those reported earlier by Raphael et al. (2011). This should be expected, as many aspects of this analysis differ from the earlier work, apart from being based on different bookend years. First, we are now using an updated set of vegetation data (updated GNN models based on a larger sample of vegetation plots) and, for reasons of stronger model predictive accuracy, included correlated covariates in the 20-year models, which was not done previously. Second, for the current analyses, we used the same, and larger set of covariates for each of our three modeling regions, and excluded the Landsat tasseled cap variables (Wetness, Brightness, Greenness) that were included in our previous models. Third, we had a slightly larger set of murrelet locations available with which to train models. Last, we employed different, and we believe superior, criteria for separating habitat suitability Class 1 from Class 2, and Class 3 from Class 4. As a result of these differences, our new baseline estimate of higher-suitability habitat over all lands (2.5 million acres) is less than our previous estimate of 3.8 million acres. Also, the current estimate of the highest-suitability habitat (Class 4) over all lands (0.46 million acres) is less than our previous estimate of 1.7 million acres (Raphael et al. 2011), but the Class 4 difference was strongly influenced by our change to a new, and biologically-based criterion for the division point between Class 3 and Class 4. In summary, the results in this report are the product of updated data, models, and methods and provide, we believe, the best available estimates of the status and trend of marbled murrelet nesting habitat in the NWFP area.

## Implications of Results

The Northwest Forest Plan was designed, among its many objectives, to provide habitat conditions that support a viable and well-distributed population of marbled murrelets. The Plan is a long term strategy that is expected to reach its full potential after many decades when previously-cutover forest stands within federal reserves mature and begin functioning as suitable habitat. In the short term, the objective is to conserve all remaining habitat, and to that end the NWFP has conserved to date the large majority (greater than 97 percent) of suitable marbled murrelet nesting habitat that was present on the federal lands NWFP management at the inception of the plan in 1994. Some habitat loss did occur in federal lands, both reserved and nonreserved, during the period of analysis, due to fire, harvest, and natural disturbances. While some future losses due to wildfire and natural disturbances are likely, harvest losses within federal reserves should drop or cease, with the completion of the ‘grandfathered’ timber sales approved prior NWFP implementation, but harvested after 1993. Over 90 percent of currently higher-suitability habitat on federal lands occurs within the various reserve land use allocations, but whether this continues is highly dependent on future management and political decisions.

We used a bookend approach to assess gains and losses in higher-suitability habitat. We cannot be certain that all gains are real, as some changes may be due to mapping error and other “noise” in the Landsat-based imagery that would cause erroneous estimates. While there is some uncertainty about gains and net change, we believe that a real loss in habitat has occurred from 1993 to 2012. Based on our bookend data, the rate of loss of higher-suitability habitat on reserved lands has been about 2.5 percent over the 20-year period (due mostly to fire, especially in Oregon; Table 2-12). However, rate of loss of higher-suitability habitat has been about 10 times greater (26.6 percent) on nonfederal lands, due mostly to timber harvest (Table 2-13). Conservation of the threatened murrelet is not possible if such losses continue at this rate into the future.

If the amount of higher-suitability habitat for murrelets is to be maintained at its current level, and given that almost half of the higher-suitability habitat is on nonfederal lands, accomplishing this goal will require significant contributions from nonfederal lands. Over time, as habitat on federal reserved lands increases in quality, less reliance on nonfederal lands may be warranted. Thus, currently, there are limits on the extent to which the NWFP can protect remaining suitable habitat and prevent its ongoing loss.

We found that the highest-suitability habitat (Class 4) comprised a relatively small proportion (about 20 percent) of all higher-suitability nesting habitat (Classes 3 plus 4). Class 4 includes areas with suitability scores equaling or exceeding the average condition for the murrelet presence sites used to train our models. To the extent which murrelets might preferentially nest in this highest-suitability habitat, our estimates of the total amount of suitable habitat available to murrelets, as represented by Classes 3 plus 4, may be optimistic.

We estimated a loss of about 34,000 acres of higher-suitability habitat (Classes 3 and 4) from federal reserves over the 20 years from 1993 to 2012 (Table 2-6). If that rate continued for 50 years (through 2042), the total loss would be about 85,000 acres.

There were also over 2 million acres of federally reserved lands in Class 2 condition (i.e., young forest, 37 percent of all habitat capable reserved lands). Given time, much of this has the potential to develop into more suitable nesting habitat depending on site conditions, presence of older trees, future management and other factors. If 5 percent of the nearly 2 million Class 2 acres developed into higher-suitability condition over the next 50 years, that would be more than enough—about 100,000 acres—to balance a loss of 85,000 acres. One must consider, though, that losses of our highest-suitability habitat (Class 4) would not be balanced by gains in lower classes of suitability represented by acres that just cross over the habitat suitability threshold. In addition, it can take more than 100 years for Class 2 habitat to become Class 3 and more than 200 years to become Class 4. The development of stands with old-growth characteristics necessary for murrelets is expected to take at least 100 to 200 years from the time of regeneration (USFWS 1997). For the many younger stands in the murrelet range that were clear-cut harvested in the past century, the benefits of habitat development are far into the future. However, if management for late-successional and old-growth forests continues, projections show substantial increases of forest exceeding 150 years in age by 2050 on western federal lands (Mills and Zhou 2003). Shorter-term gains in habitat quality may occur as older forest fills in around existing suitable habitat and reduces edge and fragmentation effects in existing habitat, prior to the older forest developing the large limbs, nest platforms, and other characteristics of murrelet nesting habitat.

Over the long run it is not unreasonable to expect to see some net increase in total amount of higher-suitability habitat, however in the short term conservation of the higher-suitability habitat (Classes 3 and 4) is essential. If losses of suitable habitat are reduced, old forest suitable for nesting is allowed to develop, and fragmentation of older forest is reduced throughout the reserved federal lands, then meeting murrelet population objectives will be more certain. Given declining murrelet population trends as well as habitat losses, in many areas, it is uncertain whether their populations will persist to benefit from potential future increases in habitat suitability. This underscores the need to arrest the loss of suitable habitat on all lands, especially on nonfederal lands and in the relatively near term (3-5 decades).

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**Table 2-1.** Variables used as input to Maxent.

Abbreviation	Description	Unit	Source
AGE_DOM_BA_NO_REM	Stand age	Years	GNN
CANCOV_CON	Canopy cover of all conifers	Percent	GNN
CANCOV_HDW	Canopy cover of all hardwoods	Percent	GNN
DDI	Diameter diversity index: measure of structural diversity of a forest stand, based on tree densities in different DBH classes (5-24 cm, 25-49 cm, 50-99 cm, and $\geq 100$ cm). See McComb et al. 2002 for details.	No units	GNN
FOG	The average value of effective precipitation from fog drip and low clouds.	Scaled to 1 unit = approx. 20 inches	Henderson et al. 2011.
MOD_OGSI_NWFP	Old growth index	Index from 1 to 100	GNN
MNDBHBA_CON	Basal-area weighted mean diameter of all live conifers	Inches	GNN
MULTISTORY_50	Percentage of 50-ha circular area classified as GNN IMAP_LAYERS (number of tree canopy layers present) equal 3	Percent	Derived from GNN
PCTMATURE_50	Percentage of 50-ha circular area classified as GNN VEGCLASS 10 (large conifer, moderate to closed canopy) or 11 (giant conifer, moderate to closed canopy)	Percent	Derived from GNN
PLATFORMS	Platforms per acre derived from GNN TPH (trees per hectare) by species and DBH (diameter at breast height) variables.	Number per acre	Derived from GNN
QMDC_DOM	Quadratic mean diameter of dominant conifer trees	Inches	GNN
SMR_PRECIP	Mean precipitation from May to September	Inches	PRISM
STANDHGT	Stand height	Feet	GNN
TPH_GE_100	Trees per ha of stems $\geq 100$ cm dbh	Number per acre	GNN

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**Table 2-2.** Accuracy assessment data for GNN attributes used as covariates in murrelet nest habitat modeling. Values presented are normalized root means square error (RMSE) and correlation coefficients ('r'), for individual GNN modeling regions. Includes 9 attributes used directly as covariates, and 2 GNN attributes which contributed toward the PCTMATURE\_50 covariates.

Covariate	GNN Attribute	Correlation coefficients by GNN Modeling Region							
		Washington Coast and Cascades		Oregon Coast Range		Oregon and California Klamath		California Coast	
		RMSE	Correlation coefficient	RMSE	Correlation coefficient	RMSE	Correlation coefficient	RMSE	Correlation coefficient
Conifer canopy cover	CANCOV_CON	0.21	0.78	0.28	0.78	0.39	0.73	0.46	0.69
Hardwood canopy cover	CANCOV_HDW	1.43	0.70	0.80	0.65	0.72	0.70	0.48	0.61
Tree diameter diversity index	DDI	0.33	0.78	0.34	0.77	0.37	0.69	0.35	0.61
Basal-area weighted mean diameter at breast height (DBH), all conifers	MNDBHBA_CON	0.43	0.72	0.47	0.73	0.52	0.52	0.66	0.39
Quadratic mean DBH, dominant and codominant conifers	QMDC_DOM	0.44	0.71	0.47	0.74	0.56	0.51	0.70	0.37
	<i>Contributors:</i>								
<b>PCTMATURE_50</b>	QMDA_DOM	0.40	0.75	0.40	0.79	0.51	0.55	0.54	0.51
	CANCOV	0.19	0.75	0.19	0.80	0.26	0.75	0.19	0.78
Old growth structure index	MOD_OGSI_NWFP	0.51	0.68	0.50	0.70	0.57	0.51	0.59	0.42
Stand height, dominant and codominant trees	STNDHGT	0.38	0.71	0.33	0.81	0.47	0.58	0.45	0.55

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Basal-area weighted stand age, excluding remnant trees	AGE_DOM_BA_NO_REM	0.55	0.77	0.55	0.76	0.51	0.59	0.64	0.46
Conifer trees per hectare ≥100cm dbh	TPHC_GE_100	1.91	0.58	1.66	0.67	1.70	0.52	2.28	0.48
Total Inventory Plots Used in GNN Model Development		2937		2024		3703		975	

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**Table 2-3.** Summary statistics for 1993 covariates used in Maxent analysis for points occupied by murrelets (species) and the analysis area (global) and for 2012 global analysis.

State	Covariate <sup>a</sup>	Species (Nest and occupied)				Global (Habitat capable) 1993				Global (Habitat capable) 2012			
		Mean	SD <sup>b</sup>	Min	Max	Mean	SD <sup>b</sup>	Min	Max	Mean	SD <sup>b</sup>	Min	Max
Washington		n = 108 points				n = 10,851,100 acres <sup>c</sup>				n = 10,851,100 acres <sup>c</sup>			
	CANCOV_CON	80.6	11.6	42.6	94.2	63.0	26.4	0.0	99.9	66.2	24.0	0.0	99.9
	CANCOV_HDW	7.7	10.7	0.0	48.3	12.9	16.0	0.0	95.5	14.7	17.1	0.0	97.4
	DDI	6.3	1.5	1.8	9.4	4.0	2.1	0.0	10.0	4.0	2.0	0.0	10.0
	MNDBHBA_CON	70.4	23.9	20.2	143.6	39.6	22.3	0.0	195.0	38.8	21.3	0.0	195.0
	MULTISTORY_50	61.6	25.6	3.0	100.0	46.7	28.3	0.0	100.0	45.7	28.6	0.0	100.0
	PCTMATURE_50	40.6	25.8	0.0	97.0	18.6	23.9	0.0	100.0	17.7	24.6	0.0	100.0
	PLATFORMS	55.1	32.2	1.6	135.1	19.7	28.2	0.0	251.2	18.4	28.1	0.0	251.2
	QMDC_DOM	59.8	16.5	16.1	92.0	35.9	20.3	0.0	122.3	35.7	19.4	0.0	136.4
	TPHC_GE_100	10.8	8.8	0.0	37.4	3.3	6.6	0.0	91.4	3.2	6.5	0.0	91.4
	AGE_DOM_BA_NO_REM	183.6	94.8	23.4	376.0	95.1	89.0	0.0	629.0	92.4	89.7	0.0	647.0
	MOD_OGSI_NWFP	40.7	22.9	0.0	83.0	16.6	20.4	0.0	90.0	15.2	20.0	0.0	93.0
	STNDHGT	30.2	7.9	8.4	50.2	20.3	9.7	0.0	65.1	20.4	8.9	0.0	68.0
	FOG	0.9	0.6	0.0	2.3	0.5	0.5	-0.1	3.1	n/a	n/a	n/a	n/a
	JULY_MAXT	71.2	2.5	63.9	77.2	71.9	3.7	56.4	81.1	n/a	n/a	n/a	n/a
	SMR_PRECIP	3.3	1.1	1.4	6.1	3.1	1.2	0.8	7.5	n/a	n/a	n/a	n/a
Oregon		n = 134 points				n = 6,610,400 acres <sup>c</sup>				n = 6,610,400 acres <sup>c</sup>			
	CANCOV_CON	64.0	14.5	16.6	87.9	52.5	22.5	0.0	100.0	54.8	23.0	0.0	100.0
	CANCOV_HDW	19.6	16.4	0.0	70.8	22.9	18.6	0.0	99.1	22.1	18.3	0.0	99.1
	DDI	5.8	1.4	1.7	8.1	4.0	1.9	0.0	10.0	3.8	1.8	0.0	10.0
	MNDBHBA_CON	75.5	22.6	22.0	127.8	44.4	24.9	0.0	325.9	41.9	24.3	0.0	325.9
	MULTISTORY_50	42.7	21.7	12.0	99.0	38.3	22.0	0.0	100.0	34.6	21.4	0.0	100.0
	PCTMATURE_50	37.1	24.2	0.0	96.0	14.3	17.8	0.0	100.0	12.4	17.2	0.0	100.0
	PLATFORMS	53.5	34.8	1.3	186.4	16.0	21.9	0.0	207.7	14.5	21.8	0.0	269.9
	QMDC_DOM	68.1	21.7	18.1	120.4	38.4	22.5	0.0	259.8	36.8	22.1	0.0	259.8
	TPHC_GE_100	12.7	10.8	0.0	51.5	3.4	5.9	0.0	71.6	3.1	5.8	0.0	71.6
	AGE_DOM_BA_NO_REM	109.8	48.0	23.7	291.0	66.4	45.1	0.0	849.0	62.2	43.5	0.0	849.0
	MOD_OGSI_NWFP	39.8	20.1	0.0	78.3	15.3	17.6	0.0	89.0	12.9	16.7	0.0	92.0
	STNDHGT	35.2	10.5	7.4	58.1	20.7	10.5	0.0	69.5	20.3	10.3	0.0	67.8

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FOG	0.5	0.4	0.1	2.0	0.4	0.5	0.0	2.9	n/a	n/a	n/a	n/a
JULY_MAXT	75.1	3.9	64.2	85.6	76.4	4.9	58.0	91.9	n/a	n/a	n/a	n/a
SMR_PRECIP	2.7	0.7	1.1	4.3	2.2	0.8	0.9	5.7	n/a	n/a	n/a	n/a
California	n = 126 points				n = 3,250,100 acres <sup>c</sup>				n = 3,250,100 acres <sup>c</sup>			
CANCOV_CON	84.6	15.4	15.1	99.5	51.0	22.9	0.0	100.0	54.0	22.5	0.0	100.0
CANCOV_HDW	26.8	20.7	0.0	75.8	44.6	22.7	0.0	97.5	45.4	22.7	0.0	97.8
DDI	6.7	1.4	2.2	9.6	4.8	1.6	0.0	10.0	4.9	1.5	0.0	10.0
MNDBHBA_CON	110.2	49.6	14.8	290.3	52.6	27.7	0.0	325.9	53.3	26.5	0.0	325.9
MULTISTORY_50	73.1	19.9	18.0	100.0	64.8	23.0	0.0	100.0	68.3	22.9	0.0	100.0
PCTMATURE_50	42.7	26.7	0.0	98.0	10.4	16.5	0.0	100.0	10.2	16.1	0.0	100.0
PLATFORMS	82.8	63.6	0.2	221.0	16.0	24.8	0.0	221.0	16.0	23.8	0.0	221.0
QMDC_DOM	84.6	35.5	13.6	231.2	43.5	23.5	0.0	259.8	43.8	22.3	0.0	259.8
TPHC_GE_100	24.1	18.1	0.0	61.7	4.2	7.5	0.0	71.6	4.1	7.2	0.0	71.6
AGE_DOM_BA_NO_REM	181.5	109.1	44.9	728.3	85.4	58.2	0.0	849.0	83.9	56.0	0.0	849.0
MOD_OGSI_NWFP	49.1	24.6	3.6	88.0	20.9	18.0	0.0	88.0	20.5	17.0	0.0	88.0
STNDHGT	33.1	10.6	8.7	62.8	18.4	7.5	0.0	67.8	18.8	7.1	0.0	67.8
FOG	1.2	0.1	0.9	1.5	1.0	0.4	0.1	1.8	n/a	n/a	n/a	n/a
JULY_MAXT	74.0	4.7	65.1	85.6	80.7	7.1	57.6	99.0	n/a	n/a	n/a	n/a
SMR_PRECIP	1.1	0.3	0.5	2.0	1.4	0.7	0.3	4.1	n/a	n/a	n/a	n/a

<sup>a</sup> description of the covariates, including units, can be found in Table 2-1.

<sup>b</sup> Standard deviation

<sup>c</sup> Represents analysis area in murrelet Zones 1 and 2 in WA and Zone 1 only in OR and CA. Figures rounded to the nearest 100 acres.

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Table 2-4. Percent contribution of each covariate to Maxent model in each state.

Covariate	Percent Contribution		
	Washington	Oregon	California
CANCOV_CON	2.9	2.6	27.3
CANCOV_HDW	8.5	2.6	2.7
DDI	3	3	1.6
MNDBHBA_CON	3.5	2.4	2.5
MULTISTORY_50	2.2	4.2	1.9
PCTMATURE_50	6.4	8.7	6
PLATFORMS	10.7	3.9	16.1
QMDC_DOM	2.7	2.3	1.9
TPHC_GE_100	3.6	2.9	4.8
AGE_DOM_BA_NO_REM	1.4	2.3	1.8
MOD_OGSI_NWFP	26.4	19.3	5
STNDHGT	3.2	13.3	0.8
FOG	17.1	3.1	11.8
JULY_MAXT	6.8	3.5	5.5
SMR_PRECIP	1.8	25.8	10.3

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Table 2-5. Distribution of murrelet nesting habitat on Federal nonreserved lands, by habitat suitability class, for the baseline period (1993) and final year of analysis (2012).<sup>a</sup>

State/Province	Federal nonreserved – 1993 (1000s of acres)					Federal nonreserved – 2012 (1000s of acres)				
	Class 1	Class 2	Class 3	Class 4	Habitat Capable Total	Class 1	Class 2	Class 3	Class 4	Habitat Capable Total
Washington:										
(Maxent <sup>b</sup> score)	0–0.06	0.06–0.21	0.21–0.53	0.53–1		0–0.06	0.06–0.21	0.21–0.53	0.53–1	
Olympic Peninsula	54.7	39.2	21.3	7.2	122.4	49.6	43.3	21.9	7.5	122.4
Western Lowlands	0.3	0.2	0.0	0.0	0.5	0.3	0.2	0.0	0.0	0.5
Western Cascades	122.8	72.5	27.7	1.7	224.7	118.1	77.0	27.9	1.6	224.7
Eastern Cascades	45.6	28.6	4.8	0.0	79.1	38.6	34.7	5.7	0.1	79.1
Total	223.5	140.5	53.8	8.9	426.7	206.7	155.2	55.5	9.2	426.7
Oregon:										
(Maxent score)	0–0.04	0.04–0.18	0.18–0.51	0.51–1		0–0.04	0.04–0.18	0.18–0.51	0.51–1	
Coast Range	196.9	108.3	41.4	6.4	353.0	179.7	123.1	43.1	7.0	353.0
Willamette Valley	1.2	0.3	0.0	0.0	1.6	1.3	0.2	0.0	0.0	1.6
Western Cascades	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Klamath	121.1	78.0	19.4	2.3	220.8	127.2	74.5	17.0	2.0	220.8
Total	319.3	186.7	60.7	8.7	575.3	308.3	197.9	60.1	9.1	575.3
California:										
(Maxent score)	0–0.01	0.01–0.17	0.17–0.57	0.57–1		0–0.01	0.01–0.17	0.17–0.57	0.57–1	
Coast Range	12.7	4.8	1.1	0.5	19.1	12.9	4.8	1.0	0.4	19.1
Klamath	188.3	5.2	0.2	0.0	193.7	188.0	5.5	0.1	0.0	193.7
Total	200.9	10.0	1.2	0.5	212.7	200.9	10.3	1.1	0.4	212.7
Plan area total	743.7	337.2	115.8	18.1	1,214.7	715.9	363.4	116.8	18.7	1,214.7

<sup>a</sup> Numbers rounded to nearest 100; total computed prior to rounding.

<sup>b</sup> Maxent is a habitat suitability modeling software used in this study to model relative suitability of marbled murrelet nesting habitat.

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Table 2-6. Distribution of murrelet nesting habitat on Federal reserved lands, by habitat suitability class, for the baseline period (1993) and final year of analysis (2012).<sup>a</sup>

State/Province	Federal reserved – 1993 (1000s of acres)					Federal reserved – 2012 (1000s of acres)				
	Class 1	Class 2	Class 3	Class 4	Habitat Capable Total	Class 1	Class 2	Class 3	Class 4	Habitat Capable Total
Washington:										
( <i>Maxent</i> <sup>b</sup> score)	0–0.06	0.06–0.21	0.21–0.53	0.53–1		0–0.06	0.06–0.21	0.21–0.53	0.53–1	
Olympic Peninsula	209.3	477.0	402.8	128.7	1,217.8	203.3	498.8	388.2	127.4	1,217.8
Western Lowlands	24.2	44.1	12.2	0.4	81.0	23.9	43.5	13.3	0.4	81.0
Western Cascades	587.7	736.7	266.9	13.9	1,605.1	571.1	753.8	267.9	12.3	1,605.1
Eastern Cascades	76.3	57.2	10.6	1.4	145.5	69.9	62.7	11.4	1.4	145.5
<b>Total</b>	<b>897.5</b>	<b>1,314.9</b>	<b>692.5</b>	<b>144.5</b>	<b>3,049.4</b>	<b>868.1</b>	<b>1,358.8</b>	<b>680.9</b>	<b>141.5</b>	<b>3,049.4</b>
Oregon:										
( <i>Maxent</i> score)	0–0.04	0.04–0.18	0.18–0.51	0.51–1		0–0.04	0.04–0.18	0.18–0.51	0.51–1	
Coast Range	280.9	337.9	314.7	72.4	1,005.9	263.4	355.5	309.0	77.9	1,005.9
Willamette Valley	2.3	0.3	0.0	0.0	2.6	2.3	0.3	0.0	0.0	2.6
Western Cascades	0.5	0.3	0.0	0.0	0.8	0.5	0.3	0.0	0.0	0.8
Klamath	357.0	388.0	104.9	11.6	861.5	459.3	304.7	85.4	12.1	861.5
<b>Total</b>	<b>640.6</b>	<b>726.5</b>	<b>419.6</b>	<b>84.1</b>	<b>1,870.8</b>	<b>725.5</b>	<b>660.8</b>	<b>394.4</b>	<b>90.1</b>	<b>1,870.8</b>
California:										
( <i>Maxent</i> score)	0–0.01	0.01–0.17	0.17–0.57	0.57–1		0–0.01	0.01–0.17	0.17–0.57	0.57–1	
Coast Range	129.3	77.4	17.4	7.2	231.3	126.3	80.7	17.3	7.0	231.3
Klamath	536.6	15.5	0.2	0.0	552.3	537.4	14.8	0.2	0.0	552.3
<b>Total</b>	<b>666.0</b>	<b>92.8</b>	<b>17.6</b>	<b>7.2</b>	<b>783.6</b>	<b>663.6</b>	<b>95.4</b>	<b>17.5</b>	<b>7.0</b>	<b>783.6</b>
<b>Plan area total</b>	<b>2,204.0</b>	<b>2,134.3</b>	<b>1,129.7</b>	<b>235.7</b>	<b>5,703.7</b>	<b>2,257.3</b>	<b>2,115.0</b>	<b>1,092.7</b>	<b>238.6</b>	<b>5,703.7</b>

<sup>a</sup> Numbers rounded to nearest 100; total computed prior to rounding.

<sup>b</sup> Maxent is a habitat suitability modeling software used in this study to model relative suitability of marbled murrelet nesting habitat.

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Table 2-7. Distribution of murrelet nesting habitat on nonfederal lands, by habitat suitability class, for the baseline period (1993) and final year of analysis (2012).<sup>a</sup>

State/Province	Nonfederal – 1993 (1000s of acres)					Nonfederal – 2012 (1000s of acres)				
	Class 1	Class 2	Class 3	Class 4	Habitat Capable Total	Class 1	Class 2	Class 3	Class 4	Habitat Capable Total
Washington:										
(Maxent <sup>b</sup> score)	0--0.06	0.06-0.21	0.21-0.53	0.53-1		0--0.06	0.06-0.21	0.21-0.53	0.53-1	
Olympic Peninsula	638.8	444.4	199.3	99.5	1,381.9	718.3	451.8	139.6	72.3	1,381.9
Western Lowlands	2,793.2	1,129.0	223.8	32.4	4,178.4	3,158.6	845.6	154.1	20.1	4,178.4
Western Cascades	1,291.3	381.1	82.6	10.4	1,765.5	1,388.1	308.8	62.0	6.6	1,765.5
Eastern Cascades	36.7	11.2	1.3	0.0	49.2	32.6	15.2	1.3	0.0	49.2
Total	4,760.0	1,965.7	507.0	142.3	7,375.0	5,297.6	1,621.4	357.0	99.0	7,375.0
Oregon:										
(Maxent score)	0-0.04	0.04-0.18	0.18-0.51	0.51-1		0-0.04	0.04-0.18	0.18-0.51	0.51-1	
Coast Range	2,325.8	857.6	211.4	42.9	3,437.7	2,564.6	674.8	162.5	35.8	3,437.7
Willamette Valley	95.2	5.1	0.0	0.0	100.3	97.8	2.5	0.0	0.0	100.3
Western Cascades	3.7	0.3	0.0	0.0	4.0	3.9	0.1	0.0	0.0	4.0
Klamath	460.5	135.8	23.0	3.0	622.2	479.8	119.6	19.9	3.0	622.2
Total	2,885.2	998.8	234.5	45.8	4,164.3	3,146.2	797.0	182.3	38.8	4,164.3
California:										
(Maxent score)	0-0.01	0.01-0.17	0.17-0.57	0.57-1						
Coast Range	1,276.4	605.3	89.7	14.9	1,986.2	1,246.7	658.2	69.0	12.2	1,986.2
Klamath	236.7	29.5	1.4	0.1	267.6	234.3	31.7	1.6	0.1	267.6
Total	1,513.0	634.7	91.1	14.9	2,253.8	1,481.0	689.9	70.6	12.3	2,253.8
Plan area total	9,158.2	3,599.2	832.5	203.1	13,793.0	9,924.7	3,108.3	609.9	150.1	13,793.0

<sup>a</sup> Numbers rounded to nearest 100; total computed prior to rounding.

<sup>b</sup> Maxent is a habitat suitability modeling software used in this study to model relative suitability of marbled murrelet nesting habitat.

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Table 2-8. Distribution of murrelet nesting habitat on all lands, by habitat suitability class, for the baseline period (1993) and final year of analysis (2012).<sup>a</sup>

State/Province	All Lands – 1993 (1000s of acres)					All Lands – 2012 (1000s of acres)				
	Class 1	Class 2	Class 3	Class 4	Habitat Capable Total	Class 1	Class 2	Class 3	Class 4	Habitat Capable Total
Washington										
<i>(Maxent score)</i>	<i>0–0.06</i>	<i>0.06–0.21</i>	<i>0.21–0.53</i>	<i>0.53–1</i>		<i>0–0.06</i>	<i>0.06–0.21</i>	<i>0.21–0.53</i>	<i>0.53–1</i>	
Olympic Peninsula	902.9	960.5	623.3	235.4	2,722.1	971.2	994.0	549.7	207.2	2,722.1
Western Lowlands	2,817.7	1,173.3	236.0	32.9	4,259.9	3,182.9	889.2	167.4	20.4	4,259.9
Western Cascades	2,001.8	1,190.3	377.2	25.9	3,595.3	2,077.2	1,139.6	357.8	20.6	3,595.3
Eastern Cascades	158.6	97.0	16.7	1.5	273.8	141.1	112.7	18.5	1.5	273.8
Total	5,881.0	3,421.1	1,253.3	295.7	10,851.0	6,372.4	3,135.5	1,093.4	249.8	10,851.0
Oregon:										
<i>(Maxent score)</i>	<i>0–0.04</i>	<i>0.04–0.18</i>	<i>0.18–0.51</i>	<i>0.51–1</i>		<i>0–0.04</i>	<i>0.04–0.18</i>	<i>0.18–0.51</i>	<i>0.51–1</i>	
Coast Range	2,803.6	1,303.7	567.5	121.6	4,796.5	3,007.8	1,153.4	514.5	120.8	4,796.5
Willamette Valley	98.7	5.7	0.0	0.0	104.5	101.5	3.0	0.0	0.0	104.5
Western Cascades	4.2	0.6	0.0	0.0	4.8	4.4	0.4	0.0	0.0	4.8
Klamath	938.5	601.9	147.3	16.9	1,704.5	1,066.3	498.8	122.3	17.2	1,704.5
Total	3,845.0	1,912.0	714.8	138.6	6,610.4	4,180.0	1,655.6	636.8	137.9	6,610.4
California:										
<i>(Maxent score)</i>	<i>0–0.01</i>	<i>0.01–0.17</i>	<i>0.17–0.57</i>	<i>0.57–1</i>		<i>0–0.01</i>	<i>0.01–0.17</i>	<i>0.17–0.57</i>	<i>0.57–1</i>	
Coast Range	1,418.4	687.5	108.2	22.5	2,236.5	1,385.9	743.7	87.3	19.7	2,236.5
Klamath	961.6	50.2	1.8	0.1	1,013.6	959.7	51.9	1.9	0.1	1,013.6
Total	2,379.9	737.6	110.0	22.6	3,250.1	2,345.6	795.6	89.2	19.7	3,250.1
Plan area total	12,106.0	6,070.7	2,078.0	456.8	20,711.5	12,897.9	5,586.7	1,819.5	407.4	20,711.5

<sup>a</sup> Numbers rounded to nearest 100; total computed prior to rounding.

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Table 2-9. Acres of loss, gain and net change in marbled murrelet higher-suitability (Classes 3 plus 4) nesting habitat, and net change for the highest-suitability (Class 4) nesting habitat, from 1993 to 2012, by federal land allocation.

State and province	Federal nonreserved						Federal reserved					
	Higher-suitability (Classes 3+4)			Class 4			Higher-suitability (Classes 3+C4)			Class 4		
	Losses	Gains	Net Change Acres    %	Net Change Acres    %	Losses	Gains	Net Change Acres    %	Net Change Acres    %	Losses	Gains	Net Change Acres    %	Net Change Acres    %
<b>Washington</b>												
Olympic Peninsula	4,857	5,812	955    3.4	310    4.31	48,032	32,226	-15,806    -3.0	-1,288    -1.00				
Western Lowlands	1	0	-1    -6.2	0    0.00	3,658	4,594	936    7.4	-79    -18.09				
Western Cascades	4,933	5,096	163    0.6	-34    -2.03	44,071	43,596	-475    -0.2	-1,523    -10.98				
Eastern Cascades	739	1,693	954    19.6	51    118.04	1,824	2,647	823    6.9	-39    -2.73				
<b>Washington total</b>	<b>10,530</b>	<b>12,601</b>	<b>2,071    3.3</b>	<b>327    3.68</b>	<b>97,585</b>	<b>83,063</b>	<b>-14,522    -1.7</b>	<b>-2,930    -2.03</b>				
<b>Oregon</b>												
Coast Range	8,093	10,464	2,371    5.0	662    10.40	42,883	42,729	-154    0.0	5,532    7.64				
Willamette Valley	1	0	-1    -7.7	0    0.00	0	0	0    -50.1	0    0.00				
Western Cascades	0	0	0    0.0	0    0.00	4	3	-1    -5.0	0    0.00				
Klamath	8,720	6,121	-2,599    -12.0	-273    -11.86	41,971	22,916	-19,055    -16.3	474    4.07				
<b>Oregon total</b>	<b>16,814</b>	<b>16,585</b>	<b>-229    -0.3</b>	<b>389    4.49</b>	<b>84,858</b>	<b>65,648</b>	<b>-19,210    -3.8</b>	<b>6,006    7.15</b>				
<b>California</b>												
Coast Range	222	47	-175    -11.1	-113    -22.49	1,943	1,686	-257    -1.0	-125    -1.74				
Klamath	70	22	-48    -25.1	1    -31.25	80	60	-20    -8.9	0    0.00				
<b>California total</b>	<b>292</b>	<b>69</b>	<b>-223    -12.7</b>	<b>-115    -22.55</b>	<b>2,023</b>	<b>1,746</b>	<b>-277    -1.1</b>	<b>-125    -1.74</b>				
<b>Plan Area Total</b>	<b>27,636</b>	<b>29,255</b>	<b>1,619    1.2</b>	<b>602    3.33</b>	<b>184,466</b>	<b>150,457</b>	<b>-34,009    -2.5</b>	<b>2,952    1.27</b>				

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Table 2-10. Acres of loss, gain and net change in marbled murrelet higher-suitability (Classes 3 plus 4) nesting habitat, and net change for the highest-suitability (Class 4) nesting habitat, from 1993 to 2012, for nonfederal and all lands.

State and province	Nonfederal						All lands					
	Higher-suitability (Classes 3+4)			Class 4			Higher-suitability (Classes 3+C4)			Class 4		
	Losses	Gains	Net Change Acres    %	Net Change Acres    %	Losses	Gains	Net Change Acres    %	Net Change Acres    %	Losses	Gains	Net Change Acres    %	Net Change Acres    %
<b>Washington</b>												
Olympic Peninsula	137,349	50,398	-86,951   -29.1	-27,187   -27.34	190,238	88,436	-101,802   -11.9	-28,165   -12.0	134,437	53,379	-81,058   -30.1	-12,429   -37.8
Western Lowlands	130,778	48,785	-81,993   -32.0	-12,350   -38.09	90,551	65,843	-24,708   -6.1	-5,339   -20.6	3,220	5,033	1,813   10.0	16   1.1
Western Cascades	41,547	17,151	-24,396   -26.2	-3,782   -36.31								
Eastern Cascades	657	693	36   2.7	4   19.61								
<b>Washington total</b>	<b>310,331</b>	<b>117,027</b>	<b>-193,304   -29.8</b>	<b>-43,314   -30.43</b>	<b>418,446</b>	<b>212,691</b>	<b>-205,755   -13.3</b>	<b>-45,917   -15.5</b>				
<b>Oregon</b>												
Coast Range	120,236	64,180	-56,056   -22.0	-7,061   -16.48	171,212	117,373	-53,839   -7.8	-867   -0.7	17	5	-12   -57.6	0   0.0
Willamette Valley	16	5	-11   -66.2	0   0.00	6	3	-3   -18.1	0   0.0	63,451	38,710	-24,741   -15.1	250   1.5
Western Cascades	2	0	-2   -83.3	0   0.00								
Klamath	12,760	9,673	-3,087   -11.9	49   1.66								
<b>Oregon total</b>	<b>133,014</b>	<b>73,858</b>	<b>-59,156   -21.1</b>	<b>-7,012   -15.30</b>	<b>234,686</b>	<b>156,091</b>	<b>-78,595   -9.2</b>	<b>-617   -0.4</b>				
<b>California</b>												
Coast Range	35,801	12,503	-23,298   -22.3	-2,633   -17.72	37,966	14,236	-23,730   -18.2	-2,871   -12.7	781	904	123   6.6	-12   -17.5
Klamath	631	822	191   13.0	-11   -16.72								
<b>California total</b>	<b>36,432</b>	<b>13,325</b>	<b>-23,107   21.8</b>	<b>-2,644   -17.72</b>	<b>38,747</b>	<b>15,140</b>	<b>-23,607   -17.8</b>	<b>-2,883   -12.8</b>				
<b>Plan Area Total</b>	<b>479,777</b>	<b>204,210</b>	<b>-275,567   -26.6</b>	<b>-52,970   -26.08</b>	<b>691,879</b>	<b>383,922</b>	<b>-307,957   -12.1</b>	<b>-49,416   -10.8</b>				

**Table 2-11.** Distribution of higher-suitability (Classes 3 plus 4) murrelet nesting habitat by conservation zone, for the baseline period (1993) and final year of analysis (2012). See Figure 2-2 for map of conservation zones.

<b>Conservation Zone</b>	<b>Acres of Higher-Suitability Habitat</b>			
	<b>1993</b>	<b>2012</b>	<b>Change (acres)</b>	<b>Change (percent)</b>
Zone 1 (northern Washington)	829,525	739,407	-90,118	-10.9
Zone 2 (outer coast of Washington)	719,414	603,777	-115,638	-16.1
Zone 3 (northern and central Oregon)	662,767	610,583	-52,184	-7.9
Zone 4 (southern Oregon and northern California)	309,072	256,636	-52,436	-17.0
Zone 5 (north-central California)	14,060	16,479	+2,419	+17.2



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**Table 2-12.** Attribution of loss of marbled murrelet higher-suitability nesting habitat from 1993 to 2012 using LandTrendr disturbance data on federal lands.

State/Province	Federal Non-reserved lands				Federal Reserved lands			
	Timber harvest	Wild fire	Insect, disease	Natural disturb.	Timber harvest	Wild fire	Insect, disease	Natural disturb.
Olympic Peninsula	463	0	33	0	1,164	873	548	1,293
Western Lowlands	0	0	0	0	0	0	29	720
Western Cascades	259	0	18	0	1,154	82	225	1,289
Eastern Cascades	125	0	37	0	34	1	691	163
<b>WASHINGTON TOTAL</b>	<b>847</b>	<b>0</b>	<b>88</b>	<b>0</b>	<b>2,352</b>	<b>956</b>	<b>1,493</b>	<b>3,465</b>
Coast Range	1,426	0	39	0	1,611	56	85	18
Willamette Valley	0	0	0	0	0	0	0	0
Cascades West	0	0	0	0	0	0	0	0
Klamath Mountains	990	2,366	30	0	656	17,855	73	46
<b>OREGON TOTAL</b>	<b>2,416</b>	<b>2,366</b>	<b>69</b>	<b>0</b>	<b>2,267</b>	<b>17,911</b>	<b>158</b>	<b>64</b>
Coast	8	0	0	0	1	257	9	114
Klamath	20	0	1	0	0	0	2	11
<b>CALIFORNIA TOTAL</b>	<b>28</b>	<b>0</b>	<b>1</b>	<b>0</b>	<b>1</b>	<b>257</b>	<b>11</b>	<b>125</b>
<b>RANGE TOTAL</b>	<b>3,291</b>	<b>2,366</b>	<b>158</b>	<b>0</b>	<b>4,620</b>	<b>19,124</b>	<b>1,662</b>	<b>3,654</b>

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**Table 2-13.** Attribution of loss, in acres, of marbled murrelet higher-suitability nesting habitat from 1993 to 2012 using LandTrendr Disturbance data on nonfederal and all lands.

State/Province	Nonfederal lands				All lands			
	Timber harvest	Wild fire	Insects	Natural disturb.	Timber harvest	Wild fire	Insects	Natural disturb.
Olympic Peninsula	87,731	0	2,716	0	89,358	873	3,297	1,293
Western Lowlands	90,102	1	1,790	0	90,102	1	1,819	720
Western Cascades	22,492	200	385	0	23,905	282	628	1,289
Eastern Cascades	333	0	22	0	492	1	750	163
<b>WASHINGTON TOTAL</b>	<b>200,658</b>	<b>201</b>	<b>4,913</b>	<b>0</b>	<b>203,857</b>	<b>1,157</b>	<b>6,494</b>	<b>3,465</b>
<hr/>								
Coast Range	80,049	197	1,456	0	83,086	253	1,580	18
Willamette Valley	10	0	0	0	10	0	0	0
Cascades West	2	0	0	0	2	0	0	0
Klamath Mountains	8,004	14	128	0	9,650	20,235	231	46
<b>OREGON TOTAL</b>	<b>88,065</b>	<b>211</b>	<b>1,584</b>	<b>0</b>	<b>92,748</b>	<b>20,488</b>	<b>1,811</b>	<b>64</b>
<hr/>								
Coast	19,704	161	433	0	19,713	418	442	114
Klamath	320	0	15	0	340	0	18	11
<b>CALIFORNIA TOTAL</b>	<b>20,024</b>	<b>161</b>	<b>448</b>	<b>0</b>	<b>20,053</b>	<b>418</b>	<b>460</b>	<b>125</b>
<b>RANGE TOTAL</b>	<b>308,747</b>	<b>573</b>	<b>6,945</b>	<b>0</b>	<b>316,658</b>	<b>22,063</b>	<b>8,765</b>	<b>3,654</b>

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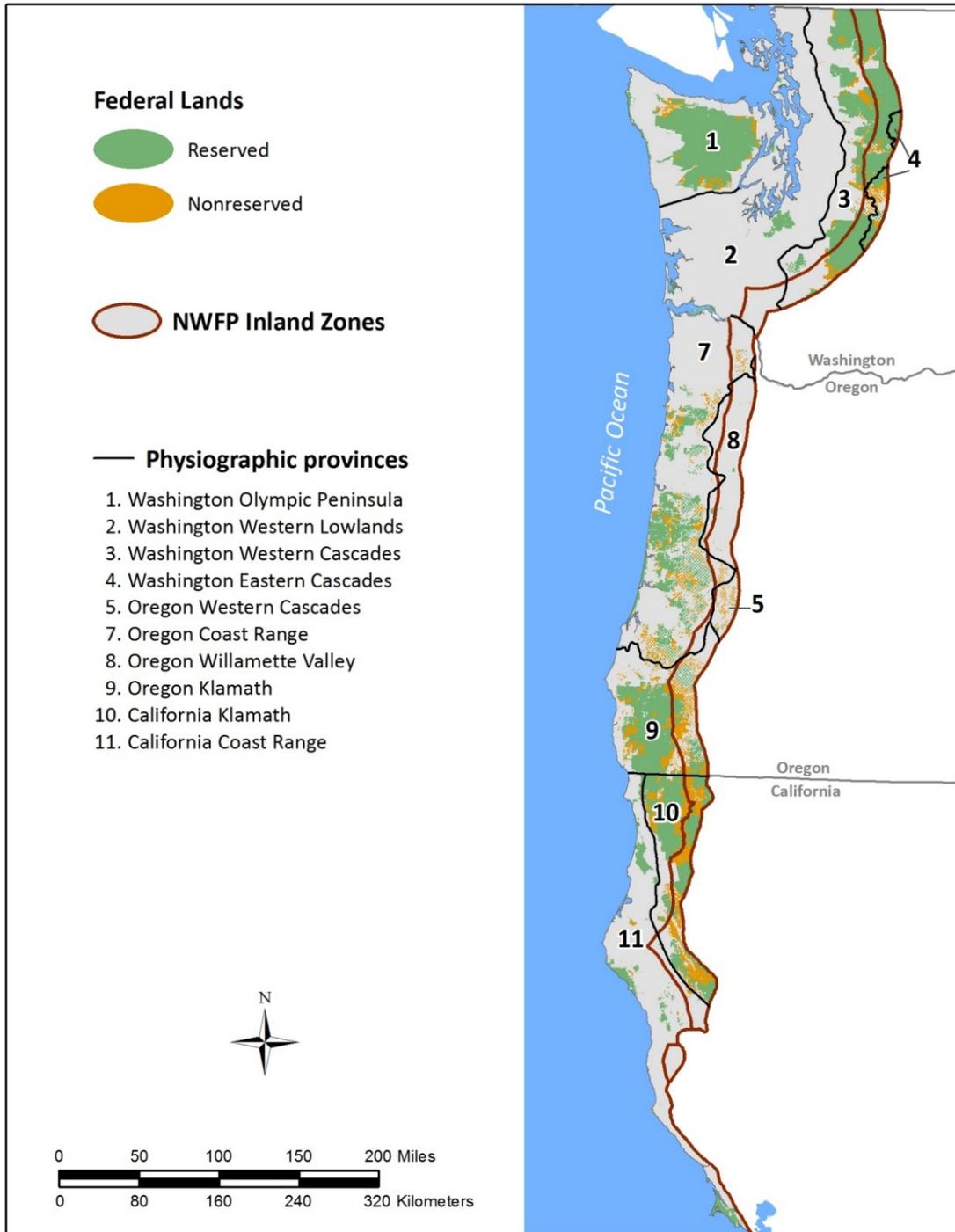
**Table 2-14.** Acres of Edge, Core-edge and Core habitat (GUIDOS classifications) by State, land ownership (federal vs. nonfederal) and Land Use Allocation (reserved vs. non-reserved) for higher-suitability marbled murrelet nesting habitat.

State	Year	Federal Nonreserved			Federal Reserved			Nonfederal		
		Edge	Core-edge	Core	Edge	Core-edge	Core	Edge	Core-edge	Core
Washington	1993	38,444	17,206	7,041	340,160	255,779	240,992	439,026	141,158	69,125
	2012	39,244	17,820	7,698	342,028	251,157	229,224	315,525	94,516	45,965
Oregon	1993	38,368	19,762	11,263	185,761	164,785	153,110	164,696	70,673	44,918
	2012	39,837	19,800	9,528	189,380	167,703	127,362	138,165	55,206	27,760
California	1993	735	542	480	7,472	6,439	10,896	38,938	27,263	39,793
	2012	664	468	403	7,580	6,265	10,686	34,204	20,704	27,979

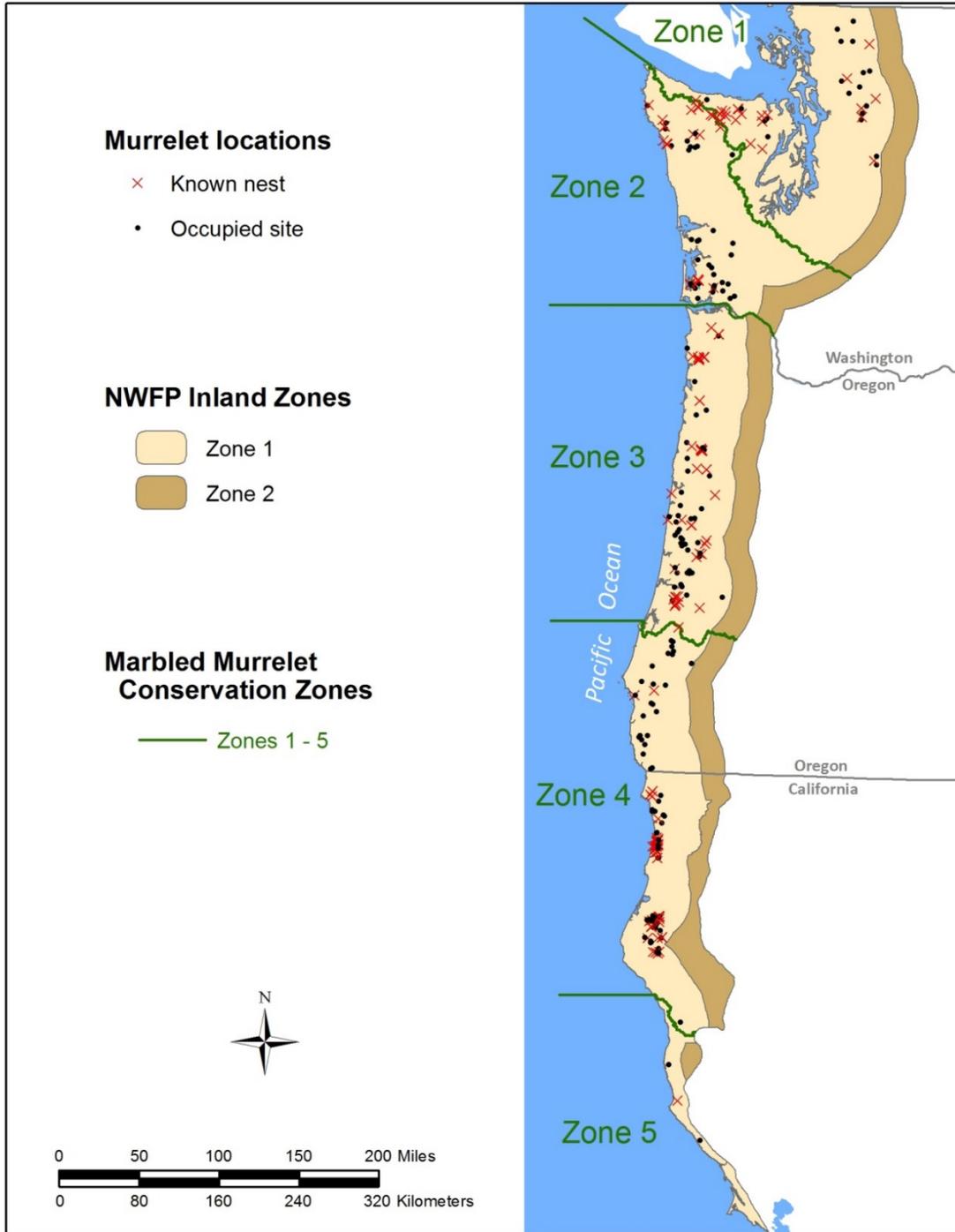
**Table 2-15.** Mean human footprint rank (see text) of higher-suitability murrelet nesting habitat in edge and core in 1993 and 2012.

State	Year	Habitat Pattern		
		Core	Core-edge	Edge
Washington	1993	2.72	2.85	3.24
	2012	2.63	2.79	3.20
Oregon	1993	3.60	3.63	3.58
	2012	3.64	3.67	3.64
California	1993	3.93	3.88	3.62
	2012	4.02	4.00	3.66

**Figure 2-1.** Locations of Northwest Forest Plan reserved and nonreserved land use allocations on federal lands within the range of the marbled murrelet, as of 2012. Also depicted are physiographic provinces as defined by the NWFP, and locations of NWFP Inland Zones, which are denoted as Zone 1 closer to the west coast and Zone 2 farther from the coast. Nonfederal lands are depicted in gray.



**Figure 2-2.** Locations of known marbled murrelet nest sites (including downy young and egg shells) and occupied sites used as training data for building habitat suitability models using Maxent software. See “Murrelet Locations” in text for definitions of nest sites and occupied sites. Also shown are the locations of the five Marbled Murrelet Conservation Zones, as defined in the U.S. Fish and Wildlife Service’s marbled murrelet recovery plan (USFWS 1997).



**Figures 2-3a to 2-3c.** Mean and 95% confidence intervals of the ratio between predicted frequencies (P/E ratio, also referred to in text as Area Adjusted Frequency --AAF) of occurrence from 25 replicated Maxent model runs in Washington (a), Oregon (b) and California (c). The vertical dashed lines show the values used to separate the continuous Maxent model output (logistic probability) into the 4 habitat suitability classes used to evaluate nesting habitat status and trends. As described in the text, these cutpoint values were computed separately for each state (modeling region), and thus differ between states

Figure 2-3a.

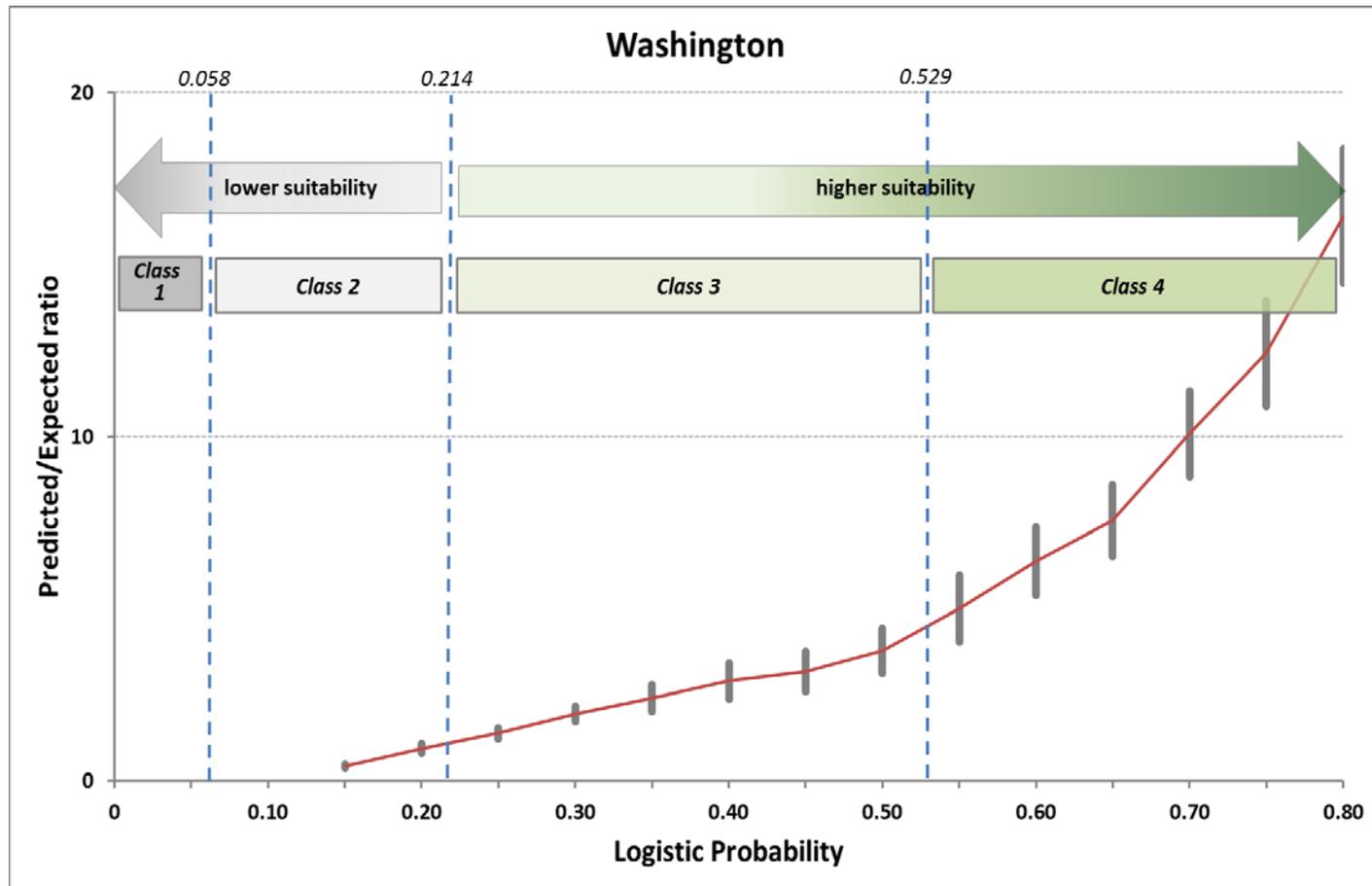


Figure 2-3b.

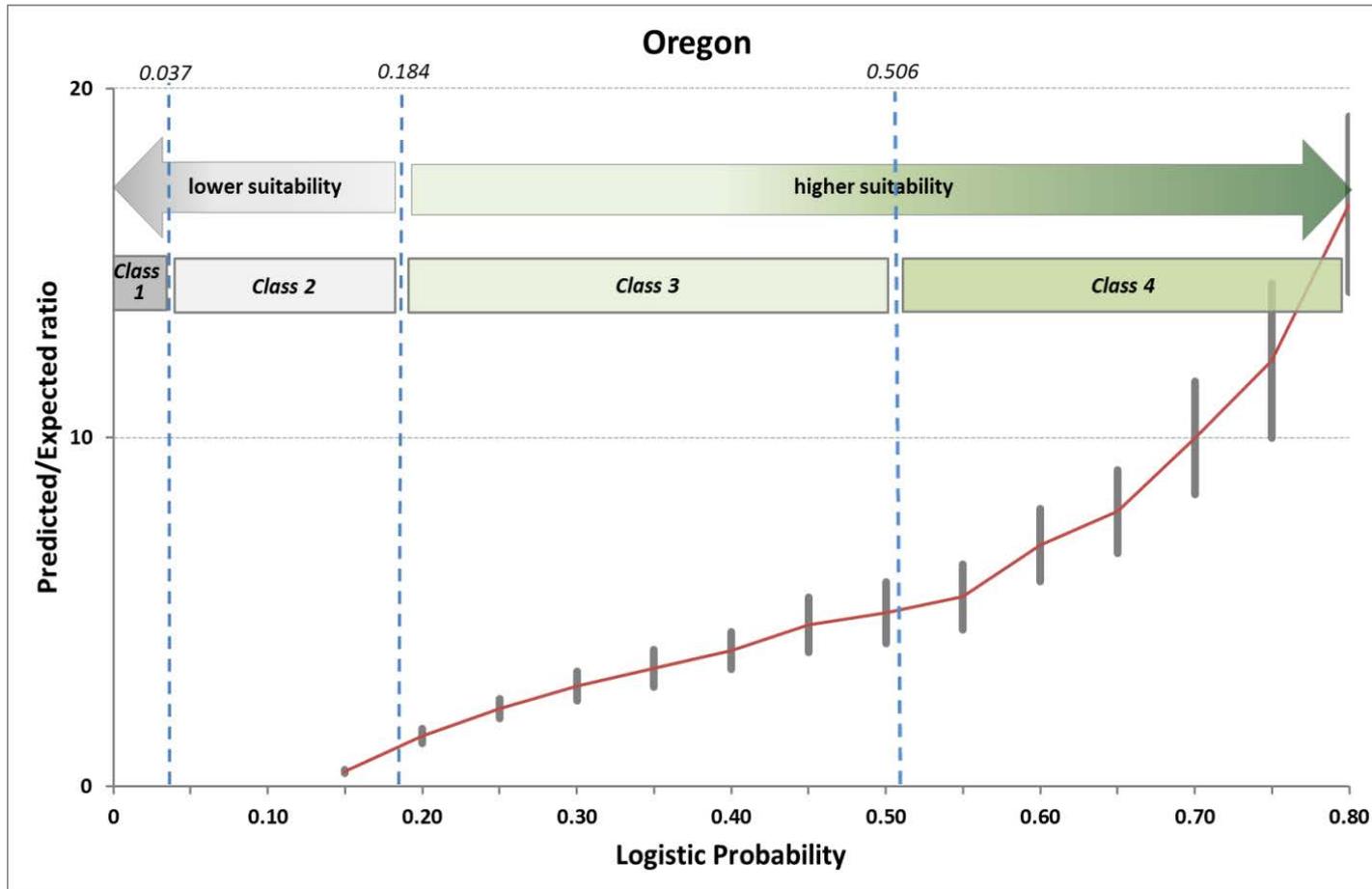
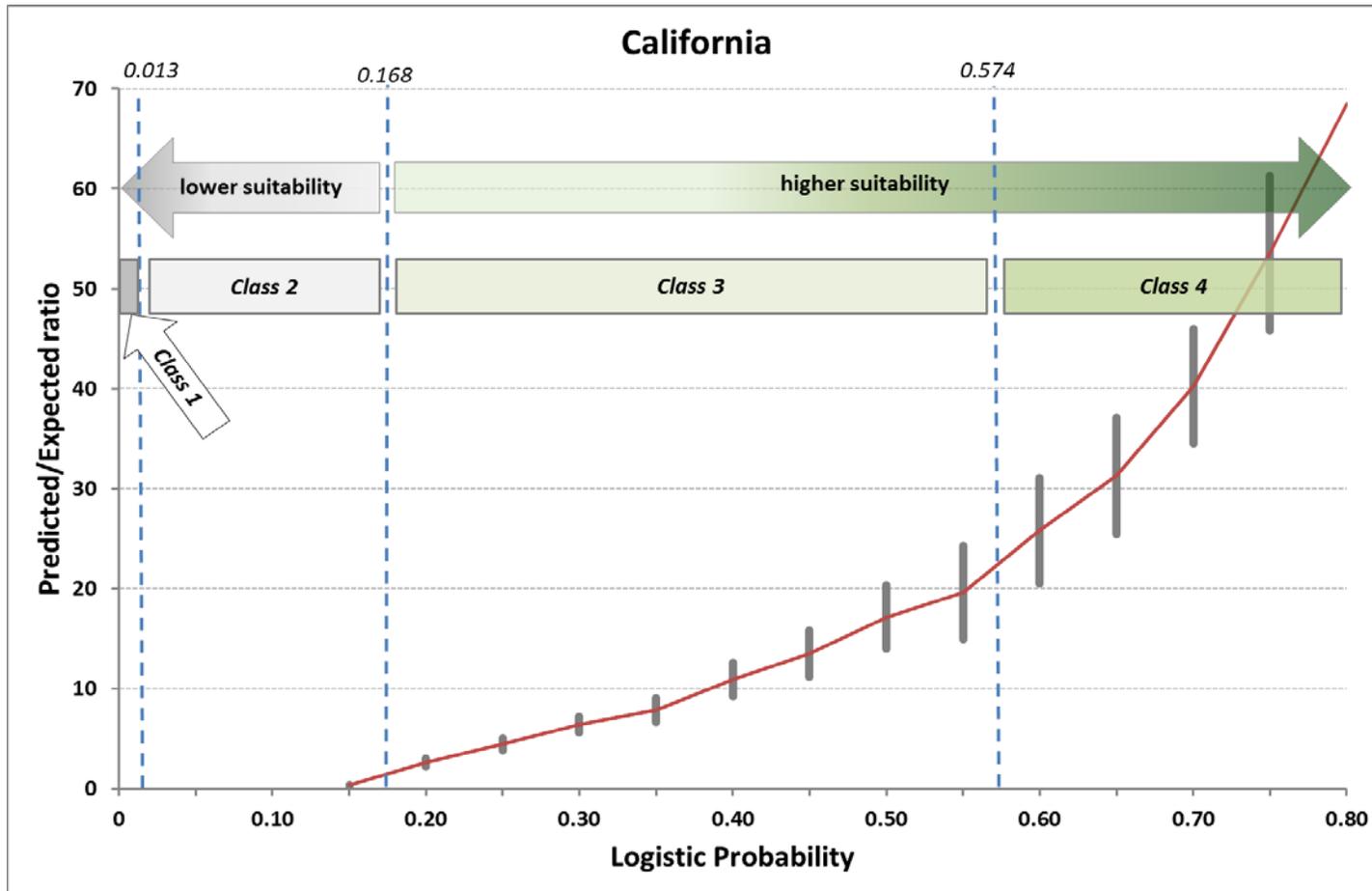
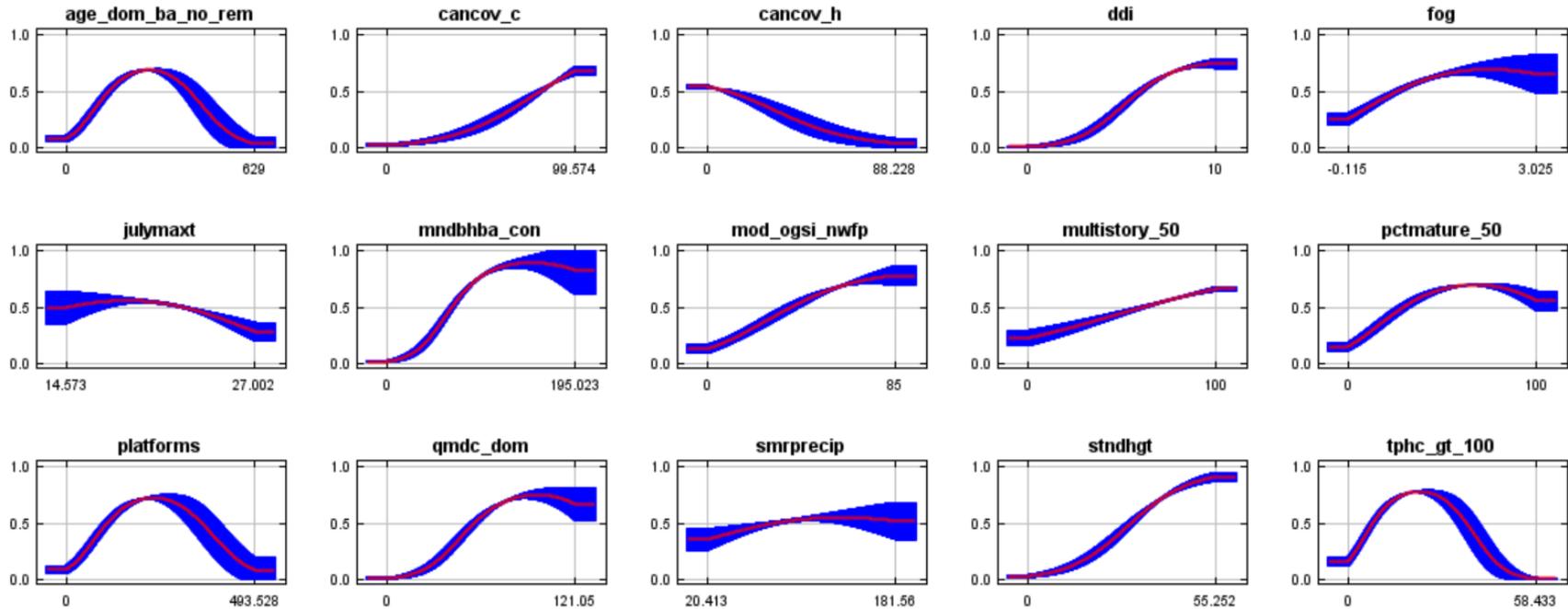


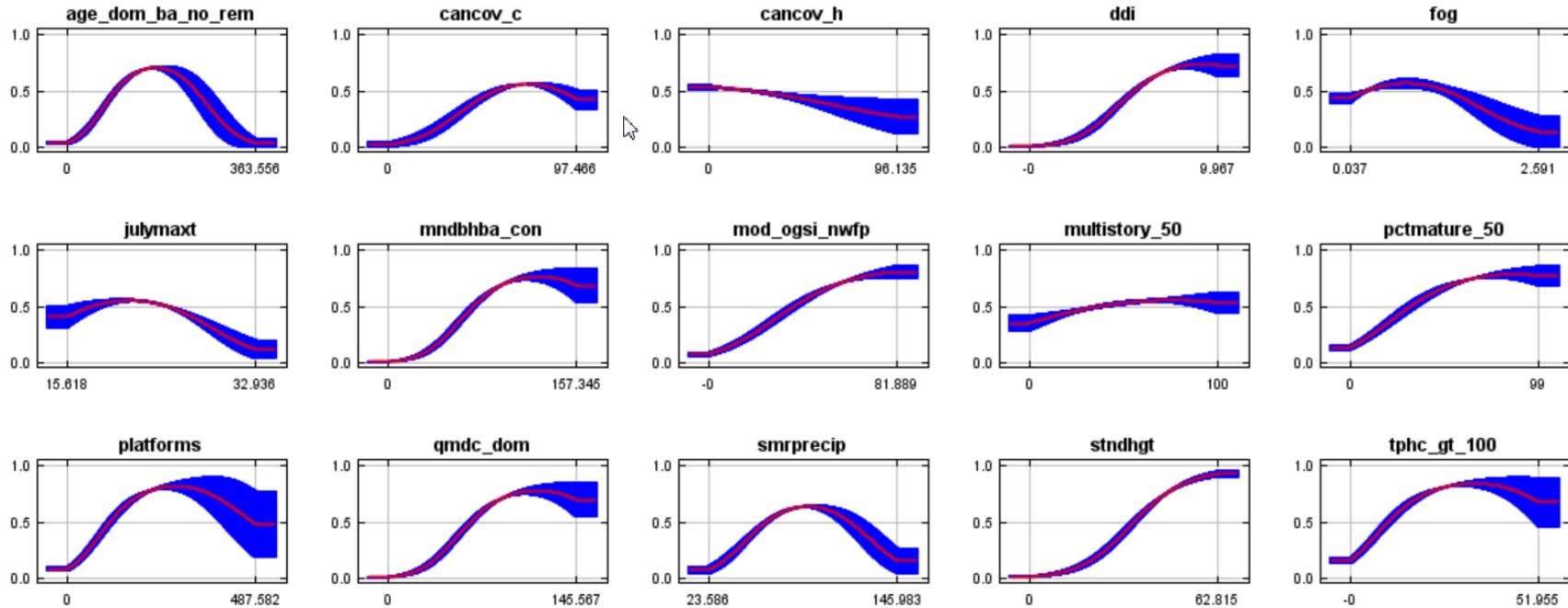
Figure 2-3c.



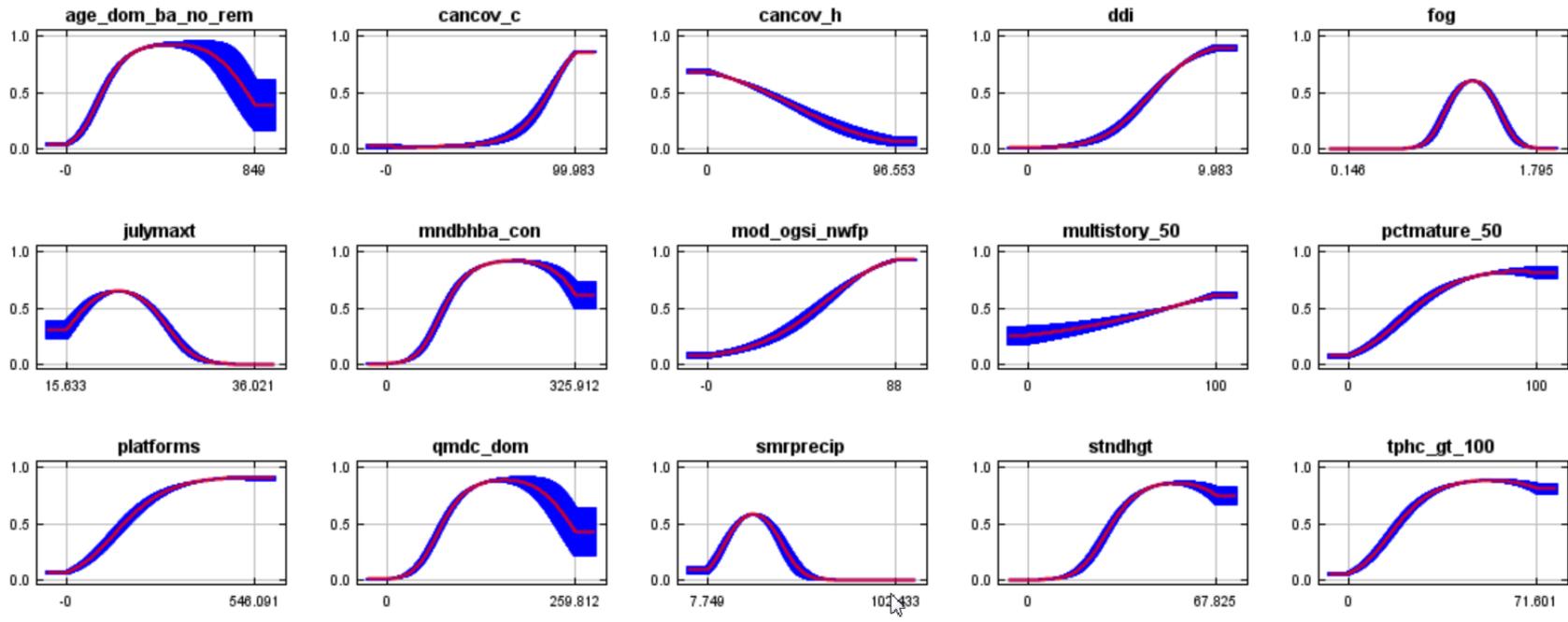
**Figure 2-4a.** Response curves showing how each variable affects the habitat suitability score in Washington. The red lines indicate mean response across 25 replicated model runs; blue shapes represent 1 SD above and below the mean. The curves show in relative terms how the habitat suitability changes across the range of covariate values (horizontal axis) found within the modeling region, when all other covariates are kept at their average value in the modeling region.



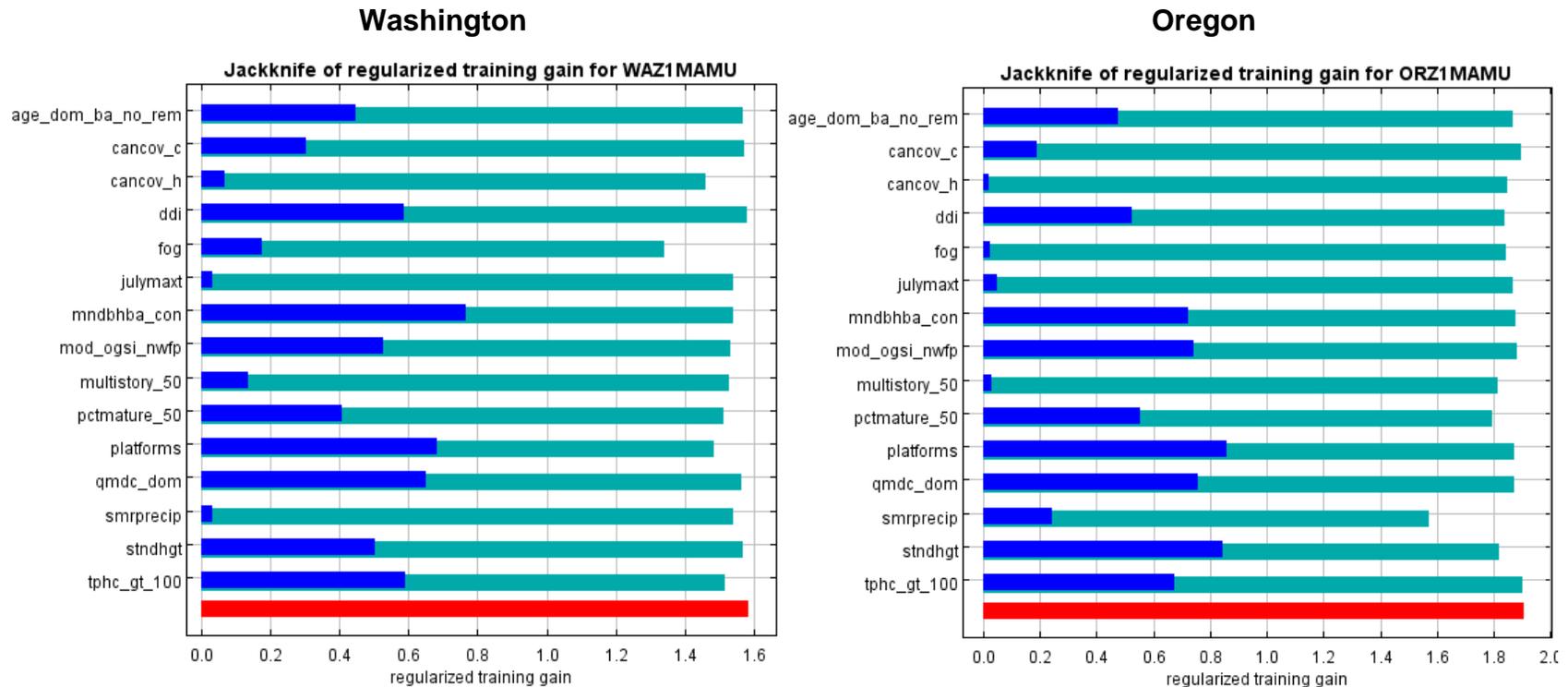
**Figure 2-4b.** Response curves showing how each variable affects the habitat suitability score in Oregon. The red lines indicate mean response across 25 replicated model runs; blue shapes represent 1 SD above and below the mean. See Figure 2-4a for additional details.



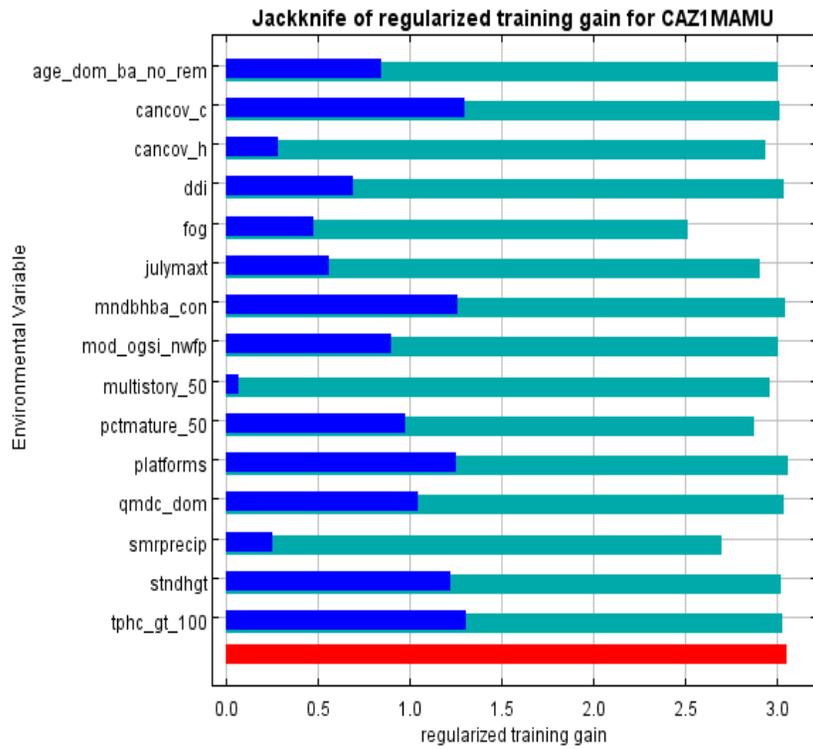
**Figure 2-4c.** Response curves showing how each variable affects the habitat suitability score in California. The red lines indicate mean response across 25 replicated model runs; blue shapes represent 1 SD above and below the mean. See Figure 2-4a for additional details.



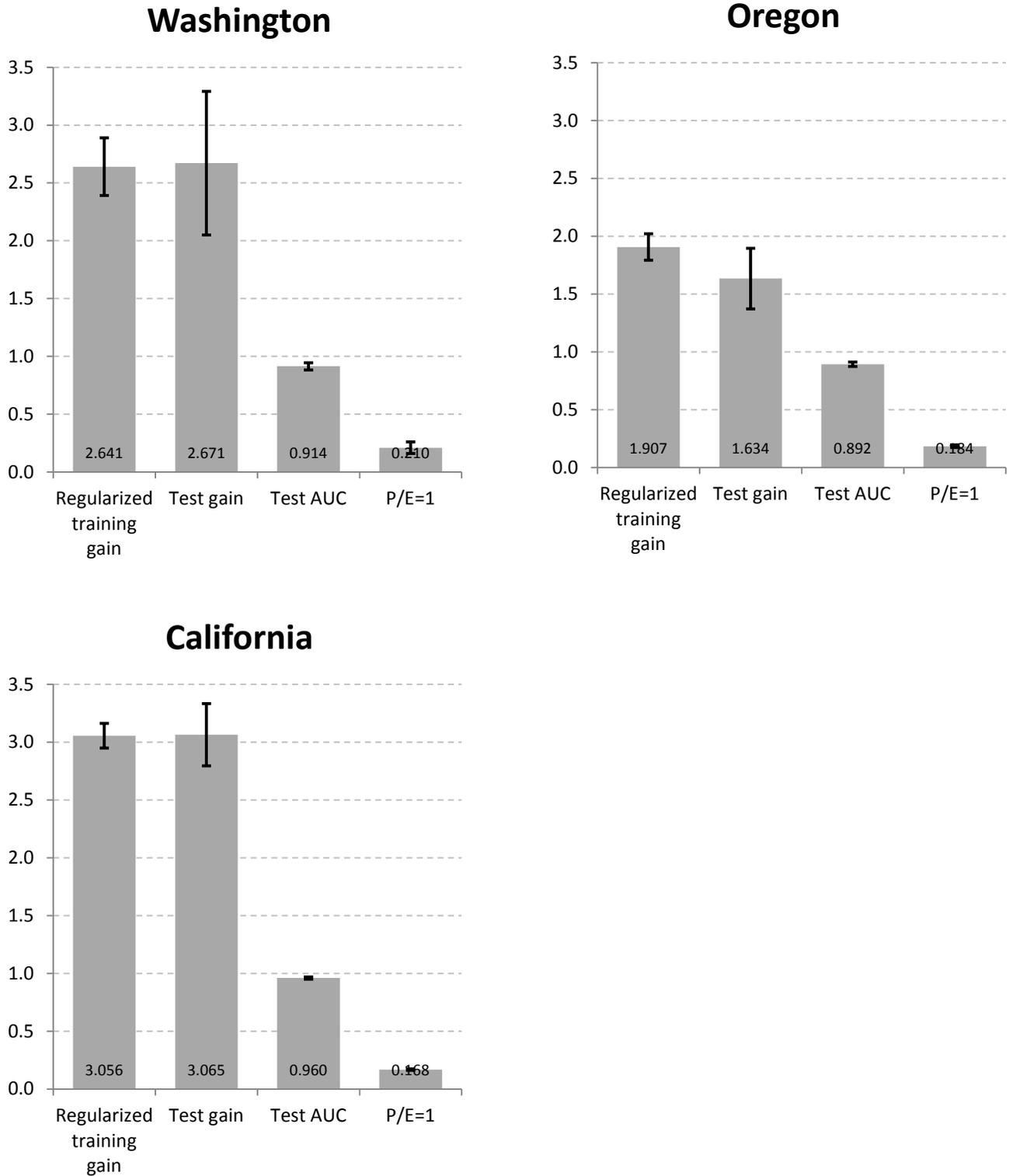
**Figure 2-5.** Contributions of environmental variables (covariates) to Maxent models of habitat suitability in Washington, Oregon, and California. The red bar indicates gain from a model with all covariates included in the model. The dark blue bars indicate gain from a model with only that covariate included. The light blue bars indicate the reduction in gain (relative to the red bar) that would occur if that covariate was removed from the model but all other covariates were included.



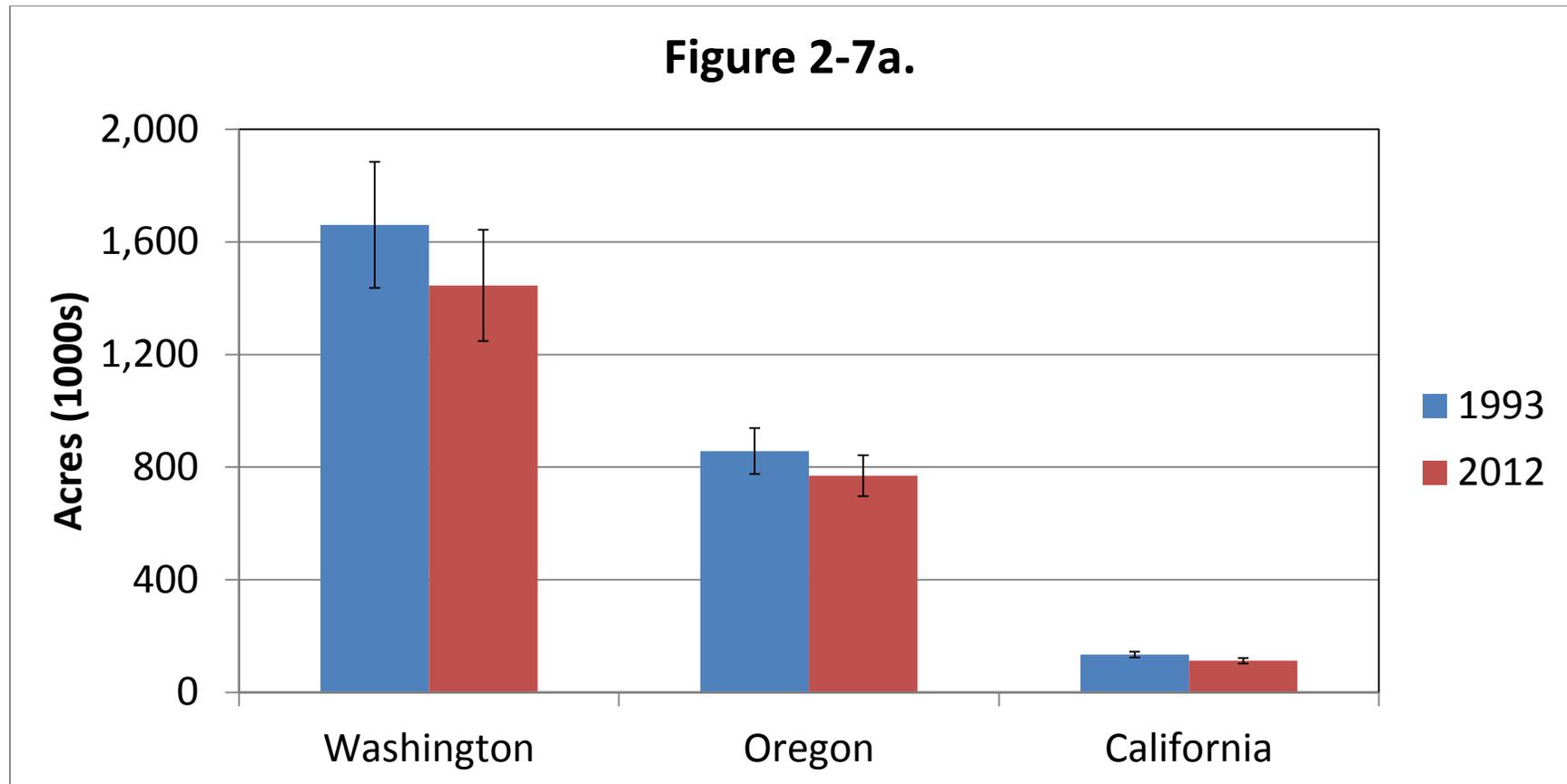
### California



**Figure 2-6.** Model performance and thresholds by model region (state). See text for explanations of performance metrics.



**Figures 2-7a to 2-7d.** Acres of higher-suitability nesting habitat (classes 3 and 4 combined) at start of NWFP (1993) and 2012, by (a) state, (b) land use allocation, (c) physiographic province, and (d) conservation zone. Bars represent 95% confidence intervals computed across 25 replicated model runs.



**Figure 2-7b.**

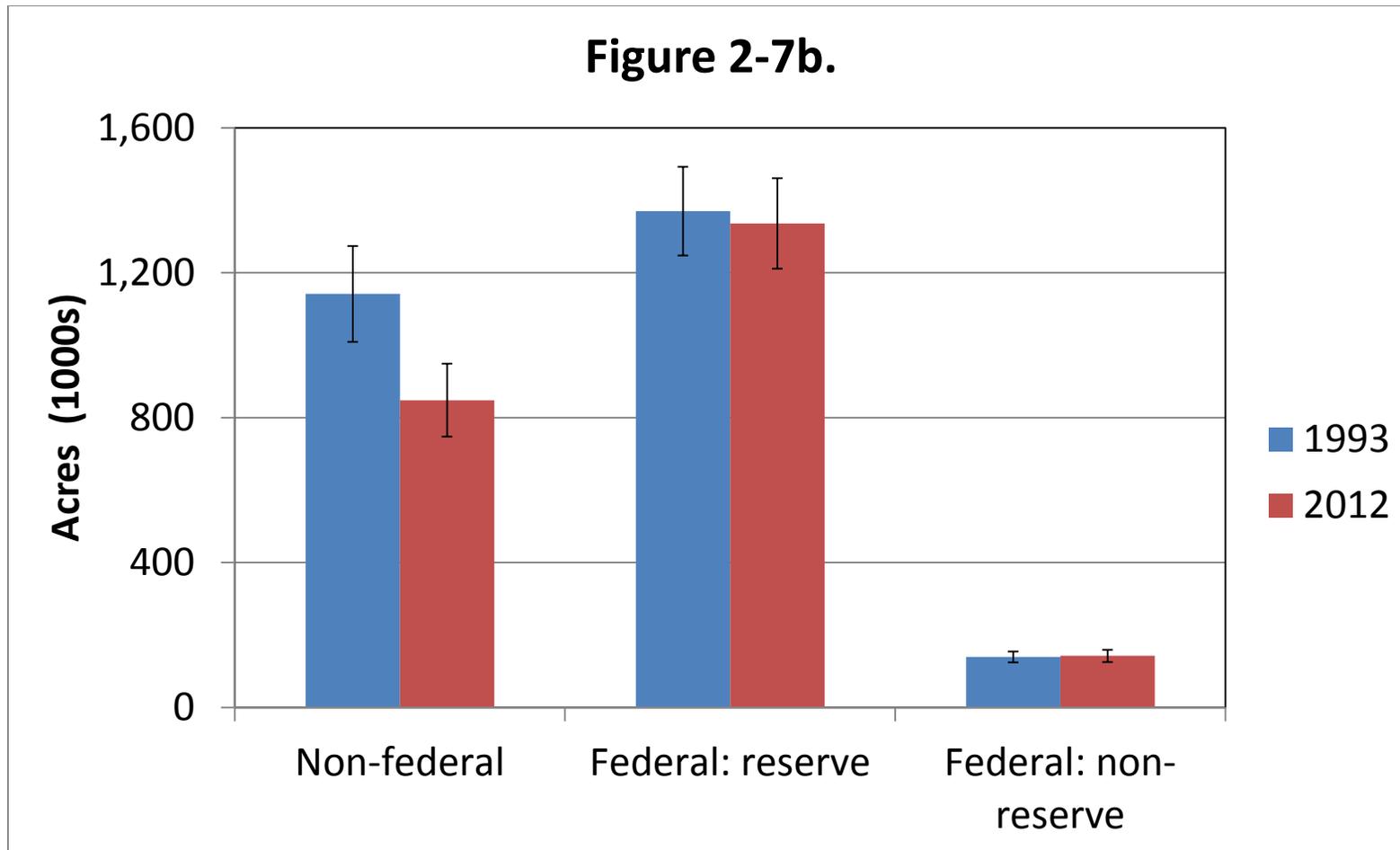
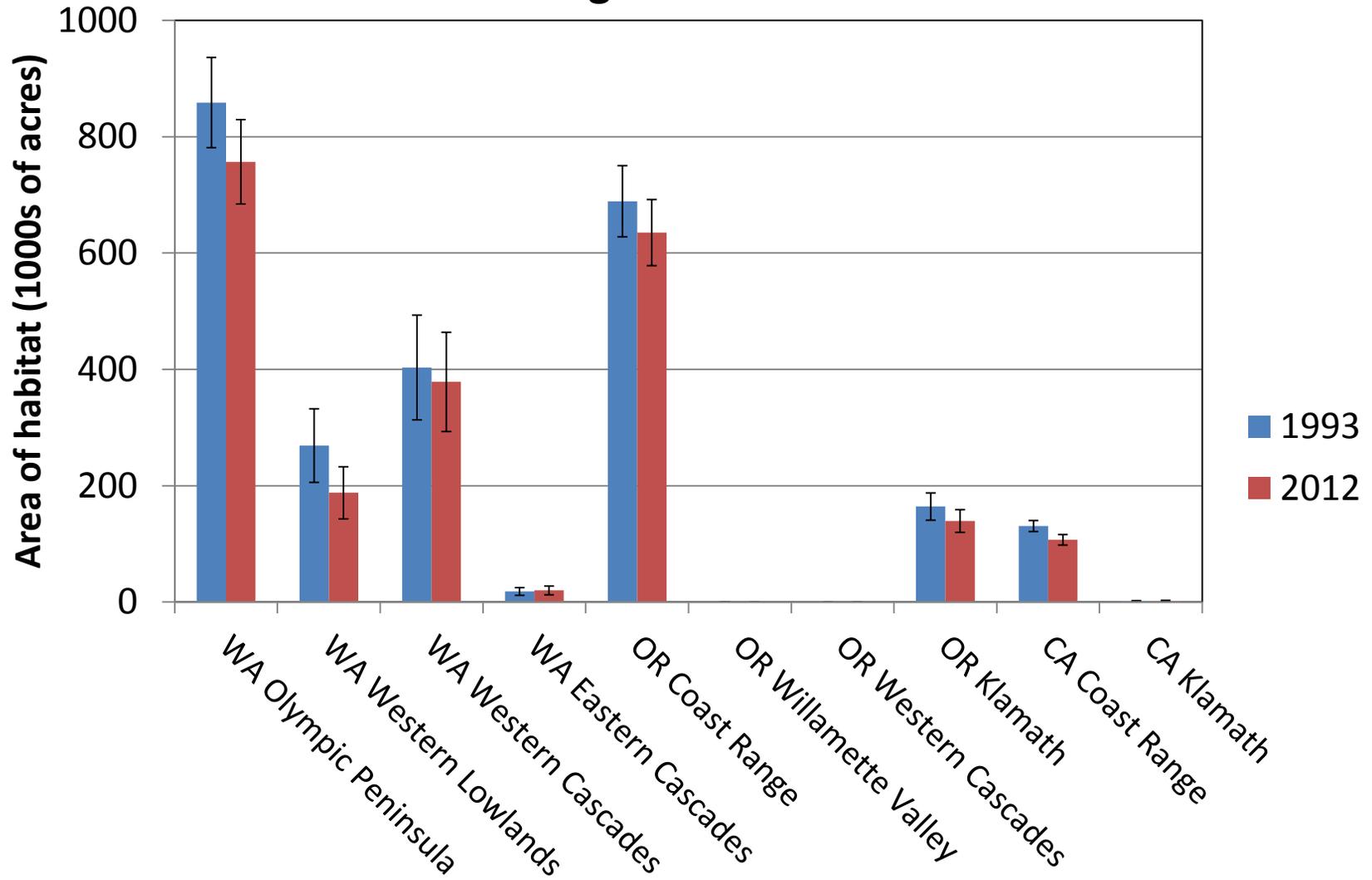
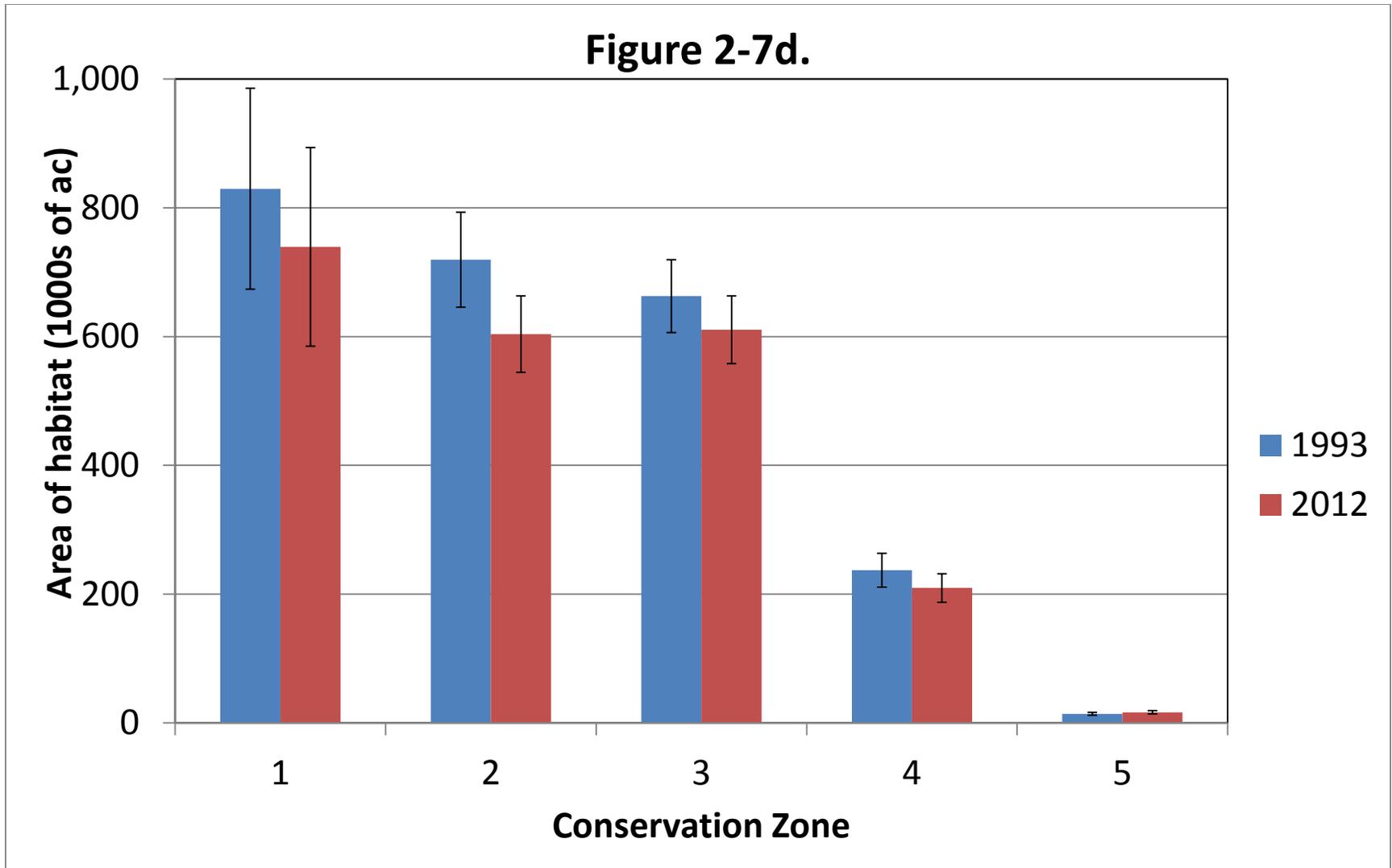


Figure 2-7c.



**Figure 2-7d.**



**Figures 2-8a to 2-8d.** Mean change, in thousands of acres, of higher-suitability nesting habitat (classes 3 plus 4 combined) from 1993 to 2012 by (a) state, (b) land use allocation, (c) physiographic province, and (d) the mean change, in thousands of acres, of only the highest-suitability habitat (Class 4) by state. Bars represent 95% confidence intervals computed based on the results of 25 replicated model runs per state model.

**Figure 2-8a.**

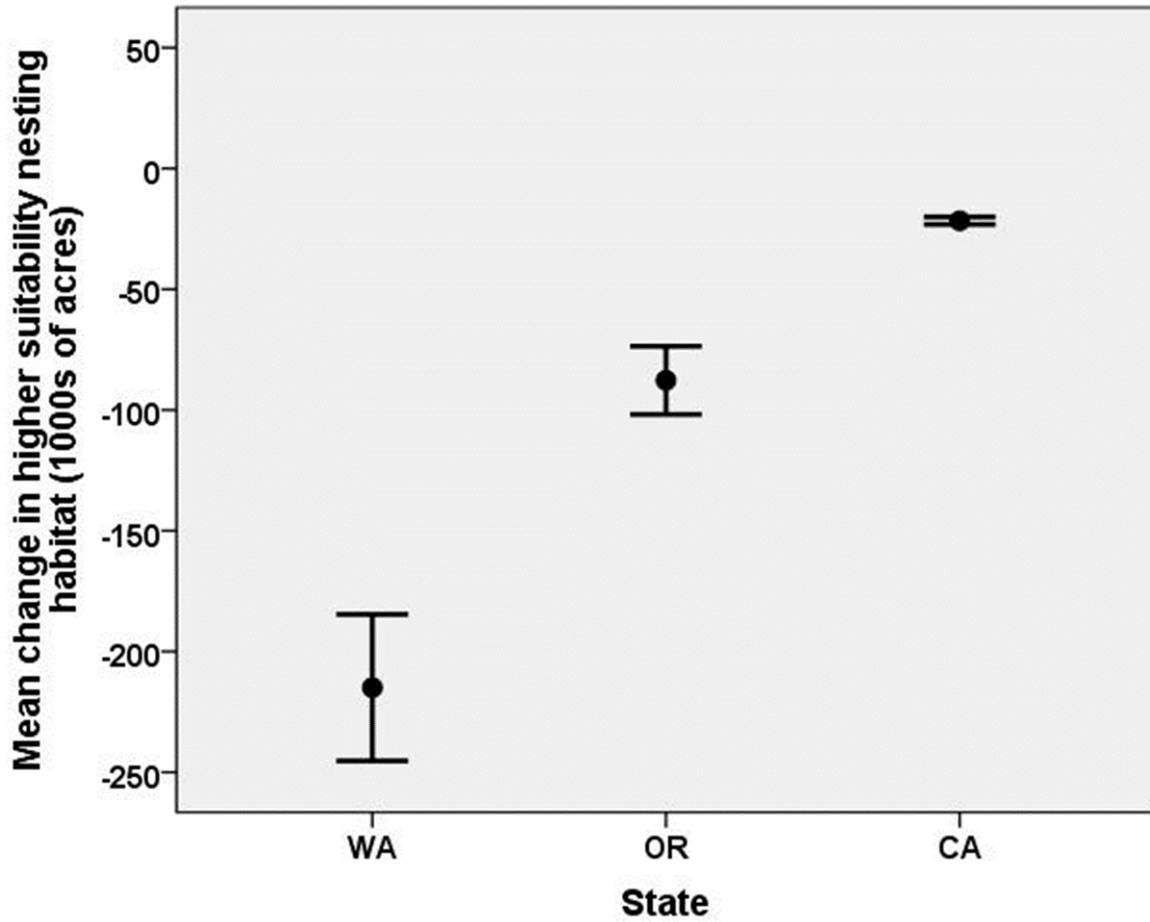


Figure 2-8b.

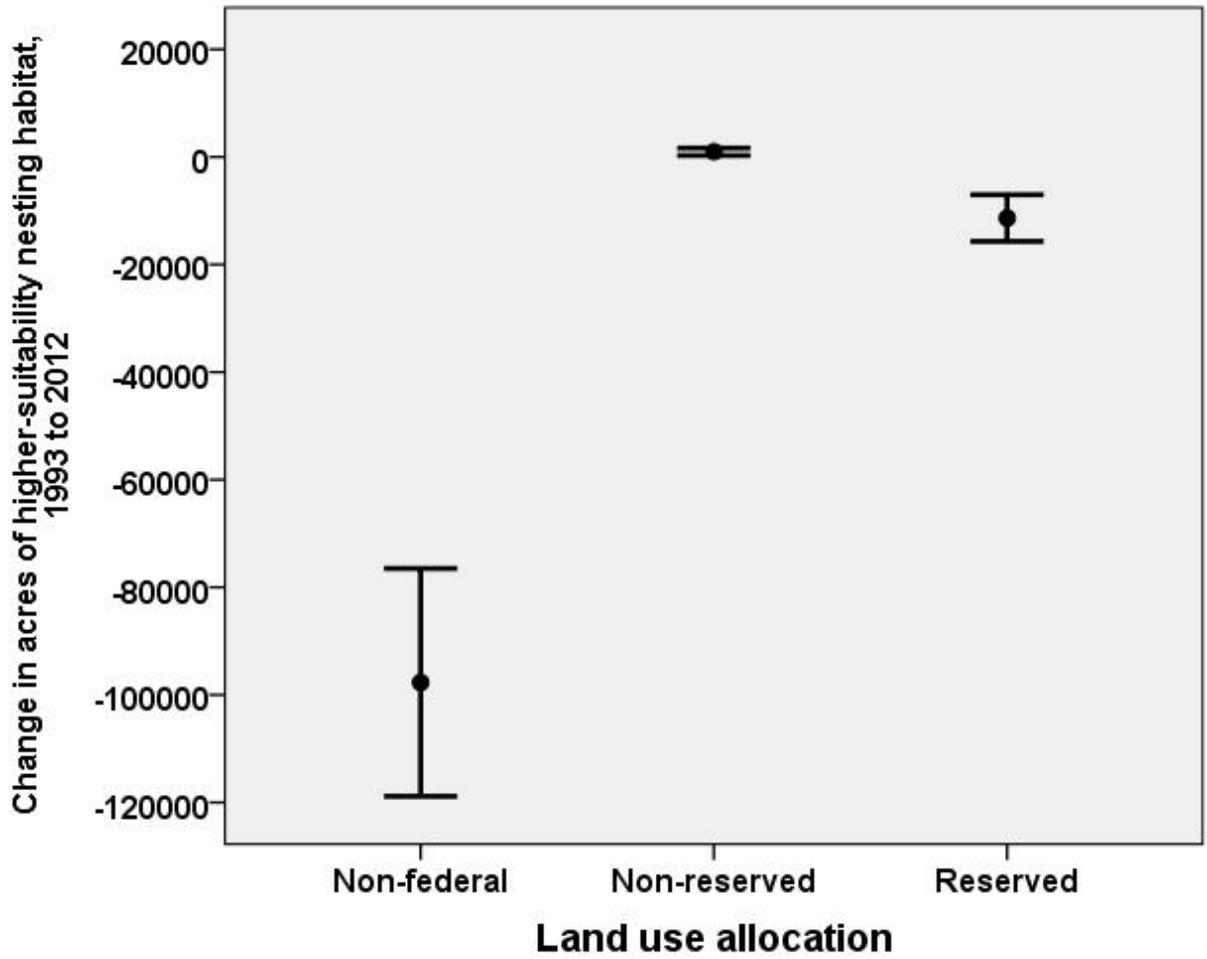


Figure 2-8c.

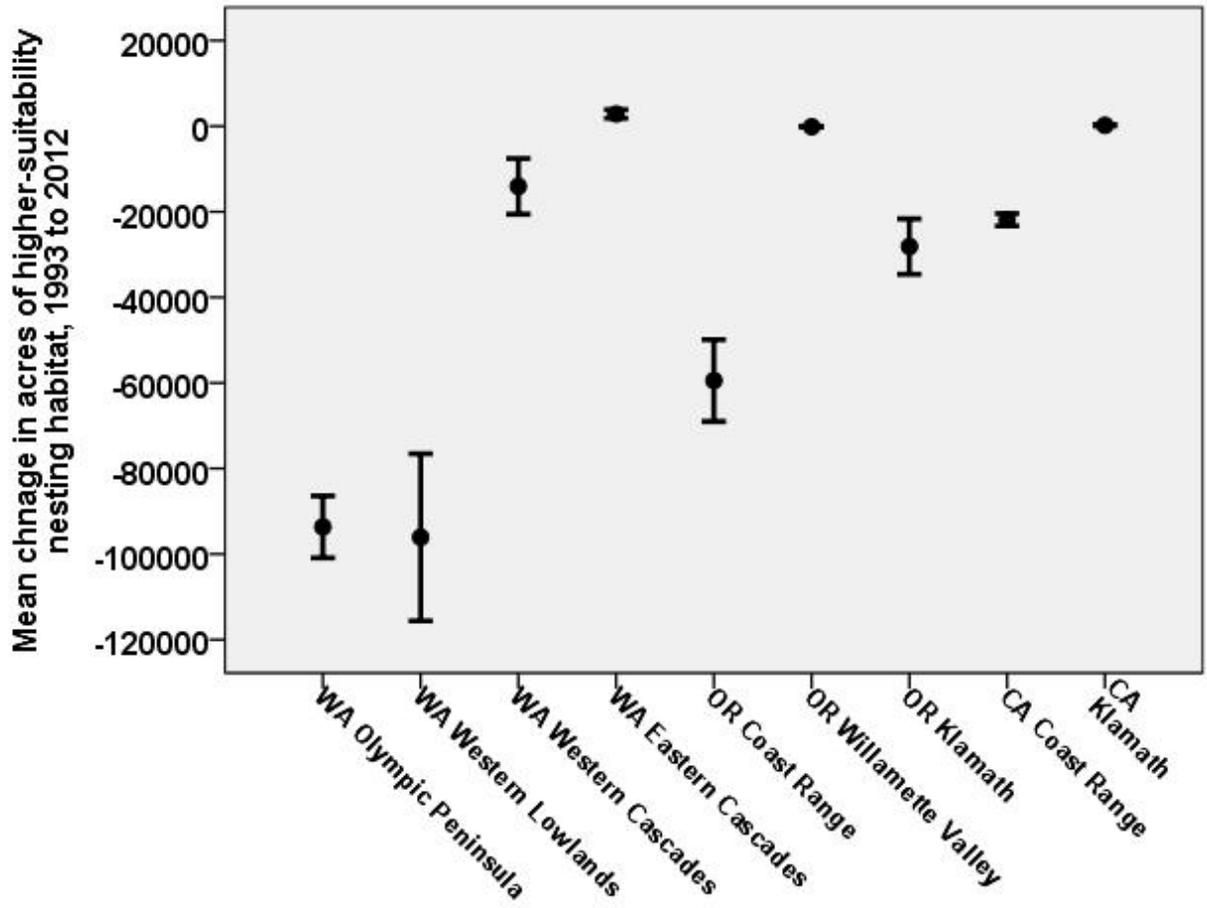
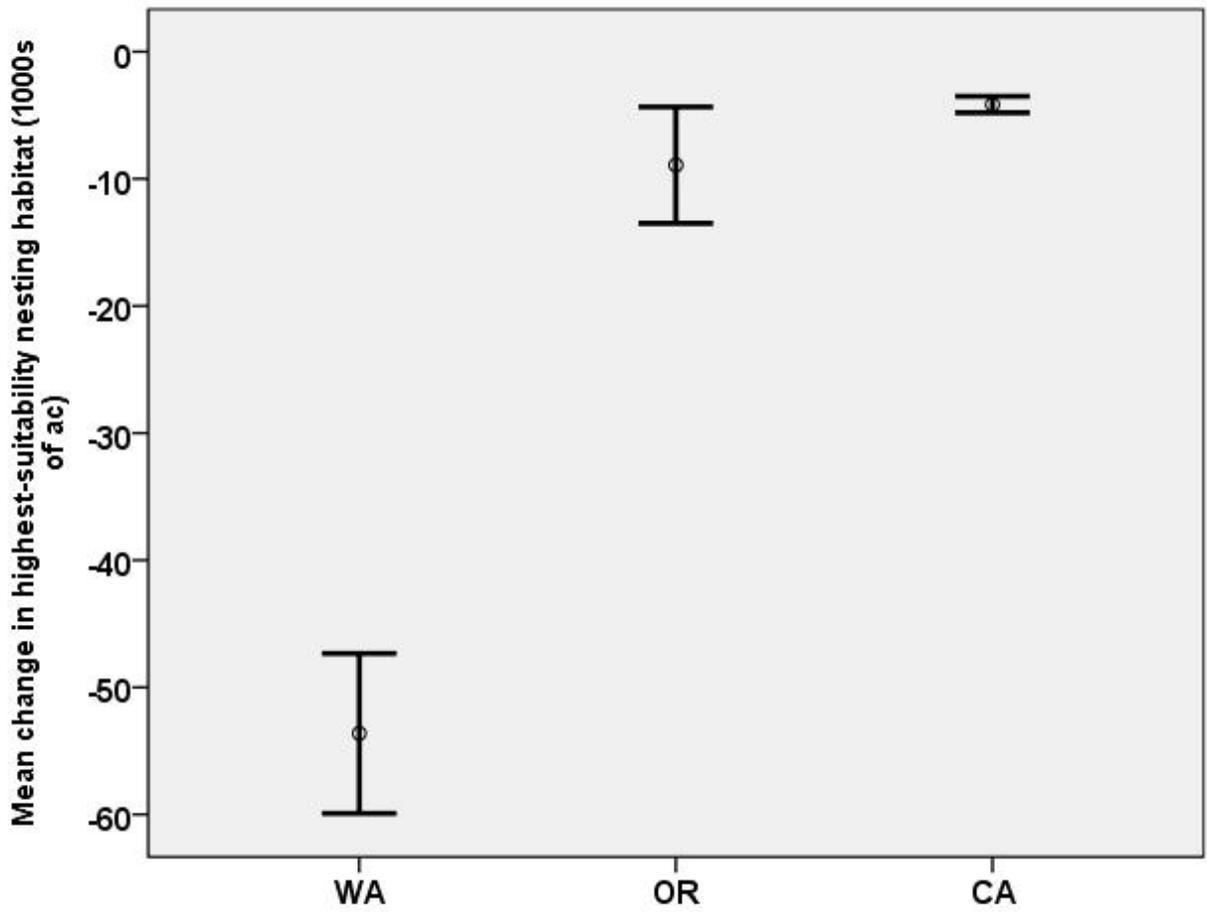


Figure 2-8d.



**Figures 2-9 to 2-11.** Habitat suitability maps for Washington, Oregon, and California, for the 2012 ‘bookend’ year, the last year of the modeling period. NWFP Inland Zones are denoted as Zone 1 closer to the west coast and Zone 2 farther away from the coast.

Figure 2-9.

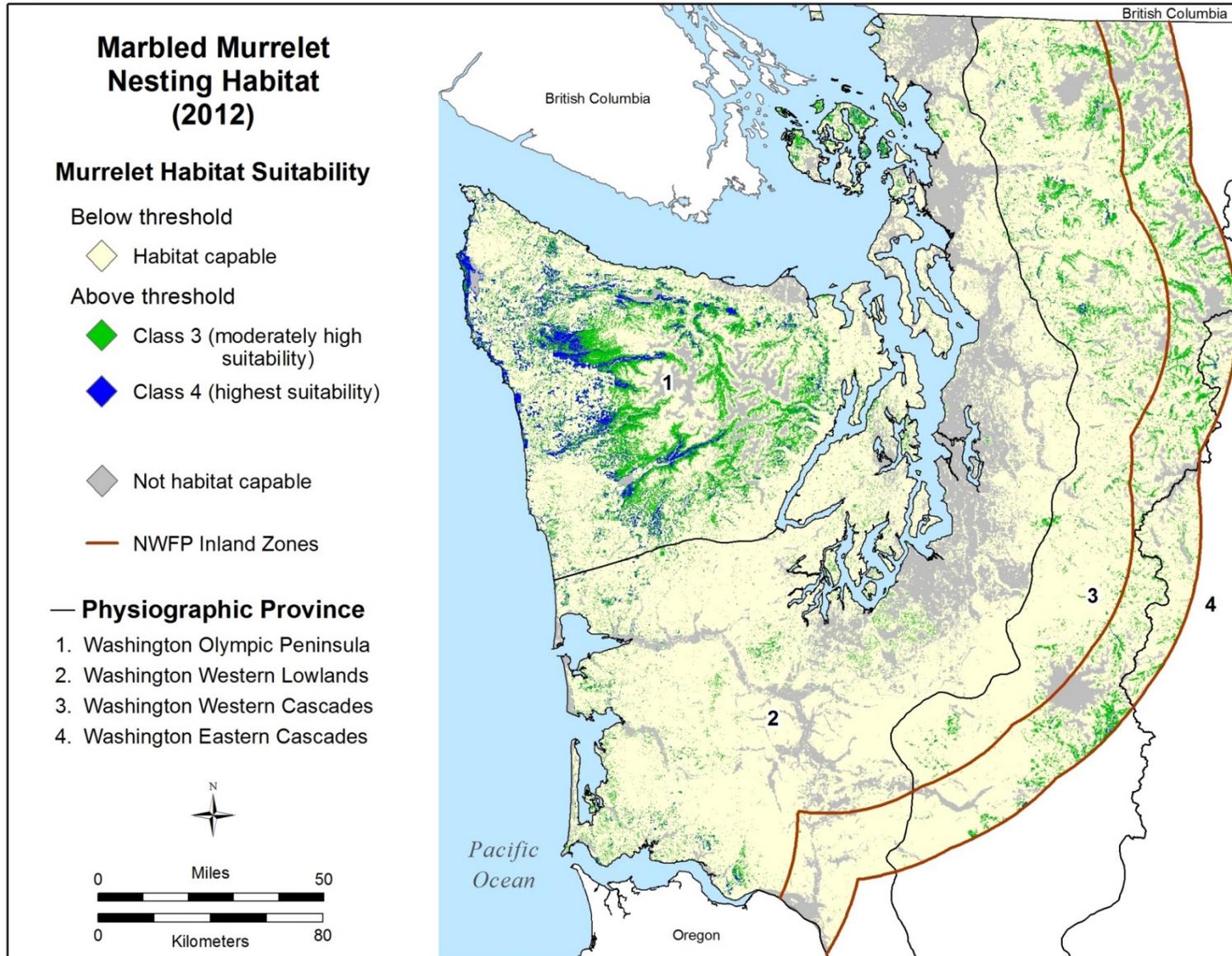


Figure 2-10.

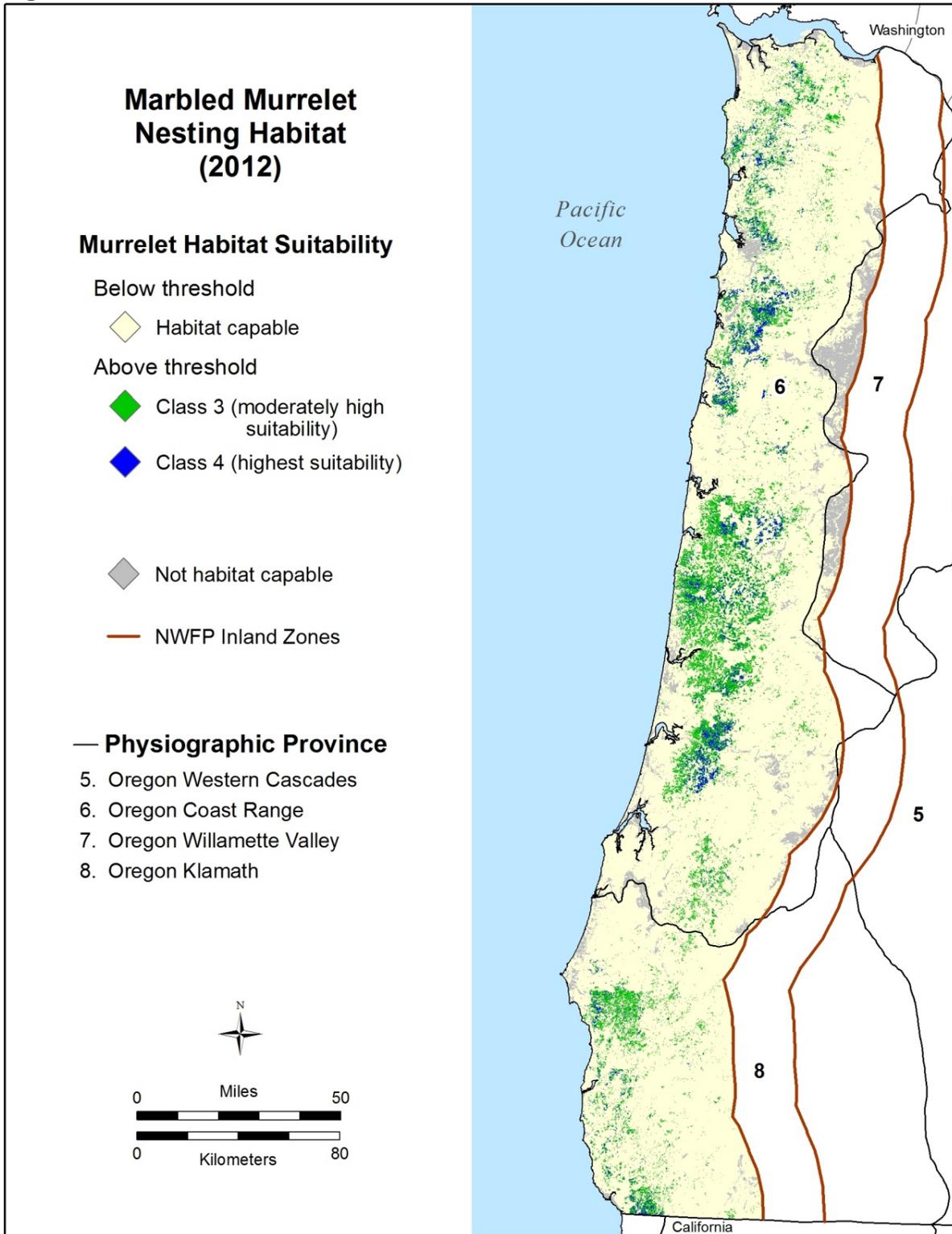
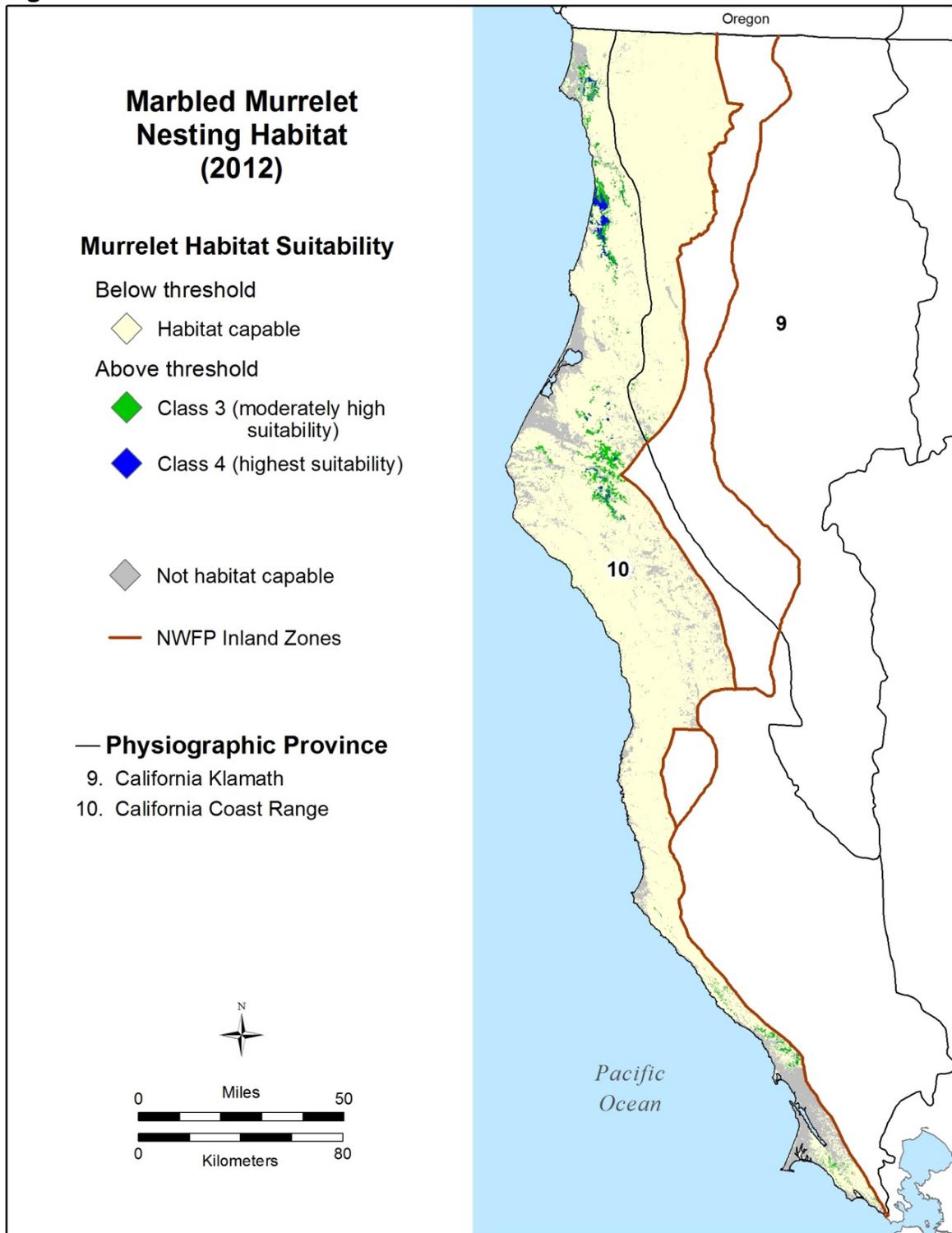
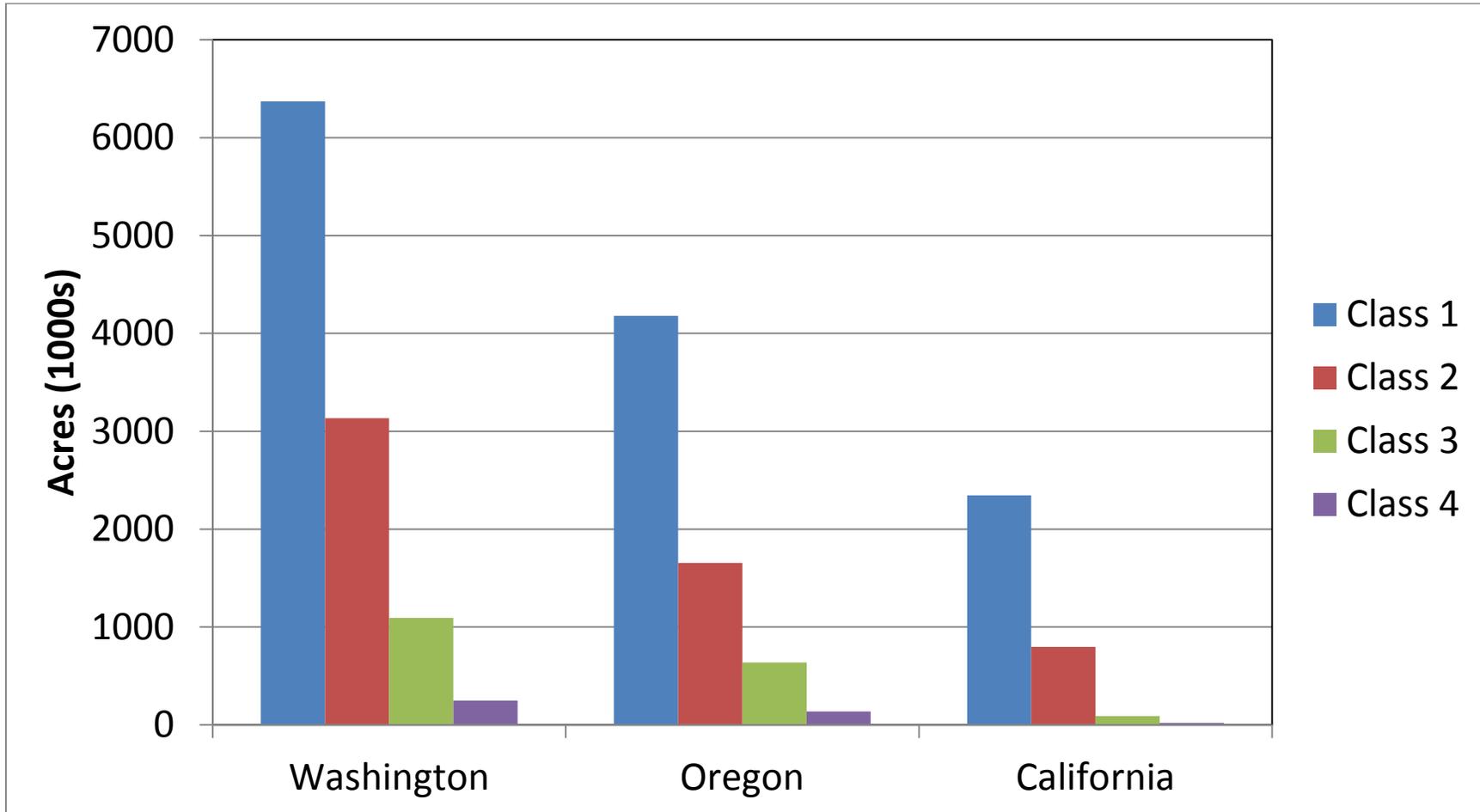


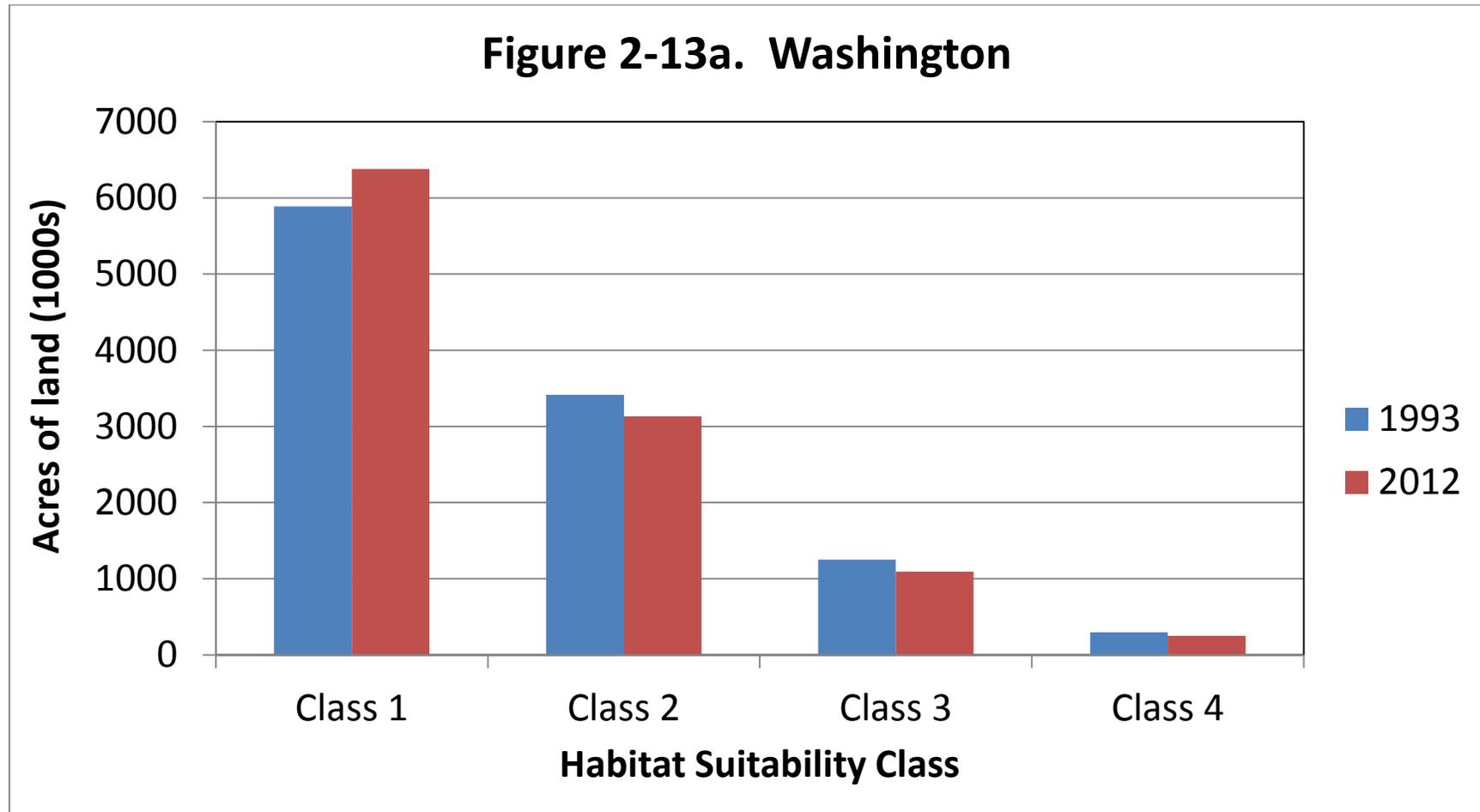
Figure 2-11.



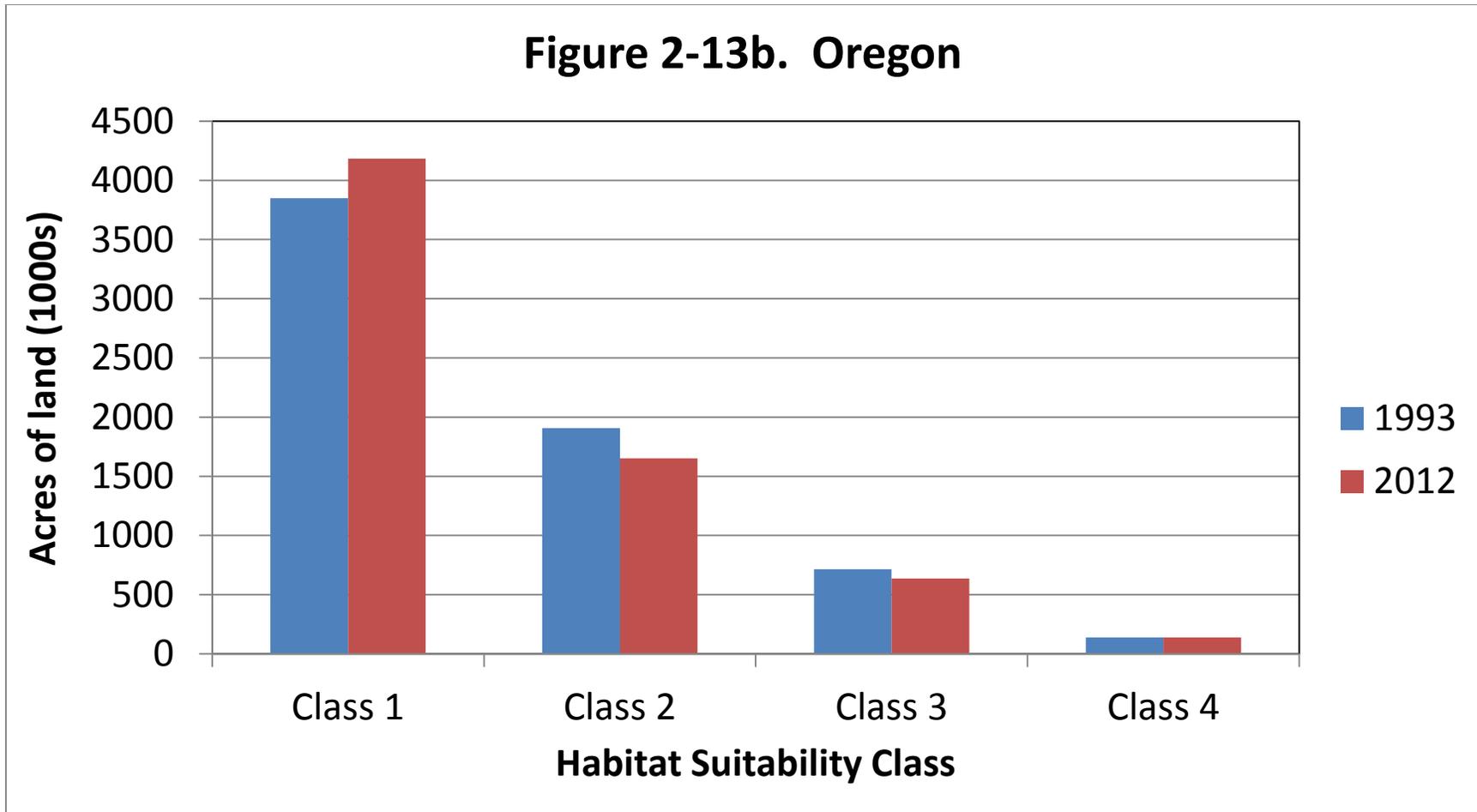
**Figure 2-12.** Distribution of habitat capable acres as classified by habitat suitability score in each state, 2012. Class 1 represents the lowest score and Class 4 the highest. See Figure 2-3 for the cutpoint values used to denote each class.



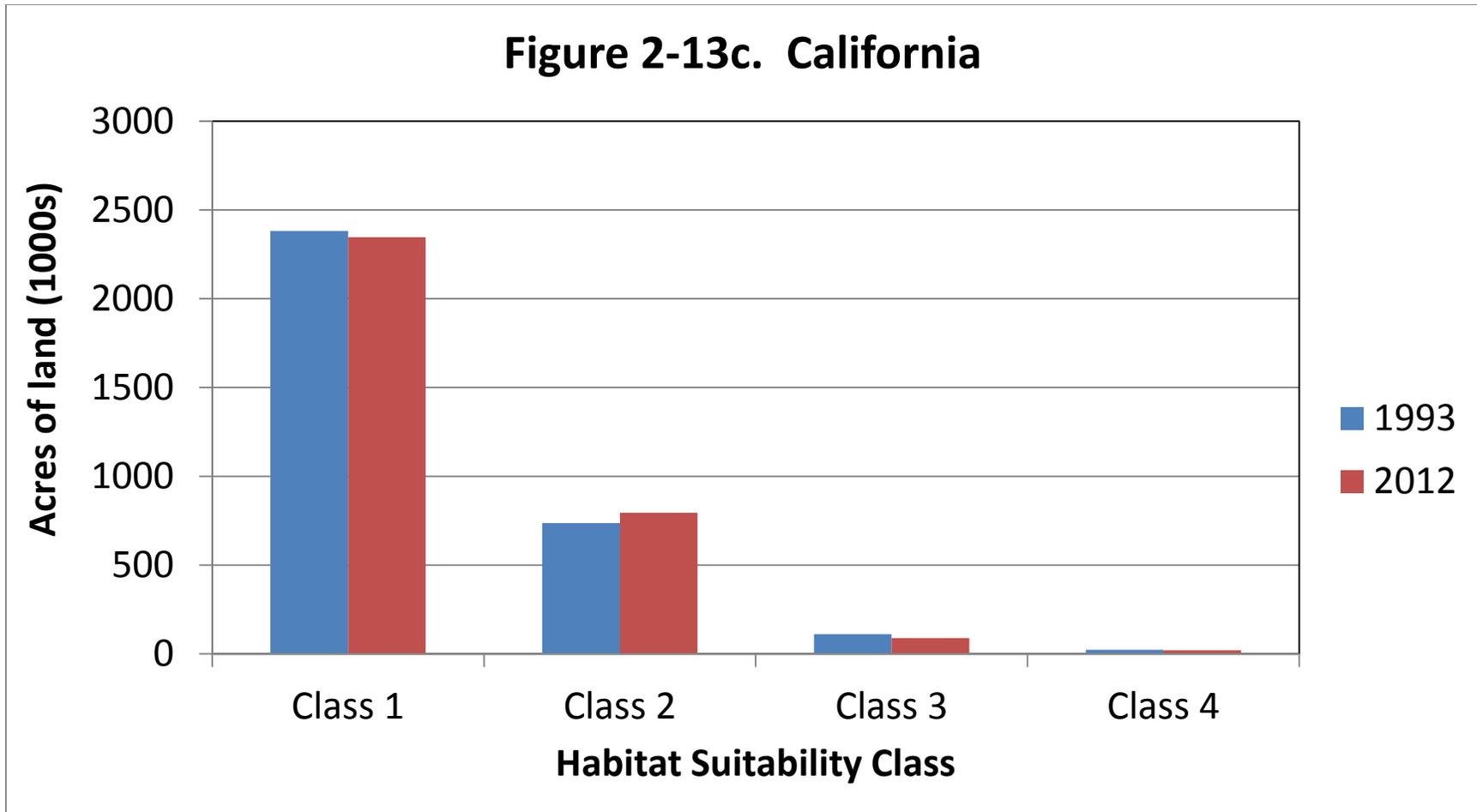
**Figures 2-13a to 2-13c.** Acres of habitat within each habitat suitability class at the baseline (1993) and at the end of the modeling period (2012), by state. Higher-suitability habitat includes class 3 (moderately high suitability) plus class 4 (highest suitability). Values are the means from 25 replicated Maxent runs in each state.



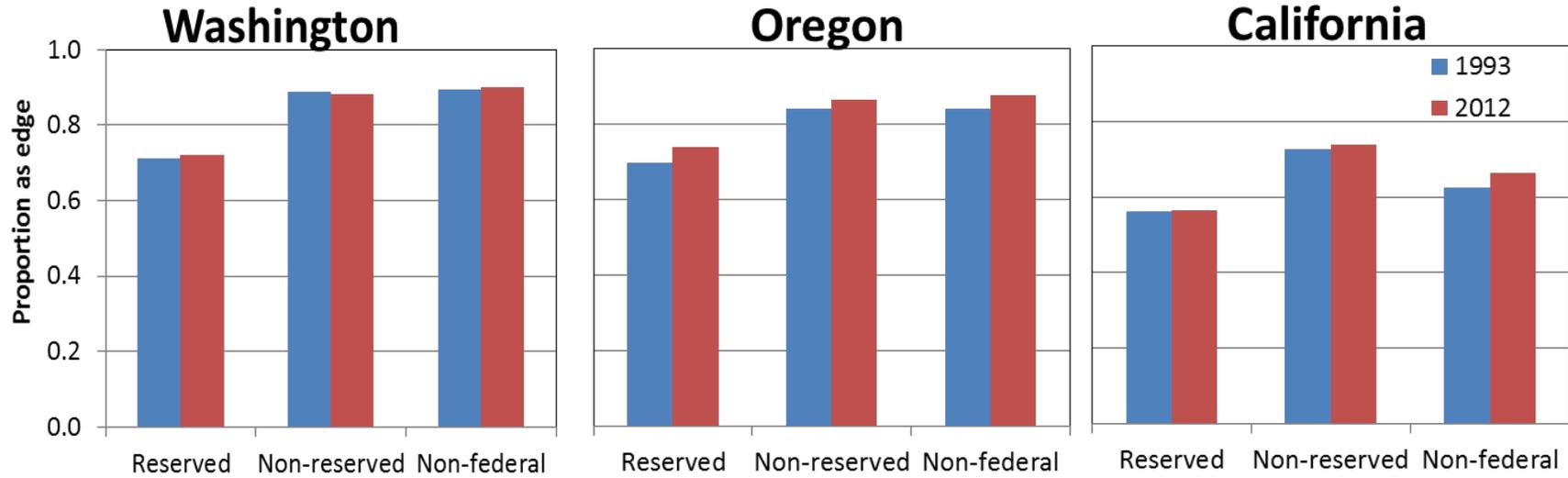
**Figure 2-13b. Oregon**



**Figure 2-13c. California**



**Figure 2-14.** Proportion of higher-suitability habitat occurring as edge habitat (versus interior habitat) in 1993 and 2012, by land use allocation.



### **Chapter 3. Factors Influencing Status and Trend of Marbled Murrelet Populations: An Integrated Perspective**

Martin G. Raphael, Andrew J. Shirk, Gary A. Falxa, Deanna Lynch, S. Kim Nelson, Scott F. Pearson, Craig Strong, Richard D. Young

*Martin G. Raphael is a Research Wildlife Biologist, USDA Forest Service, Pacific Northwest Research Station, 3625 93<sup>rd</sup> Ave. SW, Olympia WA 98512; Andrew J. Shirk is an ecologist, Climate Impacts Group, College of the Environment, University of Washington, Box 355674, Seattle, WA 98195; Gary A. Falxa is a Fish and Wildlife Biologist, U.S. Fish and Wildlife Service, Arcata Fish and Wildlife Office, 1655 Heindon Rd., Arcata, CA 95521; Deanna Lynch is a Fish and Wildlife Biologist, U.S. Fish and Wildlife Service, Washington Fish and Wildlife Office, Branch of Listing, 510 Desmond Dr., Suite 102, Lacey, WA 98503; S. Kim Nelson is a Research Wildlife Biologist, Oregon Cooperative Fish and Wildlife Research Unit, Oregon State University, Department of Fisheries and Wildlife, 104 Nash Hall, Corvallis, OR 97331; Scott F. Pearson is a senior scientist, Wildlife Science Division, Washington Department of Fish and Wildlife, 1111 Washington St. SE, 5th Floor, Olympia, WA 98501; Craig Strong is a Consultant Researcher, Crescent Coastal Research, 7700 Bailey Road, Crescent City, CA 95531; and Richard D. Young is a GIS Analyst, U.S. Fish and Wildlife Service, Pacific Regional Office, Ecological Services, 911 NE 11<sup>th</sup> Avenue, Portland, OR 97232.*

#### **Abstract**

The Northwest Forest Plan was implemented, in part, to provide habitat conditions that would contribute to the conservation and recovery of threatened species including the marbled murrelet. Effectiveness monitoring of marbled murrelet populations and nesting habitat helps inform land managers whether this objective is being met. The murrelet depends upon the marine environment to meet its foraging and roosting requirements and upon terrestrial forest to meet its nesting requirements. To assess the relative contributions of terrestrial and marine factors on murrelet population abundance and distribution, we synthesized data on the status and trend of murrelet populations, status and trend of inland nesting habitat, and status and trend of marine factors. Specifically, we initially examined the spatial and temporal correlations of marine and terrestrial factors with the spatial distribution and trend of murrelets. We then used a boosted regression tree analysis to investigate the contributions of a suite of marine and terrestrial factors to at-sea murrelet abundance. In both analyses, we found that numbers of murrelets are strongly correlated with amounts and pattern (large contiguous patches) of suitable nesting habitat and population trend is most strongly correlated with trend in nesting habitat although marine factors also contribute to this trend. Model results suggest that conservation of suitable nesting habitat is key to murrelet conservation. Conservation of habitat within reserves, as well as management actions that are designed to minimize loss of suitable habitat or improve quality of nesting habitat, will likely contribute to murrelet conservation and recovery.

## Introduction

The primary objective of the Marbled Murrelet Effectiveness Monitoring Program is to assess the degree to which land management under the Northwest Forest Plan (NWFP) is contributing to the NWFP goal of stabilizing and increasing murrelet populations by maintaining and increasing murrelet nesting habitat in the NWFP area (Madsen et al. 1999). This objective and goal were motivated by the original charter for the Forest Ecosystem Management Assessment Team (FEMAT) report that called for development of long-term [NWFP] management alternatives that would provide “maintenance and/or restoration of habitat conditions for...the marbled murrelet that will provide for viability” (FEMAT 1993:iv). The murrelet nests on forested lands but feeds, roosts, and spends the majority of its time in the marine environment. Forest managers can directly influence only the bird’s nesting habitat quantity and quality; the management of marine habitat, while important to murrelet conservation, is under the purview of management and regulatory bodies outside of the NWFP. Because the NWFP is a land-based forest ecosystem management program, the ultimate goal of the murrelet effectiveness monitoring program is to relate population trends to the amount and distribution of nesting habitat (Madsen et al. 1999). A long-term objective of the monitoring program is to “[e]xamine predictive relationships between marbled murrelets and nesting habitat conditions in the NWFP area so that trends in nesting habitat might eventually suffice as a surrogate for trends in murrelet populations” (Madsen et al. 1999).

This chapter reports on initial steps toward the long-term objective of relating nesting habitat conditions to the distribution and trend of marbled murrelet populations at sea. It builds on the findings of the at-sea population monitoring (Chapter 1 of this volume) and of inland nesting habitat (Chapter 2). Because of the potential for both marine and terrestrial influences on murrelet populations, we explored the relative contributions of both marine (foraging and survival) and terrestrial (nesting and fecundity) factors on murrelet abundance and distribution. This research is intended to help managers assess whether current management of nesting habitat on Federal lands within the NWFP area is sufficient or needs to be modified, or if agencies also need to influence management of non-Federal lands and marine factors to stabilize and increase murrelet populations. This chapter provides a brief synthesis of the results of these initial steps plus the results of the population and habitat monitoring chapters in this report, as well as a discussion of management implications of those results.

## Methods

**Univariate correlations.**—To address the influences of marine and terrestrial factors on murrelet status and trend, we first examined potential associations between the distribution of murrelets at-sea and individual factors describing the adjacent marine and terrestrial environments during the nesting season (May to September). We did this by examining simple correlations between individual marine and terrestrial factors and

murrelet populations in space and time. Our intent here was to conduct some data explorations to see if there might be relationships. For this exploratory work, we summarized estimates of higher-suitability nesting habitat for 2012 obtained from the work reported in Chapter 2. We also calculated the fragmentation of that nesting habitat by using the patch cohesion metric in Fragstats (McGarigal et al. 2012).

We hoped to assess the influence of the marbled murrelet's primary prey, forage fish, on murrelet populations. However, we found only one forage-fish data set that occurs in our area of interest (Emmett et al., NOAA Fisheries). To our knowledge, there are no other forage fish data sets available for the five Conservation Zones in the NWFP area [Conservation Zones 1 – 5 as defined by the marbled murrelet recovery plan (USFWS 1997)]. For analysis of trends in the broader study area, we examined a set of readily available physical and biological ocean factors that have been used in similar multivariate analyses in other studies (e.g., Ainley and Hyrenbach 2010) that we felt might serve as proxies for murrelet prey. Sea surface temperature and Chlorophyll concentration are factors that affect marine productivity. Cooler waters are enriched with nutrients compared with warmer waters; Chlorophyll A concentration is related to primary productivity. Our assumption is that cooler waters with enriched Chlorophyll A should support higher prey biomass than warmer waters or waters with lower Chlorophyll A. We obtained data from a season corresponding with murrelet breeding activity, but also from the previous winter thinking that there could be time lags between those factors and their ultimate influence on space use by murrelets.

We then summarized these data at several spatial levels: state (Washington, Oregon, California); Conservation Zone 1-5; and stratum (conservation zones broken down by coastal areas as described in Raphael et al. 2007, n = 9, Figure 3-1). The analyses reported here used 2 strata in each conservation zone except for Zone 1 (Strait of Juan de Fuca, San Juan Islands, and Puget Sound). We treated Conservation Zone 1 as a single stratum, rather than using the three strata used in population sampling (Figure 3-1). We did this because the complex geography of coastal waters and potential nesting habitat in Conservation Zone 1 allows birds from multiple at-sea strata to access nesting habitat throughout the zone. By comparison, in all other zones, the geography was that of a roughly linear band of coastal waters matched with a roughly linear band of potential nesting habitat inland. To assess the correlation between amount of nesting habitat with sampling strata and adjacent murrelet population size, we first adjusted for stratum area by regressing amount of habitat and number of birds with total land area of each stratum. We then saved the residuals from these regressions and computed the correlation between residual murrelet abundance and residual habitat area.

In examining trends over time, we were able to obtain data on abundance of forage fish from two transects used to sample forage fish abundance (R. Emmett, personal communication), each in proximity to two of our murrelet sampling strata; one near the southern part of Conservation Zone 2 and one near the northern end of Conservation Zone 3 (Figure 3-1). We compared annual estimates of murrelet abundance in each of these two strata to annual estimates of fish abundance over the available years 2003 to 2011.

Before doing correlations we examined each factor for spatial and temporal autocorrelations. We found weak evidence of both spatial and temporal autocorrelation for some factors in at least one geographic area; therefore, p-values are not strictly interpretable. For that reason, we do not report p-values and describe the correlation coefficients as indicators of possible trends or indications of potential relationships.

**Multivariate model.**—We then developed a quantitative, multivariate model with a goal to assess concurrently the relative strength of marine and terrestrial factors in predicting the spatial and temporal abundance of marbled murrelets. To better understand the combined influences of marine and terrestrial factors on marbled murrelet distribution and population trends at sea, we constructed a set of multivariate models that simultaneously account for each marine and terrestrial factor as well as their interactions in space and time. For this analysis, we used boosted regression tree (BRT) models, generally following methods detailed in Raphael et al. (2015) but with three changes. First, the analysis scale was expanded from 5-km nearshore segments to a sample unit comprising an entire Primary Sample Unit (PSU). Each PSU consisted of an approximately 20-km long coastal segment with both nearshore and offshore sampling areas (chapter 1 of this volume describes the sampling design). In Raphael et al. (2015), the analysis was performed at the scale of 5-km nearshore segments because those authors were interested in fine-scale habitat associations and identification of “hotspots” of murrelet abundance along the coast. This current analysis more closely matches the design of population sampling reported in Chapter 1. Second, the analysis was extended to include data through 2012. Third, the environmental attribute for “distance to shore” was dropped because it is not relevant when the full PSU is used as the sample unit (environmental attributes are shown in Table 3-1). The sample size in this analysis totaled 1,099 PSU-year combinations; numbers of PSUs varied from 81 to 94 in any particular year depending on which PSUs were sampled each year.

We calculated all covariates annually from 2000-2012 for each PSU, except for Zones 1 and 2, where we used 2001-2012 because population data was not available for 2000. Covariates varied spatially (by PSU), temporally (by year), or both spatially and temporally (Table 3-1). Covariates were also associated with either marine foraging habitat suitability or terrestrial nesting habitat suitability (Table 3-1). While the habitat status and trend analysis (Chapter 2) focused on habitat conditions for two “bookend” years (1993 and 2012), as described below we estimated the amount of suitable nesting habitat in each of the years from 2000 (or 2001) to 2012.

The first two marine covariates in Table 3-1 were based on proximity to terrestrial features that may influence observed at-sea abundance of murrelets, presumably due to effects on foraging conditions. These included the distance (km) from the PSU center to the nearest major river (defined by a flow  $> 166 \text{ ft}^3 [4.7 \text{ m}^3]$  based on USGS Enhanced River Reach Data 2.0 from 2003), and the predominant shoreline type. Shorelines were classified based on the Environmental Sensitivity Index classification system (ESI; NOAA 2002), which categorizes shorelines into 21 major classes. We

simplified these into 11 classes and then calculated the majority shoreline type within each PSU boundary. This calculation resulted in 8 types represented in our study area (Table 3-2).

The next set of marine covariates in Table 3-1 was based on oceanographic conditions that may influence prey availability (primarily forage fish) and therefore murrelet abundance at sea. Because foraging conditions within each PSU are likely to be influenced by marine conditions at broader scales, we calculated the remaining marine covariates that vary spatially based on the mean or sum (depending on the covariate) of values within a 10-km moving window. We then extracted the mean values of the moving window result within each PSU (i.e., the mean of all moving window centers that fell within the PSU).

We obtained monthly mean sea surface temperature (SST) and Chlorophyll-A concentration (ChlorA) data from NASA's Earth Observations portal (2012). Data from 2000-2002 were collected by the SeaWiFS platform and data from 2003-2012 were collected from the MODIS Aqua platform ([http://aqua.nasa.gov/about/instrument\\_modis.php](http://aqua.nasa.gov/about/instrument_modis.php)). We then calculated the mean SST (°C) and ChlorA concentration (mg/m<sup>3</sup>) within 10 km of the PSU during two seasons, summer (values from May through July) and winter (values from Dec through Feb). All data were raster images with a resolution of 0.1 degrees latitude/longitude. We selected these two seasons to examine both the immediate breeding season and the pre-breeding season's influence on murrelet distribution and abundance; prey conditions both pre-breeding and later in the breeding season appear to be important for successful breeding by murrelets (Becker et al. 2007).

We quantified marine human footprint based on a raster model of human threats to marine ecosystems (Halpern et al. 2008), including commercial shipping, pollution, commercial and recreational fishing, climate change (ocean acidification, ultraviolet radiation, and changes in sea temperature), invasive species, and benthic structures. This covariate was calculated based on the mean value within 10 km of the PSU.

To quantify bathymetric influences on murrelet abundance, we used two approaches. First, we calculated the mean depth within 10 km of the PSU based on a 250-m digital elevation model (USGS). Second, based on the same bathymetric data, we summed the area (km<sup>2</sup>) of depths suitable for foraging within 10 km of the PSU, hereafter referred to as "foraging area." Suitable foraging depths were based on a threshold (< 25m deep, except for the San Juan Islands and northern Puget Sound, for which the threshold was < 40m); the thresholds were based on natural breaks observed in the plots of murrelet abundance versus depth.

The last four marine covariates were indices of broader Pacific Ocean conditions, including the Pacific Decadal Oscillation (PDO; Mantua et al. 1997) and the El Niño Southern Oscillation (ENSO; Trenberth 1997). The PDO index (JISAO 2012) is based on variation in North Pacific SST from 1900 to the present. The Oceanic Niño Index (ONI), a measure of the state of the ENSO, is based on variation in equatorial Pacific

SST (NOAA Climate Prediction Center 2012). Both indices are calculated on a monthly basis, which we then averaged for two seasons, summer (values from May through July, representing conditions during the central murrelet nesting season) and winter (values from December through February, representing the period preceding nesting).

To quantify the terrestrial habitat influences on at-sea marbled murrelet abundance, we calculated three covariates that quantified the amount and fragmentation of nesting habitat as well as degradation by human modification. Because murrelets can commute up to at least 80 km from foraging to nesting sites (Hébert and Golightly 2008, Nelson 1997, Raphael and Bloxton, unpublished data) we calculated each of these terrestrial covariates within an 80-km moving window. We then calculated the mean of the moving window result within each PSU (i.e., all moving window centers that fell within the PSU boundary). Although our main study area did not extend north of Washington State, in some areas of Washington the 80-km window included terrestrial habitat in British Columbia, Canada. We quantified terrestrial nesting habitat based on a marbled murrelet nesting habitat spatial model produced for the United States portion of our study area (Chapter 2). This model classified nesting habitat into four classes where classes 1 and 2 were lower suitability and classes 3 and 4 were higher suitability. For our analysis we converted the suitability map into a binary model and defined the combination of classes 3 and 4 as higher-suitability nesting habitat. For British Columbia terrestrial areas, we defined nesting habitat based on areas designated by the Ministry of Forests, Lands, and Natural Resource Operations as Old Growth Management areas (FLNRO 2012). Temporal variation in higher-suitability nesting habitat was represented by modeling habitat changes between 2000 and 2012. We did this by overlaying annual locations of forest disturbance (e.g., timber harvest, fire, windthrow) obtained from the LandTrendR files as described in Chapter 2. At each year we then reclassified any habitat within the disturbed area for that year from higher to lower suitability. We also calculated the fragmentation of nesting habitat using the patch cohesion metric in Fragstats (McGarigal et al. 2012) within an 80 km radius moving window. Finally, we calculated the mean terrestrial human footprint within an 80 km moving window based on a model of anthropogenic landscape modifications, including human habitation, roads, railroads, irrigation canals, power lines, linear feature densities, agricultural land, campgrounds, highway rest stops, landfills, oil and gas development, and human induced fires (Leu et al. 2008).

We used boosted regression trees with a Poisson loss function to explore the relationship between murrelet at-sea abundance and our suite of marine and terrestrial covariates. Boosted regression tree (BRT) is a machine learning approach combining regression trees with a boosting procedure that adds new trees to the model fit to the residuals of the prior trees (Elith et al. 2008). The BRT prediction is optimized based on two main parameters, the learning rate and tree complexity. The learning rate, also called shrinkage rate, determines the contribution of each new tree added to the model, while tree complexity determines the number of nodes per tree. Following recommendations of Elith et al. (2008), we used a learning rate of 0.01 and a tree complexity of 5 throughout our analysis. The optimal number of trees was selected based on training the model to one half of the data and then assessing the fit of the

model to the remaining half. When new trees began to reduce the fit of the model to the test data, no new trees were added. Final model parameters were derived from an ensemble of all trees weighted by the learning rate.

Because our sample units were contiguous 20-km segments of coastal waters, they might exhibit spatial autocorrelation; spatial autocorrelation occurs when the covariate values of two sample units is related to their distance apart. To account for spatial autocorrelation in the BRT model residuals, we calculated a residual autocovariate term (RAC), as in Crase et al. (2012), by plotting the residuals from the BRT model to raster grid cells representing each survey segment, calculating the mean residual within a 25 km moving window, and then extracting the moving window result for each PSU grid cell. We then refit the BRT model as before but including the RAC term.

We assessed variable importance based on the number of times a variable was used for splitting weighted by the squared loss of deviance due to each split, averaged over all trees; deviance measures the loss in predictive performance due to a suboptimal model, thus reducing deviance represents improved model fit. The result was scaled such that the sum of all variable importance scores added to 1, allowing them to be interpreted as percent contributions to the final model. We assessed model performance using a ten-fold crossvalidation procedure which involved training the model on ten random subsets (90%) of the full data and then evaluating the model predictions against the portion of the data withheld (10%) from the model. All BRT models were fit in R (version 3.0, R Core Team 2012) using the “dismo” package for species distribution modeling (Hijmans et al. 2012).

We ran three models. The first was based on the entire set of PSUs covering Zones 1 – 5. Because marine productivity is driven by tides and freshwater inputs in Zone 1 and by coastal upwelling in the other four zones, we also ran separate boosted regression models for the 30 PSUs in Zone 1 and the remaining 51 to 64 PSUs (depending on year) in Zones 2-5. Thus, the sample size ranged from a low of 30 PSUs for the Zone 1 model, to 94 for the model including all 5 zones.

## Results

**Spatial correlations.**— As previously reported by Raphael (2006) and Raphael et al. (2011), we found that nearshore abundance of murrelets within the 9 geographic sampling strata is correlated with the amount of higher-suitability nesting habitat (after accounting for land area) in the adjacent terrestrial environment (partial  $r = 0.57$ ,  $r^2 = 0.324$ , Figure 3-2). If we focus on the highest habitat suitability class, (Class 4 as defined in chapter 2), we also found a positive correlation with murrelet abundance, but that correlation was weaker than that for all higher-suitability habitat ( $r^2 = 0.137$ , Figure 3-2). In both cases, there is considerable unexplained variation. We note that stratum 3.2 (central Oregon coast) has a much greater abundance of murrelets relative to amount of adjacent nesting habitat, and we see that stratum 2.2 (southern Washington coast) has a much lower abundance of murrelets (Figure 3-2). Other factors, perhaps marine conditions or geographic variation in the relationship between murrelet numbers and amount of nesting habitat, could contribute to these unexplained sources of

variation. Cohesion (an index of habitat pattern where higher values indicate more contiguous and less fragmented habitat) is strongly and positively correlated with murrelet abundance within strata (Figure 3-3).

As noted above, we would have preferred to assess the influence of spatial variation in abundance of murrelet primary prey, forage fish, on murrelet populations. However, such data do not exist for our entire study area. Instead, we examined a set of ocean indicators that might serve as proxies to murrelet prey. We also examined variation in murrelet abundance in relation to dominant ESI shoreline substrate (Table 3-2, Figure 3-4). Murrelet abundance was greater offshore of fine- to medium-grained sand beaches (substrate 3) and was also greater offshore of estuaries and marshes (substrate 10) when compared to other substrates. Correlations of murrelet abundance with sea surface temperature and concentration of Chlorophyll A were weak at the state scale and also at the stratum scale (summer Chlorophyll,  $r = -0.08$ ; winter Chlorophyll,  $r = 0.46$ ; summer SST,  $r = 0.06$ ; winter SST,  $r = -0.46$ ; forage area,  $r = -0.28$ ).

**Temporal correlations.**— As we report in Chapter 1, murrelet populations are declining in Washington, stable in Oregon and stable in California, where there is a non-significant but positive population trend slope. In Chapter 2, we report that relative change in amount of higher-suitability nesting habitat from 1993 to 2012 follows the same ranking with the greatest decline in Washington, intermediate amount of decline in Oregon, and the smallest decline in California. At the scale of geographic strata, we found a weak positive correlation between change in numbers of murrelets from 2000 to 2012 and loss of higher-suitability nesting habitat from 1993 to 2012 (Figure 3-5,  $r^2 = 0.206$ ) indicating that when nesting habitat decreases so do estimates of murrelet population abundance. We note however that there is much unexplained variation in this relationship. For example, stratum 1.0 (Puget Sound) and 2.2 (southern Washington coast) have similar habitat losses but very different murrelet losses (Figure 3-5). We found a stronger relationship if we restrict the comparison to the highest suitability habitat (Class 4,  $r^2 = 0.52$ ). These results, although weak, are consistent with our spatial results showing a correlation between murrelet abundance and amount of habitat in each stratum. We note that at the zone scale, the strongest correlation we observed between changes in murrelet abundance and amount of higher-suitability nesting habitat was in Zone 2 ( $r_s = -0.915$ ), the zone where murrelet abundance has declined at the greatest rate (Table 3-3; see Chapter 1 of this volume). We emphasize, again, that these correlations do not necessarily establish cause-effect relationships but they do support the hypothesis that nesting habitat may be the factor limiting population stabilization and recovery.

To examine possible correlations between temporal change in marine factors and murrelet abundance, we summarized data within each conservation zone and then calculated Spearman rank correlations within each conservation zone (Table 3-3). We observed evidence of negative correlations (Table 3-3) of murrelet abundance with summer Chlorophyll ( $r_s = -0.608$ ) and winter Chlorophyll ( $r_s = -0.600$ ) in Zone 1 (Puget Sound). We found positive correlations with summer Chlorophyll ( $r_s = 0.588$ ) and with summer SST ( $r_s = 0.576$ ) in Zone 2 (outer coast of Washington). In Zone 3 (Oregon-

from the Washington border south to Coos Bay), we observed a negative correlation with summer Chlorophyll ( $r_s = -0.478$ ). We found no correlations in zones 4 and 5.

We found little influence of our indirect measures of ocean productivity on murrelet distribution and abundance. However, for the two localities (Figure 3-6) where we have time series for both forage fish and murrelet data, there is a potential positive relationship (for stratum 2.2,  $r^2 = 0.64$ ; for stratum 3.1,  $r^2 = 0.44$ ; Figure 3-7). This potential relationship would suggest that our indirect measures do a poor job of predicting forage fish abundance and distribution, but this possibility needs additional investigation.

We also explored the relationship of El Niño Southern Oscillation (ENSO) and Pacific Decadal Oscillation (PDO) phases to sea surface temperature, Chlorophyll A concentration (as a measure of primary productivity), forage fish abundance, and marbled murrelet at-sea abundance in two areas (near the mouth of Willapa Bay, Washington and just south of the Columbia River mouth in Oregon, Figure 3-6) where forage fish data were available during the years of our surveys. Since 2000, when at-sea surveys began, there has been one strong El Niño (2009-10) and two strong La Niña events (2007-08 and 2010-11), most apparent during the winter (Figure 3-8). During this same period, the PDO transitioned from a warm phase prior to 2007, to a cool phase after 2007. Thus, the PDO and ENSO phases were both cool during the La Niña events in 2007-08 and 2010-11, but opposed during the 2009-10 El Niño. Surprisingly, the expectation of cooler sea surface temperature, higher primary productivity, and greater forage fish abundance during these cool phase years was not met in our two forage fish data locations. Both of these stations were located near major estuaries and the Columbia River plume, so ocean conditions and forage fish abundance may have been influenced by inputs (e.g., sediments and nutrients) from these estuarine systems (Zamon et al. 2014).

**Multivariate model.**—The boosted regression models performed very well, explaining 91.4, 95.3, and 88.9 percent of deviance and 77.1, 78.8, and 74.7 percent for cross-validated samples in the full (all Zones) model, Zone 1 model, and Zones 2-5 model, respectively. For the full model, cohesion and area of higher-suitability nesting habitat had the strongest influence, followed by the index of terrestrial human footprint, spatial autocorrelation (RAC) and ESI shore substrate type (Figure 3-9). In Zone 1, area of higher-suitability nesting habitat was by far the strongest contributor. The next highest contributor was the marine human footprint, which could reflect more intense vessel traffic and fishing pressure in that zone compared to the outer Pacific Northwest coast where the influence of that covariate was much less important. In Zones 2-5, habitat cohesion had the strongest influence, followed by terrestrial human footprint, amount of nesting habitat, spatial autocorrelation term (RAC), and ESI shoreline substrate (Figure 3-9). The remaining covariates, all marine, made relatively small contributions (less than 5 percent), in all three models, just as reported by Raphael et al. (2015). Figure 3-10 shows the shape of relationships between the top-ranked covariates and the fitted function (which is based on abundance of murrelets at each sample unit each year).

Results follow and reconfirm the general patterns we observed from our univariate correlations.

## Discussion

**Spatial variation.**—The univariate correlations we observed between murrelet abundance and terrestrial or marine factors suggest that terrestrial factors had somewhat stronger relationships, but that marine factors might be responsible for some of the unexplained variation. Although the correlations we observed do not establish a cause-effect relationship, they suggest that proximity of higher-suitability nesting habitat influences the number of murrelets that occur in the ocean adjacent to that habitat during the breeding season. If true, the amount and pattern of higher-suitability nesting habitat might set the carrying capacity for murrelets and that, in turn, would suggest that management focused on conserving and restoring murrelet nesting habitat will contribute to murrelet conservation.

We conclude from these correlations that, among individual variables, the amount and pattern of higher-suitability nesting habitat seems to have the strongest influence on murrelet abundance at the sampling stratum scale. This information fits with our knowledge about the ecology of the murrelet and the need for nesting birds to remain near nesting habitat (Burger and Waterhouse 2009). With the exception of marine human footprint, none of the marine variables we were able to measure appear to be strongly correlated with murrelet abundance. We caution though, that these results do not mean that the marine environment is not important to murrelets. The fact that our set of marine variables do not correlate with spatial trends in murrelet abundance could be the result of a lack of relationship or because our chosen set of marine variables or their scaling poorly represent aspects of the marine environment important to murrelets.

**Temporal variation.**—We found that annual variation in murrelet numbers is more strongly correlated with trend in amount of nesting habitat than trend in ocean indicators. The lack of a consistent relationship between ocean factors and murrelet populations is somewhat surprising. We know that forage fish are the primary food resource for marbled murrelets, particularly during the breeding season (Nelson 1997). Other research suggests that murrelets change their foraging patterns in response to oceanographic changes (Peery et al. 2009). The fact that we did not see a relationship between ocean productivity and murrelet populations, may be caused, in part, by our use of indirect measures of productivity (Chlorophyll A and SST) as surrogates of murrelet prey populations in our analyses.

For the limited portion of our study where temporal trend in forage fish abundance was available, we did observe a positive relationship between fish abundance and murrelet density over time. This potential relationship with forage fish is not surprising given the often strong positive relationship between forage fish abundance and the abundance of fish-eating birds (e.g., Durant et al. 2009, Furness and Tasker 2000). Murrelets likely evolved in an environment that experienced considerable environmental variability that, in turn, led to fluctuations in prey abundance and distribution. However, chronic food

scarcity can compromise long-term breeding success (Cury et al. 2011) and can also reduce adult survival in seabirds (Kitaysky et al. 2010). Long-term changes in survival and fecundity would lead to changes in the murrelet population trajectory. There is some information indicating long-term decline in murrelet diet quality in portions of its range (Becker and Beissinger 2006, Norris et al. 2007, Gutowsky et al. 2009), and effects of prey changes on murrelet reproductive success (Becker et al. 2007). Consequently, because of this potential relationship between murrelets and their prey, we recommend attempting to gain a better understanding of the relationship between critical murrelet prey resources and murrelet populations.

Our results indicate that sea surface temperature and Chlorophyll A concentration may influence year-to-year changes in murrelet abundance but in a more complex manner than our preliminary correlations could detect, at least in zones 1, 2, and 3. The signs of marine correlations did not match our expectations in zones 1 and 3. We expected a positive correlation with Chlorophyll concentration and murrelet abundance because we hypothesized that higher concentration of Chlorophyll would indicate higher productivity and hence better foraging conditions and greater numbers of murrelets. Instead, we found negative correlations in these two zones. Marine productivity in Zone 1 may be driven more by tidal flow, currents, freshwater inputs, and estuaries than upwelling, which may explain the unexpected results in that zone. Zone 3, however, resides in an upwelling system (part of the California Current System) and we are unable to explain the negative correlation in that zone. It is probable that our scale or time-frame in these correlation analyses did not capture effects of the marine productivity parameters on murrelet abundance in Zone 3. In addition, forage fish are likely to move somewhat independently of Chlorophyll A distributions, so this isn't necessarily surprising given that marbled murrelets are really using habitat that is quite inshore compared to where the majority of upwelling effects are measured (e.g., upwelling-driving Chlorophyll A dynamics close to shore where most murrelets occur may not have been accurately captured by the Chlorophyll and SST covariates).

Population responses by marbled murrelets to either marine or terrestrial factors are confounded by the murrelet's long life span, with an estimated generation time of 10 years (Burger 2002, McShane et al. 2004), and low annual reproductive output. Consequently, there is likely to be a lag between seasonal changes in marine factors (unless they are very dramatic changes reducing murrelet survival) and changes in murrelet populations. Interestingly, there seems to be a decline in murrelet abundance and distribution in response to contemporaneous loss of nesting habitat. One would expect a long-lived species like the murrelet to exhibit a lag in population response to declining habitat; birds would persist in the marine environment until they eventually die. Strong (2003), for example, suggested that declines observed in the 1990s along the Oregon coast might have resulted from losses of habitat stemming from logging in the 1980s or earlier. It is possible that murrelets might move out of an area adjacent to nesting habitat once that habitat is lost and that could explain the relationships we observed, but we have no direct evidence to support that possibility.

Despite the lack of a strong link between the PDO/ENSO phase and sea surface temperature, Chlorophyll A, or forage fish abundance, we still observed a strong correlation between forage fish and marbled murrelet abundance in these same locations (Figure 3-7). Moreover, the decrease in murrelet and forage fish abundance prior to 2007-08 followed by an increase after 2007-08 corresponds to a shift from a warm to cool phase of the PDO. This correlation is consistent with the observation that abundance of other diving seabirds in the California Current System is sensitive to the PDO phase (Ainley and Hyrenbach 2010).

**Multivariate model.**—When we considered the combined influence of both marine and terrestrial influences on the spatial and temporal abundance of murrelets, amount and pattern of higher-suitability nesting habitat seemed to have the greatest contribution in explaining variation in murrelet abundance within our study area. We found, however, that human influences, both marine (Zone 1) and terrestrial (Zones 2-5) were also important. These results reinforce the idea that forest habitat features are limiting factors in murrelet abundance and recovery, but this hypothesis will require further investigation to establish cause-effect relationships.

**Effects of Climate.**—Murrelet nesting habitat and foraging success along the Pacific coast are sensitive to climate variability (Becker et al. 2007), and climate may be contributing to the trends we have observed in murrelet abundance. The trend towards warmer, drier summers along the Pacific coast has favored increased fire frequency and intensity (Littell et al. 2009). This change may be contributing to nesting habitat loss by fire. While timber harvest was the leading cause of nesting habitat loss on nonfederal lands and all lands combined, over 60% of the habitat losses on federal lands range-wide from 1993 to 2012 were due to wildfire (due mostly to one fire event, the 2002 Biscuit fire in southern Oregon; Table 3-12 in Chapter 2). Drier summers also reduce epiphyte growth on branches, thereby degrading the suitability of platforms for nesting (Malt and Lank 2007). During winter, the trend towards increased winter precipitation and more severe storm events has increased the frequency of flooding, landslides, and windthrow (Dale et al. 2001). Warmer winter temperatures and drought stress have also increased the prevalence of tree insect and disease outbreaks. Together, climate-influenced factors (wildfire, insects/disease, and natural disturbances) contributed to the range-wide loss of nearly 27,000 acres of higher-suitability nesting habitat between 1993 and 2012, compared to losses of less than 8,000 acres from timber harvest on federal lands (Table 3-12 in Chapter 2). Climate change may already be decreasing the quality and quantity of marbled murrelet nesting habitat, and projections for continuation or even acceleration of current climate trends raises the potential for even greater impacts in the future.

In addition to influencing the quality and abundance of nesting habitat, climate variability also has a profound influence on the foraging success of seabirds, including the marbled murrelet (Grémillet and Boulinier 2009). The California Current System, encompassing the entire US Pacific coast (including zones 2-5 but excluding inland marine waters such as zone 1), has experienced a warming trend over the past 50 years. This warming is driving a shift from cool productive sub-arctic ocean conditions

toward a warm subtropical marine environment that is less productive (Di Lorenzo et al. 2005). On top of this long-term warming trend, two other sources of climate variability in the Pacific Ocean exert a strong influence on the productivity of coastal waters where the murrelet forages. The El Niño Southern Oscillation (ENSO) is a pattern of sea surface temperature anomalies in the equatorial Pacific that occurs over shorter time scales (every 3-7 years and lasting 9-18 months (Mestas-Nunez and Miller 2006). During the warm phase (El Niño), upwelling is weaker and sea surface temperature is warmer in the California Current System. Conversely, the cool phase (La Niña) is associated with greater upwelling and cooler sea surface temperature. These shifts in the intensity of upwelling influence nutrient availability in coastal waters, and therefore ENSO phases have profound effects on primary productivity depending on their intensity, with cascading effects at higher trophic levels (Thayer and Sydeman 2007). El Niño events have been associated with poor seabird survival and recruitment throughout the eastern Pacific (Hodder and Graybill 1985, Bertram et al. 2005). At longer time scales (15-30 years) the Pacific Decadal Oscillation (PDO) reflects a pattern of sea surface temperature anomalies in the north Pacific (Mestas-Nunez and Miller 2006). Similar to the ENSO cycle, the warm (positive) phase of the PDO results in weak upwelling in the California Current System and warmer sea surface temperature, while a negative PDO drives strong upwelling and cool sea surface temperature (Mantua et al. 1997). The effects of the ENSO and PDO cycles on upwelling, sea surface temperature, and primary productivity are additive. However, some species respond more strongly to either the ENSO or PDO phases, but not both (Sydeman et al. 2009, Black et al. 2011). In addition, the local effect of these broad regional ocean trends is highly modified by undersea topography (which affects the strength and pattern of upwelling), complex trophic interactions, species migrations tracking suitable water temperatures and prey, and food web impacts from commercial fisheries harvest (Doney et al. 2012).

### **Management Implications**

Our finding that amount and distribution of higher-suitability nesting habitat are the primary factors influencing abundance and trend of murrelet populations suggests that land managers should focus particular attention on forest practices that will conserve and restore that habitat. Maintaining the system of late-successional reserves continues to be critical to conservation and restoration of marbled murrelet nesting habitat on federal lands. But even on those reserved lands there are risks to murrelet habitat. Fire and other natural disturbances are already the main cause of nesting habitat loss on federal lands. As described above, climate changes will likely result in loss of existing murrelet habitat due to increased fire frequency and severity along with increased severity of storms resulting in increased windthrow. During the first 20 years of the NWFP, we documented fire as the main cause of nesting habitat loss on federal lands. Given this finding, predictions of increased fire risk in the future, and the value of higher-suitability nesting habitat, which takes a long time to replace, management plans may want to prioritize protection of nesting habitat in reserves, including NWFP late-successional reserves, wilderness, and National Parks.

In Chapter 2 of this volume, we found that a relatively high proportion (typically two-thirds or more) of suitable nesting habitat occurs as small patches (lacking interior forest conditions that are more than 90 m from a patch edge) or as edges of larger habitat patches. In this chapter, we found that nesting habitat cohesion, which is the inverse of habitat fragmentation, is a strong predictor of murrelet abundance and trends. This result is not surprising because murrelets prefer larger patches, which also tend to have fewer nest predators (Malt and Lank 2007, Raphael et al. 2002). A key feature of the NWFP is a network of late-successional reserves that have the management objective of protecting and enhancing late-successional forest ecosystems, which serve as habitat for late-successional forest species, including the murrelet. These reserves contain both older and younger forests, and over time, as more mature habitat develops around existing older forest in reserves, patch size should increase, and fragmentation and the prevalence of edges should decrease within reserves. However, it can take many decades for murrelet nesting habitat to develop, and in the short-term, protection of existing habitat will continue to be critical to minimize habitat losses, both within and outside of late-successional reserves.

Near-term murrelet conservation should also consider habitat loss caused by windthrow. Windthrow is a natural phenomenon and an important process in coastal forests of the Pacific Northwest, but it can be highly influenced by human activities. Clearcut or heavy thinning harvests can increase the amount of windthrow on the landscape dramatically. This effect depends on complex interactions between biotic (e.g., forest age and condition) and abiotic (e.g., slope and aspect) factors operating at different spatial and temporal scales (Sinton et al. 2000). Portions of forests can also be lost to windthrow after lighter thinning, but the magnitude of the effect depends on factors including topography and tree height-to-diameter ratios (Harrington et al. 2005, Roberts et al. 2007, Wilson and Puettmann 2007). Thus, thinning operations may accelerate the creation of forest conditions suitable to murrelet nesting in the long term (e.g., Maguire et al. 1994), but have short term negative impacts to murrelets to consider in management decisions (McShane et al. 2004).

Forest practices, natural forest disturbance and the interaction between these factors can increase the amount of forest edge. Increased edge resulting from forest fragmentation appears to have negative effects on murrelets. Malt and Lank (2007) found that murrelet nest sites at timber harvest edges had lower moss abundance than interior and natural edge nests sites (stream corridors and avalanche chutes) due to stronger winds, higher temperature variability, and lower moisture retention. Moss is an important nest substrate on large branches for murrelets in much of the NWFP area, therefore management actions adjacent to suitable murrelet nesting habitat can have implications for murrelets. Another negative impact to murrelets associated with edges, especially those that occur between clearcuts or large openings and forests, is increased nest depredation rates (Masselink 2001, Marzluff et al. 2004, Marzluff and Neatherlin 2006). This is especially true when edges are near human development such as campgrounds (Marzluff and Neatherlin 2006) or include berry producing plants such as elderberry (*Sambucus* sp.; Masselink 2001).

One conservation measure that is commonly used to minimize negative effects of forest edges is to provide forested buffers (USFWS 1997). The murrelet recovery plan includes as a short-term recovery action maintaining and enhancing buffer habitat around occupied nesting habitat, and suggests minimum buffer widths of 300-600 feet in this situation (USFWS 1997). Buffers around suitable nesting habitat (whether determined to be occupied or not) would help reduce fragmentation, risk of windthrow loss, and potentially reduce nest predation risk (USFWS 1997). Buffers are particularly important in the near-term while larger blocks of habitat develop on reserved lands. The details of such buffers are beyond the scope of this report. However, if not already accomplished, development and implementation of forest management practices that protect (short-term) and develop (long-term) suitable murrelet nesting habitat on NWFP lands within the murrelet range would be beneficial. For such practices, minimizing short term impacts, such as by avoiding harvest of suitable nesting habitat, providing buffers round suitable nesting habitat to minimize edge effects of management actions (such as from thinning or clearcuts), and minimizing fragmentation of suitable habitat, will likely improve the status of this threatened species.

As described in Chapter 2, a substantial amount of suitable nesting habitat occurs on state and private lands. The loss of habitat on those lands is occurring at a much more rapid rate than on Federal lands. Because of the strong relationship between murrelet populations and nesting habitat and because recovery of murrelet populations will likely require contributions of nesting habitat on state and private lands, at least in the short-term (as discussed in the murrelet recovery plan), there is a need for incentives for private forest landowners to avoid fragmentation and loss of high quality nesting habitat and to maintain blocks of interior nesting habitat on the landscape as well as buffers adjacent to suitable habitat on federal and state lands.

Another conservation measure worthy of additional investigation is the management of potential nest predators. In central California (outside the NWFP area), an aggressive program to deter nest predation has been enacted to try and improve marbled murrelet nesting success (Henry and Peery 2010). Similar programs in fragmented or in suitable nesting habitat near human populations or campgrounds may prove beneficial. We emphasize that this is a conservation measure that we have little understanding of as to its population-level effectiveness. Consequently, predator management actions should be conducted in a research framework so that their effectiveness can be evaluated.

### **Summary conclusions**

Our most prominent finding in combining results of Chapters 1 and 2 into multivariate models is the strong relationship between stands of cohesive and higher-suitability nesting habitat and the distribution and trends of murrelets at sea. Areas of greatest habitat loss over the time series tended to be the areas with greatest murrelet decline. Habitat loss occurred in all zones, with greater loss documented in Washington, where the steepest declines in murrelet populations occurred. Though we found no prominent effects of marine factors on murrelet distribution and population trend in this analysis, our analyses were limited by the unavailability of direct measures of murrelet prey

abundance, and selection of appropriate marine parameters and temporal frames remains a complex problem that warrants further investigation. We did a preliminary analysis on a number of factors that did not appear to have direct links to murrelet abundance, but we did find what appears to be a promising relationship between temporal trends in murrelet abundance with forage fish abundance, and that this may indicate an important marine causal factor in nesting success that remains to be quantified (e.g., Ainley et al. 2015). Incorporating fish sampling with ongoing population surveys at sea would be a good next step to further explore this relationship.

Several points bear repeating: (1) loss of higher-suitability habitat has been relatively low on Federal land compared to non-federal land since creation of the Northwest Forest Plan; (2) marbled murrelet declines are not related to the small loss of higher suitability habitat on Federal lands, but could be related to the lack of buffers and heavy thinning adjacent to murrelet habitat in the LSRs; and (3) there appears to be a strong relationship between murrelet population declines and the large loss of higher suitability habitat on non-federal land, especially in Zone 2.

Among the factors we investigated, nesting habitat factors were the best predictors of marbled murrelet population distribution and trends at sea. However, there was unexplained variance which implies that habitat monitoring alone is not sufficient at this time for predicting murrelet population trends. Also, given the breadth of forest types within the study area, the relationship between murrelets numbers and nesting habitat conditions may vary geographically. Therefore, population monitoring continues to be an essential element of measuring the effectiveness of murrelet conservation and restoration efforts, while the monitoring program continues to evaluate predictive models relating murrelet population distribution and trends to terrestrial nesting habitat and marine conditions. The primary hypothesis emerging from our work is that marbled murrelet distribution and trend in the breeding season is largely determined by amount and trend of suitable nesting habitat. If this is true, and if amounts of habitat increase in the future as currently unsuitable habitat matures within federal reserves, then we should see a concomitant increase in murrelet population size. This will be an important test of our hypothesis and continued monitoring is needed to complete this test.

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**Table 3-1.** Description, abbreviation, variability (spatially, temporal, or both spatial and temporal) and habitat component (marine or terrestrial) represented for each covariate evaluated.

<b>Covariate</b>	<b>Abbreviation</b>	<b>Variability</b>	<b>Habitat</b>
Distance to major river (m)	DistToMajorRiver	Spatial	Marine
ESI shoreline substrate type	ESI ShoreType	Spatial	Marine
Chlorophyll A summer (May-July) (mg/m <sup>3</sup> )	ChlorA_summer	Spatial & Temporal	Marine
Chlorophyll A winter (Dec-Feb) (mg/m <sup>3</sup> )	ChlorA_winter	Spatial & Temporal	Marine
Sea surface temperature summer (May-July) (C)	SST_summer	Spatial & Temporal	Marine
Sea surface temperature winter (Dec-Feb) (C)	SST_winter	Spatial & Temporal	Marine
Marine human footprint	MarHumanFoot	Spatial	Marine
Depth (m)	Depth	Spatial	Marine
Foraging area (km <sup>2</sup> )	ForagingArea	Spatial	Marine

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Oceanographic Nino index summer (May-July)	ONI_summer	Temporal	Marine
Oceanographic Nino index winter (Dec- Feb)	ONI_winter	Temporal	Marine
Pacific Decadal Oscillation index summer (May-July)	PDO_summer	Temporal	Marine
Pacific Decadal Oscillation index winter (Dec-Feb)	PDO_winter	Temporal	Marine
Nesting habitat area	NestingHabitat	Spatial & Temporal	Terrestrial
Nesting habitat cohesion	NestHabitatCohesion	Spatial & Temporal	Terrestrial
Terrestrial human footprint	TerrHumanFootprint	Spatial	Terrestrial

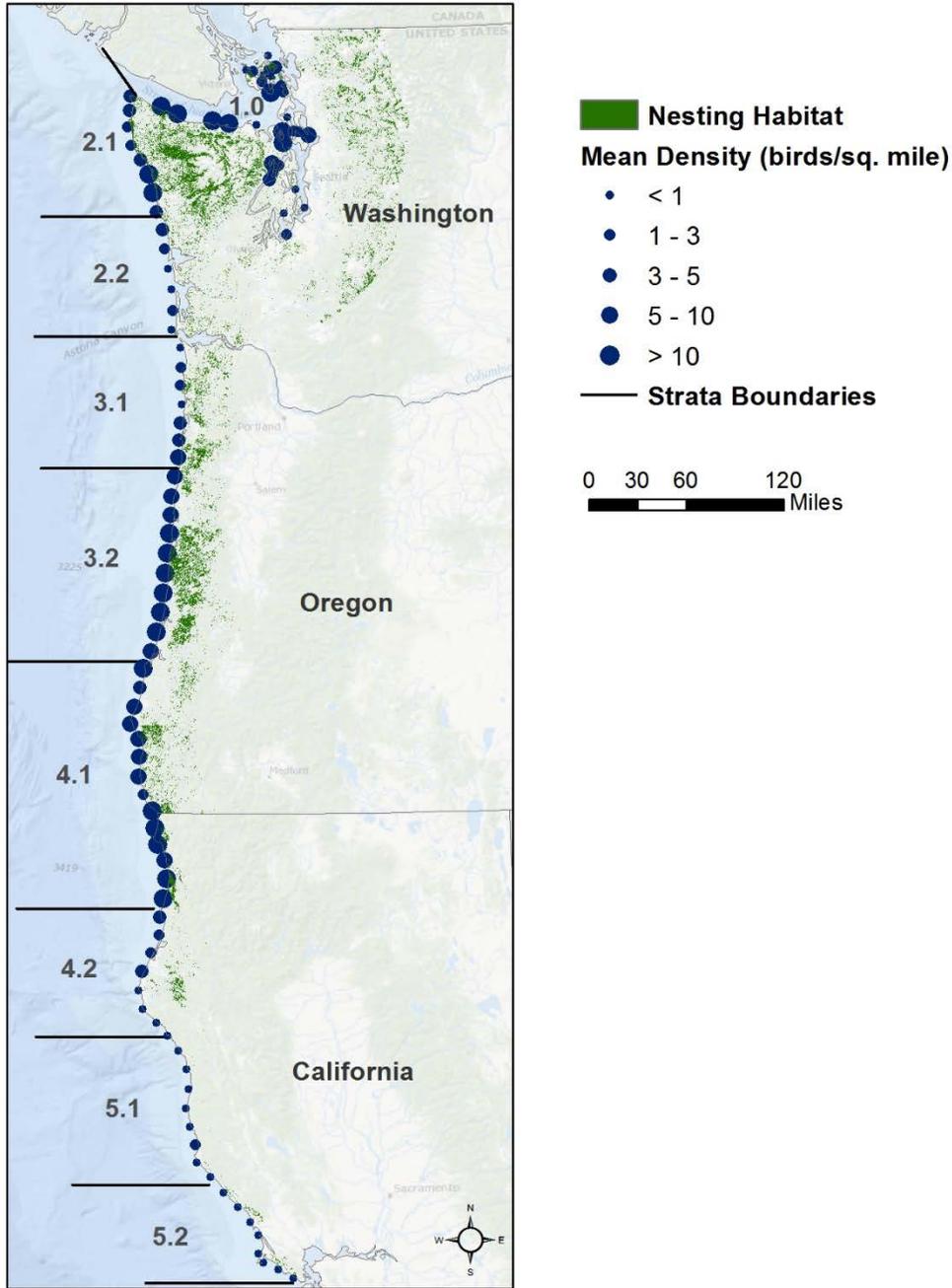
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**Table 3-2.** Dominant ESI shoreline substrate types (NOAA 2002) within murrelet sample units (n = 95).

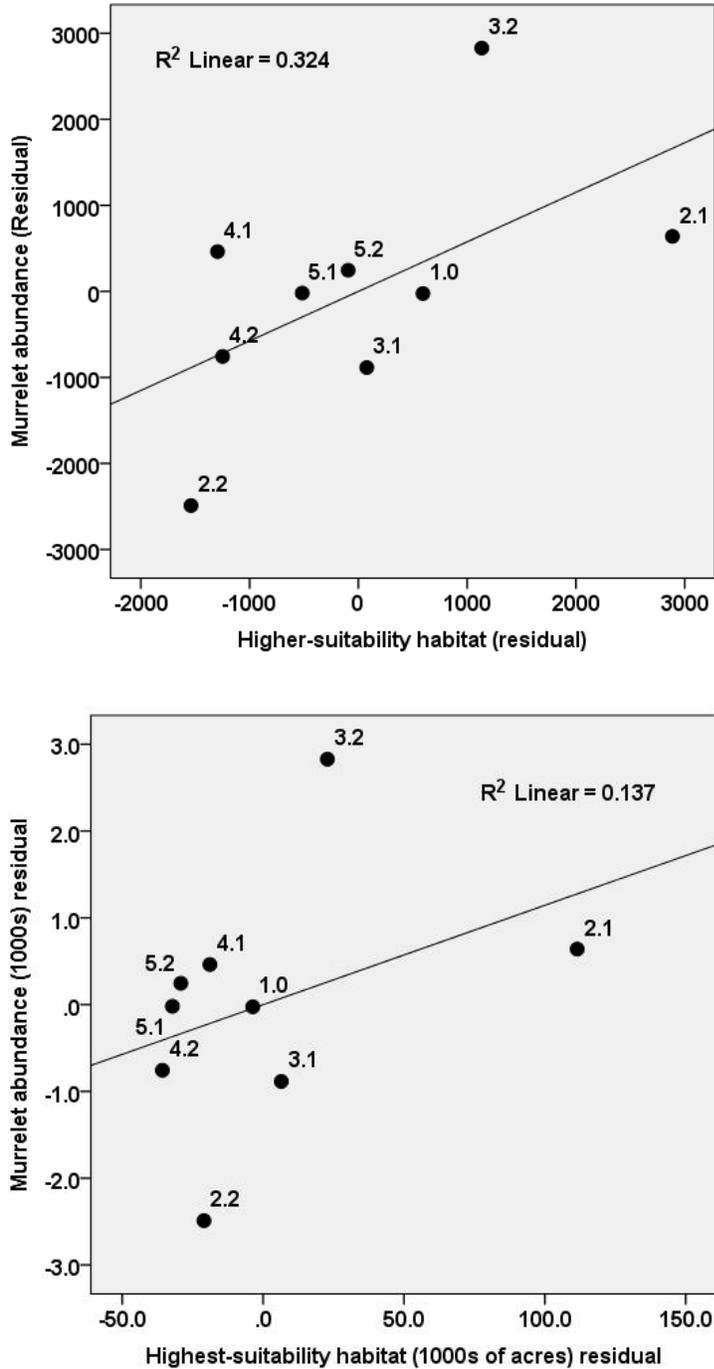
Description	Percent occurrence
Exposed rocky shores	5.3
Exposed scarps and wave-cut platforms in bedrock, mud or clay	21.1
Fine- to medium-grained sand beaches	26.3
Mixed sand and gravel beaches	3.2
Gravel beaches	34.7
Exposed tidal flats	2.1
Sheltered tidal flats and vegetated low bands	3.2
Estuaries, marshes, swamps, and wetlands	4.2

**Table 3-3.** Spearman rank correlations between abundance of marbled murrelets and covariates describing nesting habitat and ocean conditions over time (2000 to 2012).

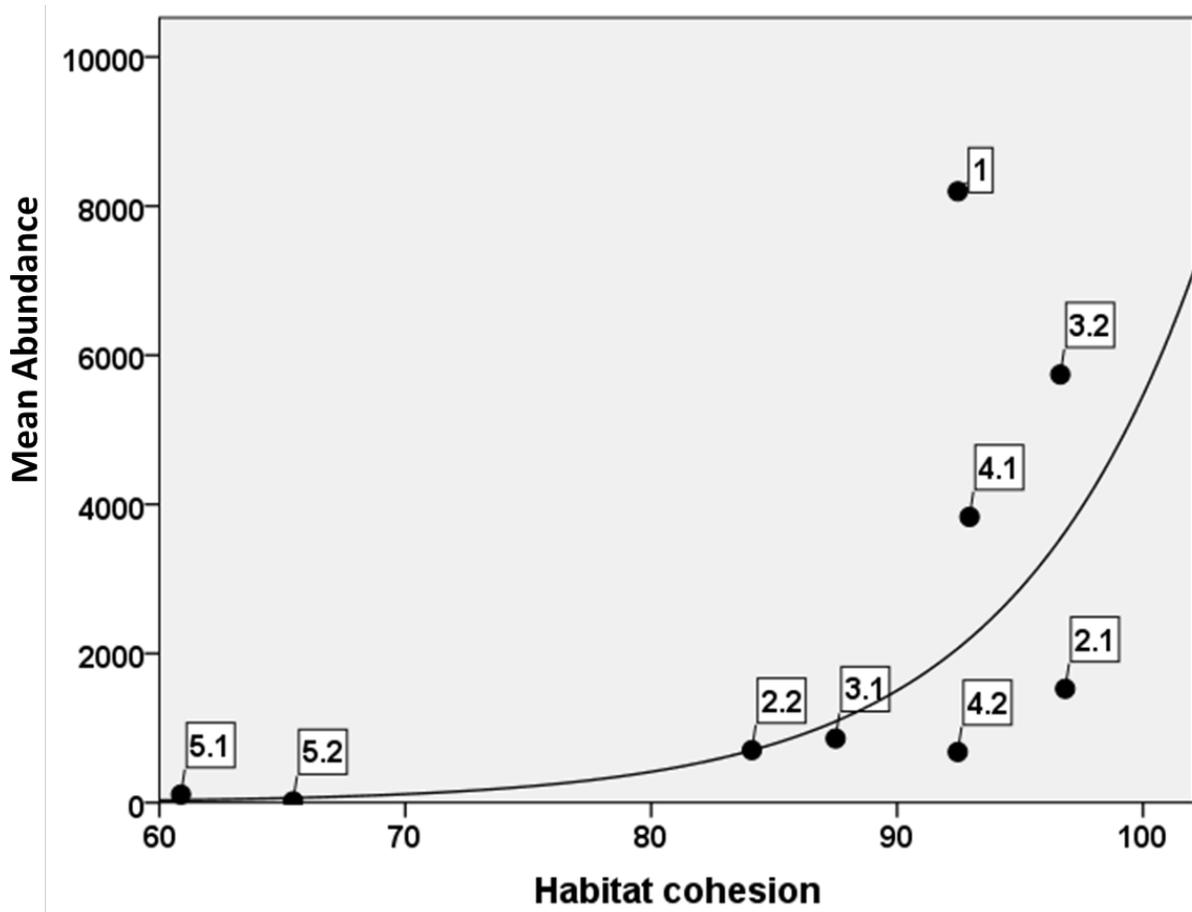
Covariate	Statistic	Zone 1	Zone 2	Zone 3	Zone 4	Zone 5
Nesthab_area_mean	Correlation Coefficient	0.350	0.915	0.071	0.082	-
	N (years)	12	10	13	13	9
ChlorA_summer_mean	Correlation Coefficient	-0.608	0.588	-	0.126	-
	N (years)	12	10	13	13	9
ChlorA_winter_mean	Correlation Coefficient	-0.600	-0.321	-	0.005	0.117
	N (years)	10	10	13	13	9
SST_summer_mean	Correlation Coefficient	0.189	0.576	-	-	0.050
	N (years)	12	10	13	12	9
SST_winter_mean	Correlation Coefficient	0.469	0.370	0.077	-	0.050
	N (years)	12	10	13	13	9



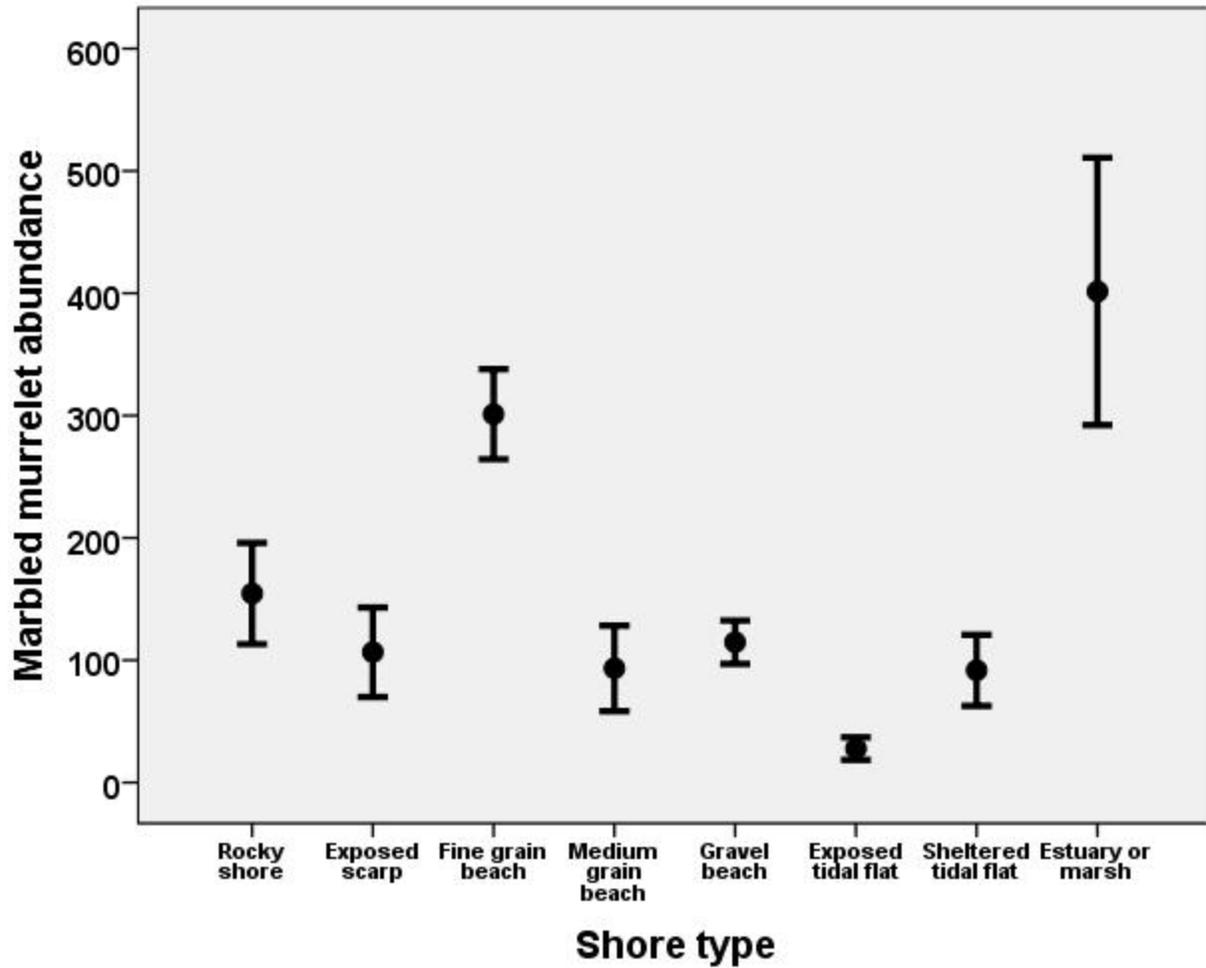
**Figure 3-1.** Locations of sampling strata within each Conservation Zone. For this analysis, the three strata within Zone 1 (1.1, 1.2, and 1.3) were lumped and labeled as stratum 1.0. The size of the blue dots corresponds to mean density of murrelets (2000-2012 for Oregon and California, 2001-2012 for Washington) in each 20-km long Primary Sample Unit.



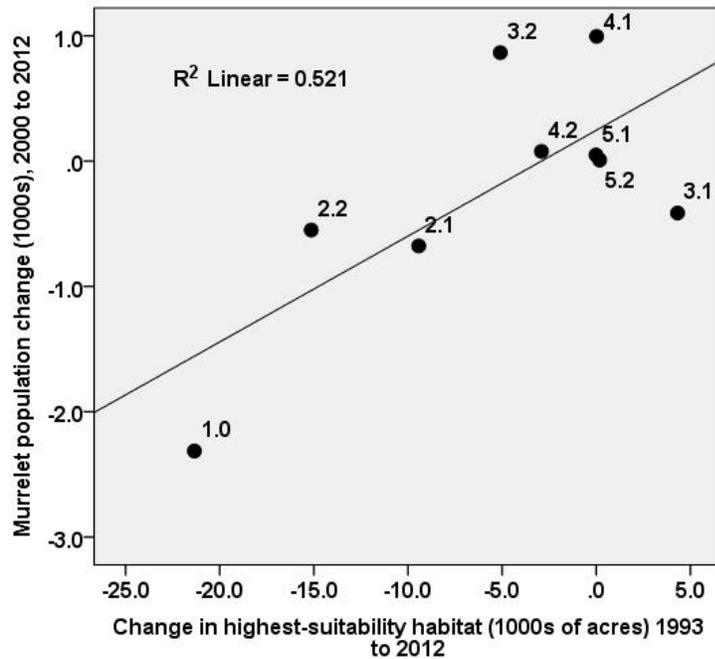
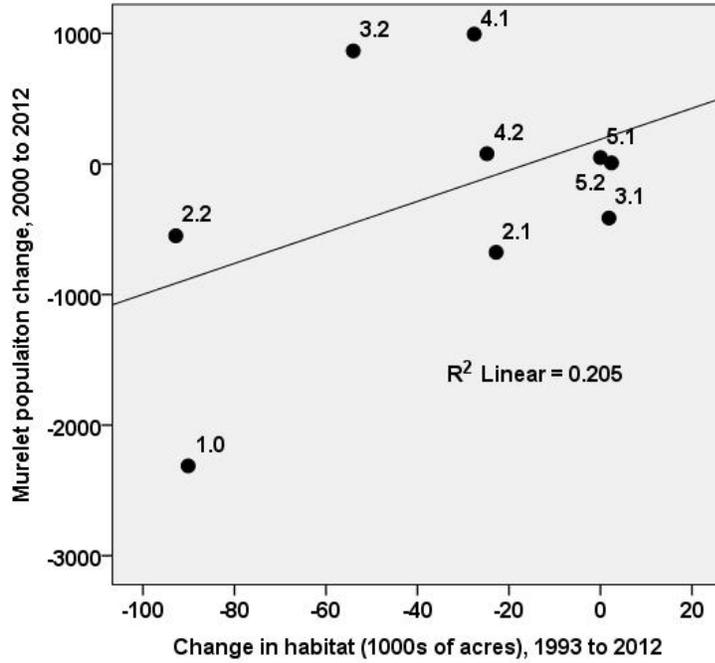
**Figure 3-2.** Relationship between residual mean abundance of marbled murrelets sampled at sea and residual amount of higher-suitability nesting habitat (suitability classes 3 plus 4; top figure) and residual amount of highest-suitability habitat (suitability class 4; lower figure) within geographic strata (as denoted by numbers above each point), after accounting for land area of each stratum. See chapters 1 and 2 for sources of data and figure 3-1 for locations of strata.



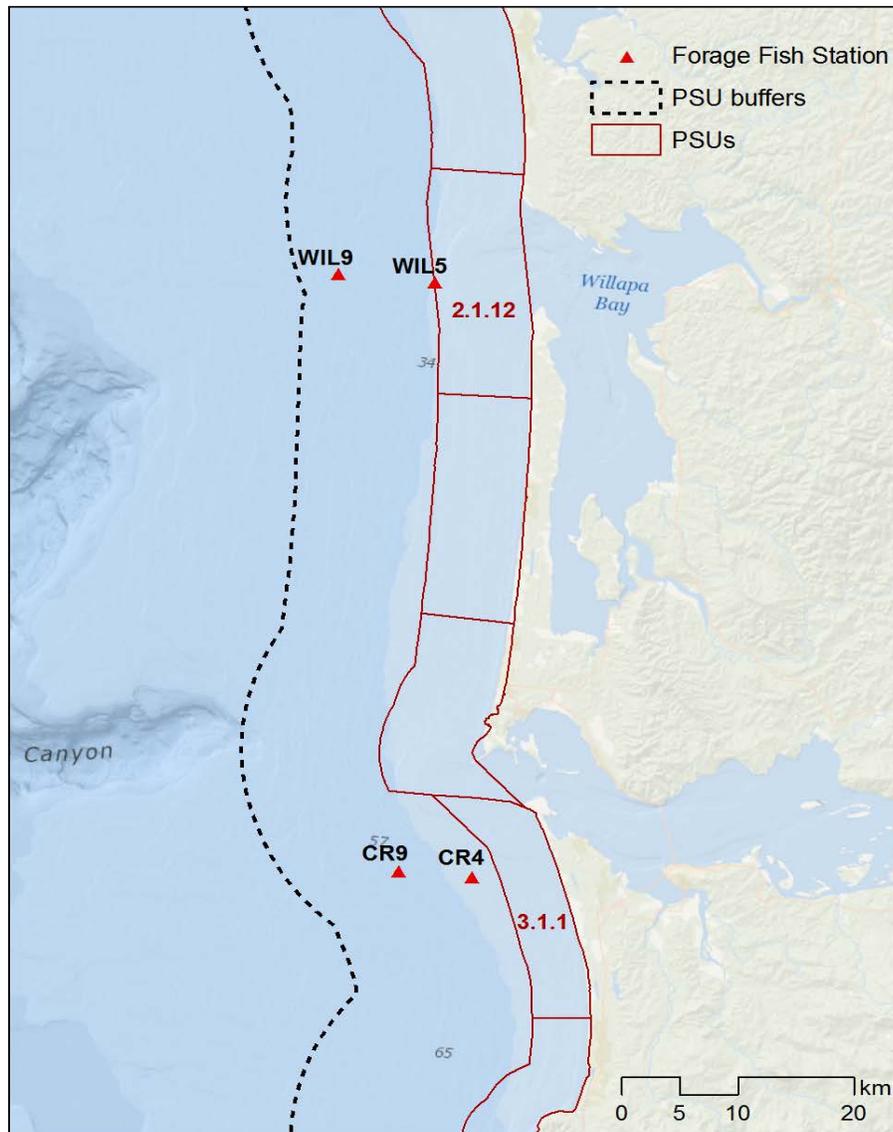
**Figure 3-3.** Relationship between mean abundance of marbled murrelets sampled at sea and cohesion of higher-suitability nesting habitat within geographic strata (numbers above points, see Figure 3-1 for locations). See chapters 1 and 2 for sources of data.



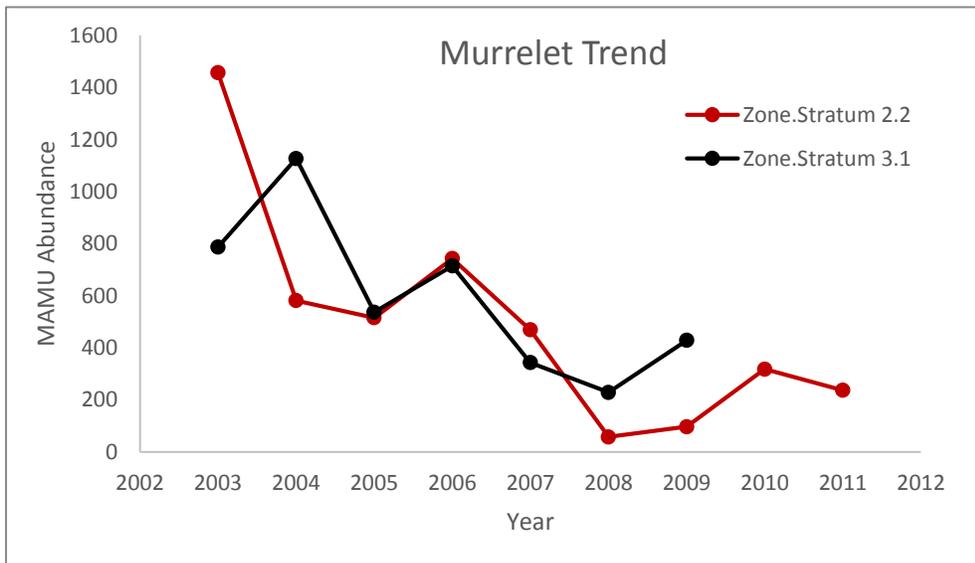
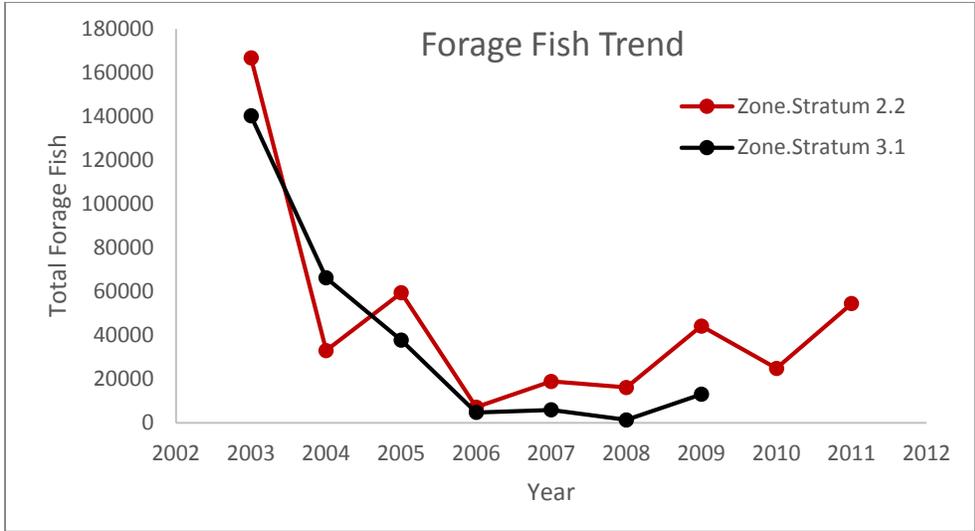
**Figure 3-4.** Mean abundance (+/- 95% confidence interval) of marbled murrelets from 2000 to 2012 (2001 to 2012 for Zones 1 and 2) by dominant ESI shoreline substrate type (NOAA 2002) in each sample unit. Prevalence of shore types is given in Table 3-2.



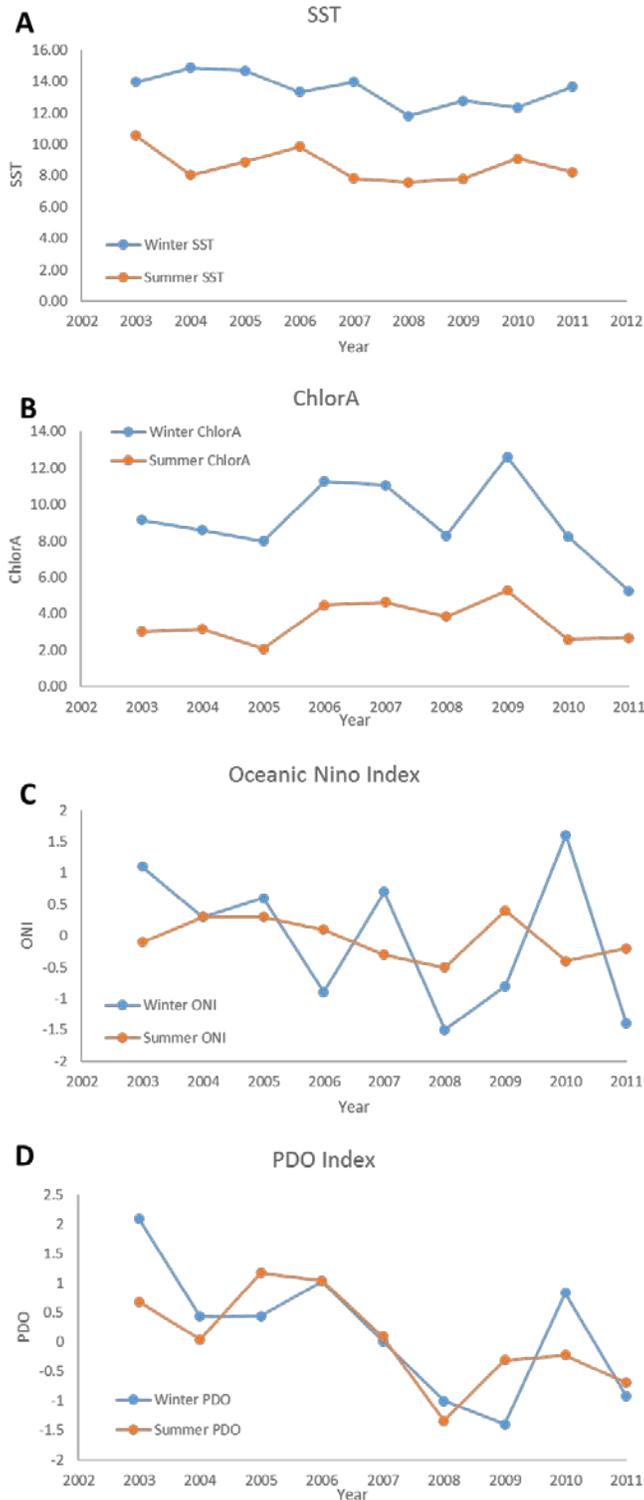
**Figure 3-5.** Relationship between changes in murrelet numbers (year 2000 or 2001 to 2012) and losses of higher-suitability nesting habitat (upper figure) and losses of highest-suitability nesting habitat (lower figure) from year 1993 to 2012. Each point represents a geographic sampling stratum as labeled by numbers above points (see Figure 3-10 for locations of each stratum).



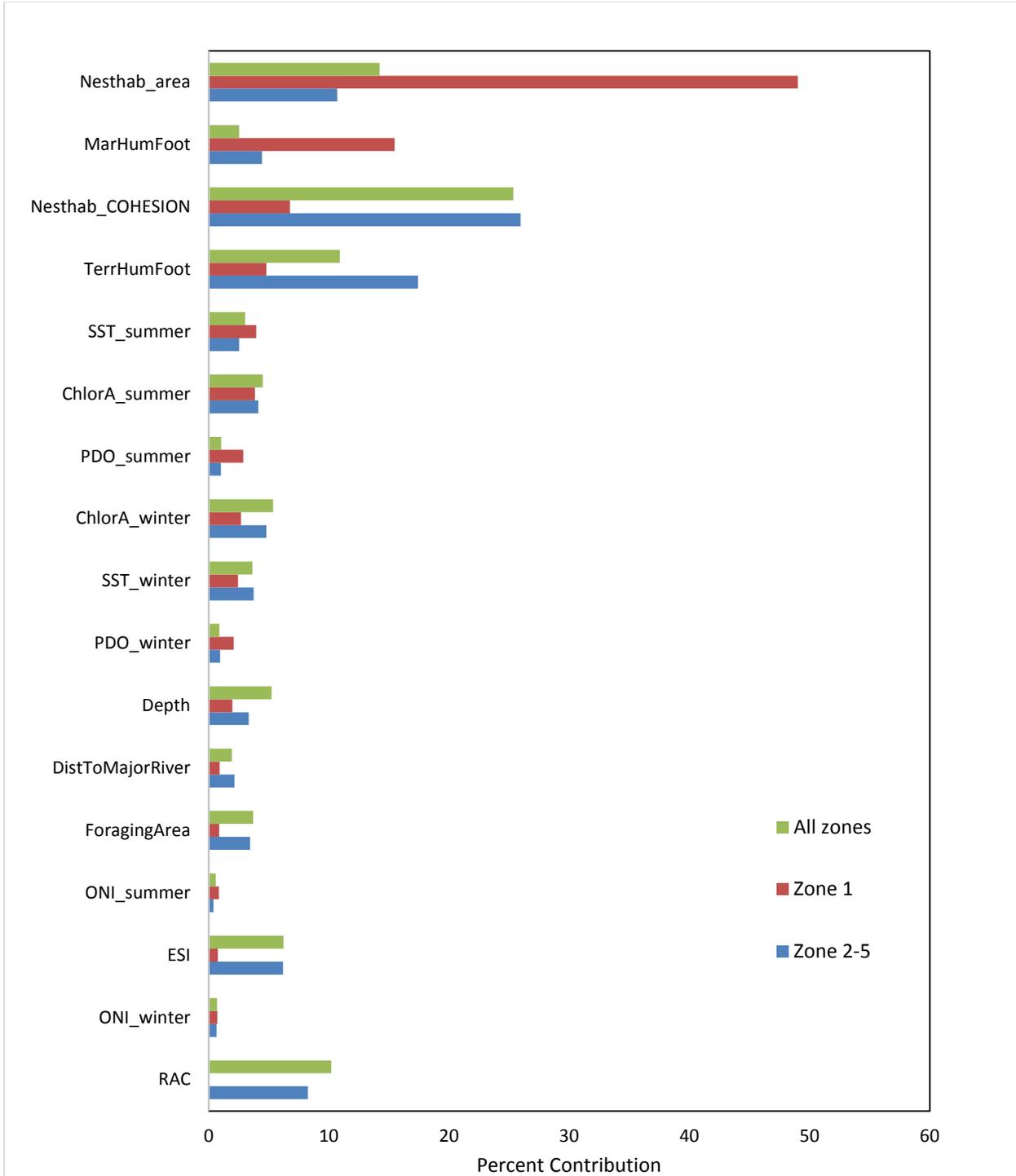
**Figure 3-6.** Location of forage fish survey stations (red triangles) offshore of Willapa Bay (WIL), WA, and the Columbia River (CR), Oregon. The population survey units (PSU) where murrelet at-sea surveys were conducted are boxed in red. The outer black dashed line represents the 20-km buffer used to calculate sea surface temperature and Chlorophyll A concentration.



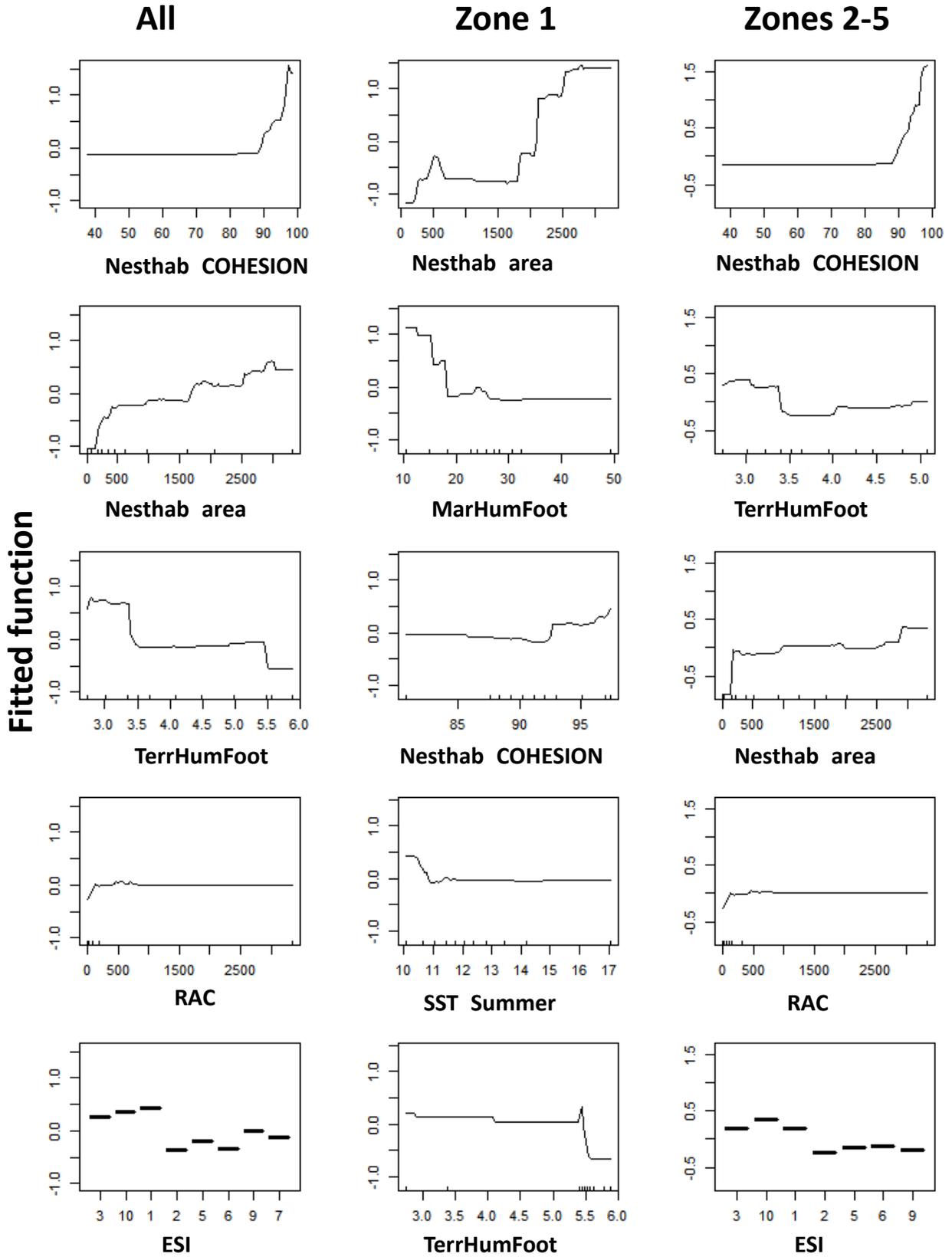
**Figure 3-7.** Comparison of trends in forage fish abundance and marbled murrelet abundance in waters near Willapa Bay, WA (Stratum 2.2) and south of the Columbia River mouth, OR (Stratum 3.1), years 2003 to 2011. Years presented are those for which forage fish abundance was available.



**Figure 3-8.** Sea surface temperature (SST, **A**) Chlorophyll A concentration (ChlorA, **B**) Oceanic Nino Index (ONI, **C**), and Pacific Decadal Oscillation (PDO, **D**) trends in winter (blue line) and summer (orange line) from 2003-2011. SST and ChlorA values were based on data from zones proximate to the forage fish sampling stations in Figure 3-6.



**Figure 3-9.** Relative influence of marine and terrestrial covariates in boosted regression models of marbled murrelet abundance in 20-km sample units within all zones, within Puget Sound (Zone 1) and along the Pacific Northwest Coast (Zones 2-5). See Table 3-1 for information about each covariate, except for RAC, which is described in the Methods text.



**Figure 3-10.** Response curves (fitted function based on marbled murrelet abundance) for the top 5 covariates (based on contribution to fitted models) depicted in Figure 3-9. Models are based on all sample units (left column), sample units in Puget Sound only (Zone 1, center column), and sample units from the outer coast (Zones 2 – 5, right column). See Table 3-1 for information about each covariate, except for RAC, which is described in the Methods text of this chapter.