Introduction

Hierarchy theory has proved a useful organizing concept in the field of landscape ecology (O'Neill et al. 1986). The concept of scale is inherent in the theory in that broad-scale, long-term patterns and processes constrain localized, short-term patterns and processes. Vegetation has been shown to be associated with a variety of environmental conditions including climate and elevation (broad-scale), as well as disturbance, succession and local topography (local scale).

We hypothesize that different local variables will emerge as important predictors of plant community types among ecoregions with differing climatic constraints. We also hypothesize that our scale of observation will dictate how well we are able to observe correlations with the localized variables. (i.e., at broad scales, climate and elevation will dominate our models, while at localized scales, we will be able to observe the effects of local topography, disturbance and imagery variables.)

<u>Methods</u>

Study Areas

We chose these two ecoregions in Oregon because of their contrasting climates. The coast range has a strong maritime influence, and a Mediterranean type climate with cool, wet winters and warm, dry summers. The East Cascades climate is more continental. It is colder in the winters, and hotter in the summers, and drier year round. The topography of both ecoregions is mountainous (sea level to 1248m in the Coast Range and 842-2578m in the East Cascades).

We built four models per ecoregion. One includes all of the available plots covering the area. The other three are built from small-scale subsamples of the plots, selected from within 40,000ha squares (20km x 20km).

Plot Data

Our database includes plots from several different vegetation survey programs. These include the Forest Inventory and Analysis program with the US Forest Service, Current Vegetation Survey from the US Forest Service and Bureau of Land Management, and Ecology plots from a variety of sources. Each plot is assigned to an Ecological System (NatureServe), based on species composition, within our database.

Spatial Data

Spatial data used in model building and mapping include variables describing climate (Type = C), topography (Type = T), soil parent material (Type = S), disturbance history (Type = D), location (Type = L), and imagery (LANDSAT and derivatives, Type = I). All data are mapped at 30m resolution. Climate data are derived from DAYMET modeled climate, and are resampled from their original resolution (1km) to 30m to allow mapping.

Statistics

We built 20-variable random forest models (Breiman 2001) within the R statistical package, and then created pixel-by-pixel model predictions from our spatial data. We selected variables using a backwards selection procedure, starting with 121 predictor variables from our spatial data. We used the variable importance measure of 'mean decrease accuracy' to select which variable to eliminate at each step. Mean decrease accuracy is calculated by measuring model prediction accuracy with and without randomly permuting each explanatory variable in turn. Local models were built with the best 20 variables. Ecoregional models contained 35 (Coast Range) and 26 (East Cascades) variables. Only the top 10 variables are shown here.

Random Forest models are an extension of Classification and Regression Trees. Our 'forest' models were created by building 500 classification trees, each from a different (random) subset of plots and explanatory variables. Model predictions are an aggregate of the prediction of each tree. Each tree 'votes' for a particular ecological system. The system with the most 'votes' is the prediction of the forest as a whole.

oforonooc

Breiman, L. 2001. Random Forests. Machine Learning 45:5-32. NatureServe 2008. International Ecological Classification Standard: Terrestrial Ecological Classifications. NatureServe Central Databases. Arlington, VA, U.S.A. O'Neill, R.V., D. L. DeAngelis, J.B. Waide and T.F.H. Allen. 1986. A hierarchical concept of Ecosystems. Princeton University Press, Princeton, New Jersey. Ohmann, J.L. and M.G. Gregory. 2002. Predictive mapping of forest composition and structure with direct gradient analysis and nearest neighbor imputation in coastal Oregon, U.S.A.Canadian Journal of Forest Research, 32:725-741.

Coast Ra

STRATU

CVPI AUGMA ANNH[CONTPF ANNFROS

Total E

California Coastal Close California Coastal Redwo California Montane Wood Columbia Plateau Wester East Cascades Mesic Montan East Cascades Oak-Ponderosa Pine Forest and Woodland

Context-dependent hierarchy of plant community predictors from the Oregon Coast Range to the eastern slopes of the Cascade Mountains Emilie Grossmann¹, Janet Ohmann², James Kagan³, Matt Gregory¹, Heather May¹

Oregon State University, Department of Forest Ecosystems and Society ²US Forest Service, Pacific Northwest Research Station ³Oregon State University, Institute for Natural Resources



uglas-fir Fo

Aesic Silver Fir-

North Pacific Lowland Mixed Hardwood Conifer Forest and Woodland

nge Ec	oregion	<u>C</u>	<u>R 1</u>		9	<u>CR 2</u>		<u>CR 3</u>		
<u>Type</u>	Importance	Variables	Туре	<u>Importance</u>	Variables	Туре	<u>Importance</u>	Variables	Туре	<u>Importance</u>
L	0.56	TM500	1	1.51	SMRTP	С	0.97	FOG	С	1.82
Т	0.55	MTM500	1	1.44	SMRTMP	С	0.95	DIFTMP	С	1.48
С	0.54	CONTPRE	С	1.22	MTM500	1	0.95	ADTM500		1.46
L	0.54	MTM700		1.18	DIFTMP	С	0.93	MR4300		1.42
С	0.53	TC300		1.13	AUGMAXT	С	0.93	STRATUS	С	1.41
С	0.53	MTC300		1.13	MTM700		0.91	STDR4300		1.34
С	0.53	TM700	1	0.76	ADR5400	1	0.76	STDTC300		1.27
С	0.52	ADTC100		0.68	CONTPRE	С	0.73	STDTC100		1.27
С	0.52	ANNSWRAD	С	0.44	MTC300	- E	0.73	STDTC200		1.21
С	0.51	STDTC500	1	0.29	SMRPRE	С	0.7	MTC600		1.2
	0.357			0.072			0.179			0.458
	0.499			0.447			0.378			0.433
	3306			38			135			30
one Conifer Forest and Woodland Forest d and Chaparral uniper Woodland and Savanna		 Inter-Mountain Basins Aspen-Mixed Conifer Forest and Woodland Inter-Mountain Basins Mountain Mahogany Woodland and Shrubland Klamath-Siskiyou Lower Montane Serpentine Mixed Conifer Woodland Klamath-Siskiyou Upper Montane Serpentine Mixed Conifer Woodland 				 Mediterranean California Lower Montane Black Oak-Conifer Forest and Woodland Mediterranean California Mesic Mixed Conifer Forest and Woodland Mediterranean California Mixed Evergreen Forest Mediterranean California Mixed Oak Woodland 				dland North Pacific B North Pacific D North Pacific D North Pacific F

Mediterranean California Dry-Mesic Mixed Conifer Forest and Woodland Mediterranean California Subalpine Woodland

East Casca	<u>coregion</u>	<u> </u>	<u>EC 1</u>		<u>EC</u>	<u>EC 3</u>				
Variables	Types	<i>Importance</i>	Variables	Types	Importance	Variables	Types	Importance	Variables	Types
Y	L	0.56	CANOPY		1.32	R5400	1	0.66	ANNPRE	С
SLPPCT	Т	0.56	Y	L	1.08	CVPRE	С	0.56	DEM	Т
X	L	0.54	MTC200	1	0.9	TC400	1	0.49	SMRTMP	С
DEM	Т	0.54	MR5700	1	0.86	MTC400	1	0.49	DECMINT	С
MTM400	1	0.54	R5400	1	0.79	SMRPRE	С	0.43	SMRPRE	С
DIFTMP	С	0.53	MR5400	1.1	0.71	MR4300	1	0.08	DIFTMP	С
ADR5400	1	0.53	STDTM200	1	0.7	LACUSTRINE	S	0	ANNFROST	С
IDSURVEY	D	0.53	ADTM400	1	0.65	ORGANIC	S	0	ANNVP	С
CONTPRE	С	0.52	STDTM400	1	0.63	LANDSLIDE	S	0	ANNHDD	С
ANNVP	С	0.52	SLPPCT	Т	0.6	SEDIMENTARY	S	0	IDSURVEY	D
Total Error		0.574			0.457			0.344		
Kappa		0.301			0.27			0.396		
n		3602			74			51		
rest and Shrubland and Woodland stern Hemlock-Douglas-fir Forest ruce Forest Red-cedar-Western Hemlock Fores		North Pacific Maritin North Pacific Maritin North Pacific Maritin North Pacific Maritin	ne Coastal Sand Dune ar ne Dry-Mesic Douglas-fir ne Mesic Subalpine Park ne Mesic-Wet Douglas-fir ain Hemlock Forest	nd Strand -Western He land r-Western H	emlock Forest No No emlock Forest No No No	orth Pacific Wooded Volcanic Flor orthern California Mesic Subalpin orthern Rocky Mountain Dry-Mesi orthern Rocky Mountain Pondero orthern Rocky Mountain Subalpin	wage e Woodland c Montane sa Pine Wo e Woodland	d Mixed Conifer Forest	Rocky Mountain Lodgepo Rocky Mountain Poor-Sit Rocky Mountain Subalpin Sierra Nevada Subalpine	ole Pine For te Lodgepol ne Dry-Mesi Lodgepole sert Wester

Rocky Mountain Aspen Forest and Woodland

Discussion

Coast Range

In the Coast Range, climate variables dominated the ecoregional model, but imagery variables were important locally. This was consistent with our original hypothesis.

In CR_3, there were pronounced effects from modeling with a local sample. North Pacific Broadleaf Landslide Forest and Shrubland was more prevalent.

East Cascades

In the East Cascades, the overall pattern was not as clear. There were two strong imagery variables within the ecoregional model, and local model structure was variable.

EC_1 and EC_2 contained imagery variables, but EC_3 contained nearly all climate variables. This could possibly be due to a strong elevation gradient here that would likely correspond with a very strong climatic

EC_3 also contained two Ecological Systems in greater proportions than the ecoregion model (Inter-Mountain Basins Mountain Mahogany Woodland and Shrubland and), probably because these were relatively abundant in the local sample.

The stronger presence of imagery variables in the local models (often) yielded maps with more fine scale spatial pattern.

Local sampling sometimes led to the loss of some ecological systems (Columbia Plateau Western Juniper Woodland in EC_2), or gains in others (e.g., the landslide type in CR_3).

The fact that imagery variables were unimportant at the ecoregional scale in the coast range, but were important in the east cascades may relate to differences in the variability in these data between the regions. The coast range is uniformly dark green and spectral reflectance is only weakly correlated with forest composition at ecoregional scales (Ohmann and Gregory 2002).

Conclusions

Observational scale affects model structure and spatial predictions (sometimes strongly).

Spatial patterns in vegetation types that correlate with imagery are (often) best modeled locally.

Spatial patterns in vegetation types relating to climate are more easily observed at ecoregional scales.

Different variables emerge in each model.



This work was conducted while building the US Geological Survey's Gap Analysis Program (GAP) vegetation map for the area. Many thanks to GAP for funding.

ruce-Fir Forest and Woodland ne Forest and Woodland hite Pine-White Fir Woodland