Observations on State of the Art Modeling of Vegetation under Climate Change



Summary of Previous Efforts

Reference	Method	Domain	Species	Time	Grain	Models / scenarios
lverson et al. 2008	Machine learning (randomForest, bagging trees, single decision tree) to model spp. abundances using FIA data and env data	Eastern U.S.	134 tree species	2100	20 km	HadleyCM3, GFDL CM2.1, PCM A1, B1, ave. across emissions scenarios
Potter et al. 2010	17 env factors reduced with PCA, correlated with FIA presence	North America	200 tree species	2050 2100	4 km	Hadley, PCM A1, B1
Coops and Waring 2011	Id climate limitations to Douglas fir growth for 1950-75 with process-based model (3-PG), use decision tree and FIA data to predict presence.	Western U.S.	15 tree species	2011 - 2040 2041 - 2070 2071 - 2100	1 km	CGCM3 downscaled using CLIMATE-WNA A2, B1
McKenney et al. 2011	BIOMAP generates statistical distributions for bioclimatic variables where species are. Locations that fall within some portion of the reference distribution are retained.	North America	130 tree species	2011-2040 2041-2070 2071-2100	10 km	CCCMA) v. CGCM2 v. GCM3.1 CSIRO v. CSIRO-Mk2.0 v. CSIRO-MK3.5 NCAR v. PCM v. CCSM3.0
Crookston et al. 2010	For. Veg. Sim. model change in species composition and growth by (1) linking mortality and regen.to climate (2) linking site index to climate and modifying growth rates, and (3) changing growth rates due to climate-induced genetic responses.	Western U.S.	74 tree species	2030, 2060, 2090 (10 yr periods)		CGCM3, GFDLCM21, HADCM3 A1B, A2, B1, B2
Morin et al.						

Summary of Climate Model and Scenario Predictions

Scenarios

A1 - high emissions – which assume that the current emission trends continue for the next several decades without modification (ca 3x pre-industrial)

B1 - significant conservation and reduction of CO2 emissions (ca 2x pre-industrial)

Average climate conditions in the eastern US: currently and for four future scenarios: Hadley A1fi, PCM B1, and average A1fi and B1 for Hadley, PCM, and GFDL

Variable	Current	Hadley high	PCM low	Ave high	Ave low
PPT (mm)	1027	1118	1082	1066	1083
PPTMAYSEP (mm)	499	498	536	485	515
TJAN (C)	-0.9	4.7	0.9	3.5	1.5
TJUL (C)	24.0	32.4	26.1	31.4	27.4
JULJANDIFF (C)	25.0	27.6	25.2	27.9	25.9
TMAYSEP (C)	21.1	29.0	23.2	27.9	24.4
TAVG (C)	12.1	19.1	14.2	17.8	15.1

Relatively warm - HadleyCM3 A1 Relatively cool - PCM B1

Summary of Climate Model and Scenario Predictions

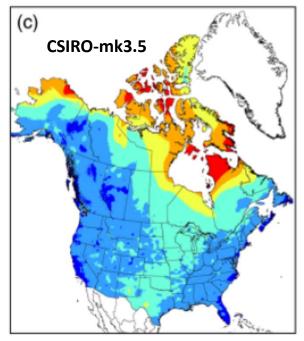
Scenarios McKenney et al. 2011

A2 - assumes rapid population growth, a reduction in forested land, and increasing levels of pollution and GHG emissions

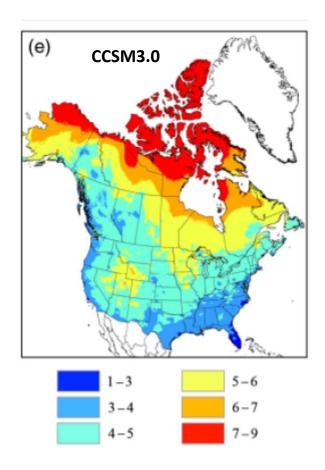
Table 1 Details on the AOGCM versions used in this study

Developer	TAR-version	AR4-version	Major improvements between versions
Canadian Centre for Climate Modeling and Analysis (CCCMA), Canada	CGCM2	CGCM3.1	Horizonal resolution increased from 608 to 680 cells More levels in the vertical Improved land surface module, which includes 3 soil layers, a snow layer, and a canopy layer Improved convection algorithm More detailed solar radiative heating module
Commonwealth Scientific and Industrial Research Organisation (CSIRO), Australia	CSIRO-Mk2.0	CSIRO-MK3.5	Improved water vapour transport algorithm Horizonal resolution increased from 528 to 2613 cells More levels in the vertical Improved land surface module, which includes 6 soil layers, 3 snow layers, and a land cover type Improved convection algorithm New prognostic cloud scheme; allows model to generate its own physically based cloud properties, based on cloud water and cloud ice Improved water vapour transport using Semi- Lagrangian algorithm
National Center for Atmospheric Reaearch (NCAR), USA	PCM	CCSM3.0	Horizonal resolution increased from 1118 to 4368 cells More levels in the vertical Greater detail in land–atmosphere flux components New treatments of cloud processes Improved aerosol radiative forcing Improved ocean mixed layer processes More realistic sea ice dynamics Many others (see reference)

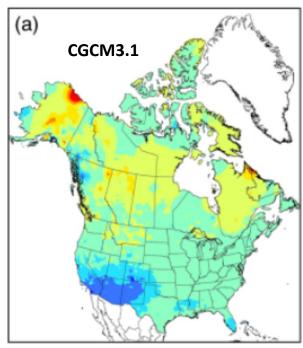
(a) CGCM3.1

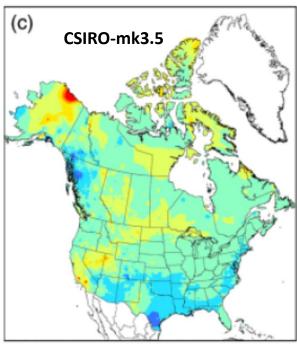


Summary of Climate Model and Scenario Predictions

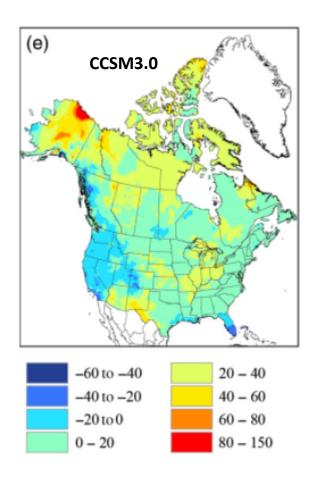


Differences between current (1971–2000) and future (2071–2100) mean annual temperature (deg C)





Summary of Climate Model and Scenario Predictions



Differences between current (1971–2000) and future (2071–2100) annual precipitation (expressed as a percentage of current values)

Plant species will respond in one of three ways to changes that push their current habitat out of their climatic tolerance limits (Davis et al. 2005):

- 1) adaptation
- 2) migration (range shift), or
- 3) extirpation

Plant species will respond in one of three ways to changes that push their current habitat out of their climatic tolerance limits (Davis et al. 2005):

- 1) adaptation
- 2) migration (range shift), or
- 3) extirpation

Where will suitable habitat be located under climate change?

• climate/habitat suitability modeling

Plant species will respond in one of three ways to changes that push their current habitat out of their climatic tolerance limits (Davis et al. 2005):

- 1) adaptation
- 2) migration (range shift), or
- 3) extirpation

Where will suitable habitat be located under climate change?

• climate/habitat suitability modeling

Variables used to predict current and future tree species habitat

- In the first					
Climate ^a					
TAVG	Mean annual temperature (°C)				
TJAN	Mean January temperature (°C)				
TJUL		Mean July temperature (°C)			
TMAYSEP		Mean May-September temperature (°C)			
PPT	Annual precipitation (r	nm)			
PPTMAYSEP	Mean May-September	Mean May-September precipitation (mm)			
JULJANDIFF	Mean difference betwee	Mean difference between July and January			
	temperature (°C)				
Elevation ^b					
ELV_CV	Elevation coefficient of	f variation			
ELV_MAX	Maximum elevation (n				
ELV_MEAN	Average elevation (m)	-7			
ELV_MIN	Minimum elevation (m)			
ELV_RANGE	Range of elevation (m				
Soil class ^c					
ALFISOL	Alfisol (%)				
ARIDISOL	Aridisol (%)	Strongest predictors:			
ENTISOL	Entisol (%)				
HISTOSOL	Histosol (%)	Temperature			
INCEPTSOL	Inceptisol (%)	PPTMAY-SEPT			
MOLLISOL	Mollisol (%)	SLOPE			
SPODOSOL	Spodosol (%)	010.1			
ULTISOL	Ultisol (%)	PPT			
VERTISOL	Vertisol (%)	ORD (soil prod)			
	vertisor (%)	Soil texture			
Soil property ^d					
BD	Soil bulk density (g/cn				
CLAY		Percent clay (<0.002 mm size)			
KFFACT	Soil erodibility factor,				
		erosion to water movement)			
NO10	Percent soil passing sid				
NO200		Percent soil passing sieve no. 200 (fine)			
OM		Organic matter content (% by weight)			
ORD		Potential soil productivity (m ³ timber/ha)			
PERM	~	Soil permeability rate (cm/h)			
PH		Soil pH			
ROCKDEP		Depth to bedrock (cm)			
SLOPE		Soil slope (%) of a soil component			
TAWC	Total available water c	apacity (cm, to 152 cm)			
Land use and fragmen	tatione				
FRAG	Fragmentation index (I	Riitters et al. (2002))			
AGRICULT	Cropland (%)	1			
FOREST	Forest land (%)				
NONFOREST	Nonforest land (%)	Iverson et al. 2008			
WATER	Water (%)	1ve15011 et al. 2006			

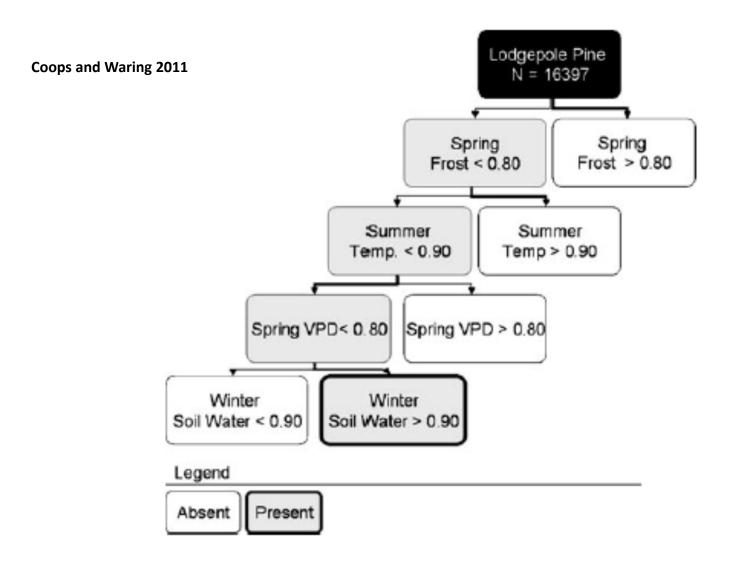
Predictors

Table 1.—Spatial environmental variables included in the Multivariate Spatio-Temporal Clustering (MSTC) analysis.

Category	Spatial environmental variable
Soil	
	Plant-available water capacity
	Bulk density of soil
	Kjeldahl soil nitrogen
	Organic matter in soil
Temperature	
	In the coldest quarter
	In the warmest quarter
	Diurnal temperature difference
	Biotemperature
	Solar insolation
Precipitation	
	In the driest quarter
	In the wettest quarter
	In the warmest quarter
	In the coldest quarter
	Ratio of precipitation to evapotranspiration
Topography	
	Compound topographic index
	(convexness or concavity)
Growing season	
	Length in integer months

Potter et al. 2010

Predictors



Decision tree developed to predict presence and absence of lodgepole pine, based on the maximum effect of the four seasonal climate modifiers

Plant species will respond in one of three ways to changes that push their current habitat out of their climatic tolerance limits (Davis et al. 2005):

- 1) adaptation
- 2) migration (range shift), or
- 3) extirpation

Where will suitable habitat be located under climate change?

climate/habitat suitability modeling

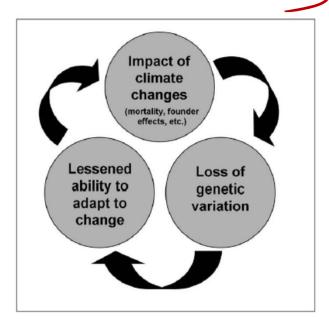
Can the population get to the newly suitable habitats?

- Dispersal ability of species
- Geographic Resistance
 - **❖** Distance from current to new habitat
 - Topography
 - Land facets
 - **Vegetation fragmentation**
 - Land use



Plant species will respond in one of three ways to changes that push their current habitat out of their climatic tolerance limits (Davis et al. 2005):

- 1) adaptation
- 2) migration (range shift), or
- 3) extirpation

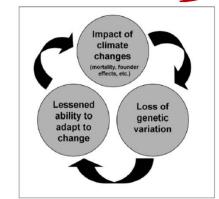


Plant species will respond in one of three ways to changes that push their current habitat out of their climatic tolerance limits (Davis et al. 2005):

- 1) adaptation
- 2) migration (range shift), or
- 3) extirpation

Table 2.—Factors, both intrinsic and extrinsic to a species or population of forest trees, that increase its risk of extinction, extirpation, or genetic degradation.

Intrinsic factors	Extrinsic factors		
Limited range	Extensive fragmentation		
Small/disjunct populations	Pest/pathogen infestation		
Limited to high elevations	Large shift of range with climate change		
Long lifespan	Exploitation		
Long time to reproduction	Exposure to atmospheric deposition		
Low fecundity	Geographic dispersal barriers ^a		
Physical habitat specialization	Anthropogenic dispersal barriers ^a		
Limited seed/pollen dispersal	Exposure to sea-level rise ^{a,b}		
Low species-wide genetic variation			
Late successional species			
Dependence on specific disturbance regime ^a			
Reliance on interspecific interactions ^a			
Sensitivity to temperature and precipitation change ^a			
Lack of phenological flexibility ^a			



a From Young et al. (2009).

^b Not applicable to the Appalachian Mountains.

Modeling Approaches

Bioclimate Envelope Models

- Iverson et al. 2008
- Potter et al. 2010
- McKenney et al. 2011

Simulation Models

- Demographic Models
 - **❖** Forest Vegetation Simulator (Crookston et al. 2010)
 - ❖ FIRE-BGC V2 (Keene et al.)

Hybrid Models

• 3PG / Climate envelope (Coops and Waring 2010)

Modeling Approaches

Bioclimate Envelope Models

- Iverson et al. 2008
- Potter et al. 2010
- McKenney et al. 2011

Simulation Models

- Demographic Models
 - **❖** Forest Vegetation Simulator (Crookston et al. 2010)
 - ❖ FIRE-BGC V2 (Keene et al.)

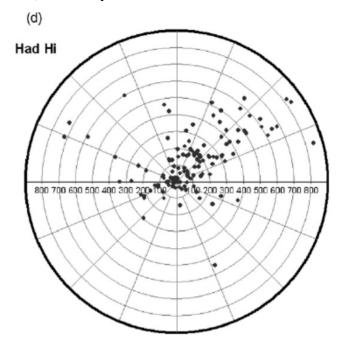
Hybrid Models

• 3PG / Climate envelope (Coops and Waring 2010)

"In this approach, we cannot include changes in land use and land cover likely to occur in the next 100 years, or disturbances such as pests, pathogens, natural disasters, and other human activities. Coupling these outputs with process-based ecosystem dynamics models which include disturbance would be a productive line of research." Iverson et al. 2008

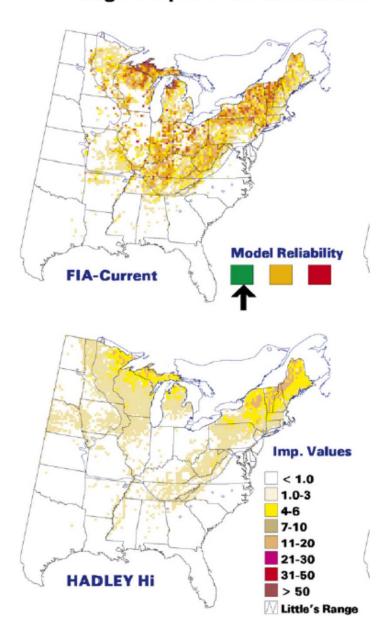
Results: Iverson et al. 2008

- 55% of species increase in habitat by >=2%
- 14% of species decrease in habitat by >=2%
- Considering importance value leads to more declines: 66 species increase, 54 decrease, 14 no change.
- Species severely diminished: black spruce, mountain maple, butternut, paper birch, quaking aspen, balsam poplar, balsam fir, northern white cedar, black maple, red spruce, white spruce.



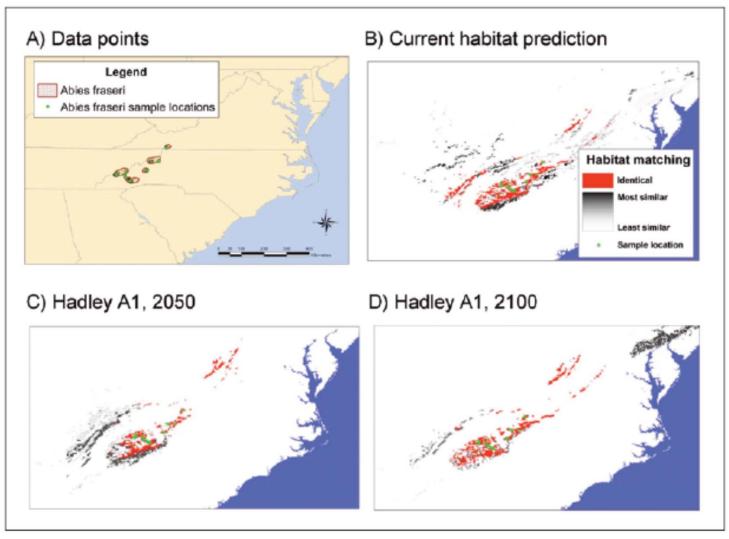
Potential changes in distance and direction of mean centers of suitable habitat
(26 species > 400 km)

sugar maple - Acer saccharum



Results: Potter et al. 2010

Predictions are sometime surprising!



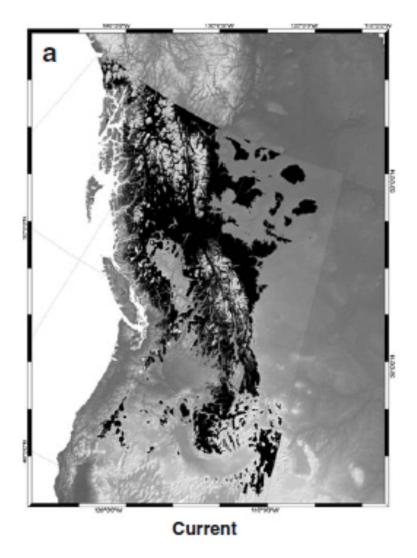
Fraser Fir

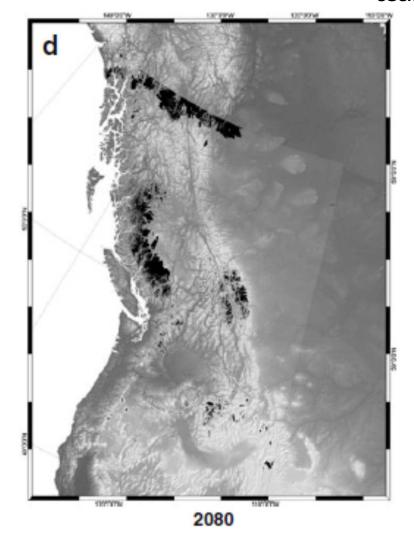
Results: Coops and Waring 2011

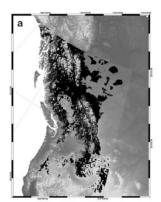
Lodgepole pine

Sites with significant spring frost, summer temperatures averaging <15°C and soils that fully recharged from snowmelt were most likely to support lodgepole pine.

CGCM2





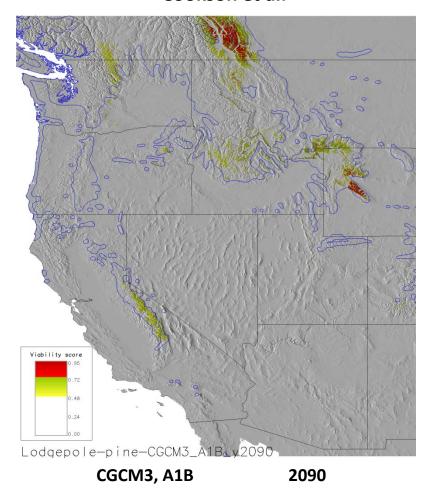


Results: Lodgepole Pine

Coops and Waring 2011

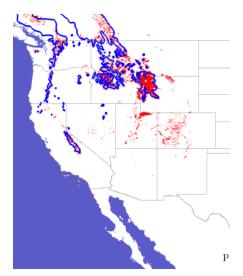
CGCM2 2080

Cookson et al.

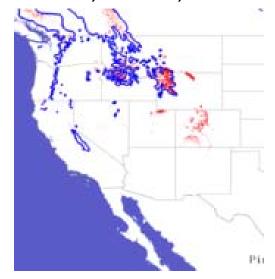


Results: Whitebark pine

Hargroves et al.

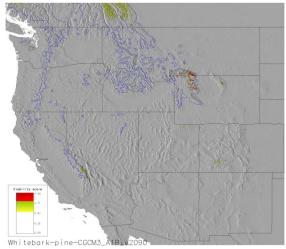


PCM, Scenario A1, 2100

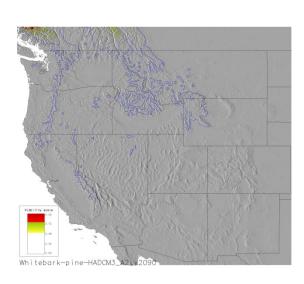


Hadley, Scenario A1, 2100

Cookson et al.



CGCM3, A1B, 2090



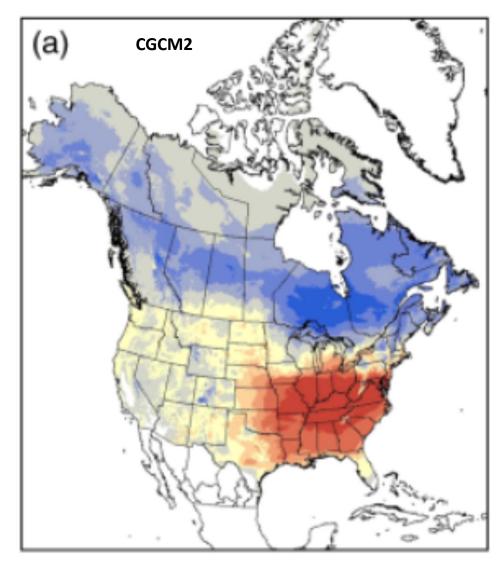
Hadley CM3, A2,2090



Hadley + CCMA-GCM/2, 2090
Warell et al. 2007

Results: McKinney et al. 2011





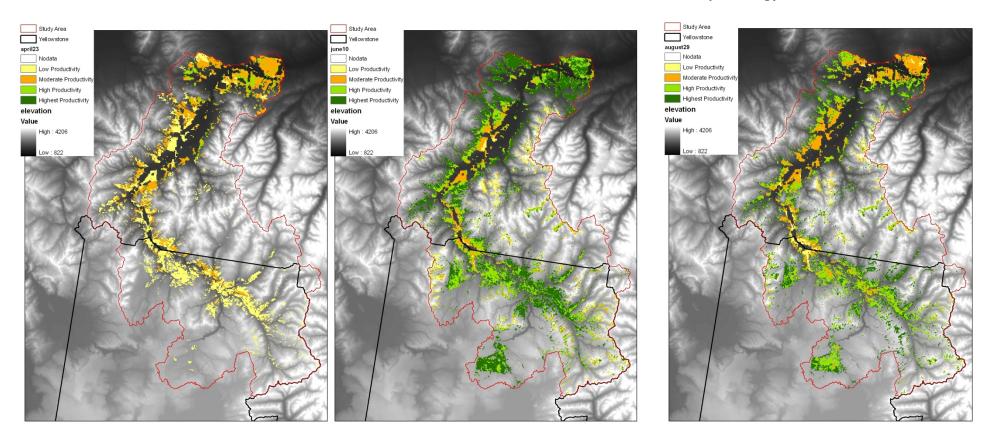
Differences between current (1971–2000) and future (2071–2100) tree climate envelope richness (i.e., number of tree species).

Conclusions

- Rather than duplicate existing efforts, we should synthesize their results in ways that are relevant to our collaborators??? Or not?
- This should include synthesis of projected climate change and response of tree species and ecological system types.
- We can add value to these by additional analyses of change in habitat area, role of disturbance, dispersal ability, landscape resistance under land use change.
- We can also do finer resolution modeling for select species/types of high interest to collaborators (e.g., WBP).

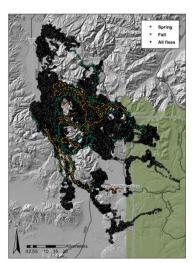
Patch Dynamics of Grassland Phenology - Nate

- Spatial dynamics of "green flush"
- Climate predictors of phenology
- Land use modification of phenology



April 23, 2010 June 10, 2010 August 29, 2010

Evaluating alternative approaches to identifying wildlife corridors - Meredith



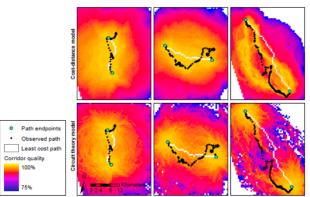


Figure 9. Best (a) cost-distance model and (b) circuit theory model outputs for spring migration of sample individuals. Outputs are from the model variant excluding climate-related covariates.

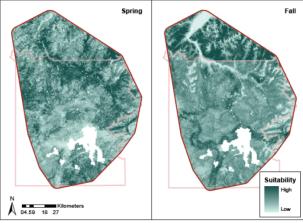
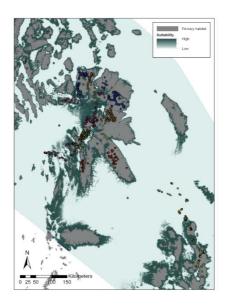
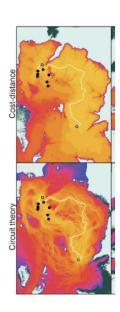


Figure 13. Application of best habitat suitability models from Madison Valley study area to Northern Range study area in (a) spring and (b) fall.

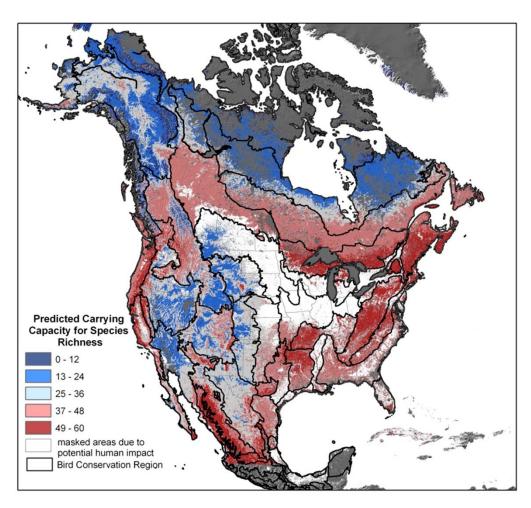




Next Step:

Nate phenology + Meredith elk connectivity + climate change

Carrying Capacity for Species Richness for Landbirds



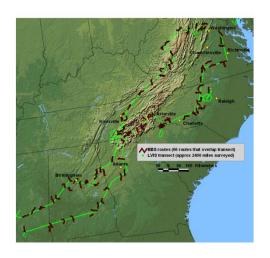
 $S_K = 27.042 \text{ aGPP} - 0.004 \text{ aGPP}^2 - 19.425 \%SCV + 0.005 PET$

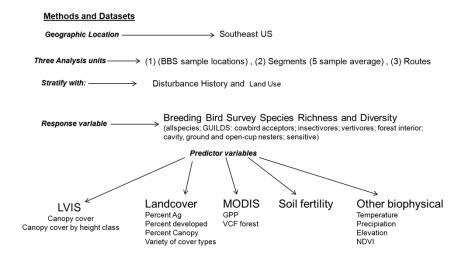
Hansen et al. in press. Global Ecology and Biogeography

%SCV: Interannual variation in GPP

PET: Potential evapotranspiration

GPP, Canopy Structure, Land Use: Bird abundance and Diversity





Statistical Analysis



