

Assessing the Transferability of Ecological Models to New Settings

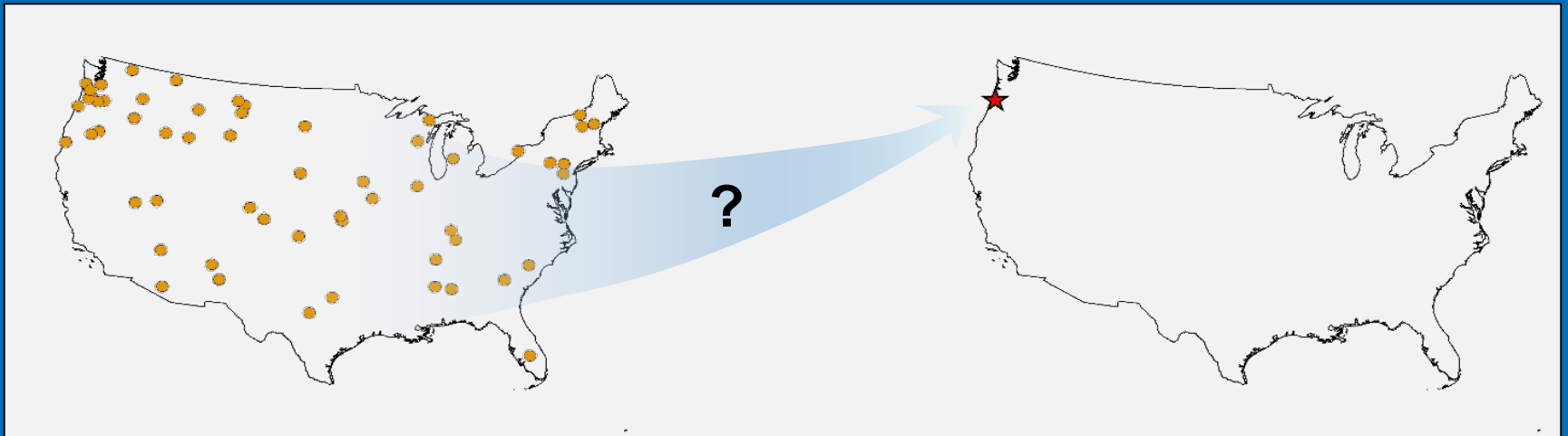
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³Integrated Statistics, NOAA Fisheries, Falmouth, MA

⁴Bloomsburg Univ., Bloomsburg, PA



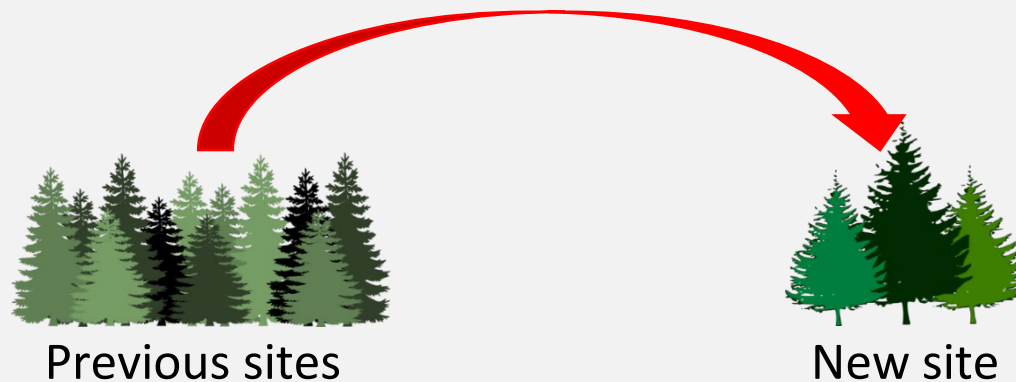
Problem:

Models useful to calculate ecological endpoints at data-poor locations,
aka, model transfer

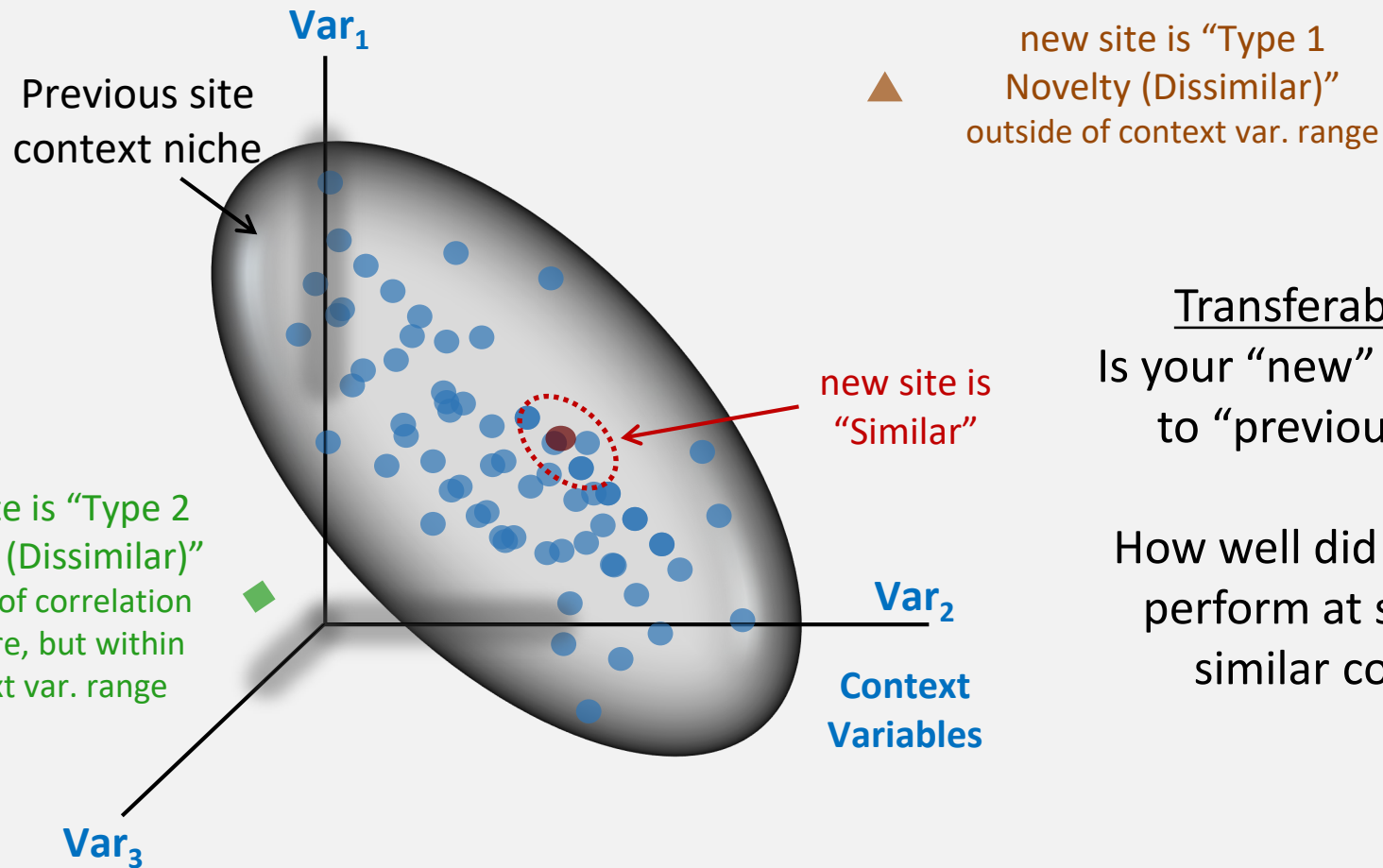
Model transfer practices often ad-hoc, poorly documented

- can result in inaccurate or indefensible results
 - lack of transparency → lack of trust
 - wasted resources

Goal: A practical methodology to assess
whether a model can be defensibly transferred to a new site



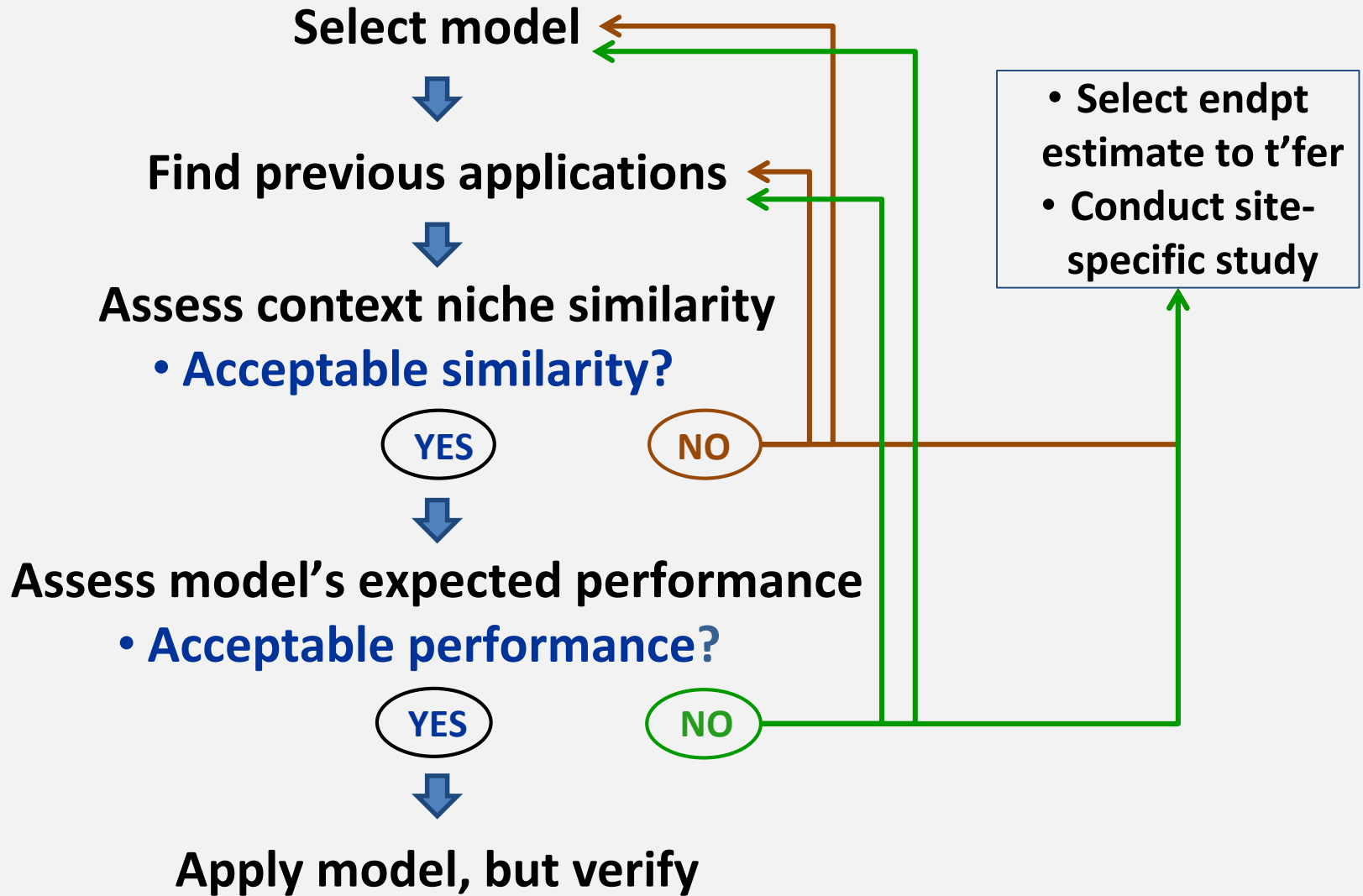
Key concept: a model that worked at “previous” sites will work at a “new” site if the previous & new sites have the same *context niche*



Transferability Q’s
Is your “new” site similar, to “previous” sites?

How well did the model perform at sites with similar context?

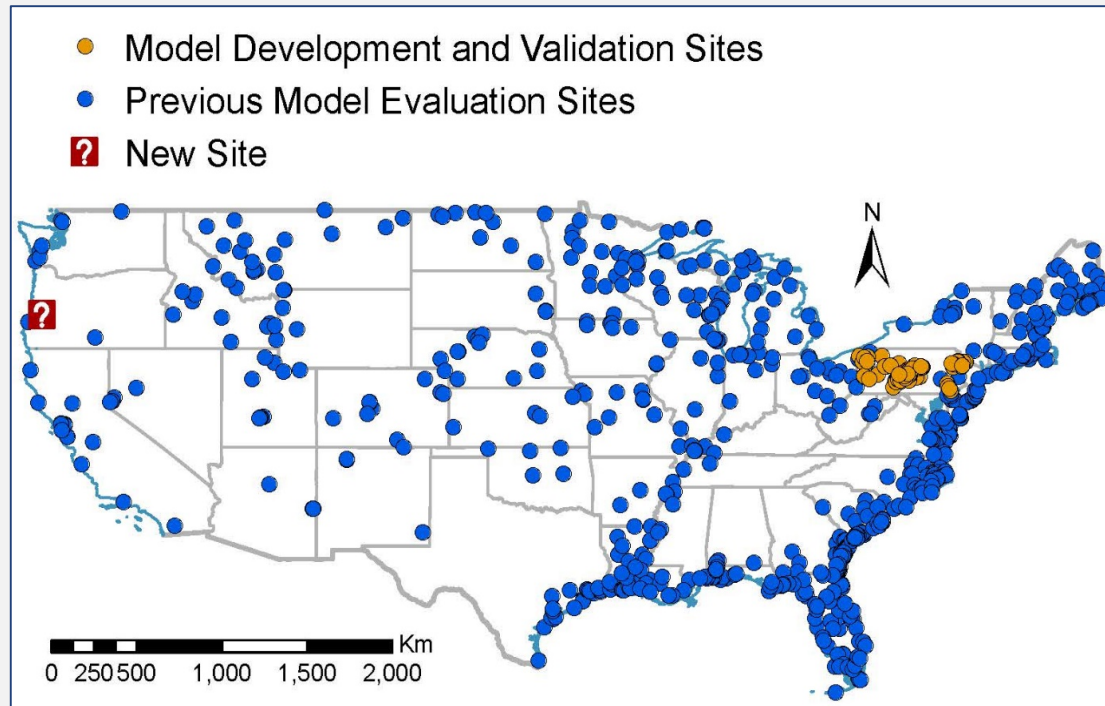
Model Transferability Assessment Framework



Example: Should the *CoC Model*¹ be used to assess wetland condition in a southern Oregon wetland?

$$\text{CoC} = 8.16 - 0.73(\text{pH}) - 0.29(\text{LDI})$$

- CoC² – wetland condition index based on plant spp.
- LDI³ - landscape development index (disturbance)



¹Moon et al. (2017)

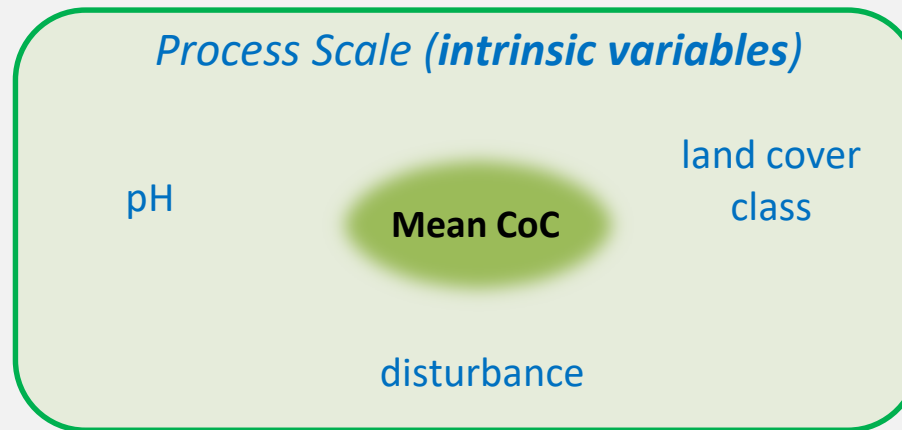
²Lopez & Fennessy (2002)

³Brown & Vivas (2005) & Lane et al. (2012)

Assess Context Similarity (1)

Select Model Context Variables: Intrinsic & Extrinsic

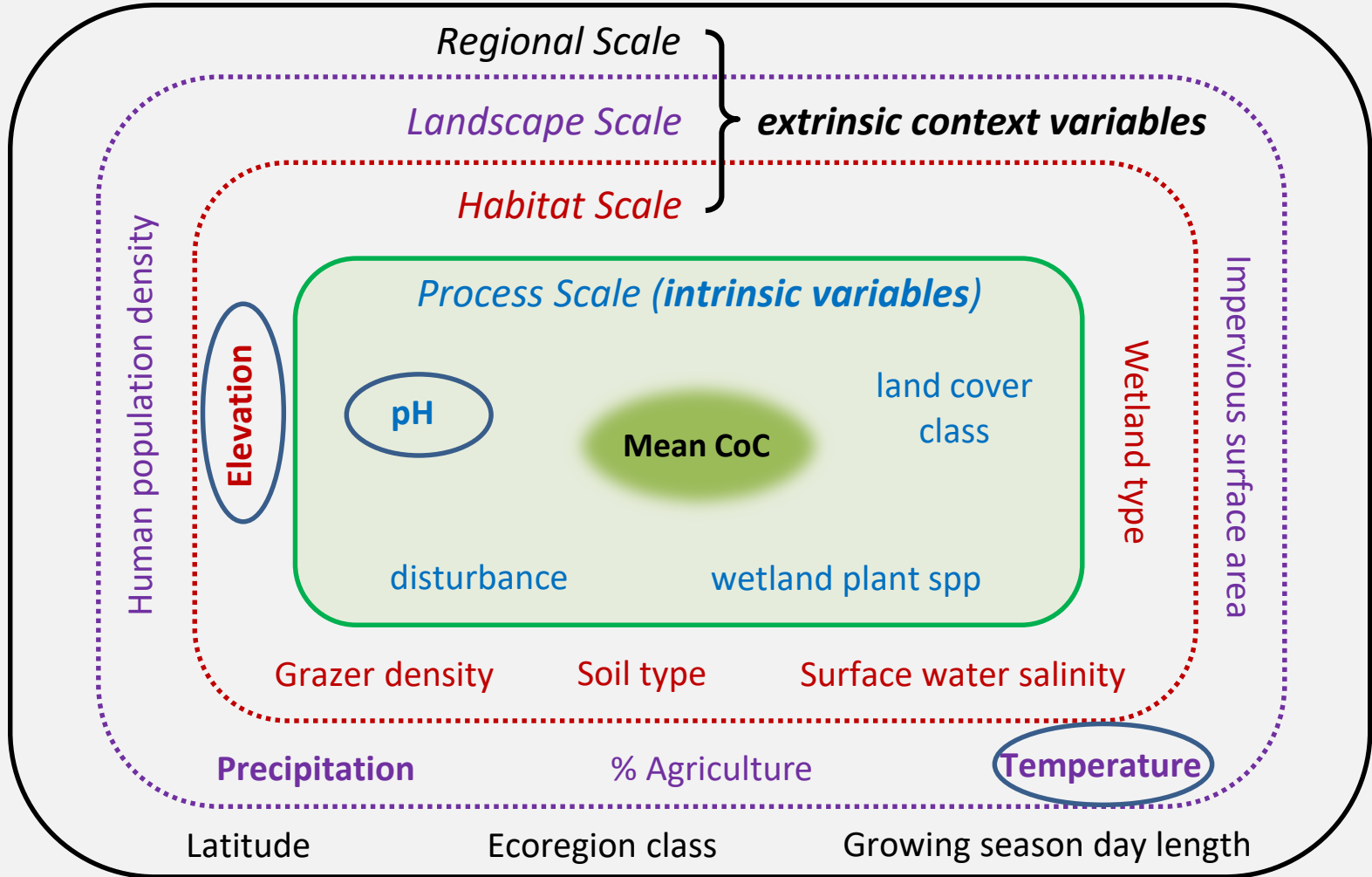
Model Context Variables = biophysical variables that are reasonably expected to influence the process being modeled



Intrinsic Context Variables:
predictor variables
in CoC model

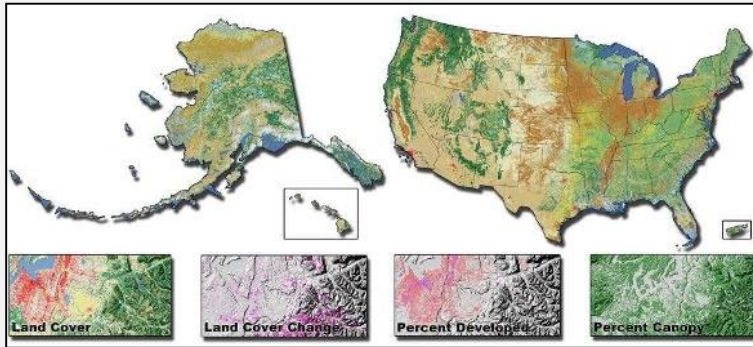
Assess Context Similarity (1)

Select Model Context Variables: Intrinsic & Extrinsic

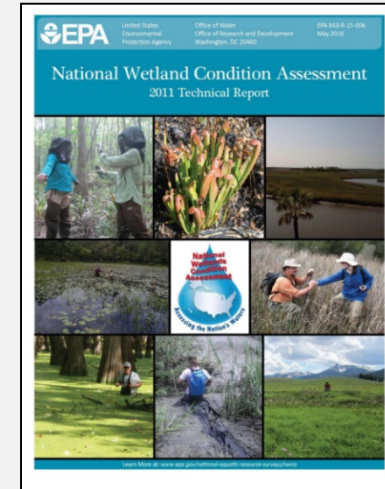


Assess Context Similarity (2)

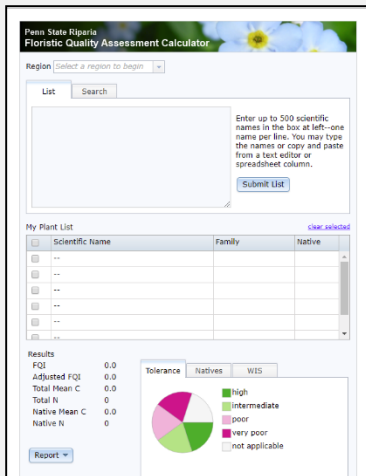
Quantifying Context Variables



USGS National Land Cover Database

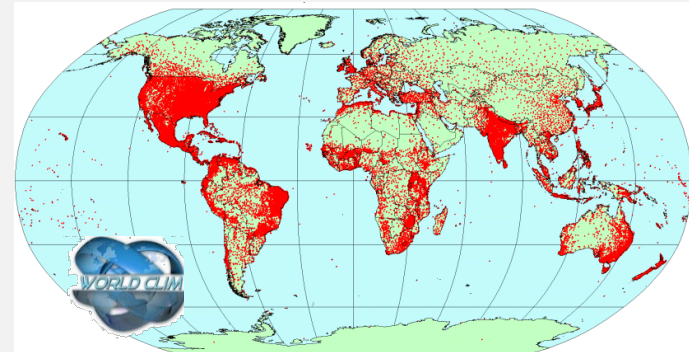


US EPA National Wetland Condition Assessment 2011



The screenshot shows the 'Penn State Riparia Floristic Quality Assessment Calculator' interface. It includes a 'Region' dropdown menu, a 'List' search box, and a text area for entering up to 500 scientific names. Below this is a table with columns for 'Scientific Name', 'Family', and 'Native'. The 'Results' section displays 'FQI' (0.0), 'Adjusted FQI' (0.0), 'Total Mean C' (6.0), 'Total N' (0), 'Native Mean C' (0.0), and 'Native N' (0). A pie chart shows the distribution of results, with a legend for 'Tolerance' (Natives, WIS) and categories: High (green), Intermediate (yellow), Poor (red), Very poor (purple), and Not applicable (grey).

Penn State Riparia
Floristic Quality
Assessment
Calculator



WorldClim – Global climate data

Assess Context Similarity (3)

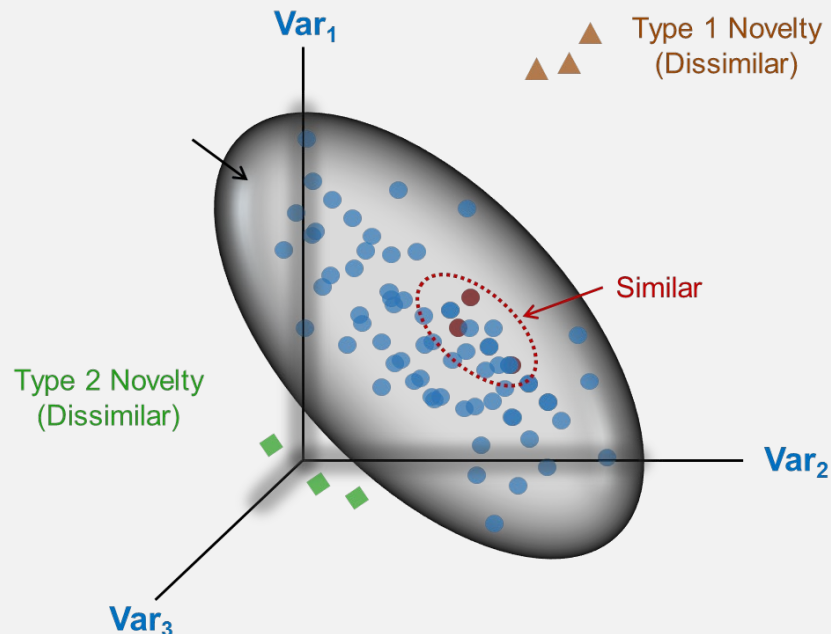
Setting Acceptability Criteria

Context Niche Similarity Analysis: ExDet¹ (Extrapolation Detection Tool)

Metrics: Novelty Type 1 & 2 dissimilarity (NT1, NT2) of “new” site

Criteria: user sets value(s) of the metrics

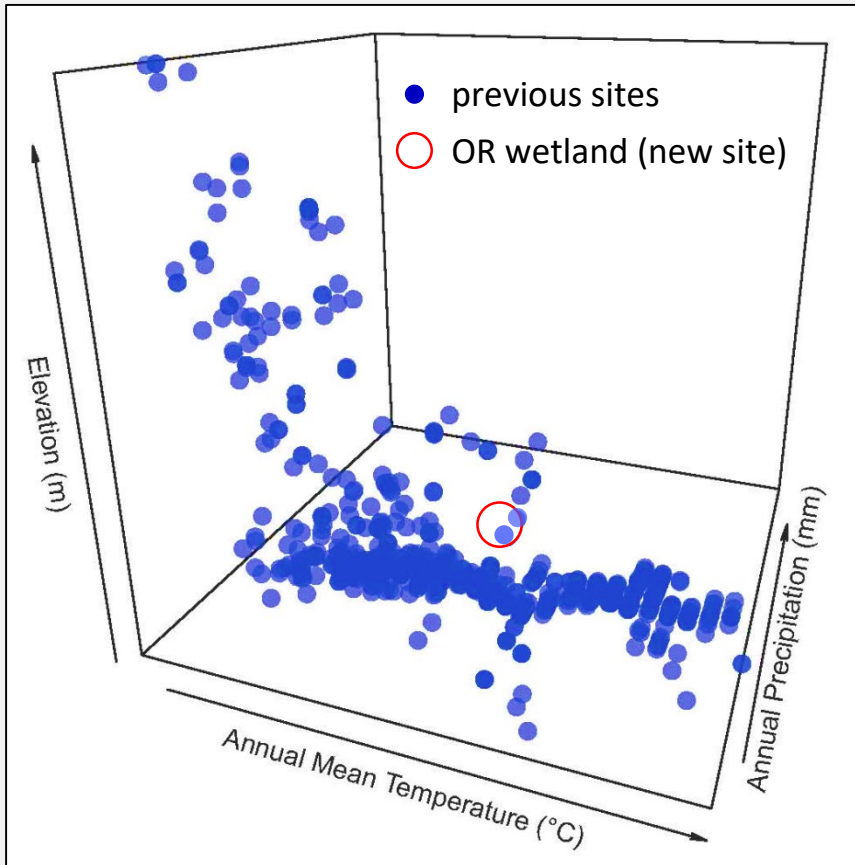
This study $NT1 \leq 1.0$ (OR site has similar context var. range)
 $NT2 \leq 1.0$ (OR site context within correlation structure)



¹Mesgaran et al (2014) & Moon et al. (2017)
<http://www.climond.org/exdet>

Assess Context Similarity (4)

Context Similarity Analysis



Results of Similarity Analysis (ExDet)

	NT1 _{OR}	NT2 _{OR}
Criteria	≤ 1	≤ 1
Observed	0	0.08

Decision:

**OR wetland site is similar
to “previous” sites**

**Proceed to determine
expected error**

If context niches not similar:

select a different model, use an estimate of endpt, or do site-specific study

Assess Model's Expected Performance (1)

Set acceptability criteria

1. Select most similar “previous” sites to compare to “new” site
 - subset of all sites; similarity metric (var. range, % of sites, etc)
2. Select “model error” and “uncertainty” metrics
 - error: f {observed vs predicted}; difference, r^2
 - uncertainty: variance in error; (SD, CoV)
3. Set tolerance thresholds for error & uncertainty metrics

Assess Model's Expected Performance (1)

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This study

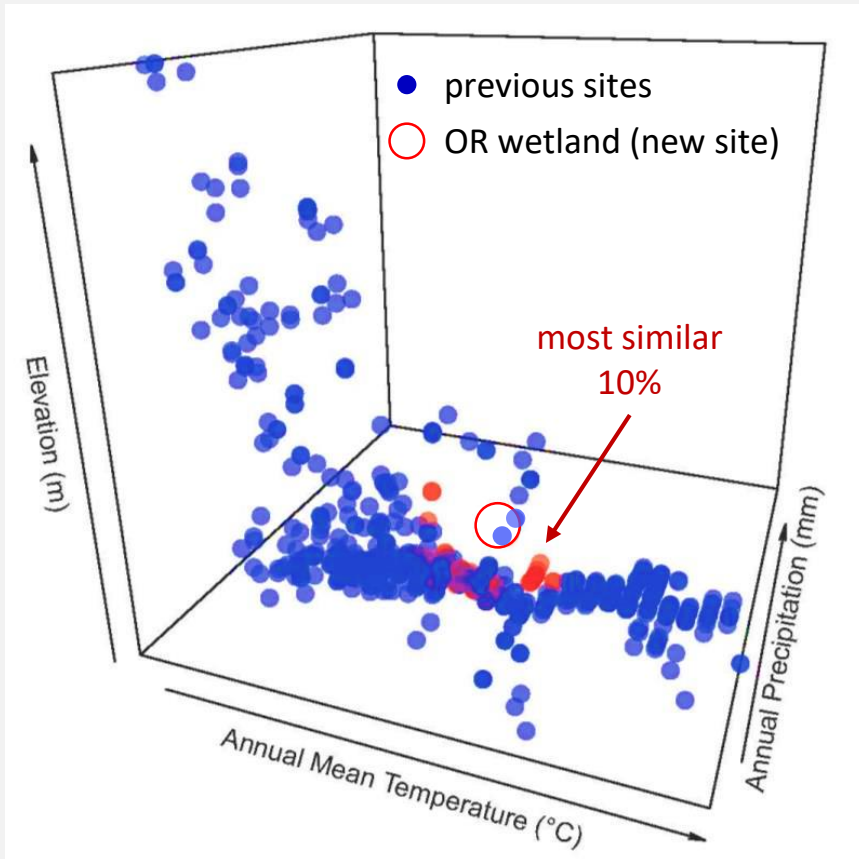
Most similar sites = 10% with context values closest to OR site

Error: mean (predicted CoC - measured CoC) ≤ 2

Uncertainty: CV of mean error $\leq 20\%$

Assess Model's Expected Performance (2)

Est. model error at "most similar" sites



Model Error \equiv
mean (measured CoC – predicted CoC)

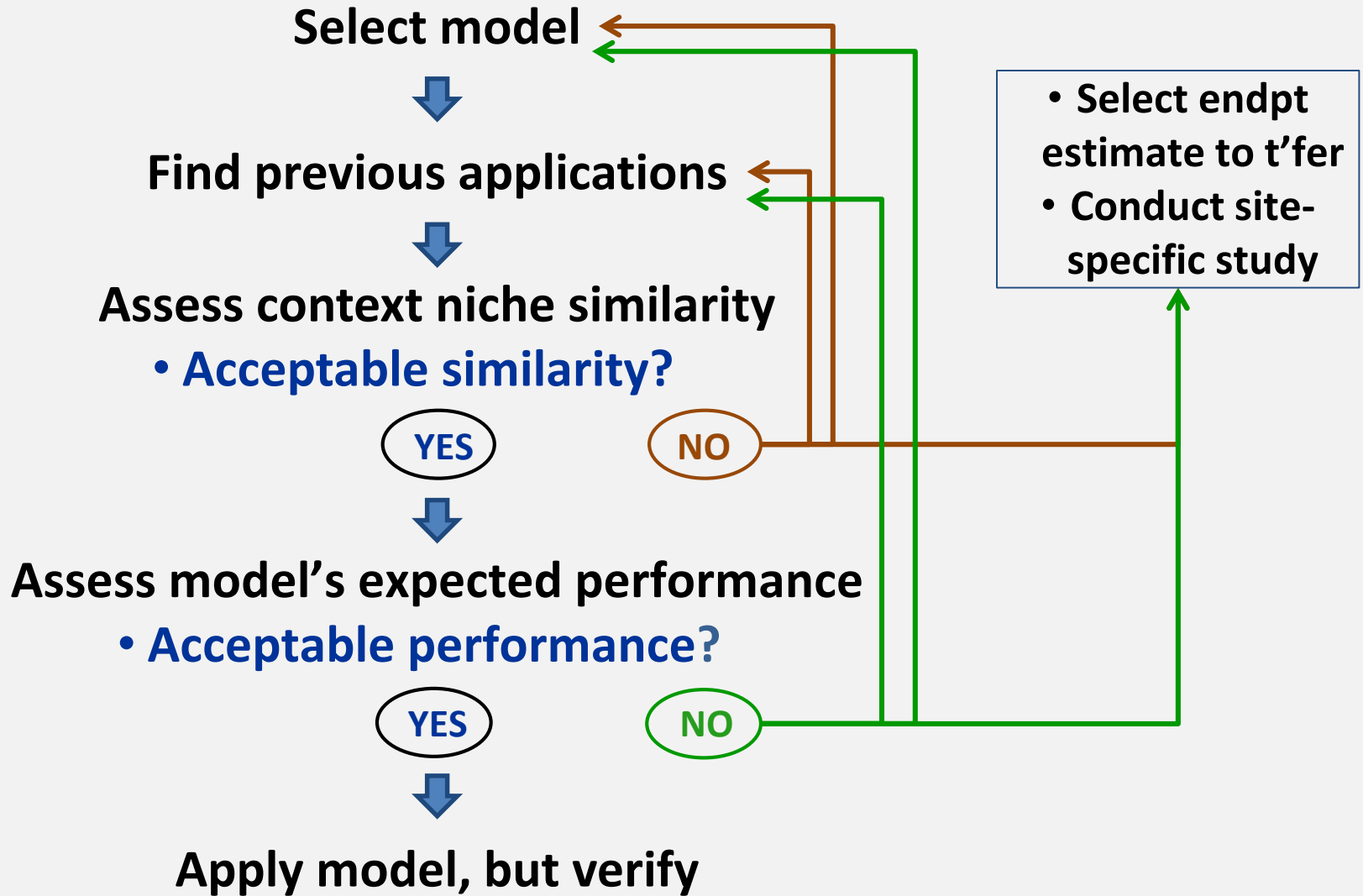
	Error (mean)	Uncertainty (CV)
Criteria	≤ 2	$\leq 20\%$
Observed	1.26	105.82

Decision:

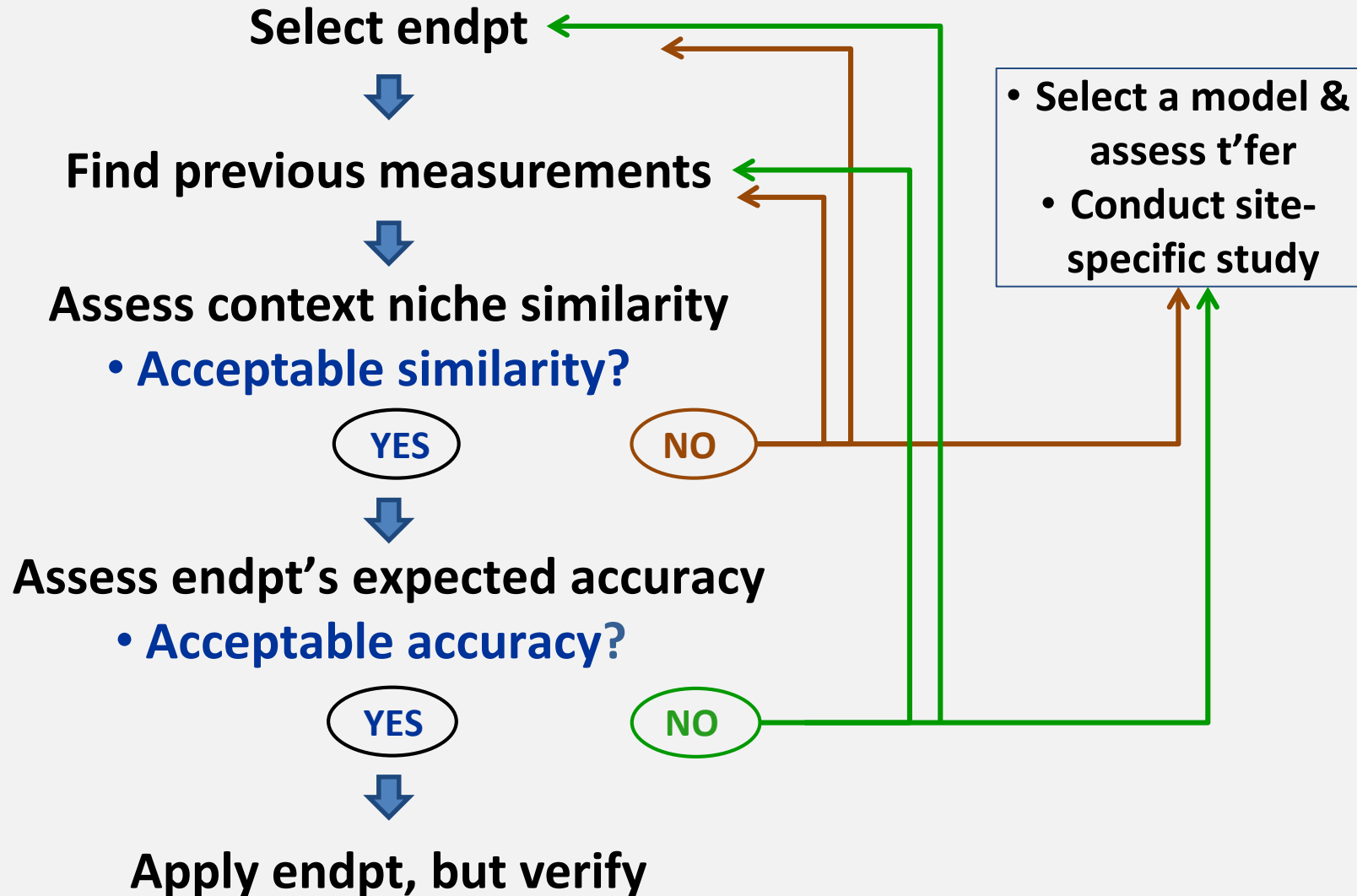
**Model can be applied at OR site,
but expect uncertainty in its
accuracy**

Invest in validation at OR site

Model Transferability Assessment Framework



Unit (or Endpt) Transfer Framework



Take-Away's

- Model transfers are necessary but fraught
- A methodology to assess a model's transferability:
 - provides consistency
 - increases legitimacy thru transparency
 - increases efficiency, reduces costs
- Framework also useful for transferability of unit values

More...

- Moon et al. (2017) *Ecosphere* 8(10):e01974
Model application niche analysis: assessing the transferability and generalizability of ecological models
- Yates et al. (2018) *Trends in Ecol. Evol.* 33:790
Outstanding challenges in the transferability of ecological models