

# NASA/NCEAS/iPlant Update

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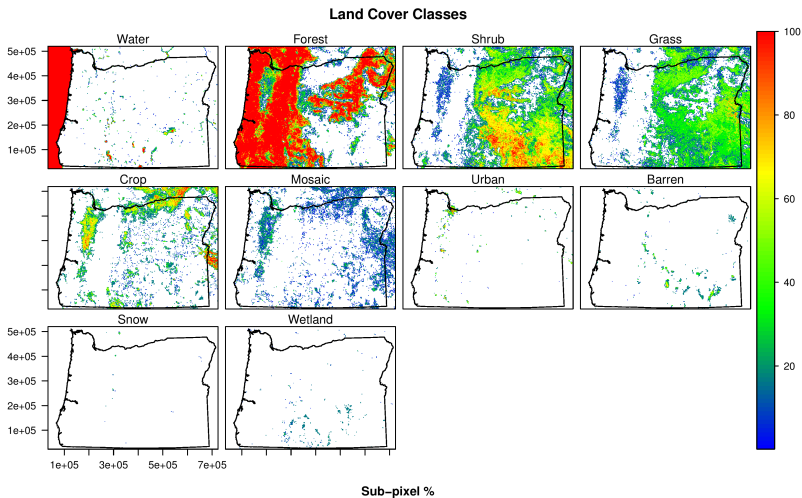
## Exploring LST-LULC interactions

We know LST is affected by Land Cover (e.g. Teuling, 2010).  
Concerns for incorporation of LST into interpolation models:

1. Land cover may introduce bias into LST-air temperature relationships
2. Biases may change throughout the year
3. Biases may be hard to estimate on a daily basis
4. Biases may be hard to estimate using only point (station) locations rather than the full grid

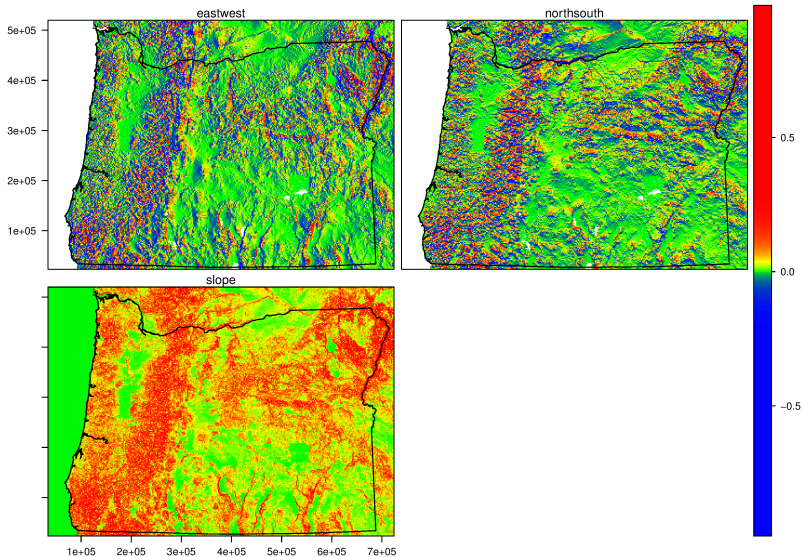
So I explored the LST-LULC relationships using the full 1km dataset (or a sample of 10k pixels for modeling, see below)

# A review of the data

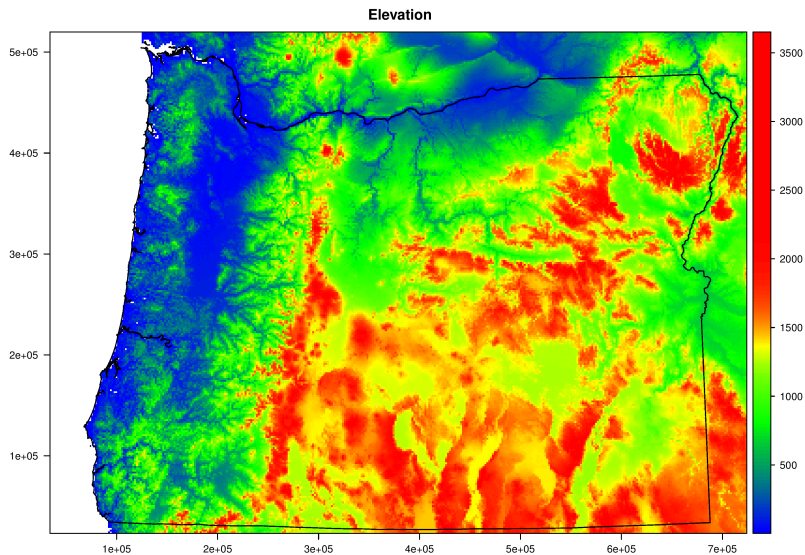


# A review of the data

## Topographic Variables



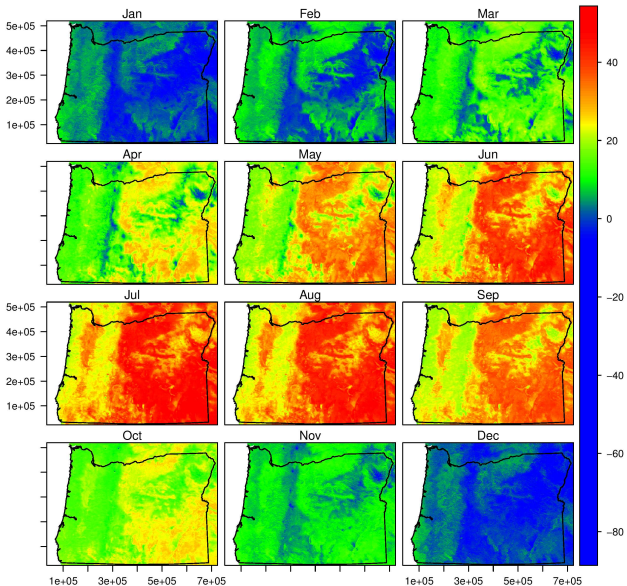
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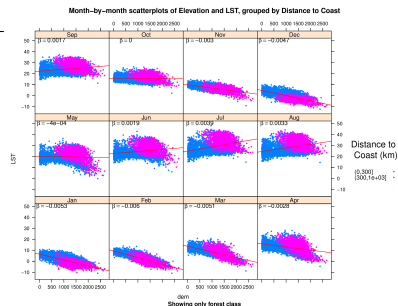
MOD11A1 Mean Monthly LST



## Sampling Bias

Stations are disproportionately in Urban and agricultural areas (surprise!).

LULC Class	Sample %	Station %
Barren	0.00	0.00
Crop	0.08	0.17
Forest	0.51	0.55
Grass	0.09	0.05
Mosaic	0.00	0.01
Shrub	0.31	0.09
Urban	0.01	0.11
Water	0.01	0.02
Wetland	0.00	0.00

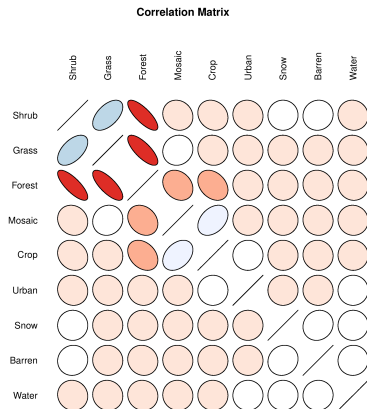


Sample % indicates the proportion of the 10k samples in each LULC class while Station % is proportion of stations in that class.



## LULC Correlations

Correlations between the sub-pixel proportions (%) of each LULC

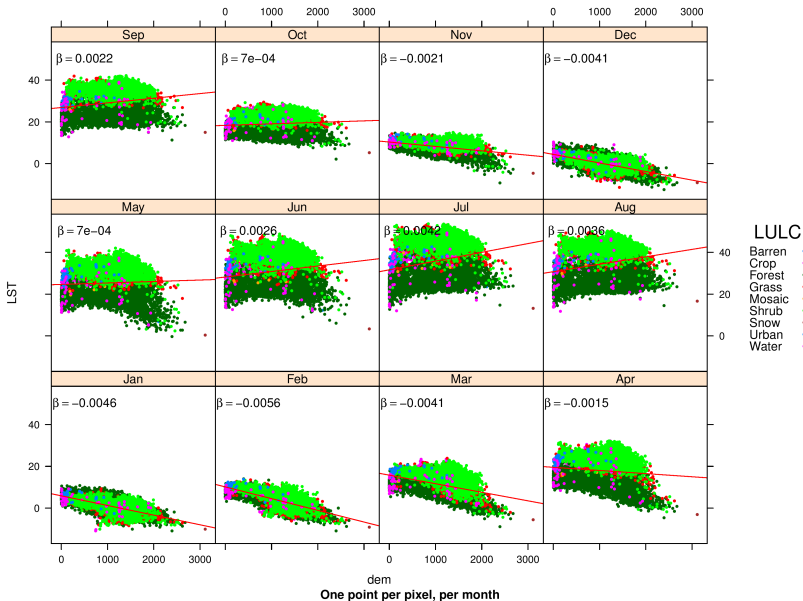


class.

Only Forest shows correlations  $> 0.6$  with any others. In regression, this class can be removed (included in the intercept) and the remaining correlations are low enough to not worry about.

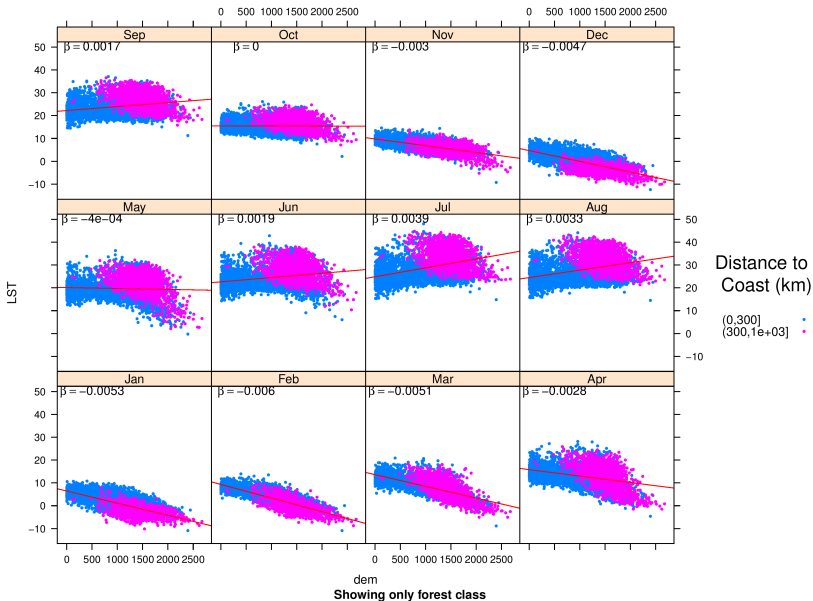
# LST-LULC-Elevation Relationships

Month-by-month scatterplots of Elevation and LST, grouped by LULC



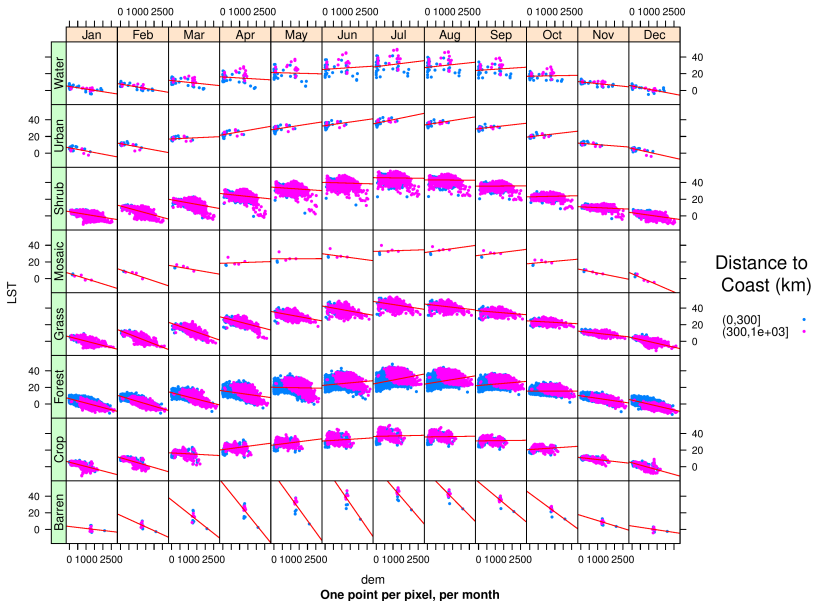
# LST-LULC-Elevation Relationships

Month-by-month scatterplots of Elevation and LST, grouped by Distance to Coast



# LST-LULC-Elevation Relationships

Month-by-month scatterplots of Elevation and LST, grouped by LULC and colored by distance to coast



## Effects of Distance to Coast - linear?

Seems unlikely that the effects of 'distance to coast' will be linear globally (especially for sites that are far inland). We need a transformation that has larger values by the coast and small values elsewhere, dropping to near zero at some point ( $\sim 100\text{km}$ ?) What are the options?

1. use linear anyways...
2. some kind of logistic curve that goes from 1 near the coast to 0 at some point inland. Hard to specify parameters (would be better to fit them, but this would require a more complex model and probably preclude use of existing packages)
3.  $\log(\text{distance})$
4.  $8 - \log(\text{distance} + 1)$  ( $8 \approx \log(\text{farthest point on earth from coast})$ )  
This is what I tried (but this certainly needs more exploration).

# Spatial Regression

$$Y(s) = X^T(s)\beta + W(s) + \epsilon(s) \quad (1)$$

$$\epsilon(s) \sim \mathcal{N}(0, \sigma^2) : \text{Measurement error} \quad (2)$$

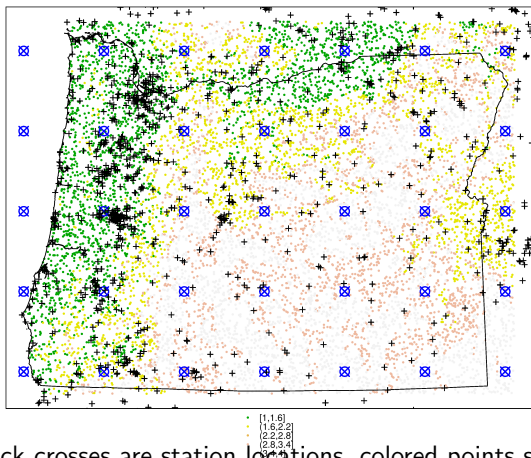
$$W(s) : \text{Zero-centered Gaussian Process} \quad (3)$$

$$(s) : \text{indexes space} \quad (4)$$

Objective: Isolate the effects of LULC on LST after accounting for topography (eastwest, northsouth, distance to coast, elevation) and spatial autocorrelation. Fit using spBayes.

## Sampling Bias

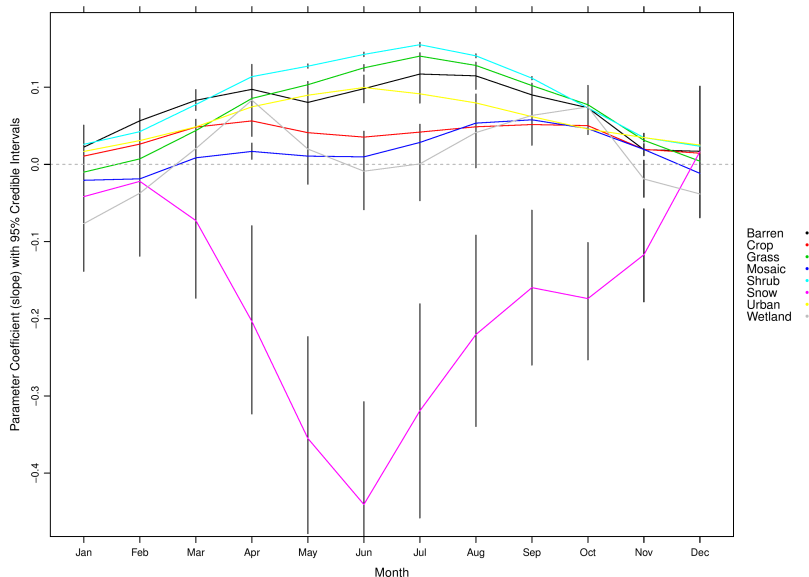
10,000 sample locations, stratified by elevation



Black crosses are station locations, colored points show the 10k random samples (stratified by elevation - highest class nearly white in color). Blue crosses show points used in spBayes 'predictive process' to estimate spatial effects.

# Estimated Parameters by month

## Effects of LULC on LST

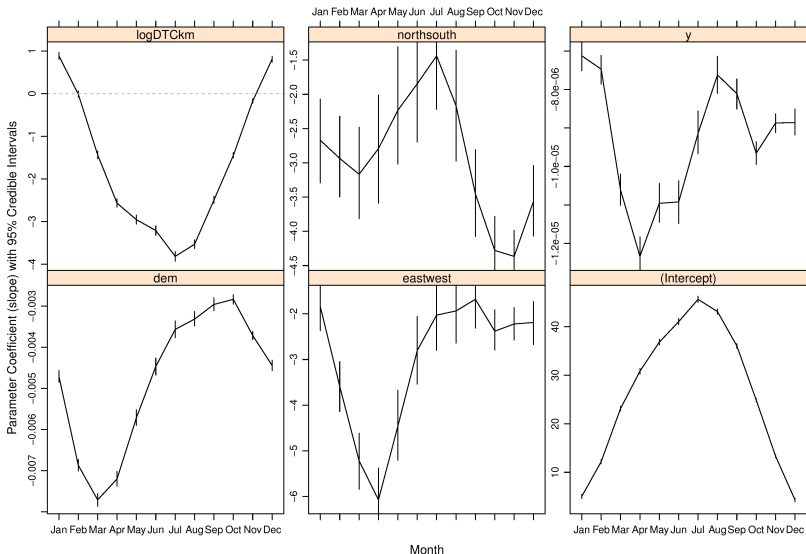


Coefficients are unstandardized and represent the change in LST expected with a 1% increase in that class from 100% Forest



# Estimated Parameters by month

## Effects of Topography on LST



**Coefficients are unstandardized. Intercept is degrees C, eastwest and northsouth range from -1 (90 degree slope) to 0 (flat) to 1 (90 degree slope), and dem is m logDTCKm is in log(km), and y (lat) is m**

# Summary

- LULC clearly affects LST measurements differently throughout the year (forests are cooler, shrub is warmest).
- Strongest biases are in summer months
- Lapse rate changes significantly throughout the year ( $< -7^{\circ}\text{C}/\text{km}$  in spring,  $> -3^{\circ}\text{C}/\text{km}$  in fall).
- Should compare these results to parameters estimated using only station locations to assess whether spatial station bias (and low  $n$ ) is biasing model fitting.

## Other thoughts

A few other things that caught me attention while working on this...

- Values in monthly LST climatologies range from  $-88^{\circ}\text{C}$  to  $54^{\circ}\text{C}$ . Need to do additional quality control?
- Need to estimate effects for all (or at least most common across region) LULC classes to predict.
- Will interaction terms in interpolation models (i.e.  $\text{lst} * \text{lulc}$ ) be sufficient to capture these patterns?
- Possible solution: use informed priors from full dataset to inform relationships at the stations (Bayesian only).