

Exploring Relationships Between Soil Structure and Climate Across the Conterminous USA

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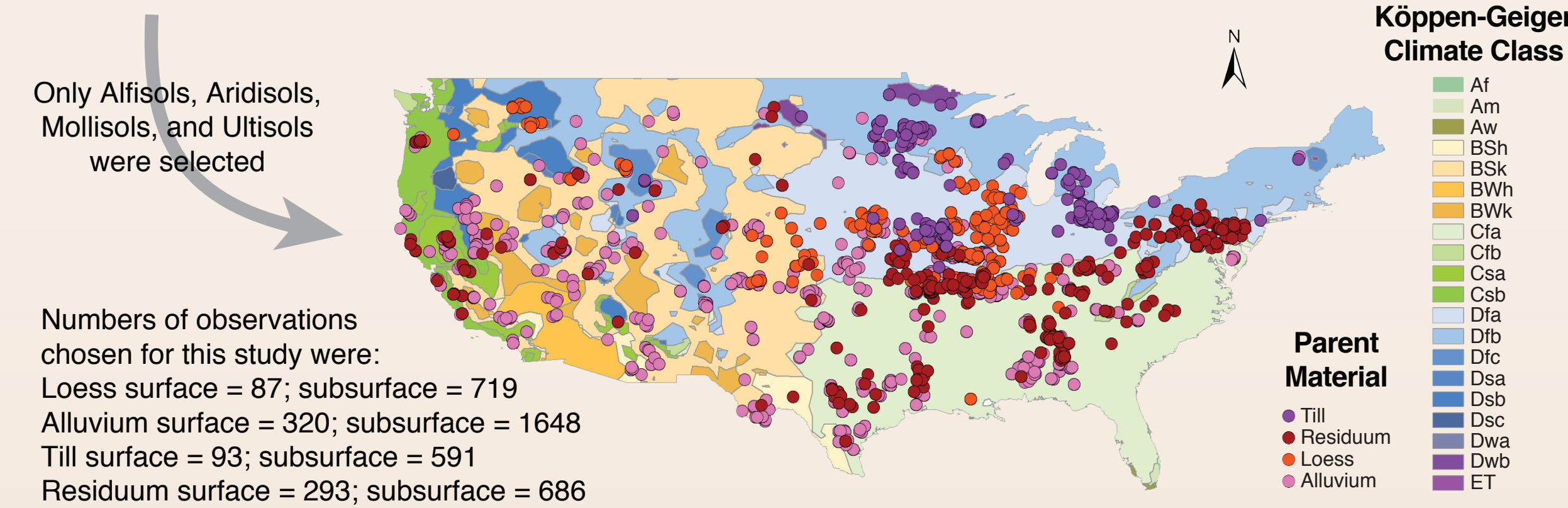
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Soil and Climate Information

We combined the USDA NRCS Soil Characterization Database with climatological information from the Köppen-Geiger climate classification and PRISM mean annual precipitation (MAP) and mean annual temperature (MAT) data to investigate relationships between soil structure and climate in the US. Surface and subsurface horizons within 4 parent materials were selected.



Grade	Value
Structureless	0.0
Weak	1.0
Weak & moderate	1.5
Moderate	2.0
Moderate & strong	2.5
Strong	3.0
Very strong	3.5

Size Class	Upper Boundary (mm)	Lower Boundary (mm)	Geometric Mean (mm)
Granular/Platy			
V. fine/thin	0.1	1	0.32
Fine/Thin	1	2	1.41
Medium	2	5	3.16
Coarse	5	10	7.07
V. coarse/thick	10	10.00	
Columnar/Prismatic/Wedge			
V. fine	0.1	10	1.00
Fine	10	20	14.14
Medium	20	50	31.62
Coarse	50	100	70.71
V. coarse	100	500	223.61
E. Coarse	500	500.00	
Angular blocky/Subangular blocky			
V. fine	0.1	5	0.71
Fine	5	10	7.07
Medium	10	20	14.14
Coarse	20	50	31.62
V. coarse	50	50.00	

Pod Type	Circularity	Orientation	Width:Height Ratio	Aspect Ratio	Roundness	Solidity
abk	0.91	88.1	0.21	1.00	0.97	0.90
col	0.35	84.0	0.19	4.37	0.07	0.82
gr	1.01	35.0	1.13	1.34	0.76	0.93
pl	0.27	5.73	4.21	6.77	0.08	0.74
pr	0.27	82.2	0.14	5.87	0.04	0.87
sbk	0.75	40.1	1.13	0.97	0.82	0.90
wq	0.48	15.4	1.94	1.00	1.00	0.89

Numbers of observations chosen for this study were:
 Loess surface = 87; subsurface = 719
 Alluvium surface = 320; subsurface = 1648
 Till surface = 93; subsurface = 591
 Residuum surface = 293; subsurface = 686

Data Analyses and Results

Occurrence frequency for categorical variables (parent material, climate class, and structure) were examined. Categorical structure variables were converted to quantitative ratio scales and MAP and MAT data were combined into EEMT to analyze relationships on a continuous scale for each type of parent material. Bivariate correlations—Pearson (continuous-continuous), correlation ratio (continuous-multichotomous), and Cramér's V (multichotomous-multichotomous)—were calculated.

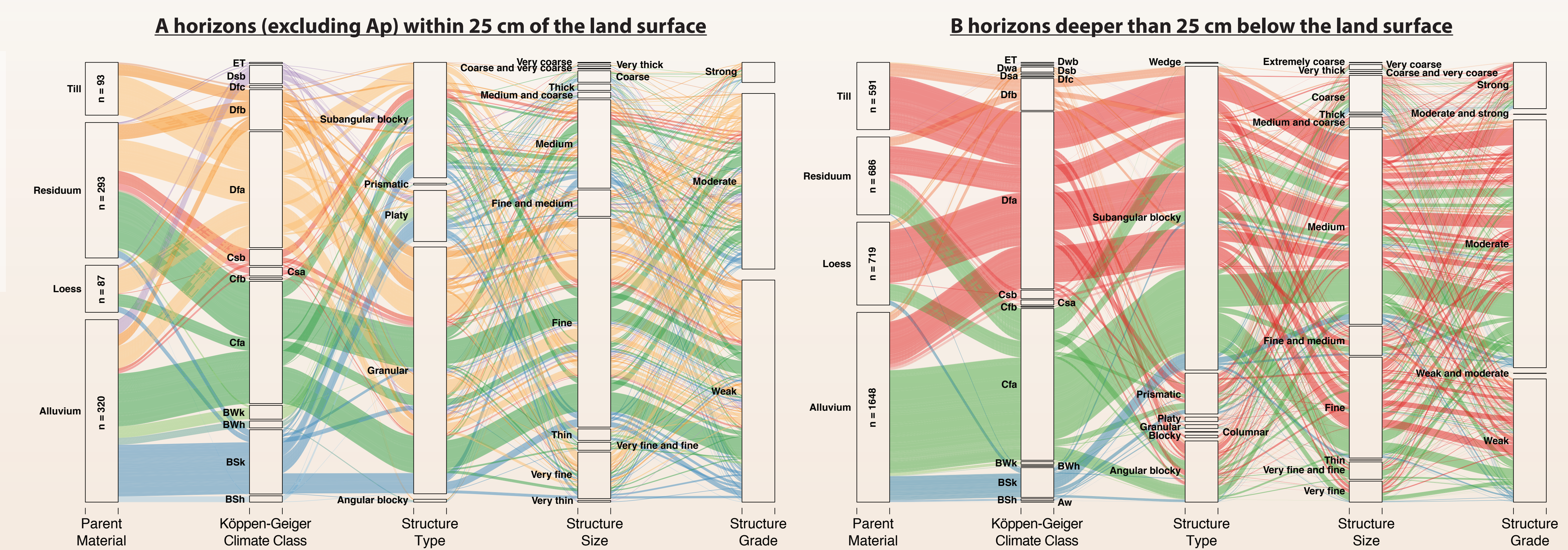
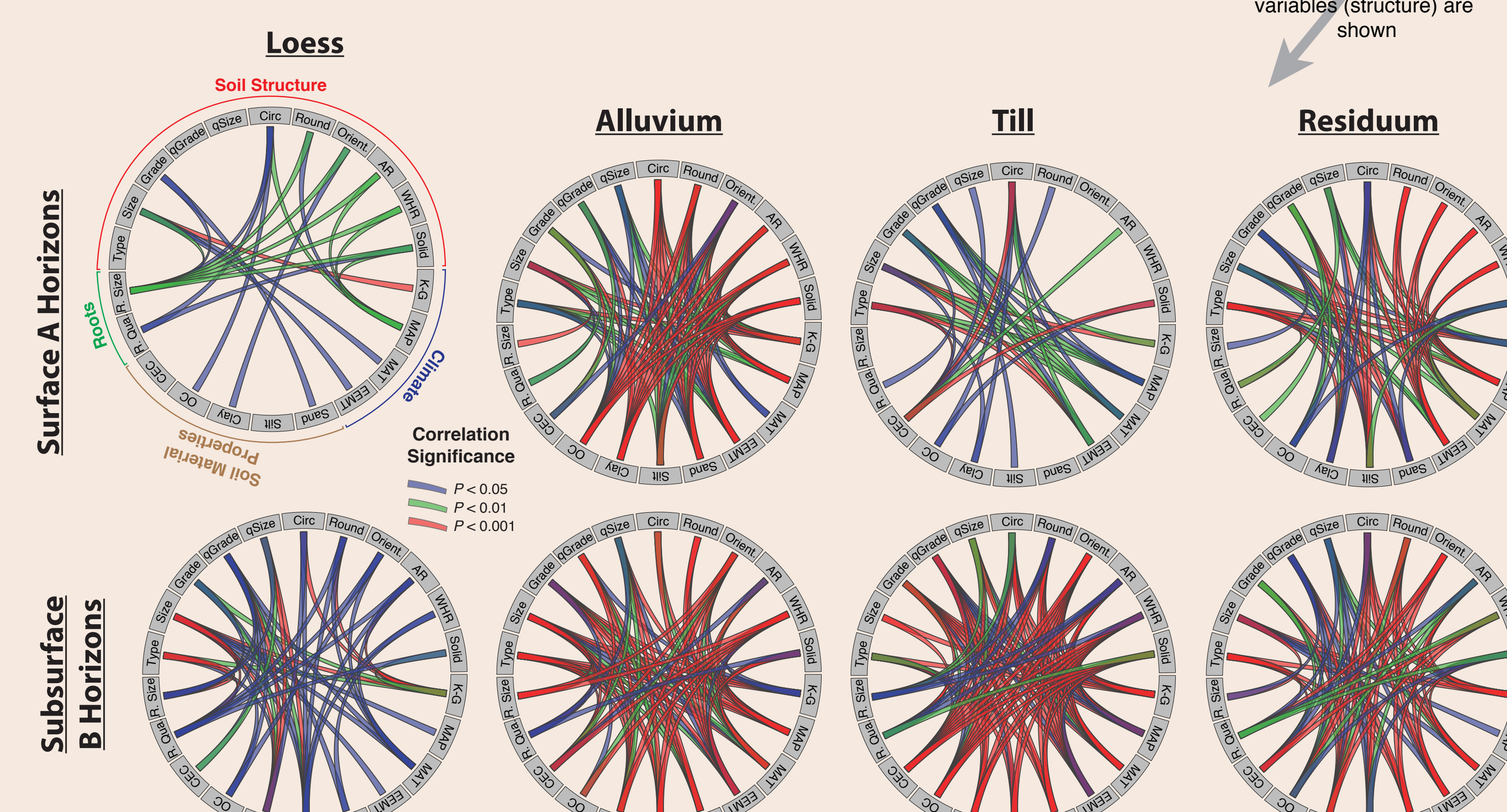
Shape parameters for each type were assigned following Mohammed et al. (2016)

Effective Energy and Mass Transfer (Rasmussen and Tabor, 2007)

$$EEMT [MJ m^2 y^{-1}] = 347.134e^{-0.5 \left[\left(\frac{MAT-21.5}{-10.1} \right)^2 + \left(\frac{MAP-4412}{1704} \right)^2 \right]}$$

Equation for quantifying grade, size, or type in horizons with compound (C) and non-compound (NC) structures

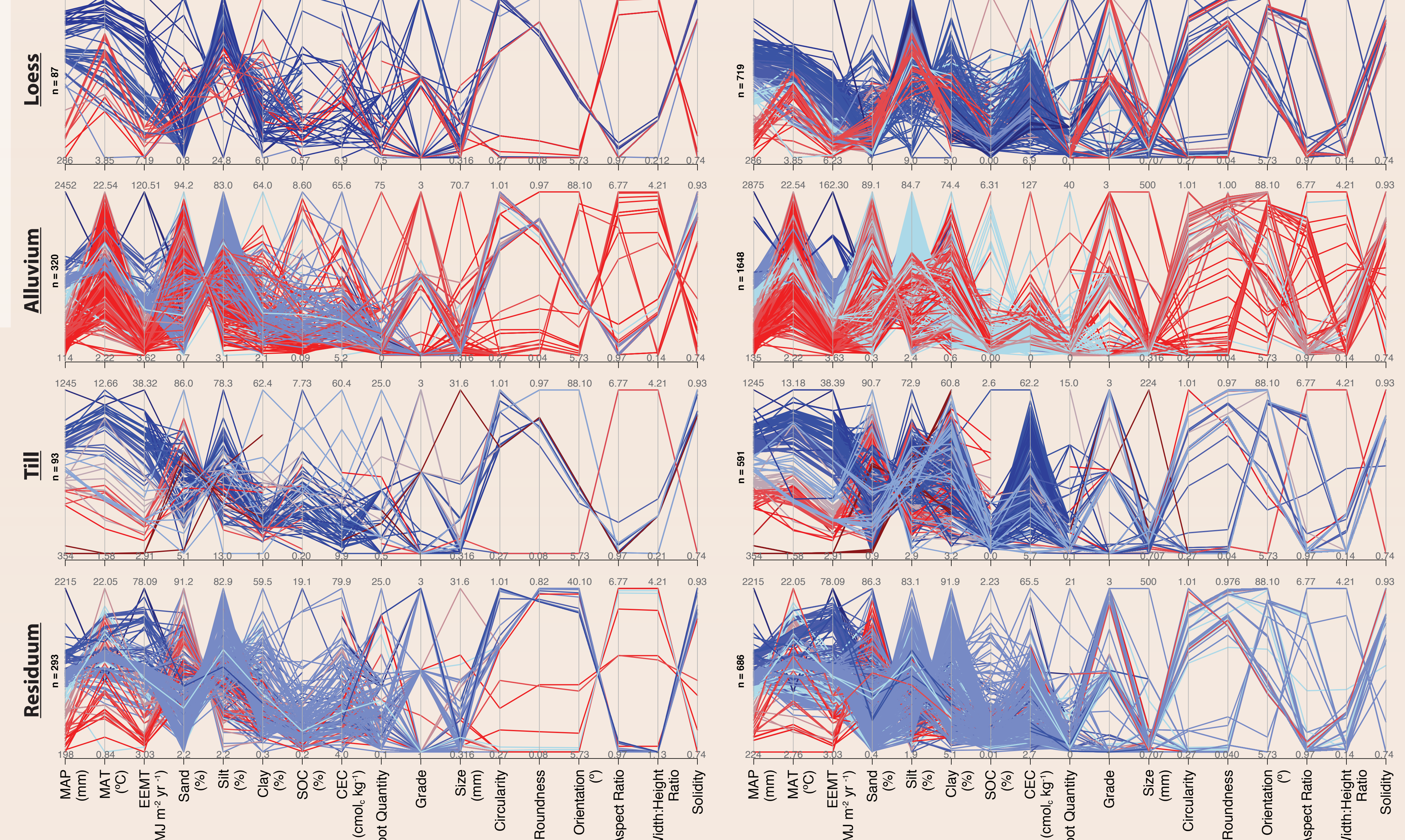
$$\sigma = \begin{cases} \frac{1}{A_{x,s}} \sum_{k=1}^m (x_k^2 AR_k f \sigma_k) & \text{only NC} \\ \frac{1}{A_{x,s}} \left[\frac{1}{Q} \sum_{j=1}^m (q_j \sigma_j) + \sum_{k=1}^m (x_k^2 AR_k f \sigma_k) \right] & \text{either C or both C and NC} \end{cases}$$



Categorical parallel coordinate plots colored by climate class

Data Analyses and Results

Quantitative variable parallel coordinate plots colored by MAP



Circular plots of variables used in this study showing correlation indices as linkages between the variables colored by significance level; only linkages between x variables (climate, material properties, and vegetation) and y variables (structure) are shown

Conclusions and Future Work

Quantitative structural parameters for grade, size, and type (i.e., shape parameters) show significant relationships to climate—especially, MAP and EEMT—for the four parent material types studied. For example, parallel coordinate plots show clear separation on shape parameters by MAP. The clarity of the relationship between climate and structure is reduced in horizons below 25 cm likely due to confounding effects from soil material properties as shown in the circular correlation plots. Roots showed significant relationships with structure suggesting that future climate change may influence structure through ecohydrological changes. Future work will examine these relationships with machine learning techniques and structural equation modeling to understand the role, importance, and interactions of the variables studied on structure.

Acknowledgements

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References

Mohammed et al. 2016: [SSSAJ] in press
 Rasmussen and Tabor 2007: [SSSAJ] 71:1719–1729