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Future Spatiotemporal Patterns of Corn Soil Organic Carbon Use Efficiency in the U.S. Corn Belt Insights from CMIP5 Projections (2010-2099)

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Objective

To assess how future climate change will affect corn yield and soil organic carbon (SOC) dynamics in the U.S. Corn Belt.

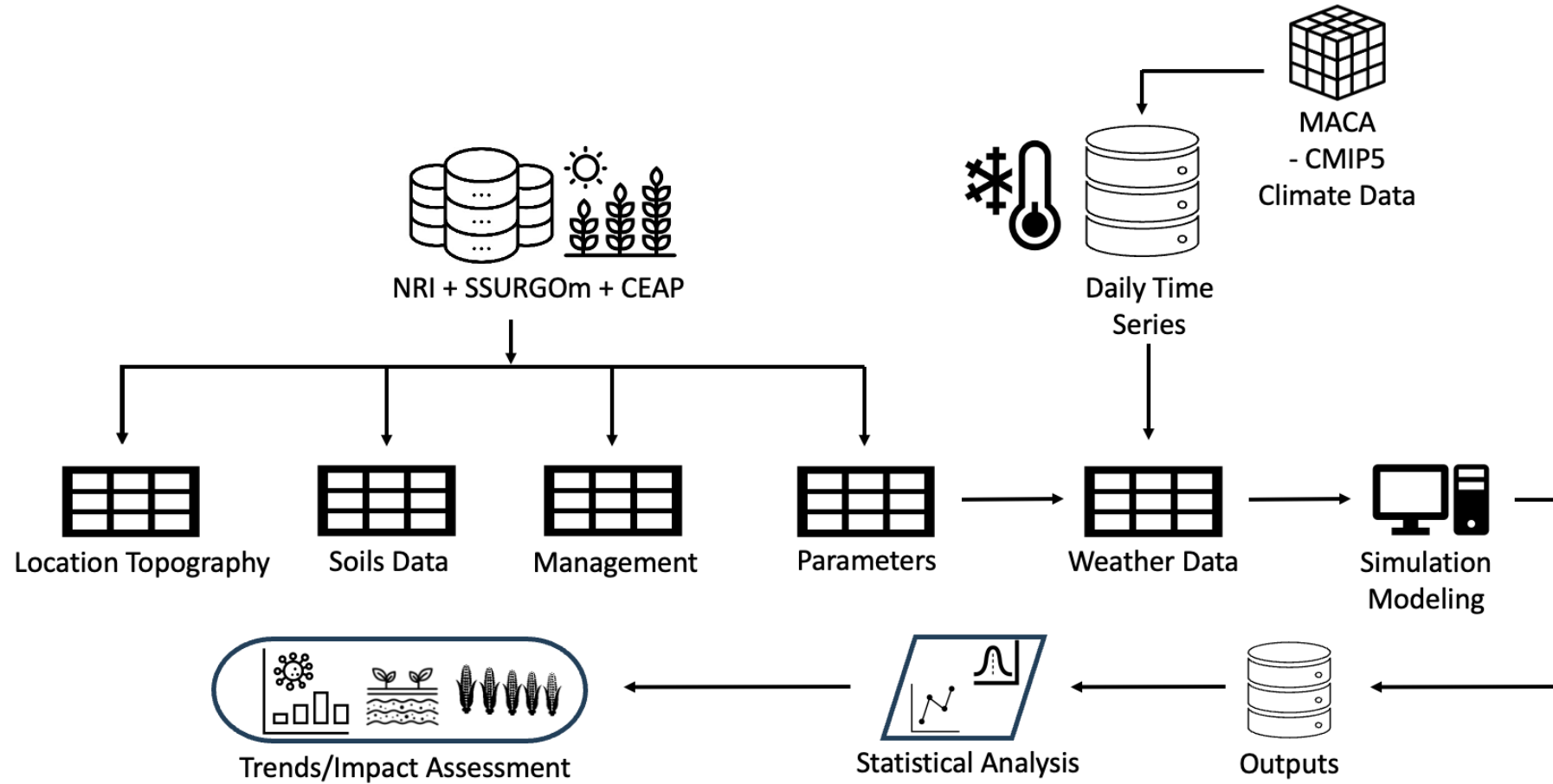


We employ the process-based APEX model as a simulation framework

Linking climate drivers to yield and SOC responses over a century.

We integrated temporal trend analysis, combining Mann–Kendall test and Sen’s slope estimator with spatial autocorrelation methods (Moran’s I and Getis–Ord G_i^*) To characterize direction and magnitude of long-term changes and spatial clustering.

MODELING FRAMEWORK



Location & Duration



Spatial scale:



Corn Belt



1020 Sites



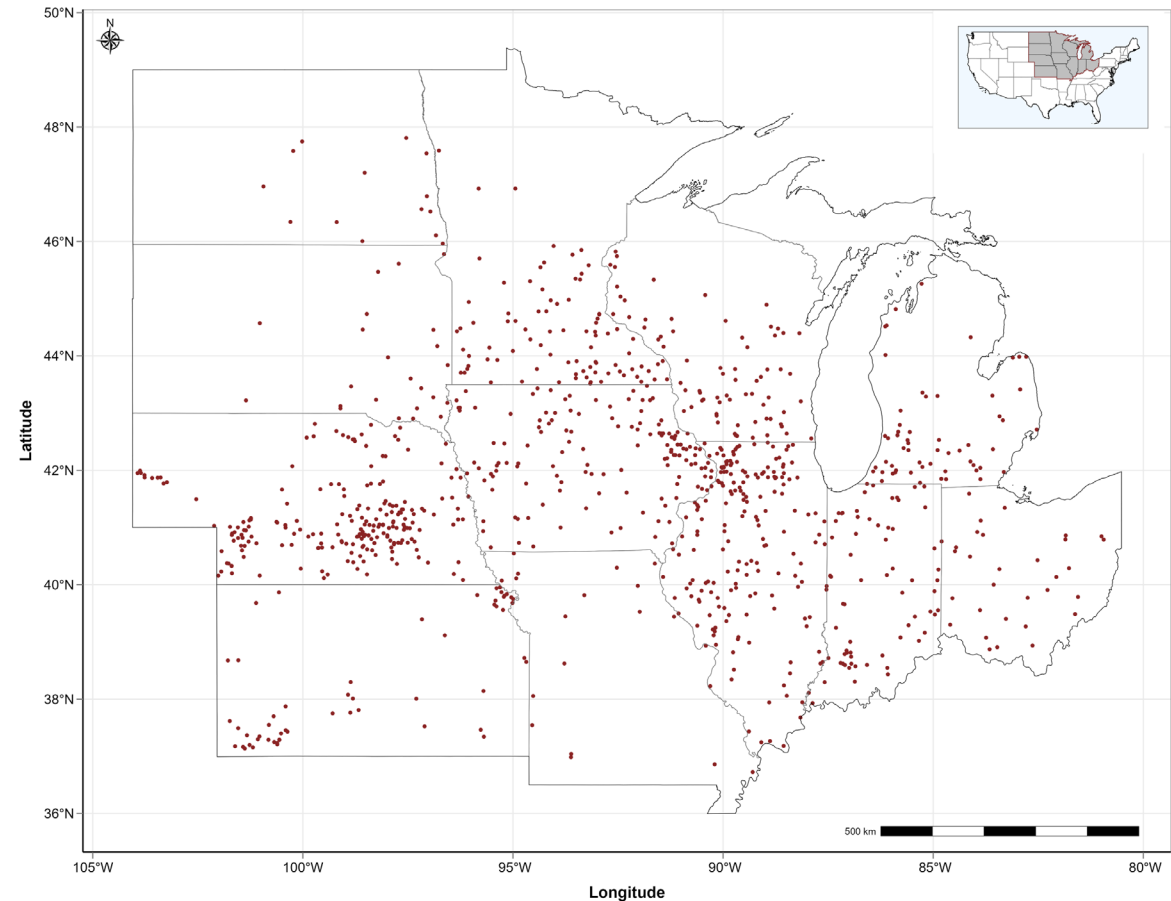
Time period



Historical: 1971–2005



Future: 2010 to 2099



G C M

GLOBAL CIRCULATION MODELS

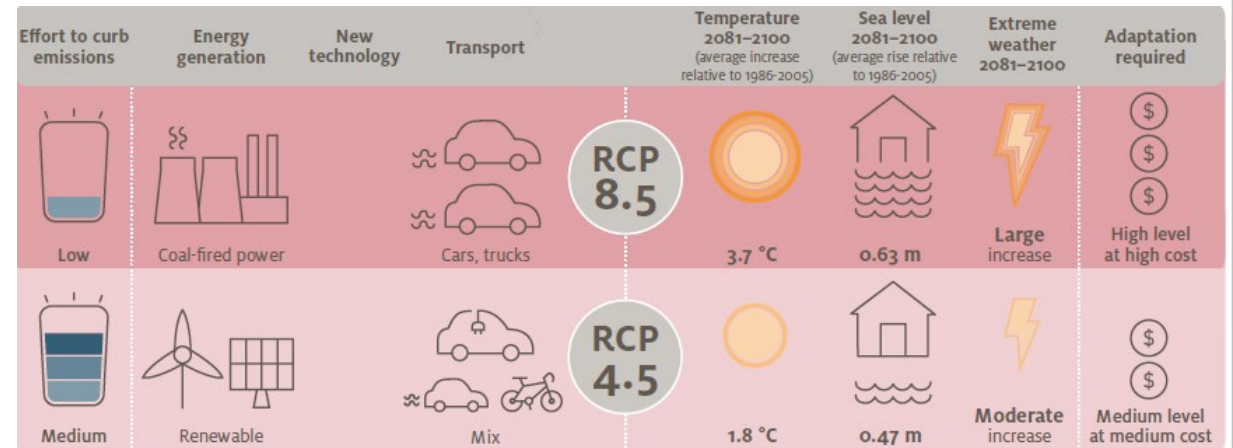
MACAv2 - METDATA

University of Idaho

HadGEM2-ES365: Met Office Hadley Centre (UK)
PSL-CM5A-MR: Institut Pierre Simon Laplace (FR)
MRI-CGCM3: Meteorological Research Institute (JP)
NorESM1-M: Norwegian Climate Center (NO)
CCSM4: National Center of Atmospheric Research (US)
CNRM-CM5: National Centre of Meteorological Research
and European Centre for Advanced
Research and Training (FR)

R C P

Representative Concentration Pathways



S O C U E

SOC Use Efficiency (kg yield / kg C):

$$SOCUE_t = \frac{Y_t}{\Delta SOC_t}$$

How many kilograms of corn grain you get for every kilogram of carbon that your soil is above or below its historical average

- For every 1 kg of carbon surplus, you get X kg of yield (+).
- For every 1 kg of carbon deficit, you are extracting X kg of yield (-).

Change in SOC from historical mean (Mg C/ha):

$$\Delta SOC_t = SOC_t - \overline{SOC}_{h(1985-2016)}$$

Temporal Analysis

The **Mann–Kendall** trend test is a non-parametric statistical method used to detect monotonic trends in time series data.

$$S = \sum_{i=1}^{n-1} \sum_{j=i+1}^n \text{sgn}(x_j - x_i)$$

- A large positive S value indicates an upward trend.
- A large negative S value indicates a downward trend.
- An S value close to zero suggests no monotonic trend.

The Mann-Kendall statistical significance of the trend is indicated through the presence of the dots

Temporal Analysis

Sen's slope is a non-parametric method used to estimate the magnitude (rate) of a linear trend in a time series.

Calculate Slopes

$$m_{ij} = \frac{x_j - x_i}{t_j - t_i}$$

Find the Median of These Slopes:

$$m = \text{Median}(m_{ij}) \quad \text{for } 1 \leq i < j \leq n$$

- Positive Sen's slope indicates an increasing trend.
- Negative Sen's slope indicates a decreasing trend.
- Slope of zero indicates no trend.

The value of the Sen's slope gives the quantitative **rate of change** per decade (color scale).

Spatial Analysis

Global Moran's I: Measures the overall spatial autocorrelation across an entire area.

$$I = \frac{N}{\sum_{i=1}^N \sum_{j=1}^N w_{ij}} \frac{\sum_{i=1}^N \sum_{j=1}^N w_{ij} (x_i - \bar{x})(x_j - \bar{x})}{\sum_{i=1}^N (x_i - \bar{x})^2}$$

Ranges from -1 to +1

- $I > 0$: Positive Spatial Autocorrelation (Clustering)
- $I < 0$: Negative Spatial Autocorrelation (Dispersion/Checkerboard Pattern)
- $I \approx 0$: No Spatial Autocorrelation (Random Distribution)

- N : The number of spatial units (sites/hexagons).
- x_i : The value of the variable at location i (e.g., Sen's slope of SOCUE).
- $\bar{x}$: The mean of the variable across all locations.
- w_{ij} : The spatial weight between location i and location j . This defines what "neighbor" means (e.g., shared border, within a certain distance).
- $\sum_{i=1}^N \sum_{j=1}^N w_{ij}$: The sum of all spatial weights (often denoted as S_0).



Spatial Analysis

Getis–Ord Gi* is a local spatial statistic used to identify statistically significant areas where high or low values cluster spatially given a specified neighborhood structure.

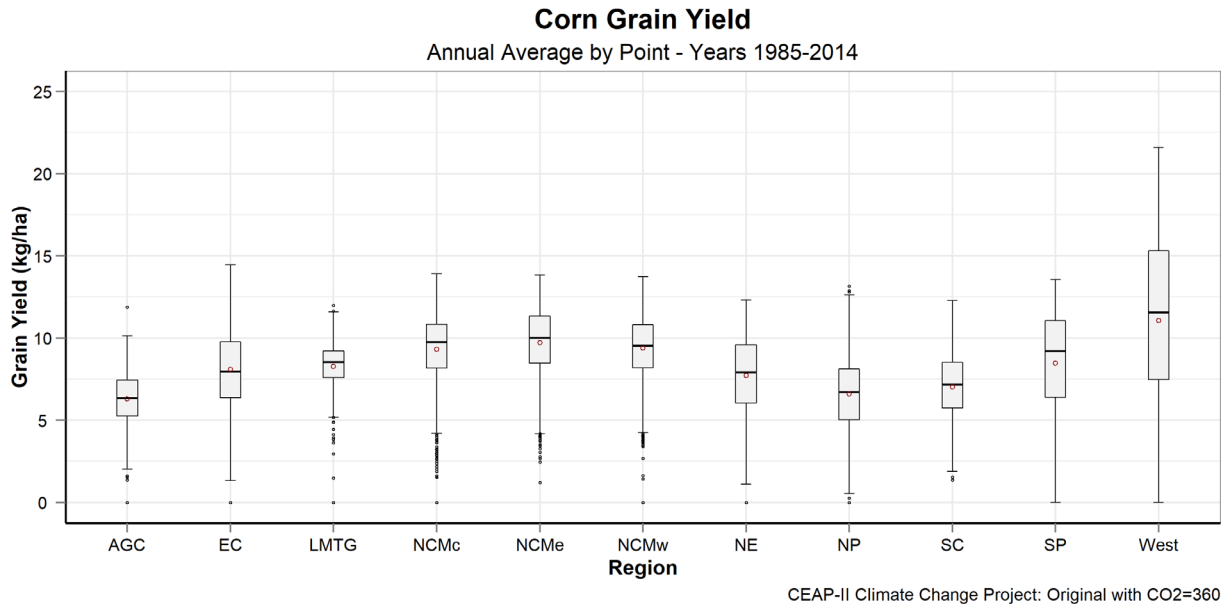
$$G_i^* = \frac{\sum_{j=1}^n w_{i,j} x_j - \bar{X} \sum_{j=1}^n w_{i,j}}{S \sqrt{\frac{n \sum_{j=1}^n w_{i,j}^2 - \left(\sum_{j=1}^n w_{i,j} \right)^2}{n-1}}}$$

Positive significant Gi*: i is in a neighborhood of high values (hotspot).

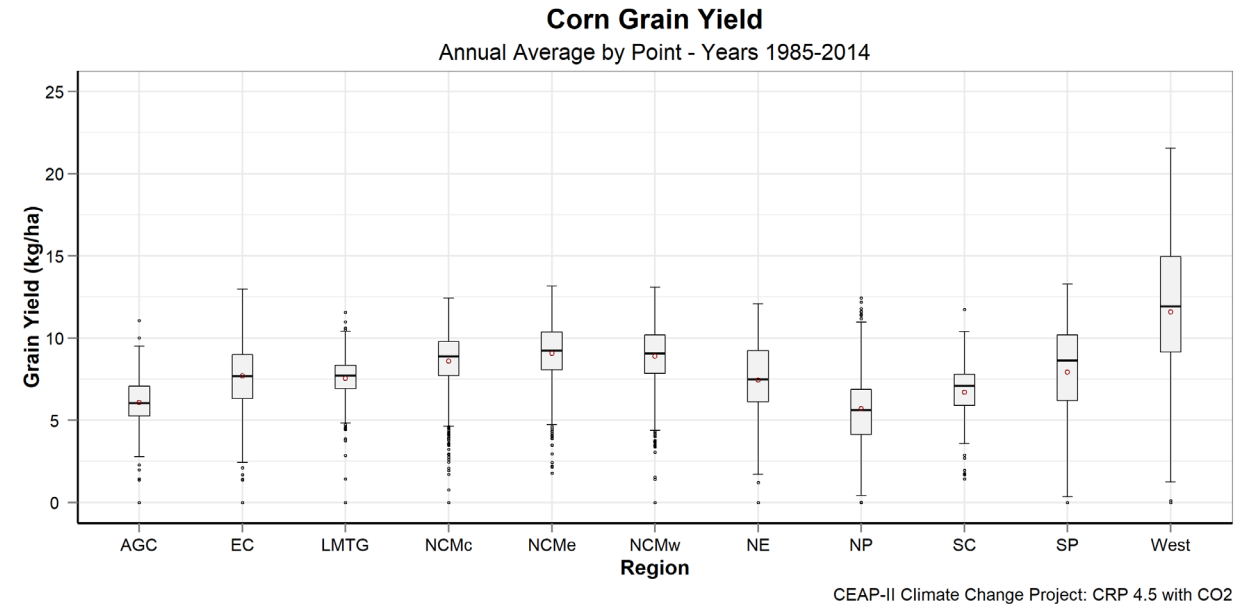
- Negative significant Gi*: i is in a neighborhood of low values (coldspot).
- Non-significant: local values are not unusually concentrated given the spatial structure.

For each location i, Gi* compares the sum of values in i's neighborhood (including i itself) to what would be expected under spatial randomness, accounting for the number/weights of neighbors.

Corn Yield Comparison by Region

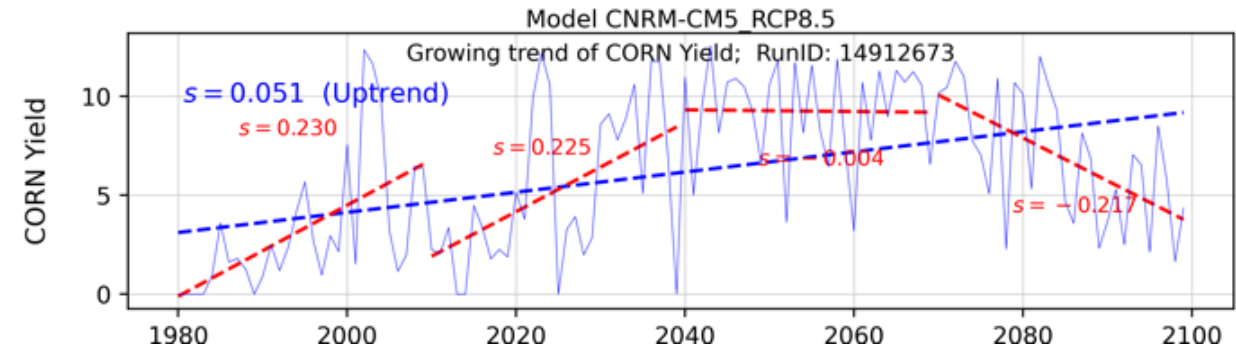
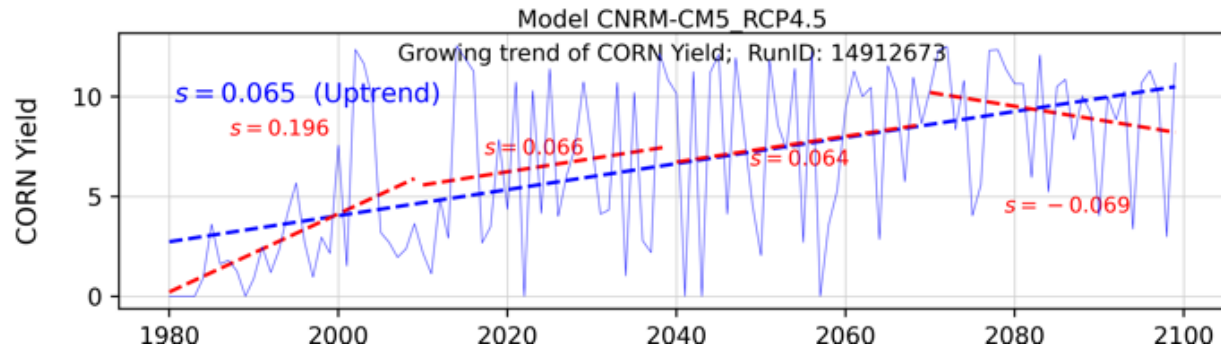
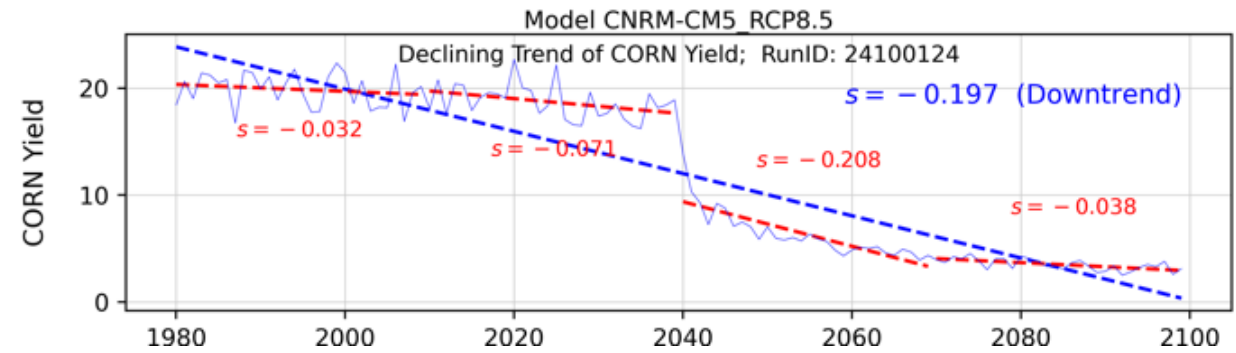
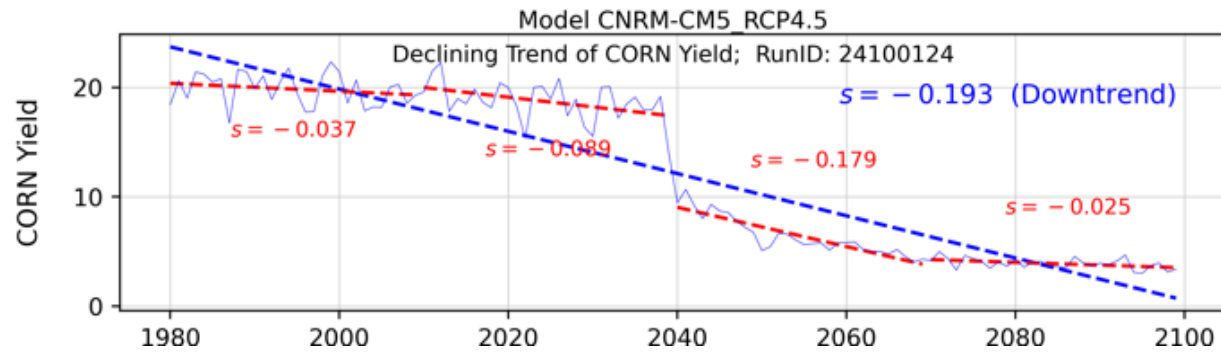


CEAP Weather Data



Historical MACA Data

Corn Yield Trends



CNRM-CM5 + RCP – 4.5

CNRM-CM5 + RCP – 8.5

S O C

RCP - 8.5

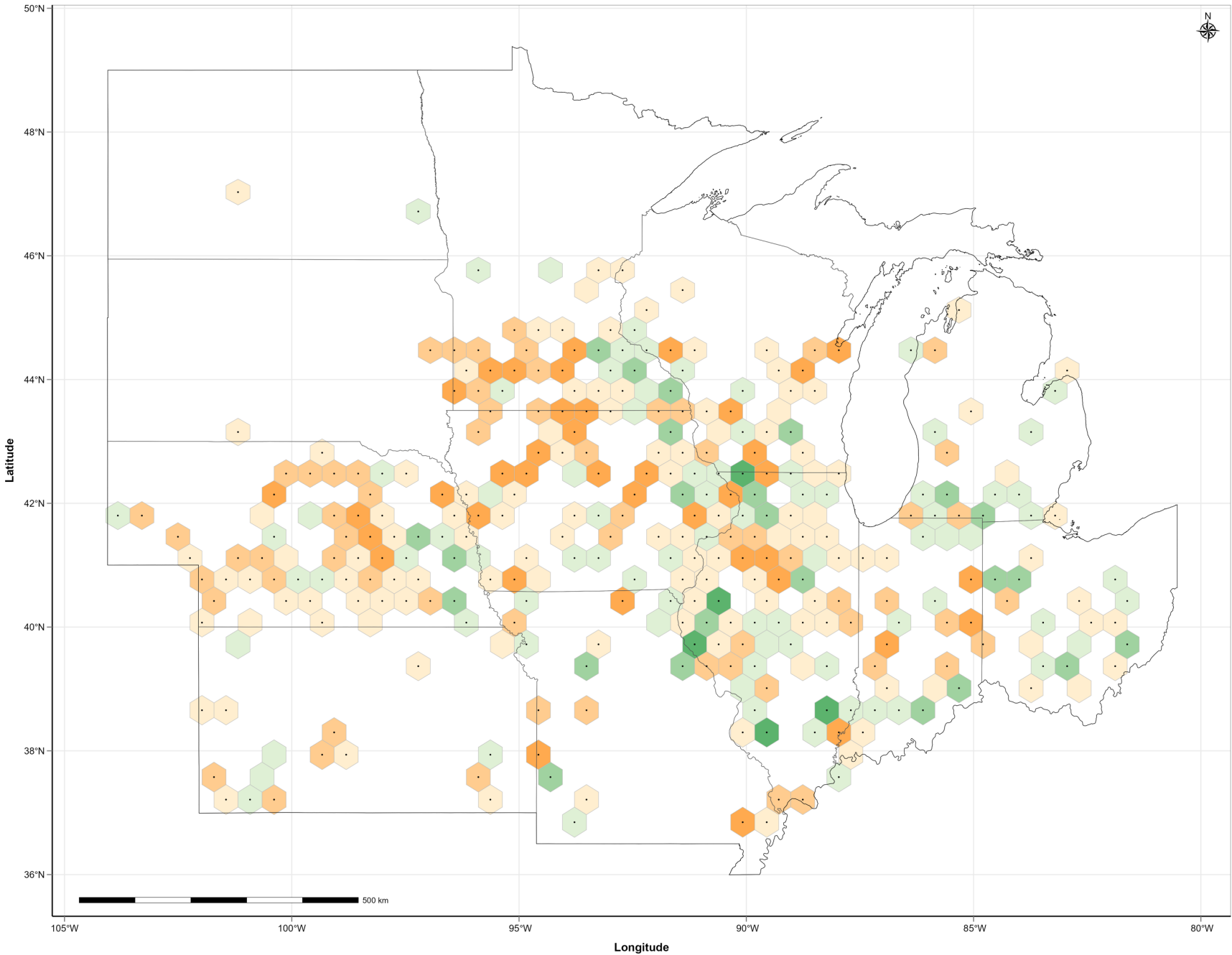
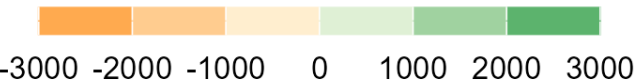
Mann-Kendall

Statistical significance of the trend

- Not Statistically significant
- Statistically Significant

Sen's Slope

Rate of change per decade
SOC (kg/ha per Decade)



S O C

RCP - 4.5

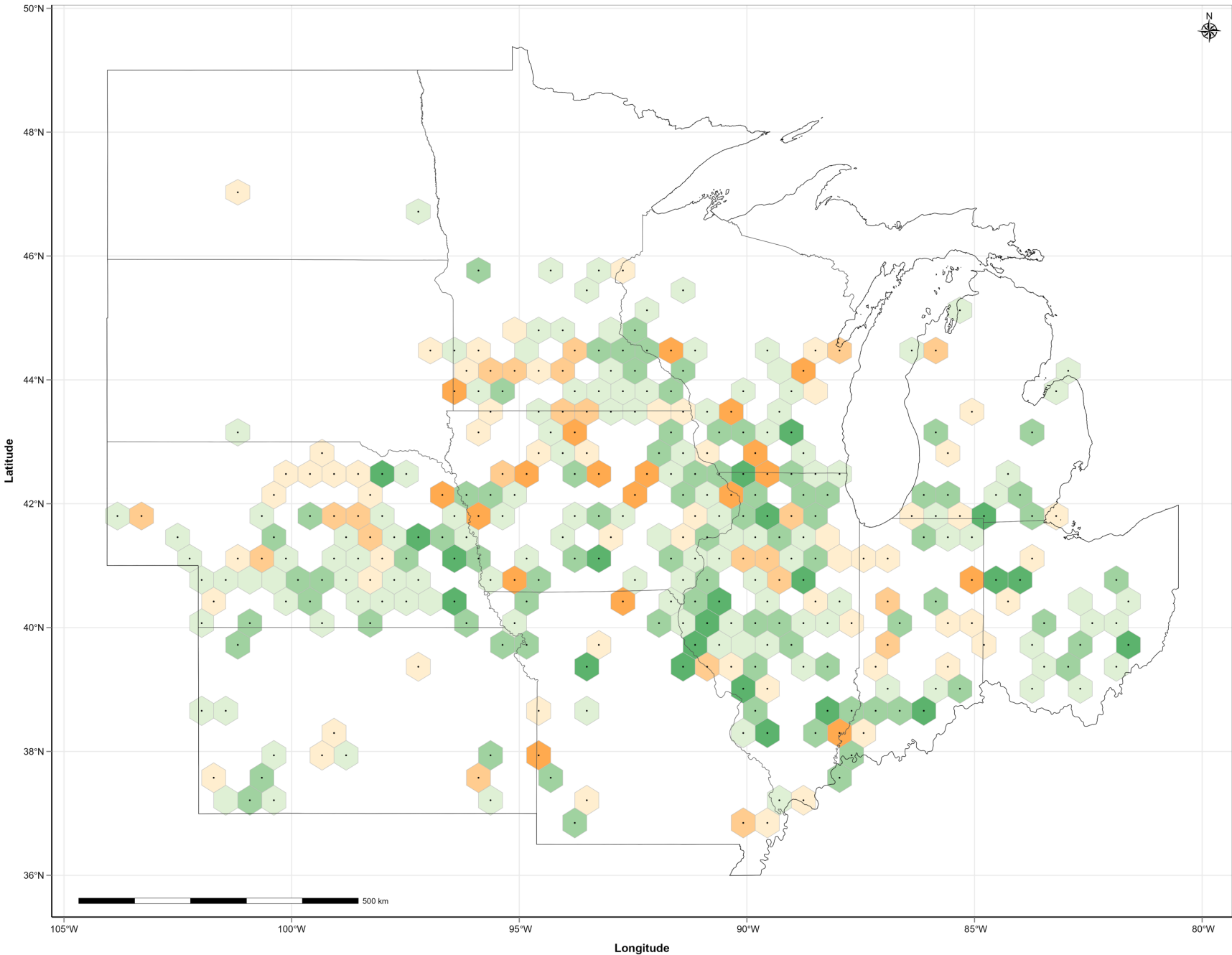
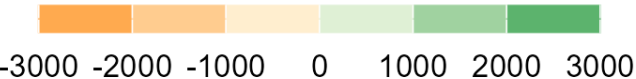
Mann-Kendall

Statistical significance of the trend

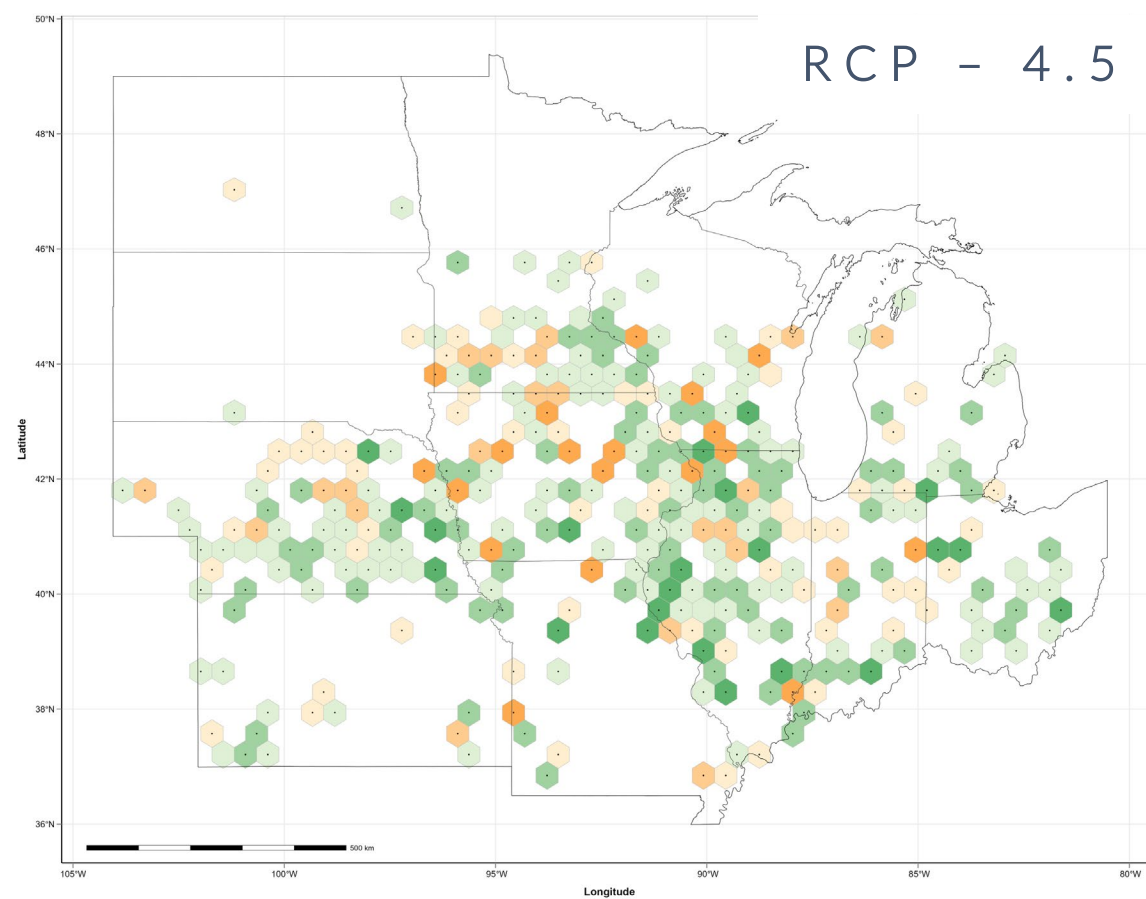
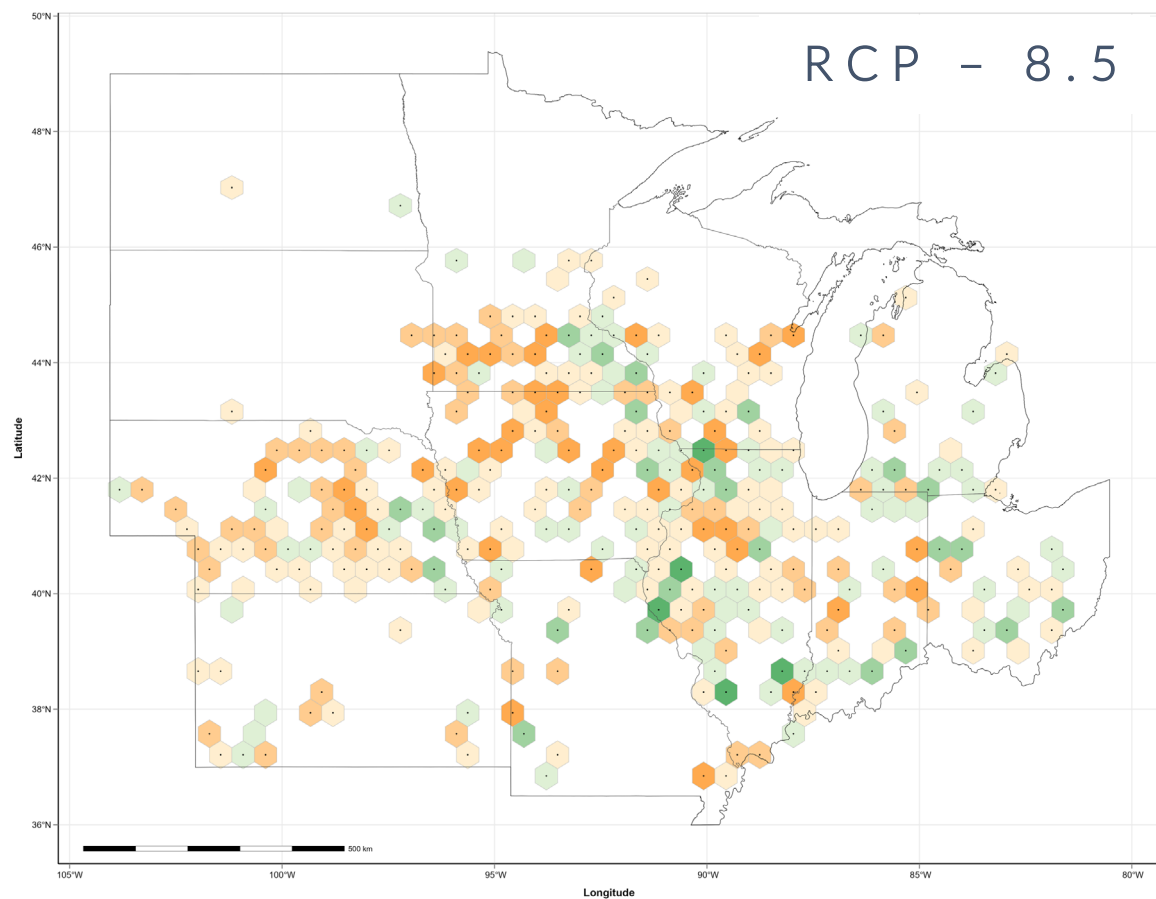
- Not Statistically significant
- Statistically Significant

Sen's Slope

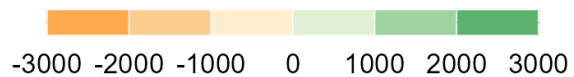
Rate of change per decade
SOC (kg/ha per Decade)



SOC



SOC (kg/ha per Decade)



Introduction

Methods

Results

Conclusion

SOCUE

RCP - 8.5



Mann-Kendall

Statistical significance of the trend



Not Statistically significant

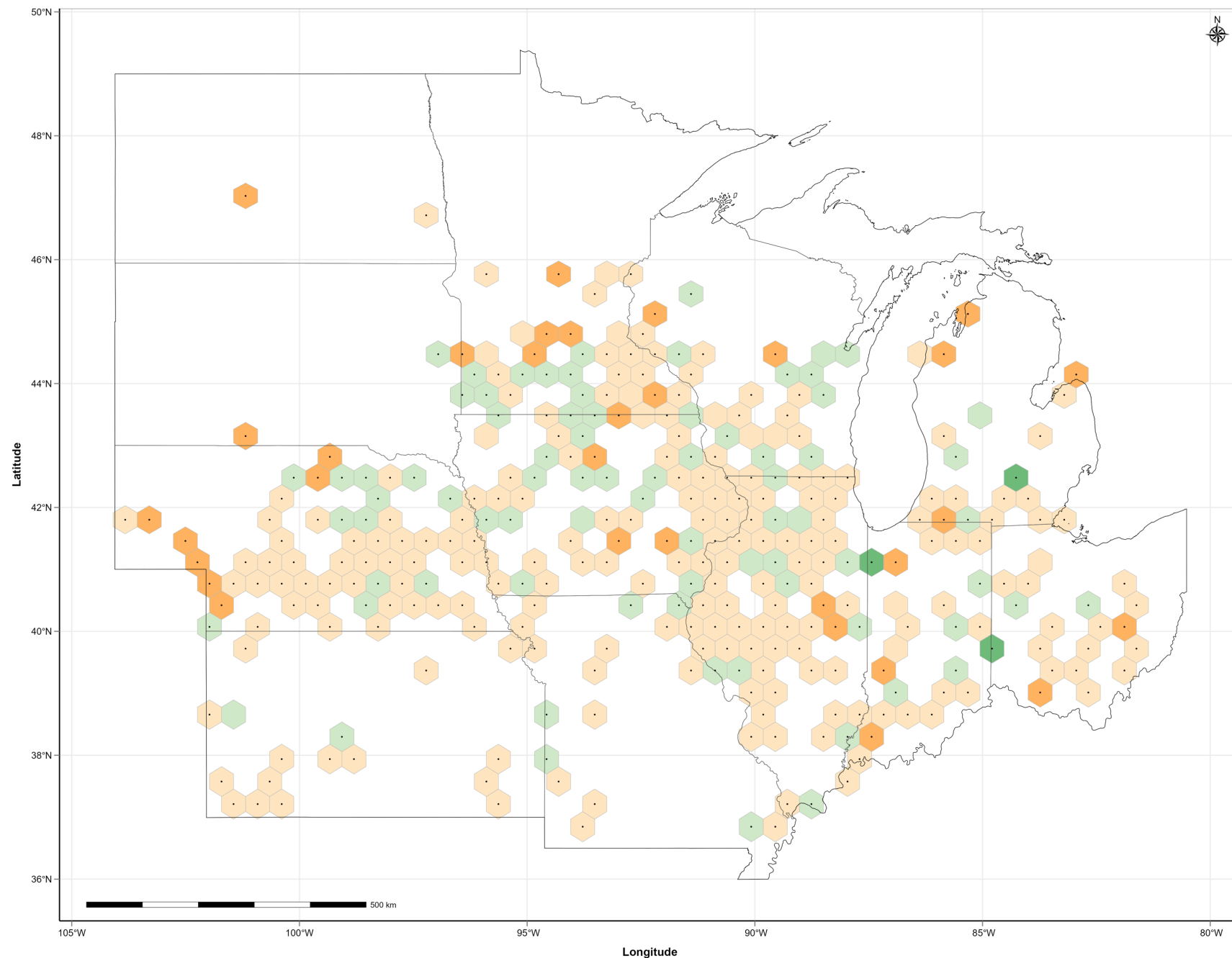
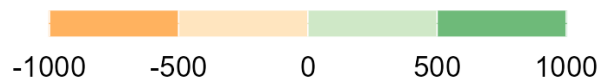


Statistically Significant

Sen's Slope

Rate of change per decade

SOCUE (kg Y/kg C per Decade)



Introduction

Methods

Results

Conclusion

SOCUE

RCP - 4.5



Mann-Kendall

Statistical significance of the trend



Not Statistically significant

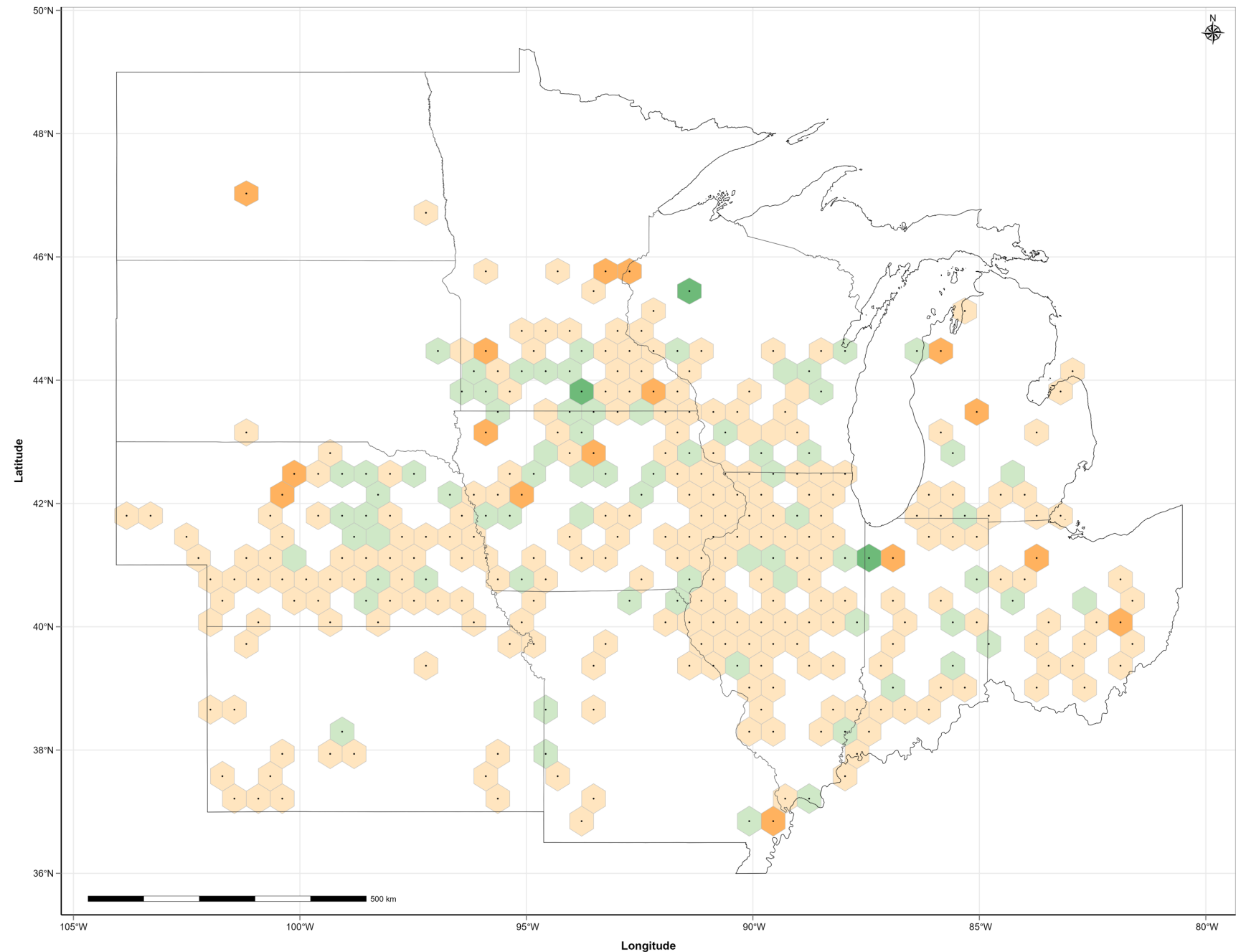
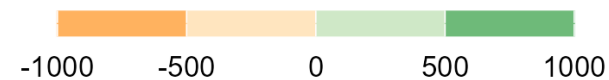


Statistically Significant

Sen's Slope

Rate of change per decade

SOCUE (kg Y/kg C per Decade)



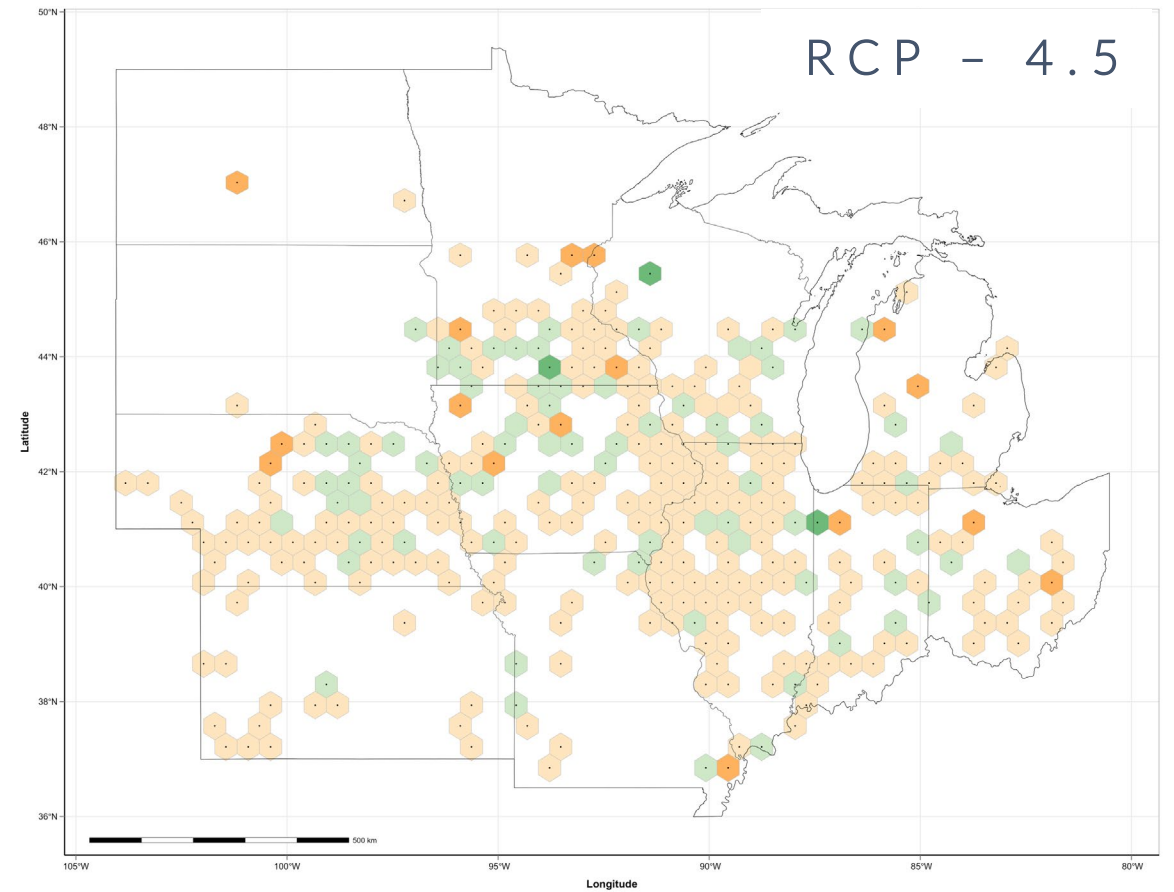
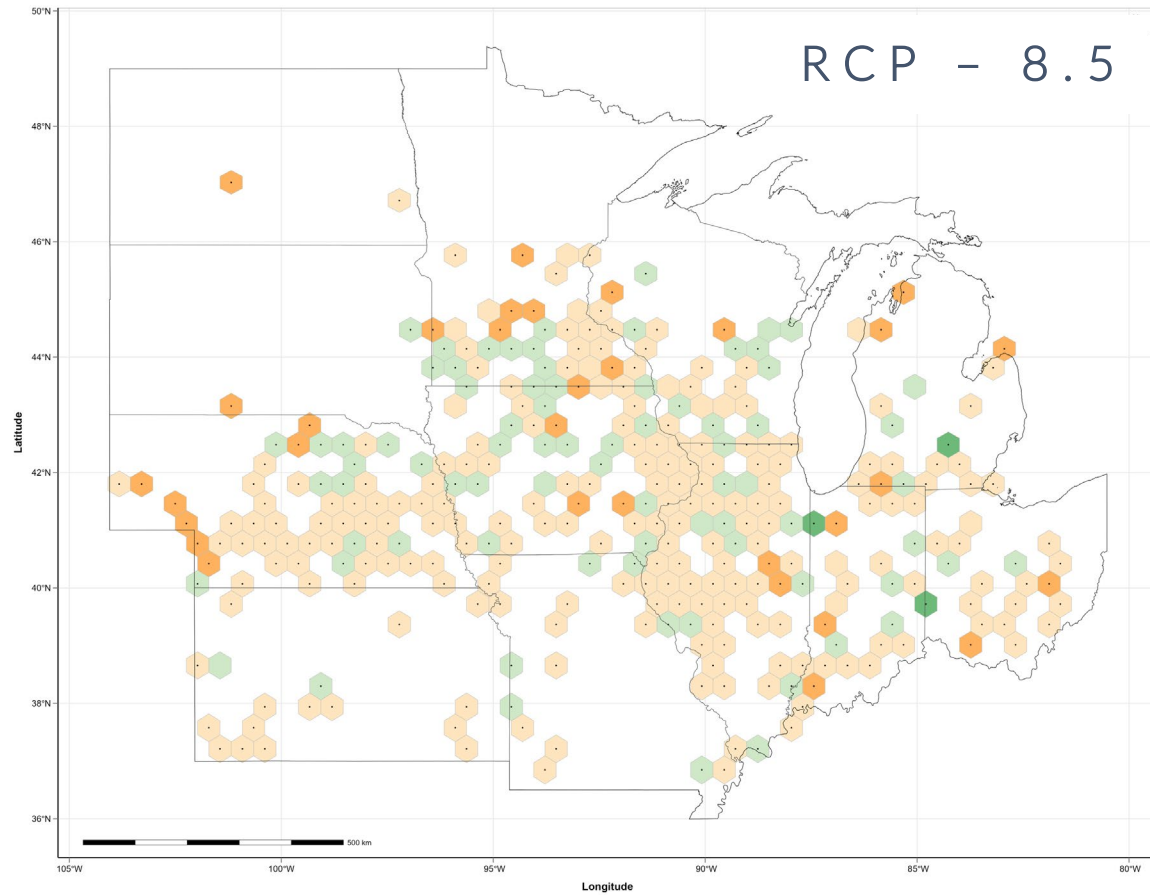
[Introduction](#)

[Methods](#)

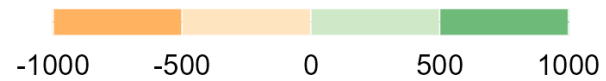
[Results](#)

[Conclusion](#)

SOCUE



SOCUE (kg Y/kg C per Decade)



Introduction

Methods

Results

Conclusion

S O C

RCP - 8.5

Global Moran's $I = -0.040$ ($p = 0.803$)

- $I > 0$: positive spatial autocorrelation
- $I < 0$: negative spatial autocorrelation
- $I \approx 0$: spatial randomness

Getis Ord G^*

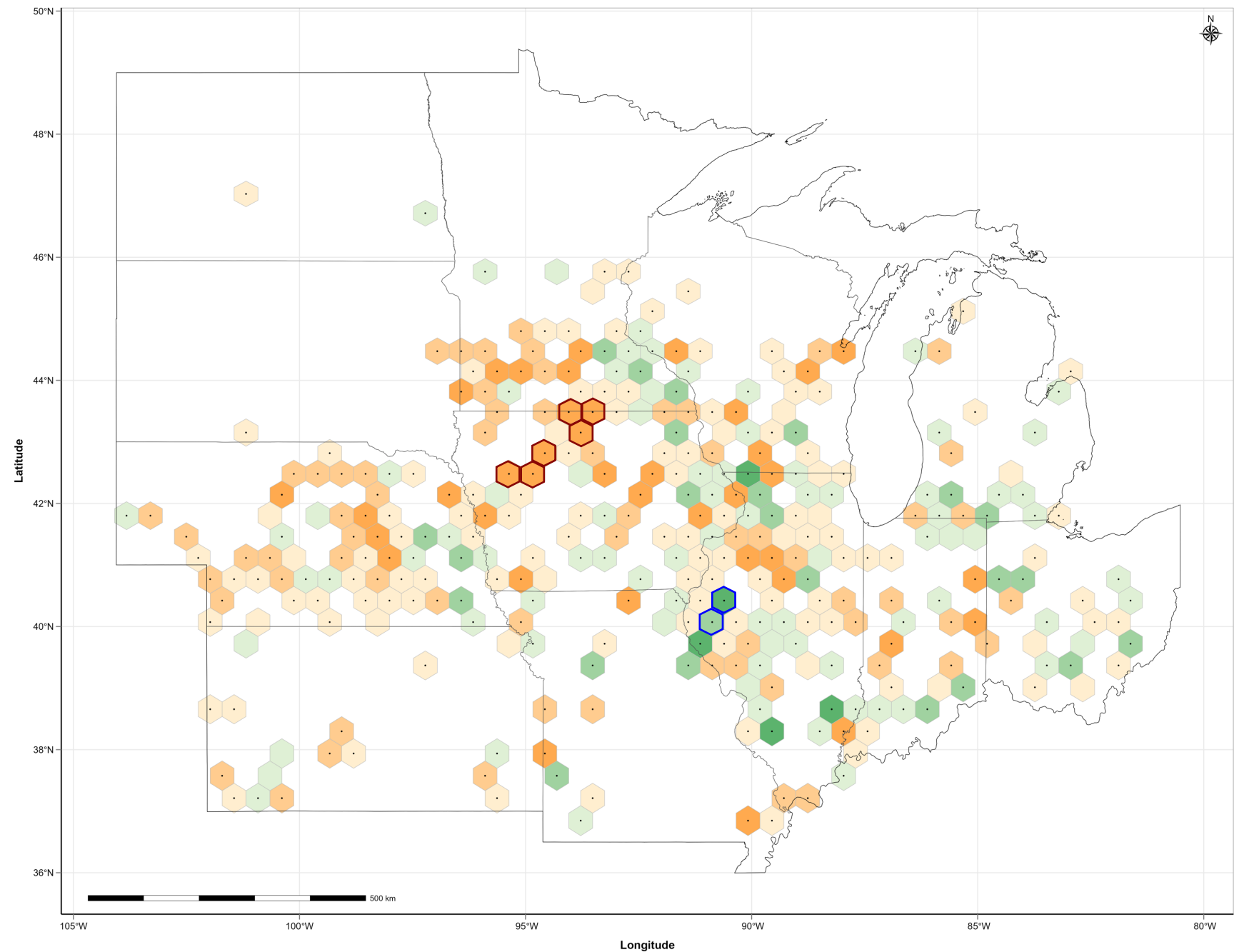
Identifies statistically significant hotspots



Hotspot (Significant positive G_i^*)



Coldspot (Significant negative G_i^*)



S O C

RCP - 4.5

Global Moran's I = 0.028 (p = 0.254)

- $I > 0$: positive spatial autocorrelation
- $I < 0$: negative spatial autocorrelation
- $I \approx 0$: spatial randomness

Getis Ord G^*

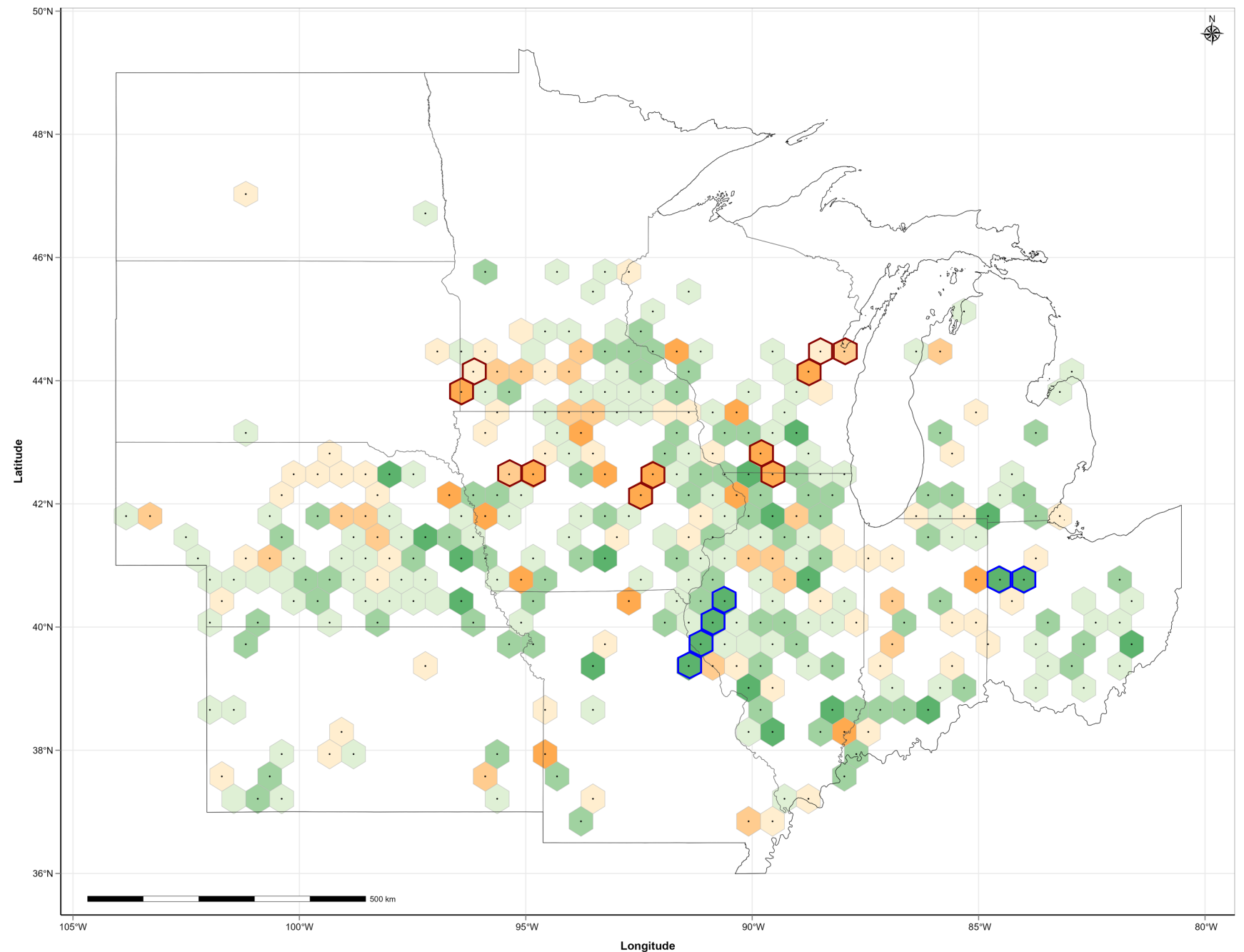
Identifies statistically significant hotspots



Hotspot (Significant positive G_i^*)



Coldspot (Significant negative G_i^*)



SOCUE

RCP - 8.5

Global Moran's $I = 0.045$ ($p = 0.126$)

- $I > 0$: positive spatial autocorrelation
- $I < 0$: negative spatial autocorrelation
- $I \approx 0$: spatial randomness

Getis Ord G^*

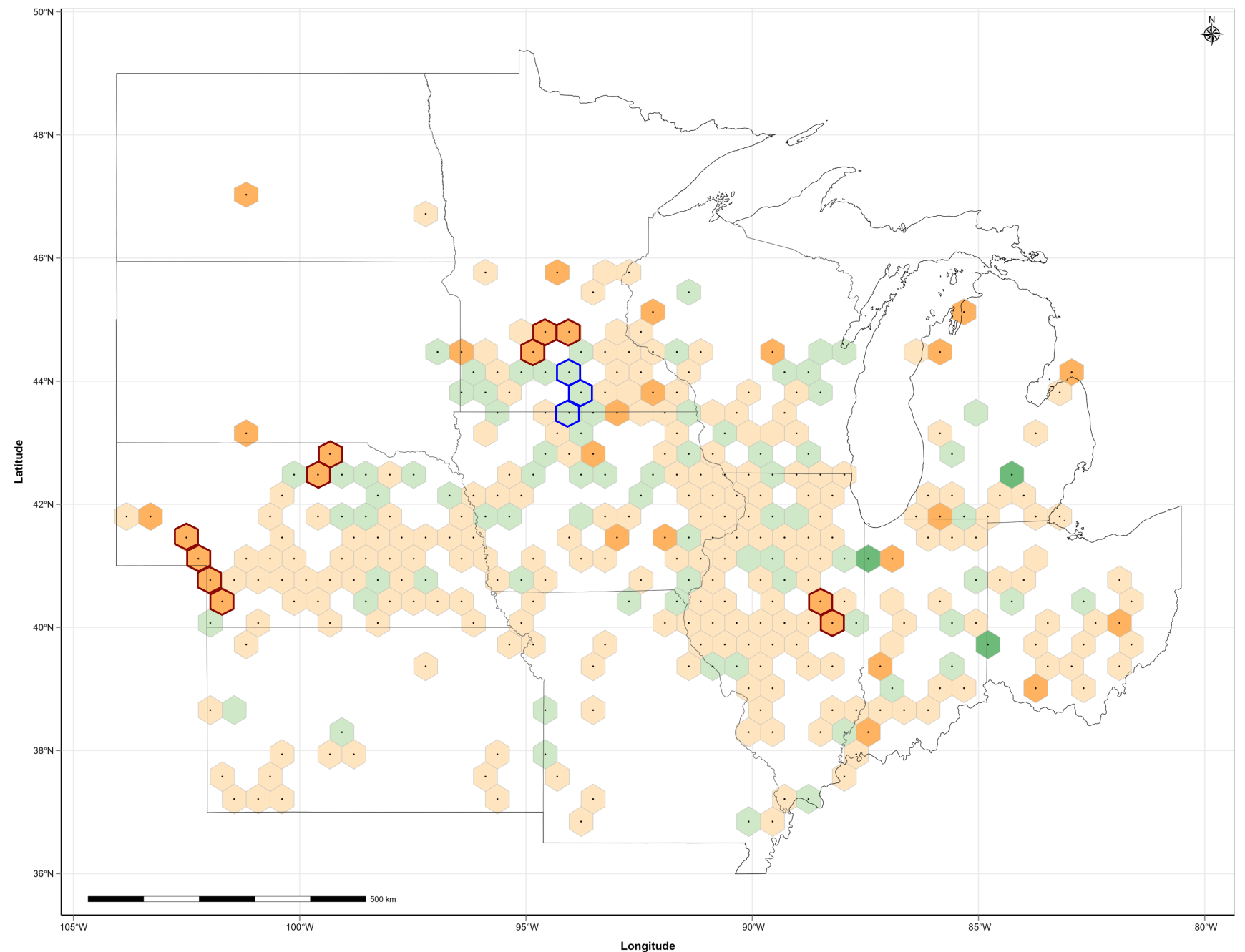
Identifies statistically significant hotspots



Hotspot (Significant positive G_i^*)



Coldspot (Significant negative G_i^*)



Introduction

Methods

Results

Conclusion

SOCUE

RCP - 4.5

Global Moran's $I = -0.028$ ($p = 0.713$)

- $I > 0$: positive spatial autocorrelation
- $I < 0$: negative spatial autocorrelation
- $I \approx 0$: spatial randomness

Getis Ord G^*

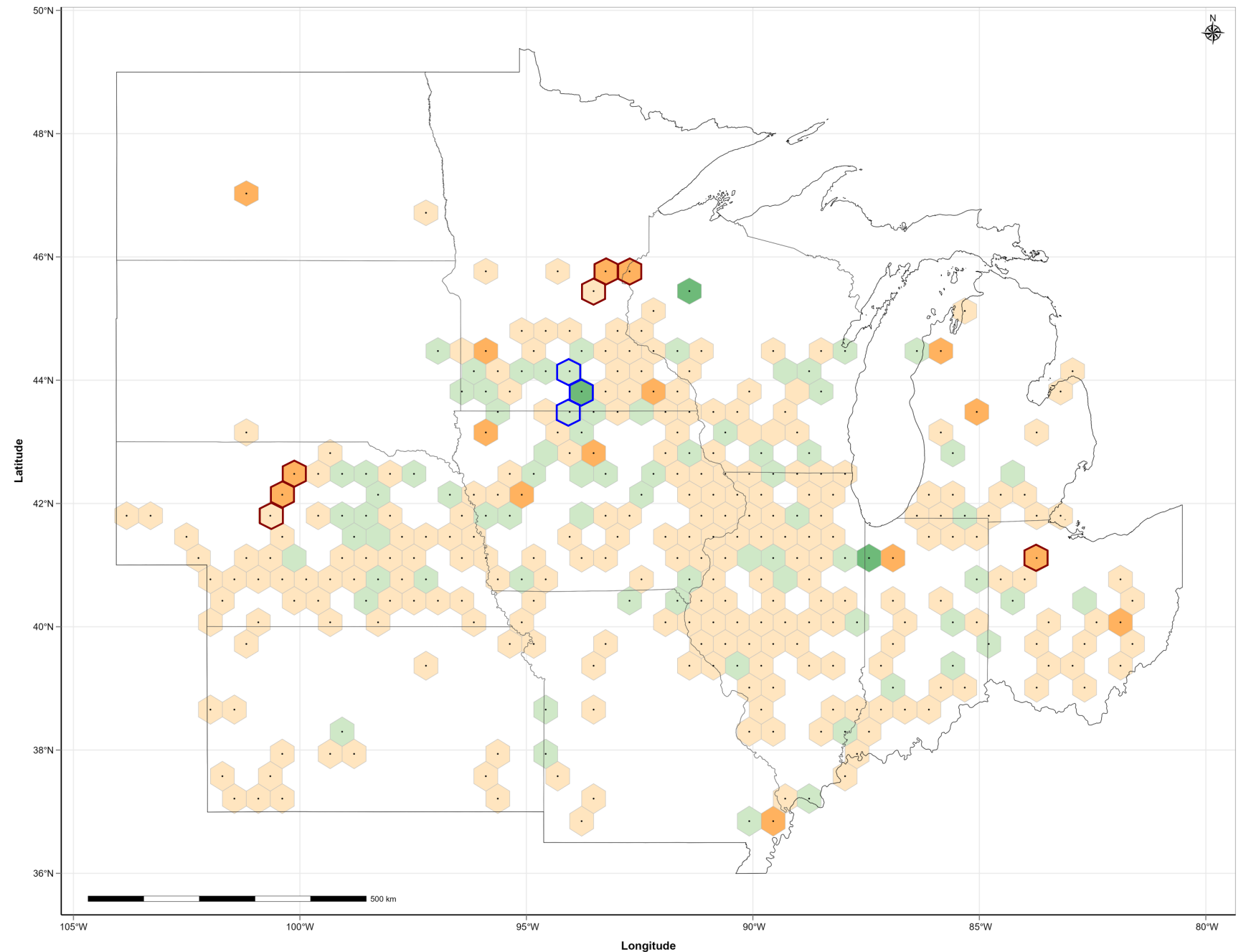
Identifies statistically significant hotspots



Hotspot (Significant positive G_i^*)



Coldspot (Significant negative G_i^*)



Introduction

Methods

Results

Conclusion

Take Home Message



1

We developed a robust, multi-model, spatio-temporal framework to quantify how crop yields respond to soil organic carbon (SOC) under climate change, using process-based model outputs, and site-by-site normalization within hexagon regions.



2

Yield–SOC coupling is spatially heterogeneous and climate-sensitive.



3

Large-scale global clustering may be weak, but local hotspots and coldspots emerge.



4

Using SOCUE non-parametric trend tests, and local spatial statistics provides a transparent, reproducible pathway to target interventions that could improve productivity and soil carbon resilience under future climates.



2025 Colorado - SWAT Conference
20-24 October - Fort Collins, Colorado, USA



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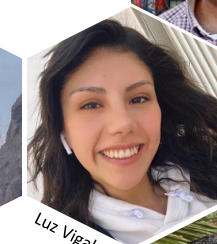
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Carlos Quezada



Hailey Schmidh



Luz Vigabriel



Luca Doro



Mousumi Deb



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NEXT



Assess socioeconomic changes



Management strategies



Cause & Effect analysis