

Are we modeling spatially varying processes or non-linear relationships?

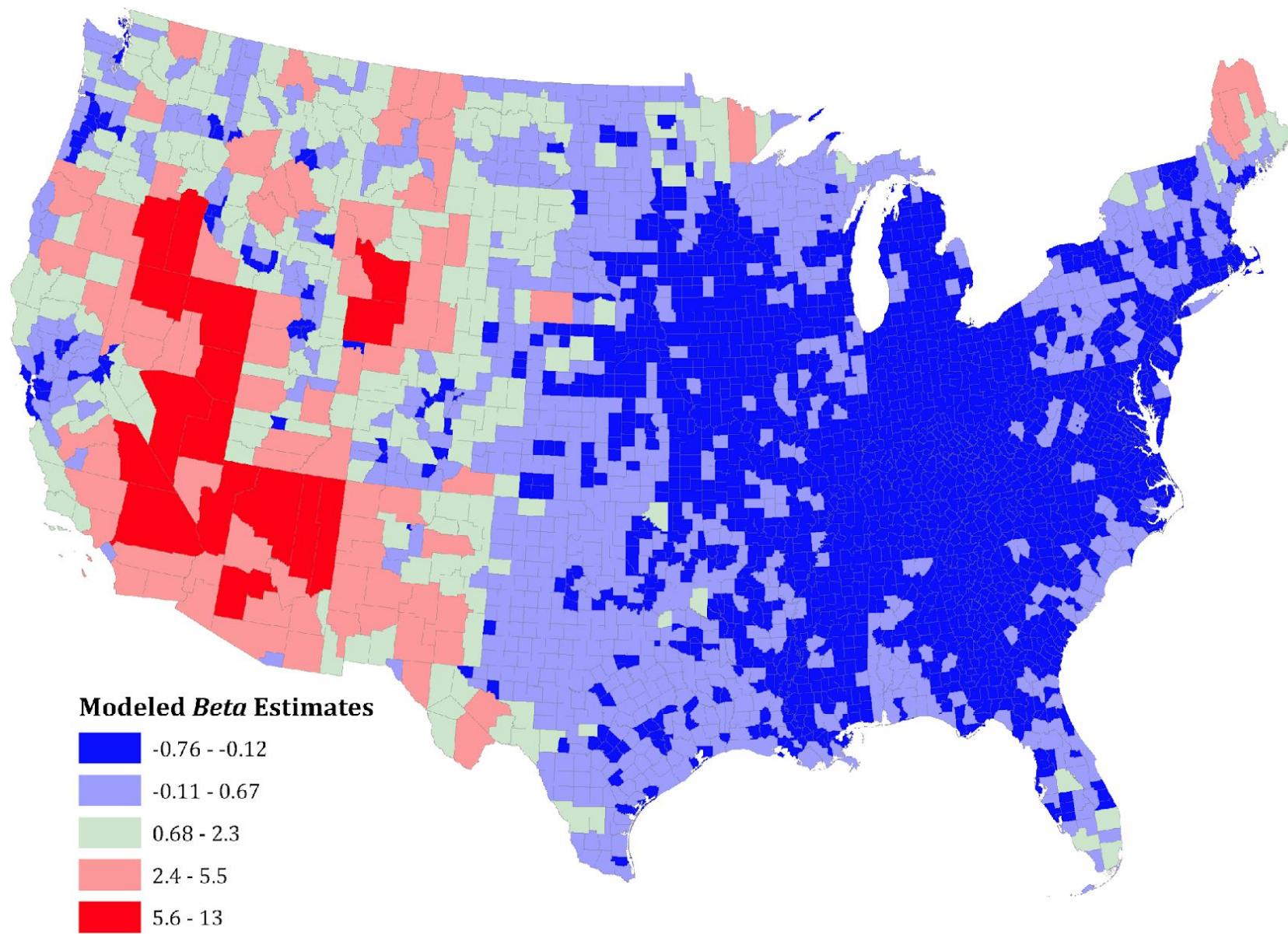
Mehak Sachdeva

Publication: Sachdeva M., Fotheringham A. S., Li Z., and Yu H., (2022) "Are we modeling spatially varying processes or non-linear relationships?" *Geographical Analysis*

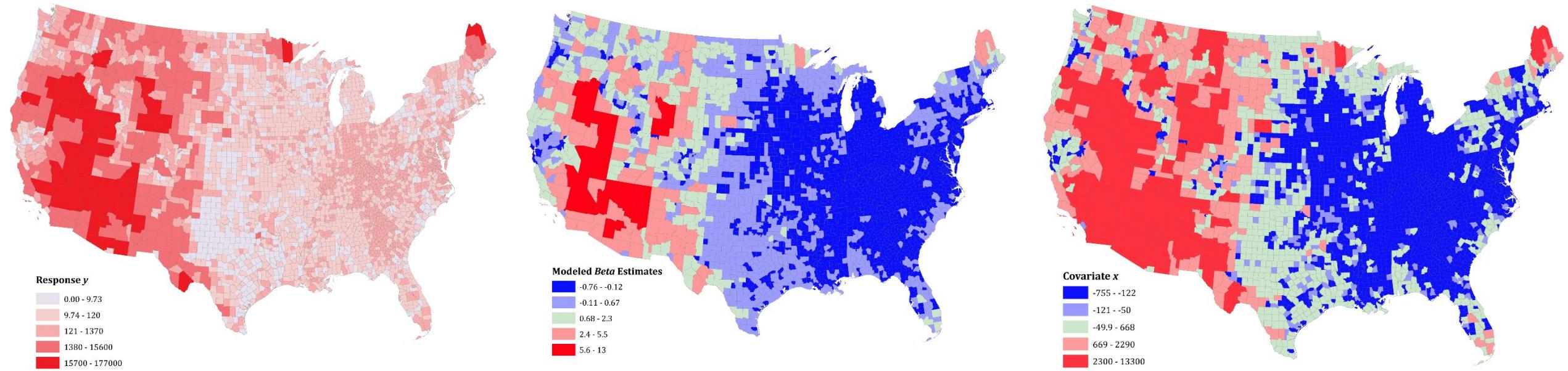
Structure

1. Problem Statement
2. Research Objective
3. Research Design
4. Intellectual Merits and Broader Impacts

1. Problem Statement

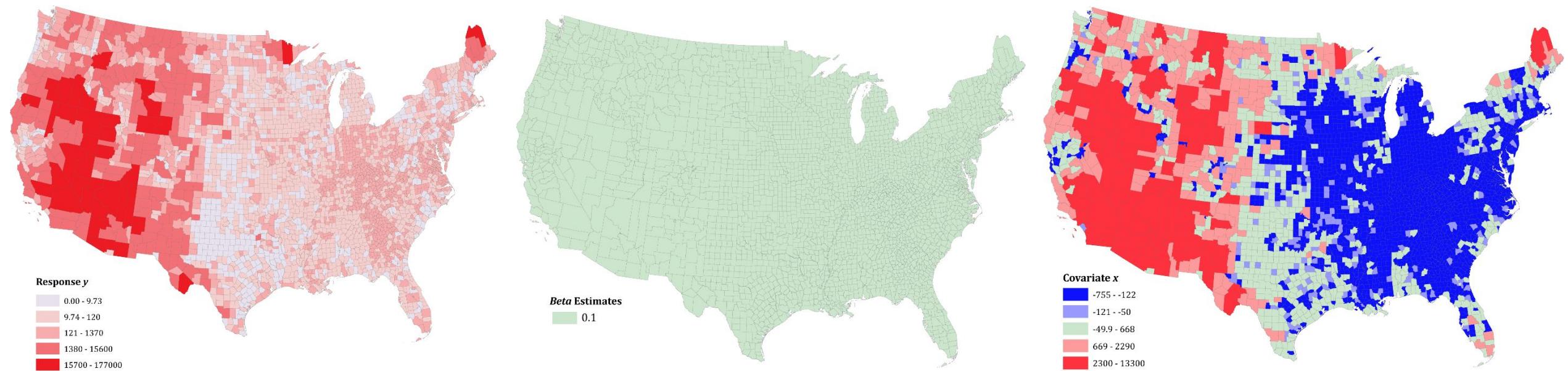


Usual Interpretation



$$y \text{ (response)} = \beta \text{ (process)} * x \text{ (covariate)}$$

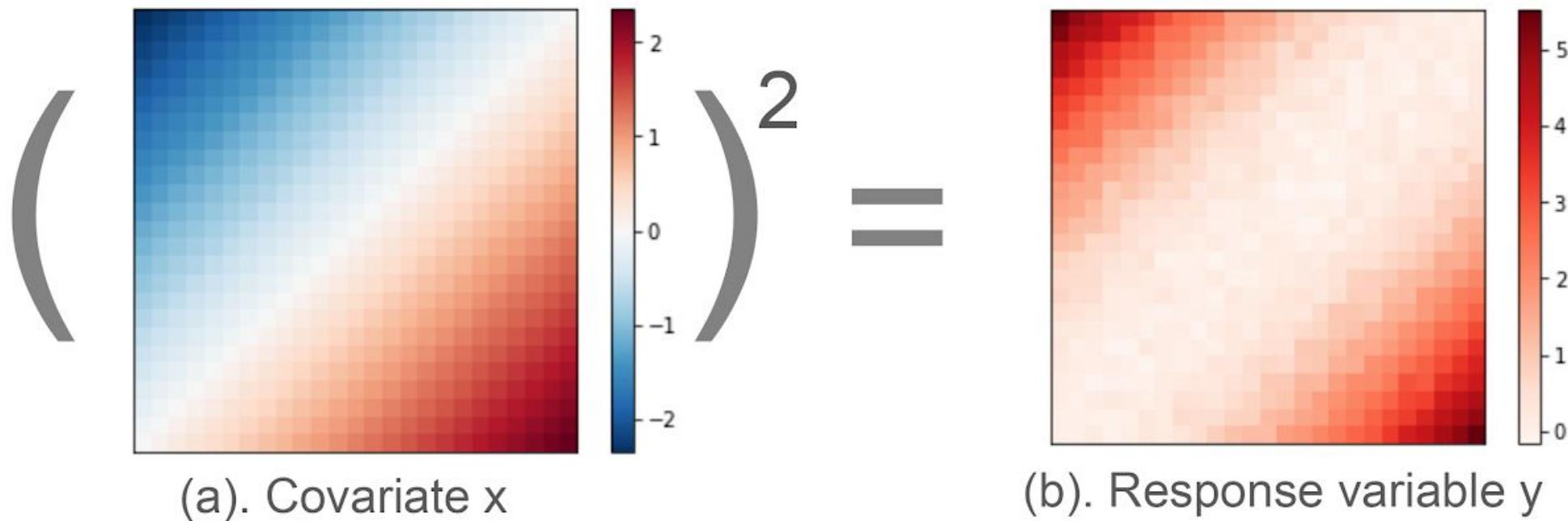
Possible Reality



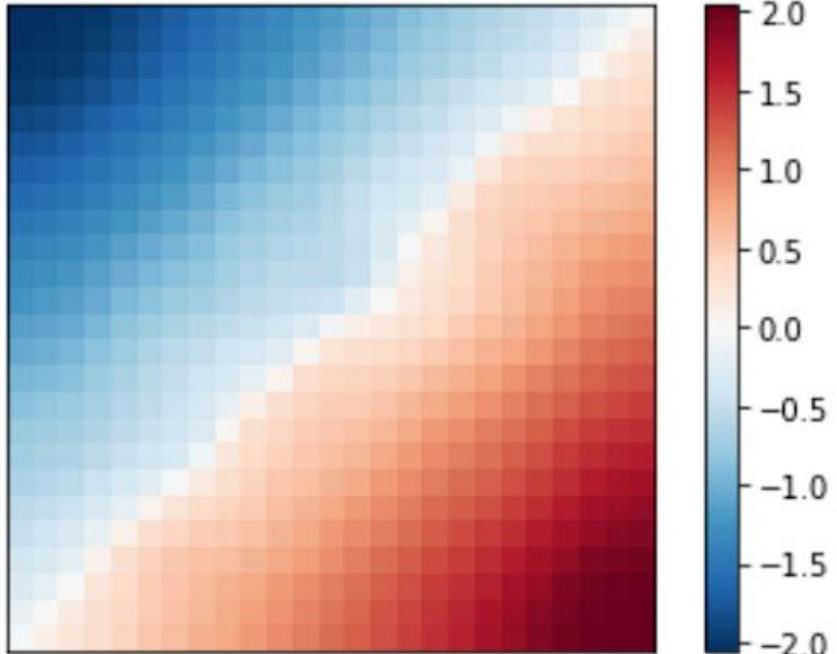
$$y \text{ (response)} = \beta \text{ (process)} * x^2 \text{ (covariate)}$$

Problem Statement

Consider a situation such that:



Problem Statement



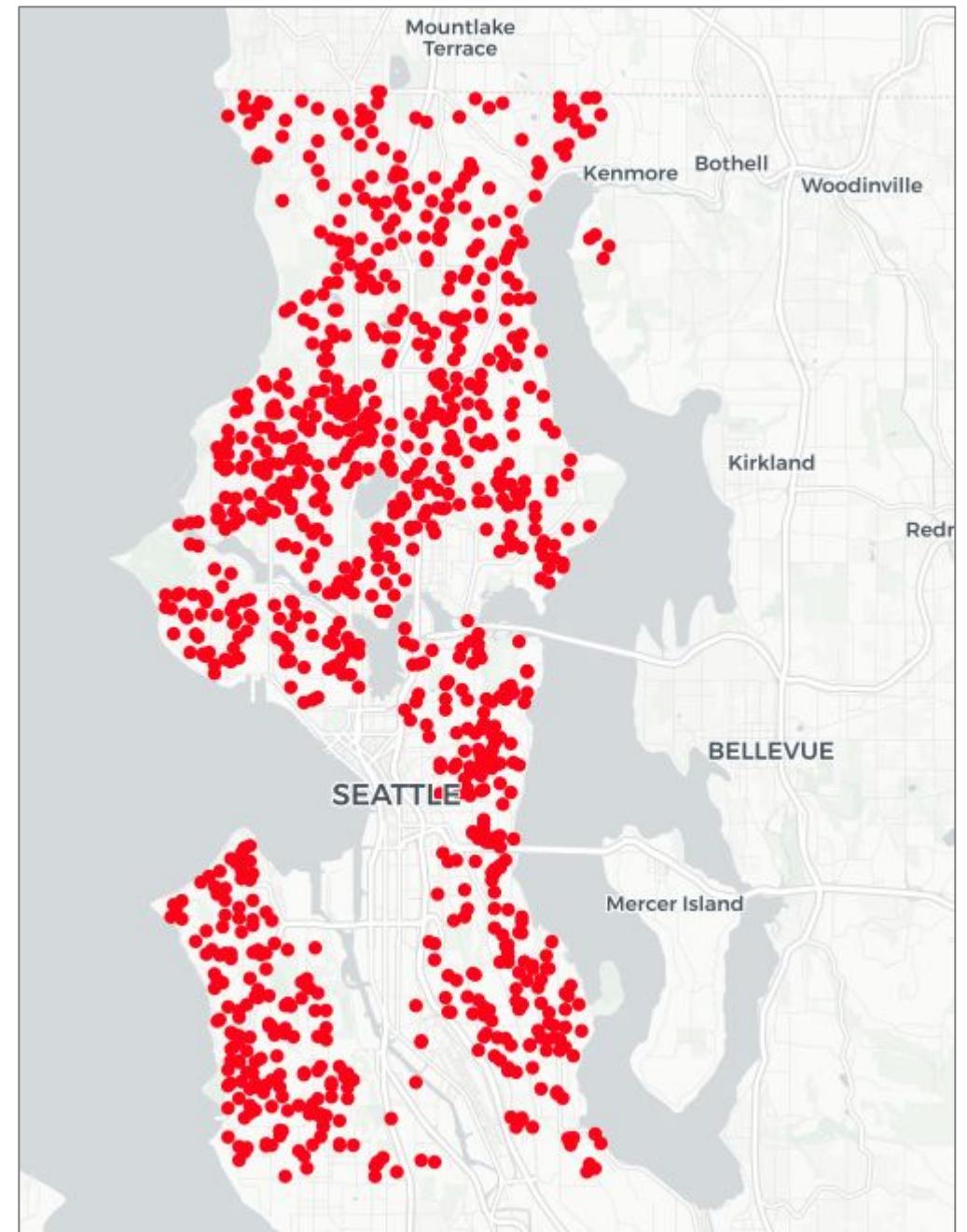
Parameter estimates from MGWR

- Without examining for the possibility of nonlinearity, the spatial variation in this surface would then be incorrectly interpreted to represent process spatial non-stationarity.
- How do we differentiate such situations from presence of non-stationarity in processes?

Background

Why is this problem important?

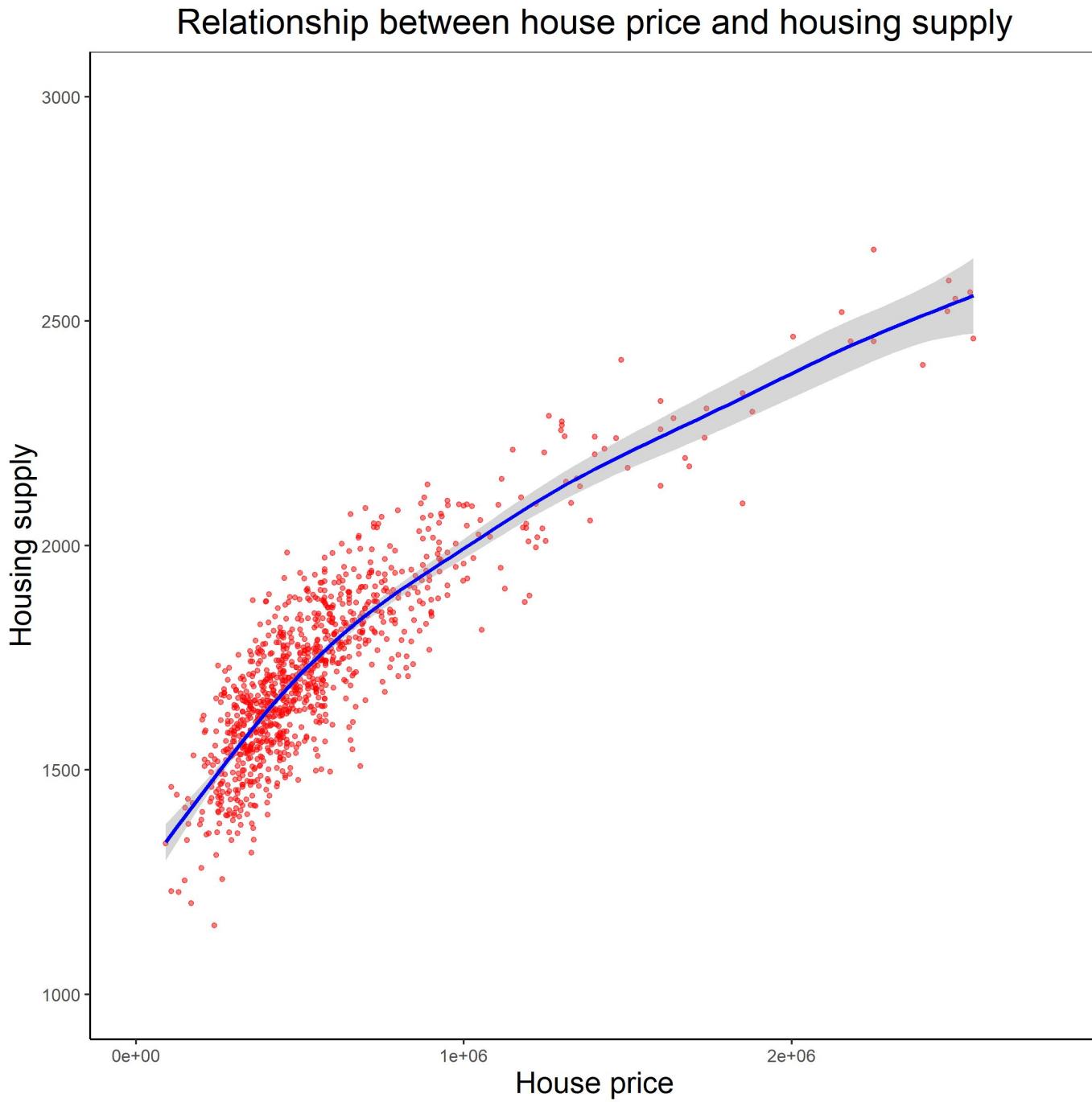
$$\text{housing supply} = \beta_0(u_i, v_i) + \beta_1(u_i, v_i)(\text{house price})$$



(Note: Hypothetical example)

Background

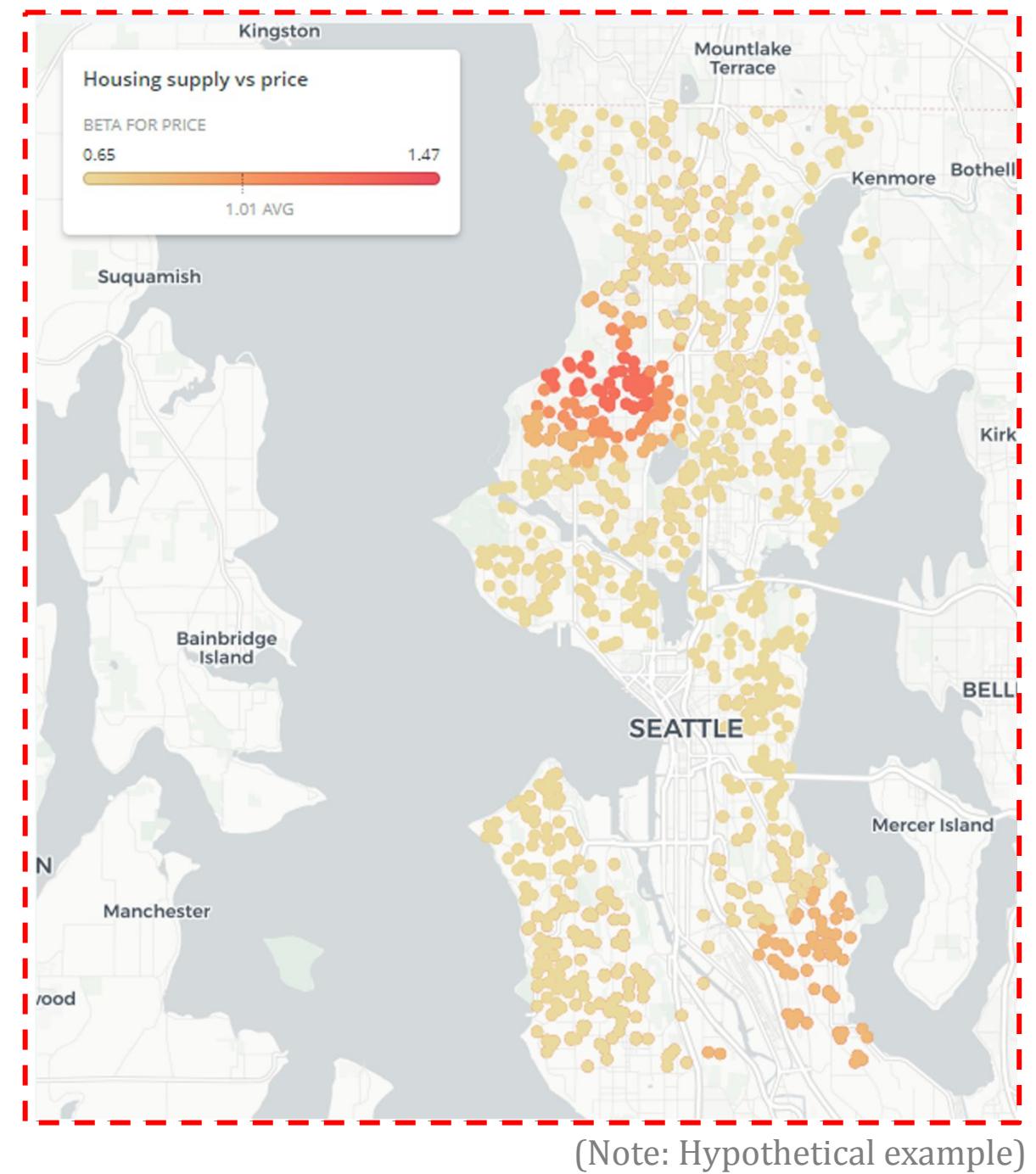
True relationship



Background

Estimated relationship

$$\text{housing supply} = \beta_0(u_i, v_i) + \beta_1(u_i, v_i) (\sqrt{\text{house price}})$$



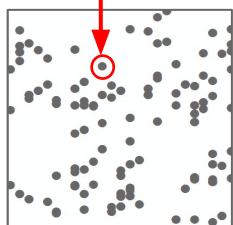
So, we need to separate these two scenarios

Modeling Processes

Process Spatial Nonstationarity

Local Models e.g. MGWR

$$y_i = \sum_j \beta_{ij}(u_i, v_i) X_{ij} + \varepsilon_i$$

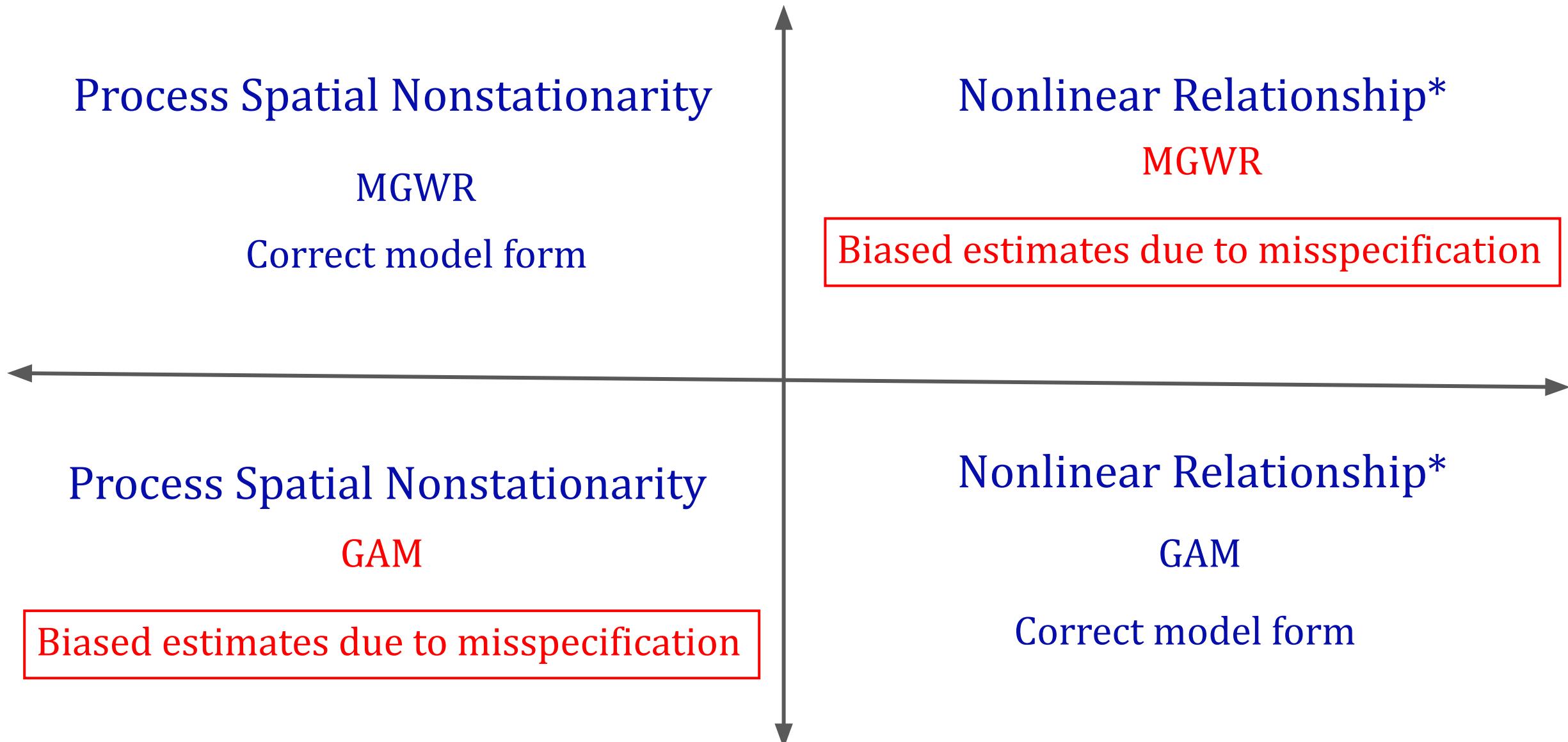


Nonlinear Relationships

Models e.g. GAM

$$y_i = \sum_{j=1}^p f_j(x_{ij})$$

Otherwise we make incorrect inferences about processes



* Assuming the covariate is spatially varying

Hence:

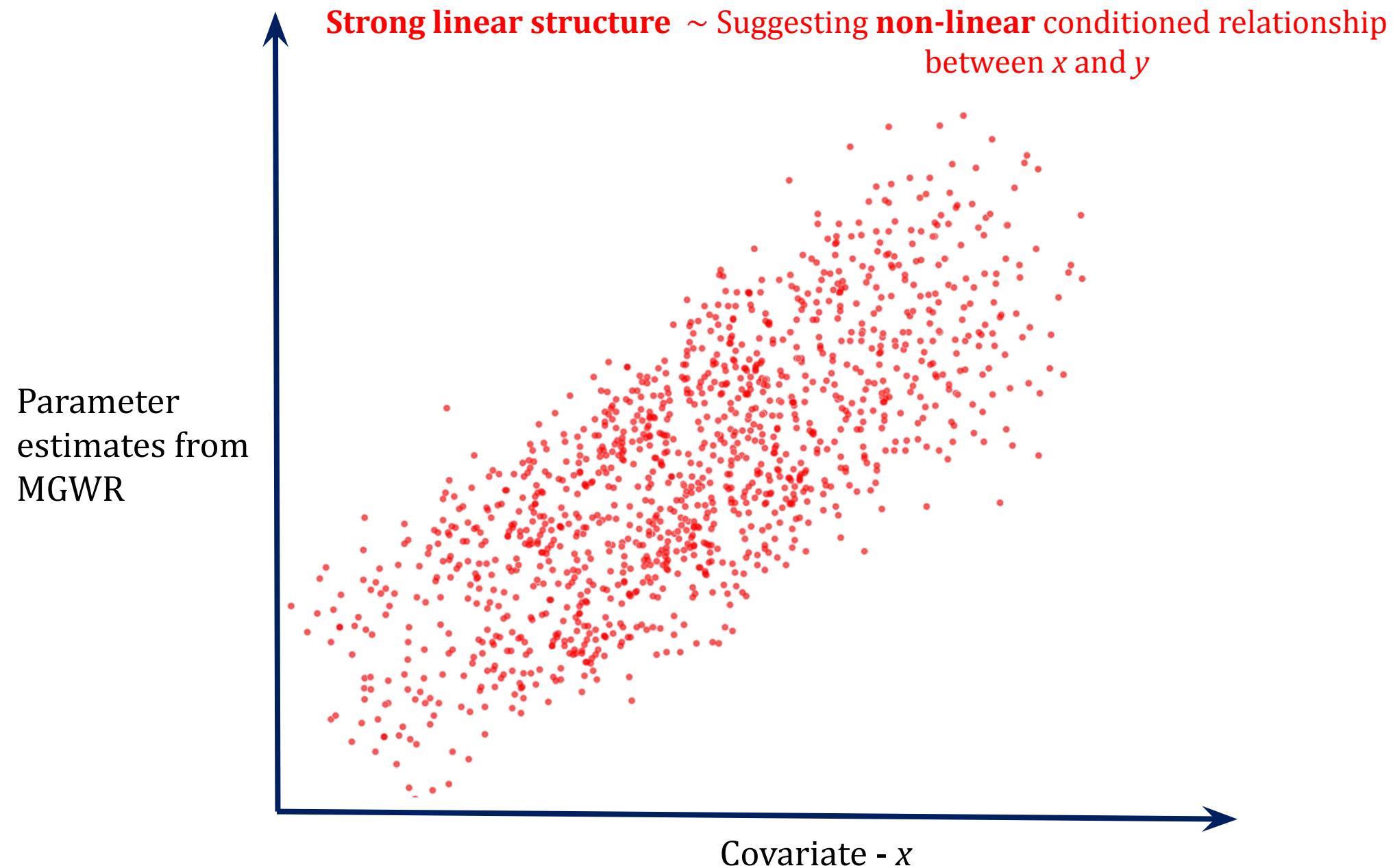
1. Is there an easy way to **detect when spatially varying parameter estimates are masking nonlinear relationships?**

2. Conversely, is there an easy way to **detect when estimated nonlinear relationships are masking process nonstationarity?**

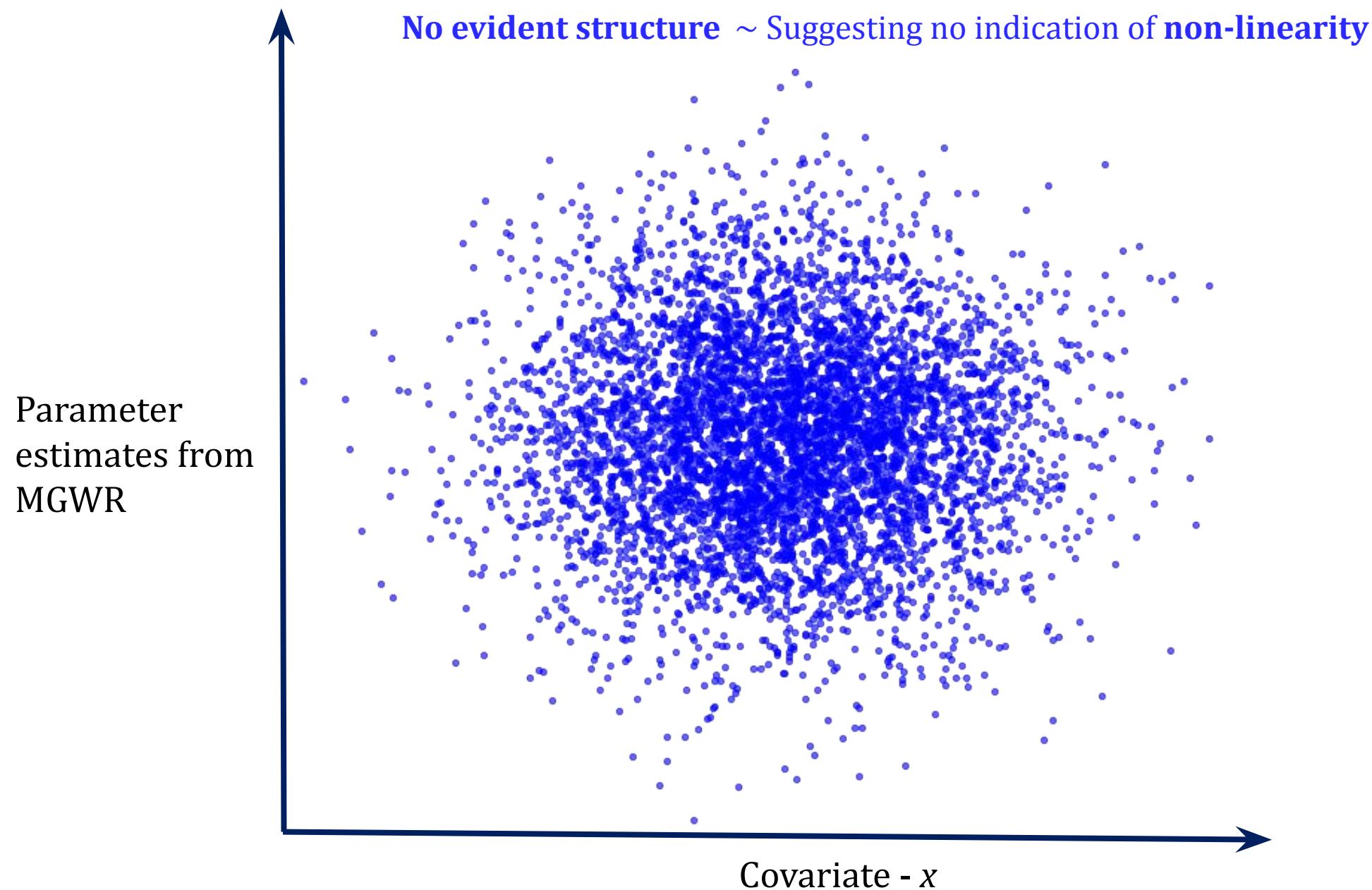
2. Research Objectives

- **Describe a simple test** to check whether parameter estimates generated by local models are manifestations of nonlinear relationships rather than spatially varying processes
- **Demonstrate feasibility of such a test** in simulation studies and empirical research with real-world noisy data

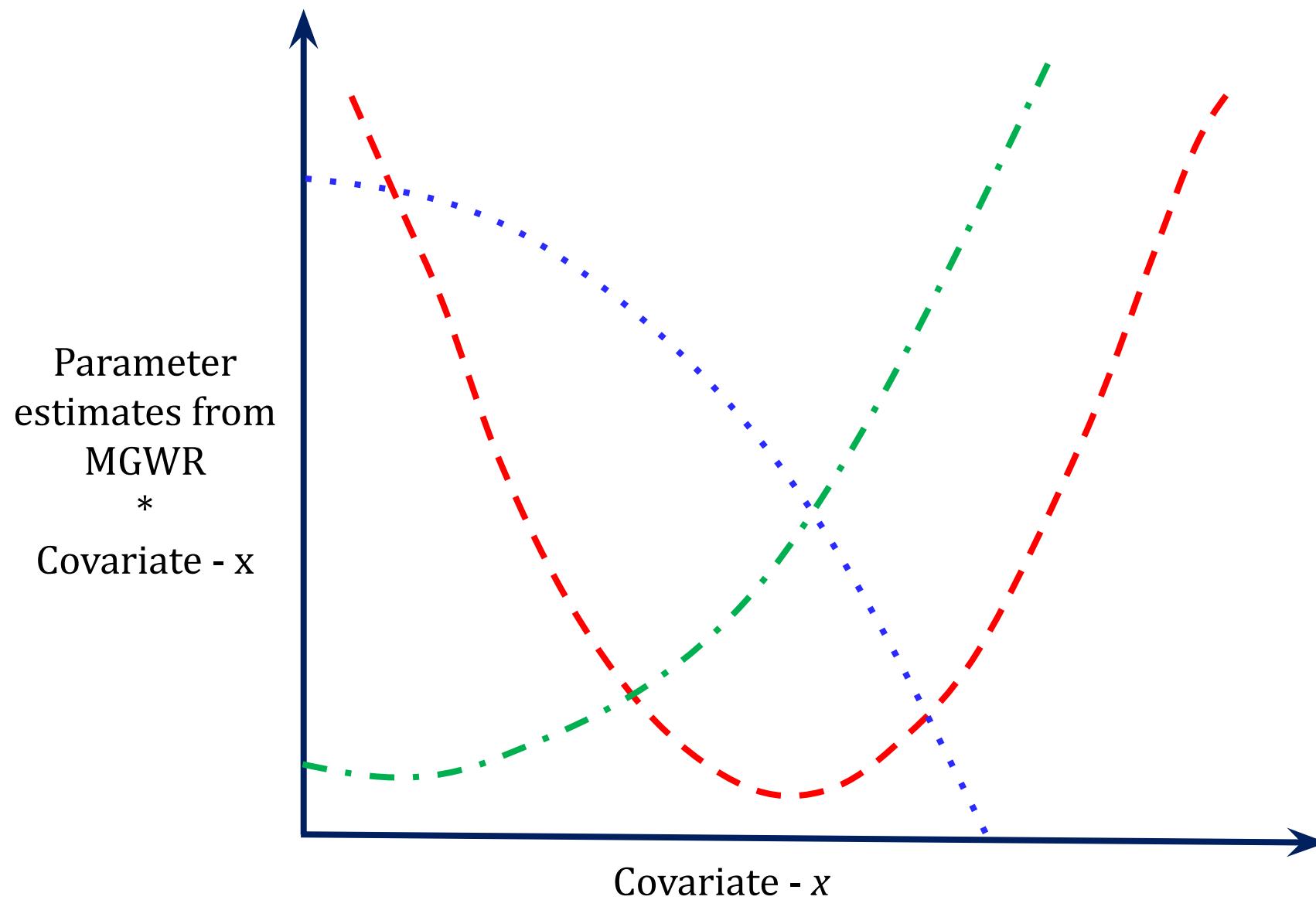
Diagnostic test described



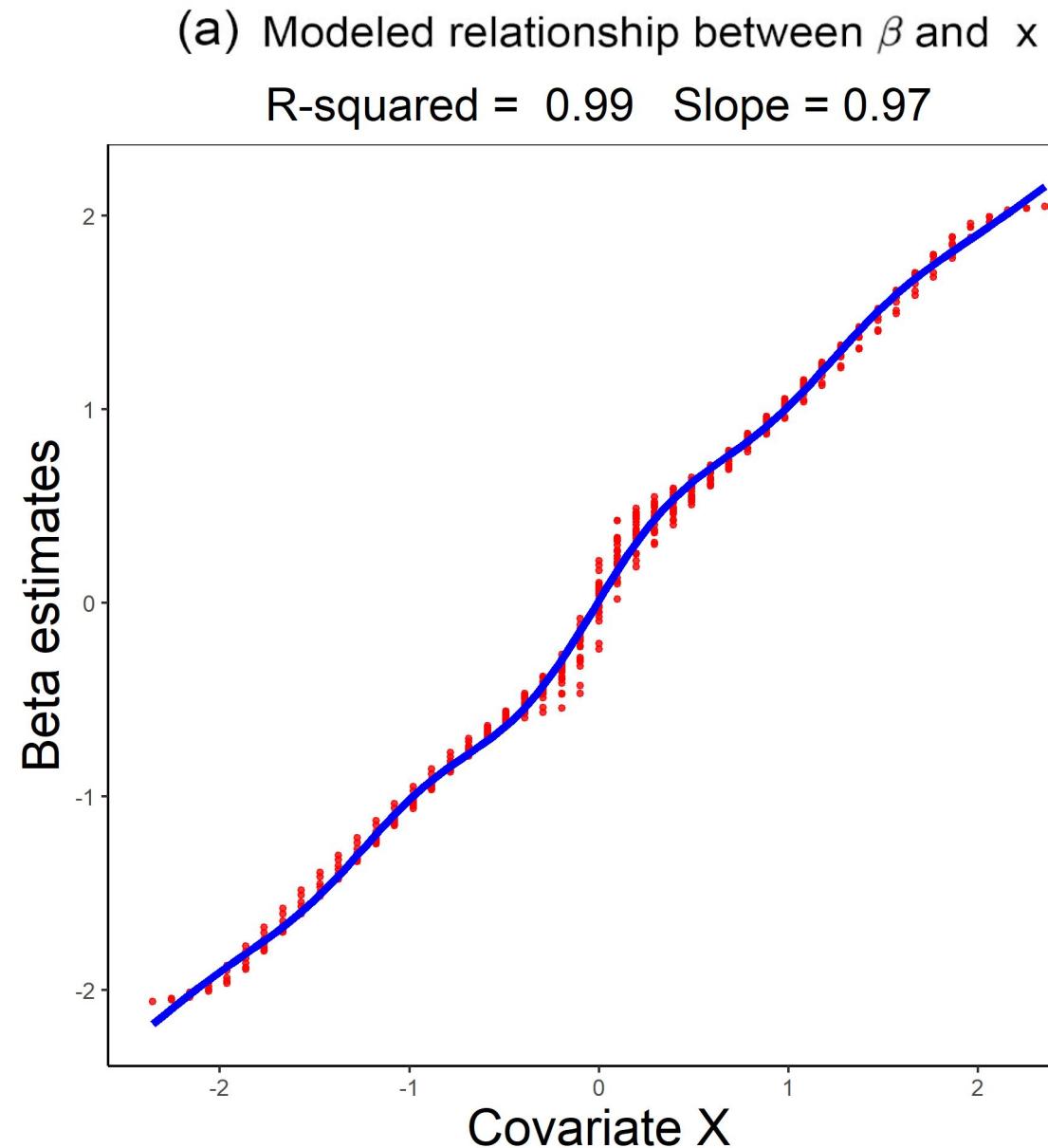
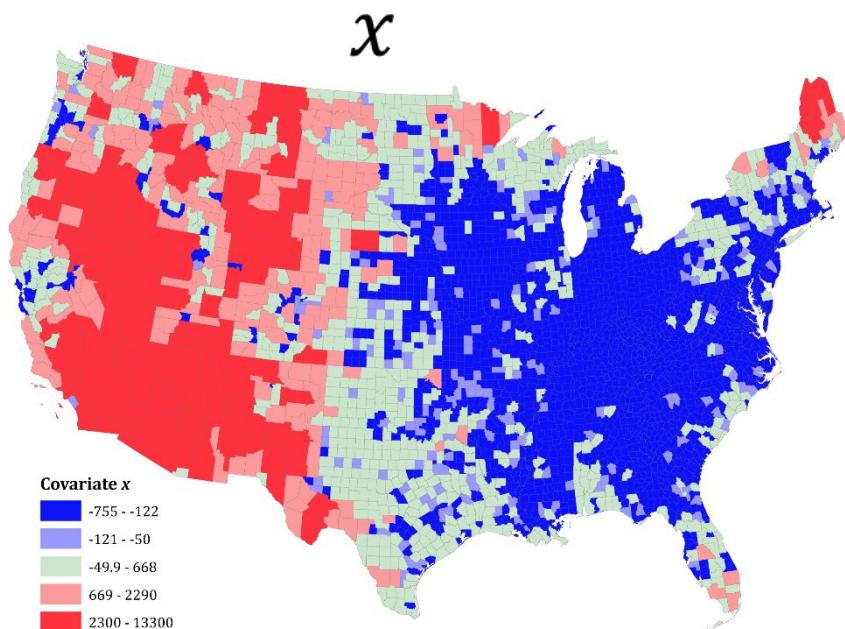
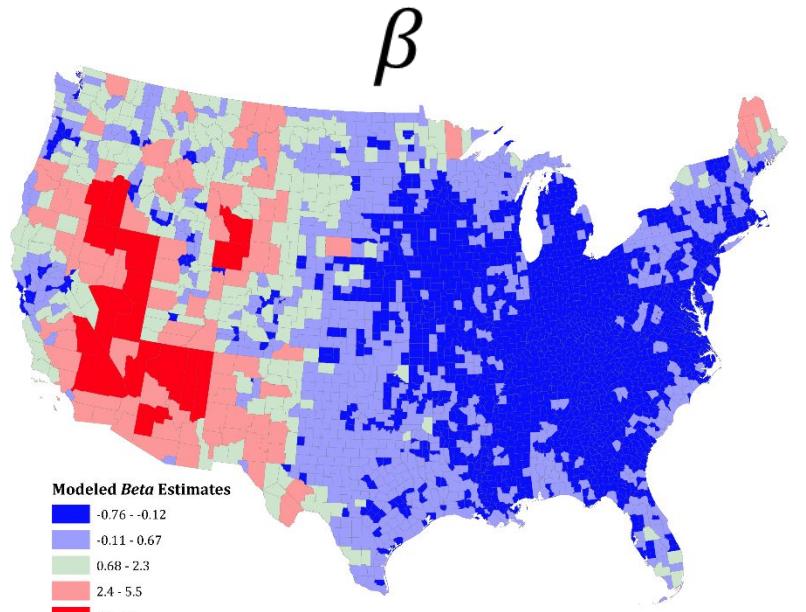
Diagnostic test described



Plot for further evidence on type of non-linearity

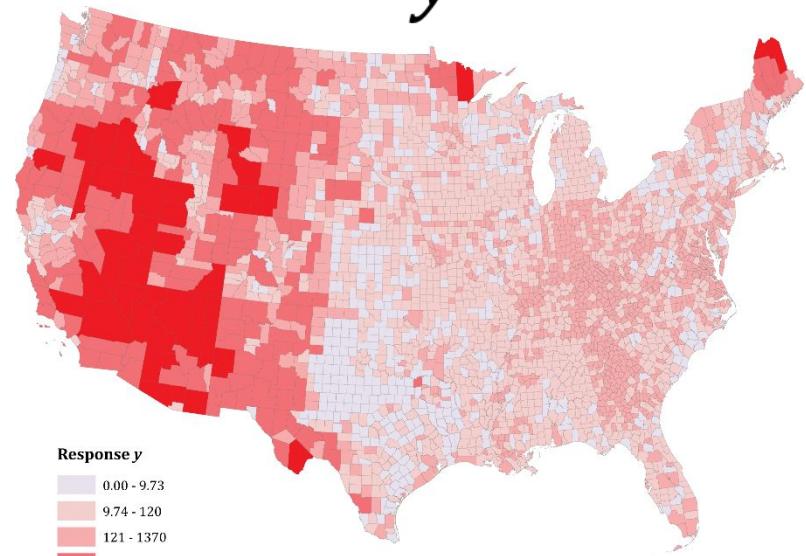


Revisiting the example:

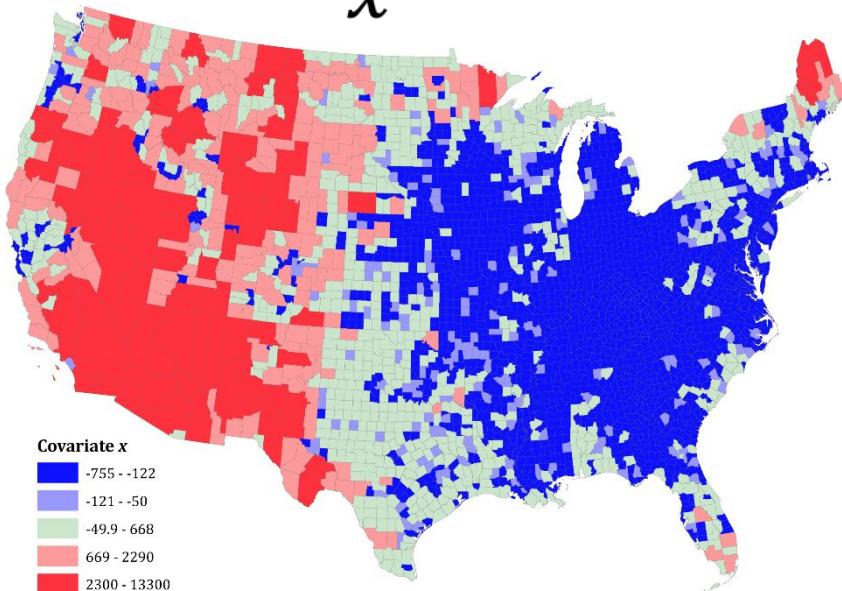


Revisiting the example:

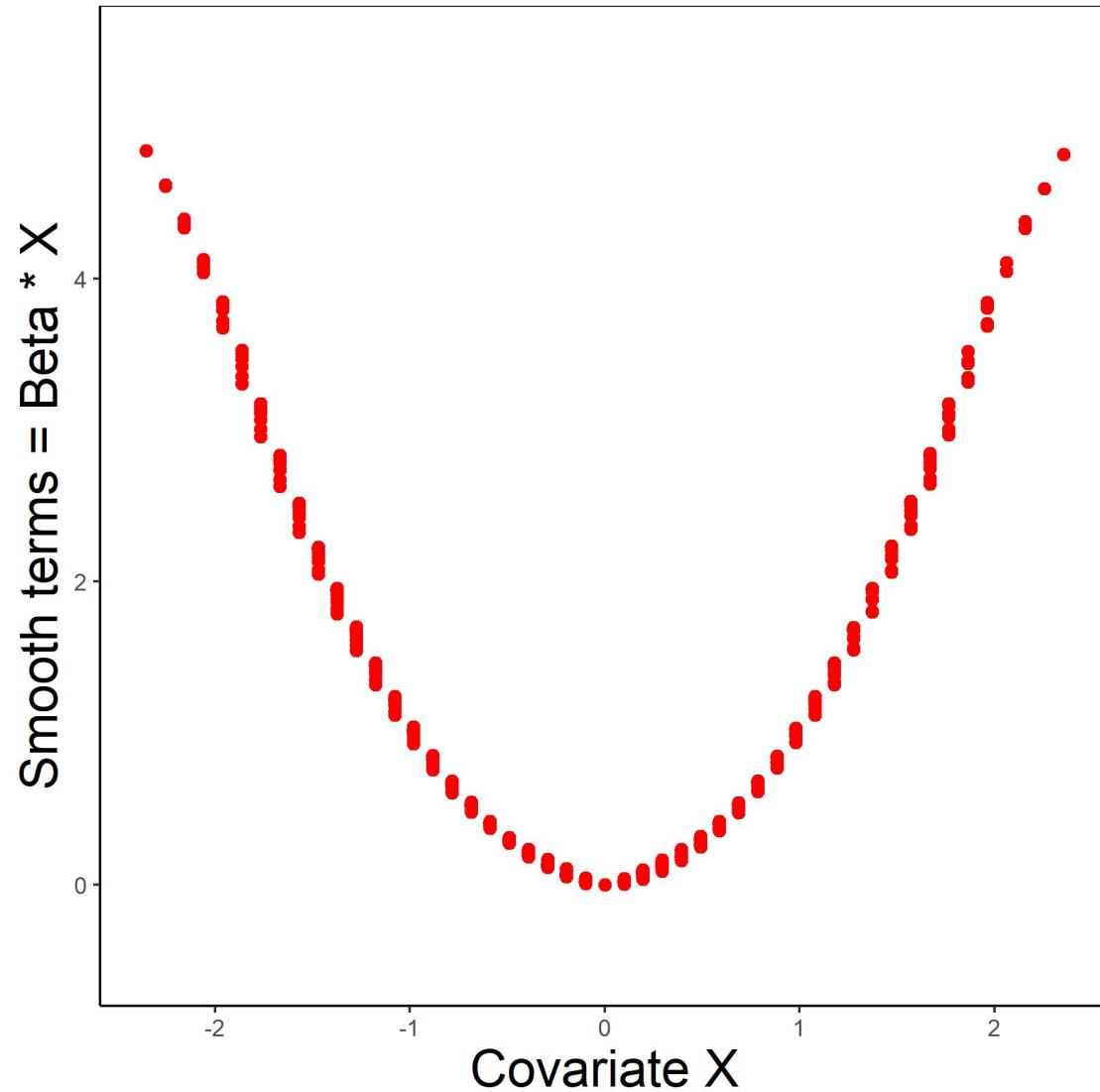
y



χ



(b) Modeled relationship between β^*x and x



3. Research Design

Check Feasibility of Diagnostic Tests in Two Extreme Scenarios

Experiment I

Processes are **spatially nonstationary** and **independent** of the covariates

$$y = \beta_1 * x_1 + \beta_2 * x_2 + \epsilon$$

β_1 and x_1 are independent
 β_2 and x_2 are independent

Experiment II

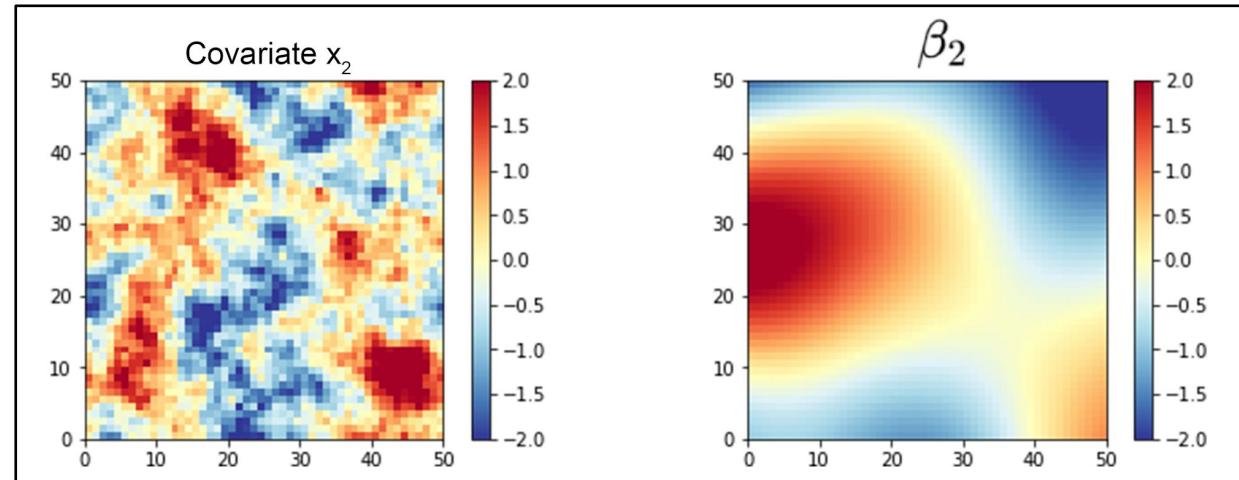
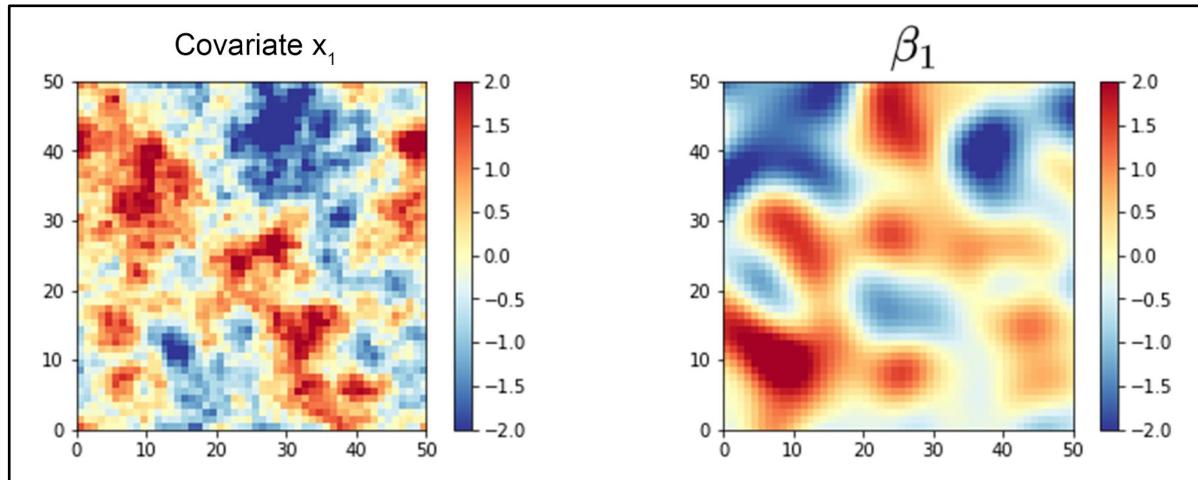
Processes are **spatially stationary** and **derived** from the covariates

$$y = \beta_1 * x_1 + \beta_2 * x_2 + \epsilon$$
$$\beta_1 = k_1 * x_1 \text{ and } \beta_2 = k_2 * x_2$$

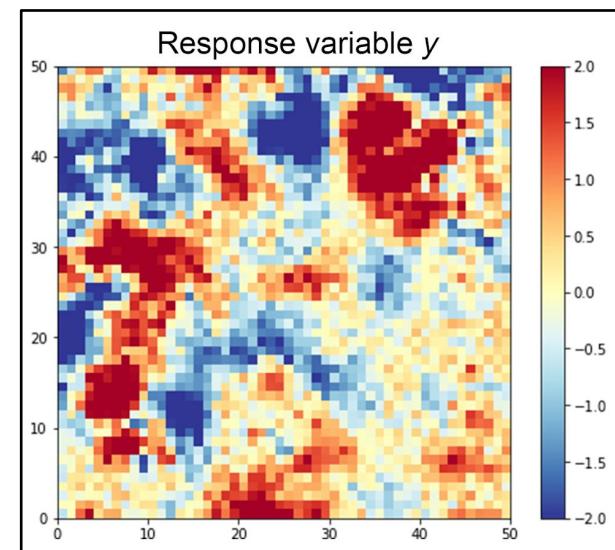
$$y = k_1 * x_1^2 + k_2 * x_2^2 + \epsilon$$

k_1 and k_2 are near constant terms

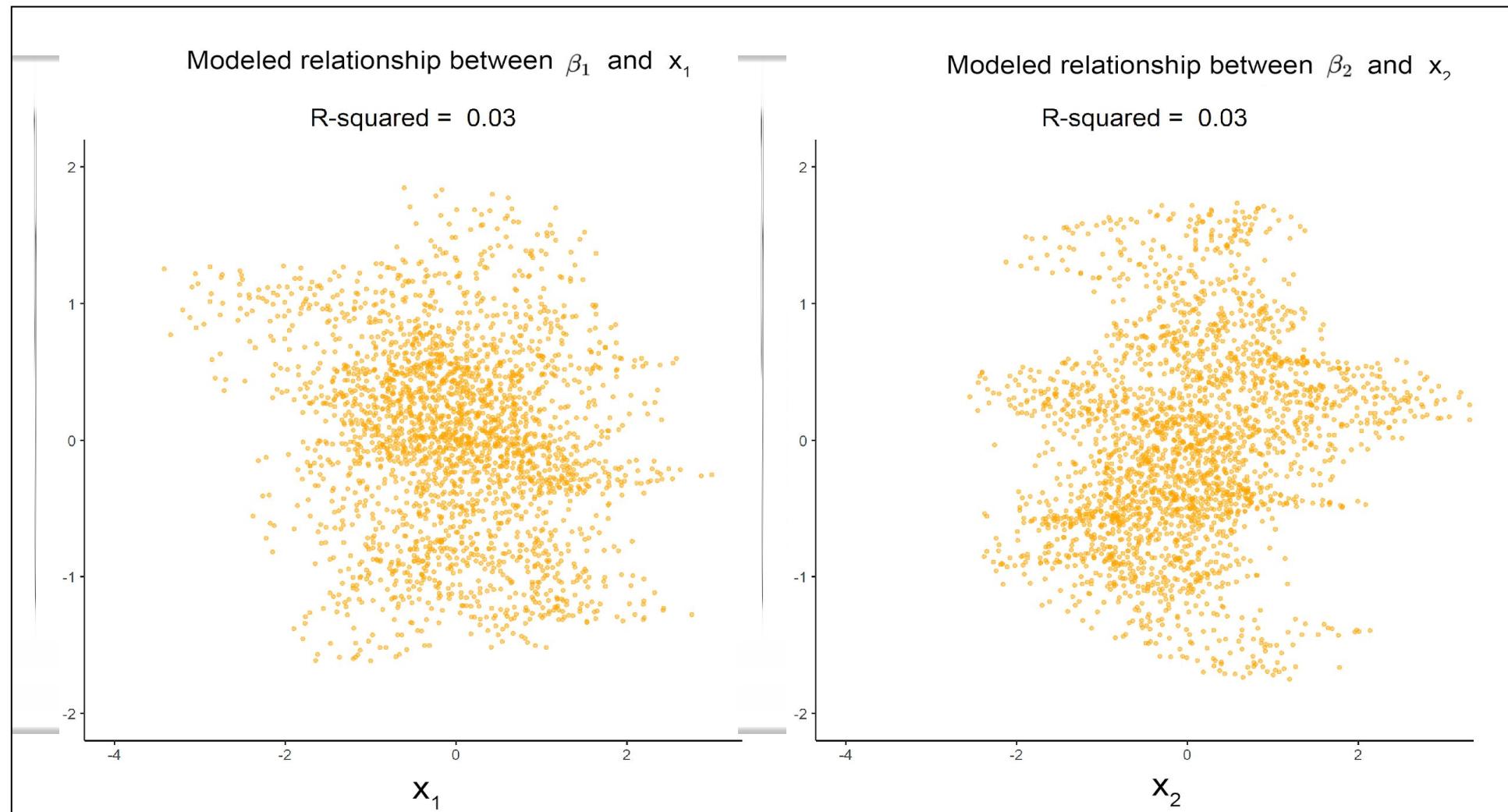
Experiment I – Non-stationarity in processes



$$y = \beta_1 * x_1 + \beta_2 * x_2 + \epsilon$$

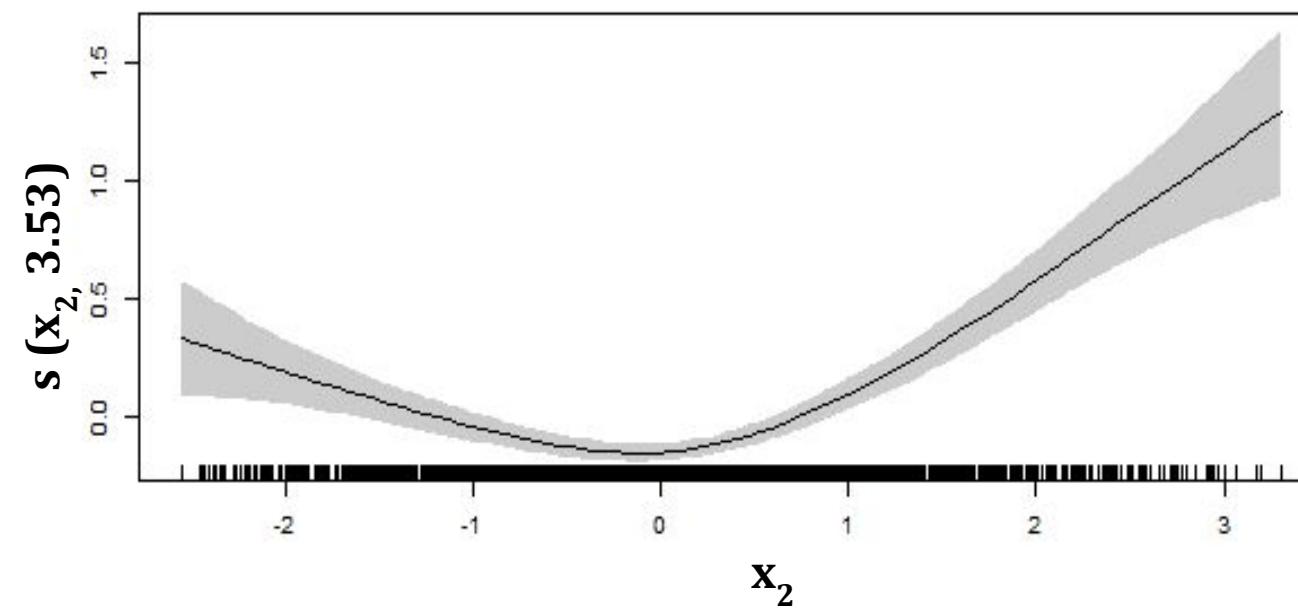
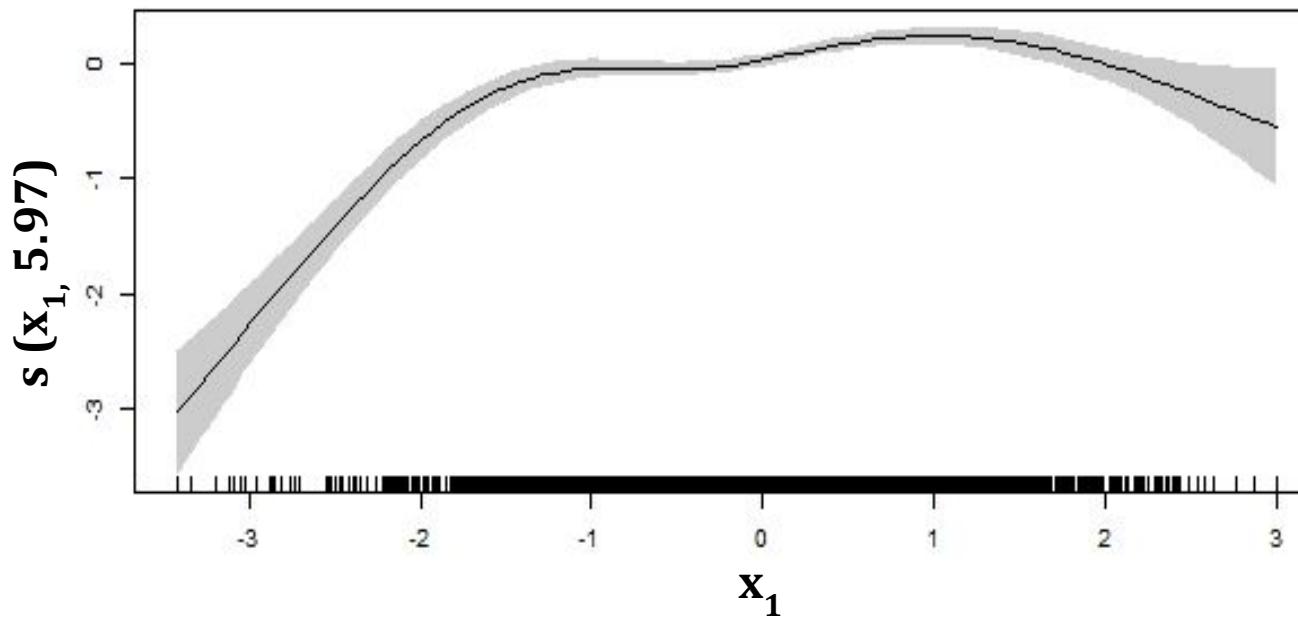


Experiment I – Diagnostic test: Detecting nonlinearity as a cause of spatially varying parameter estimates



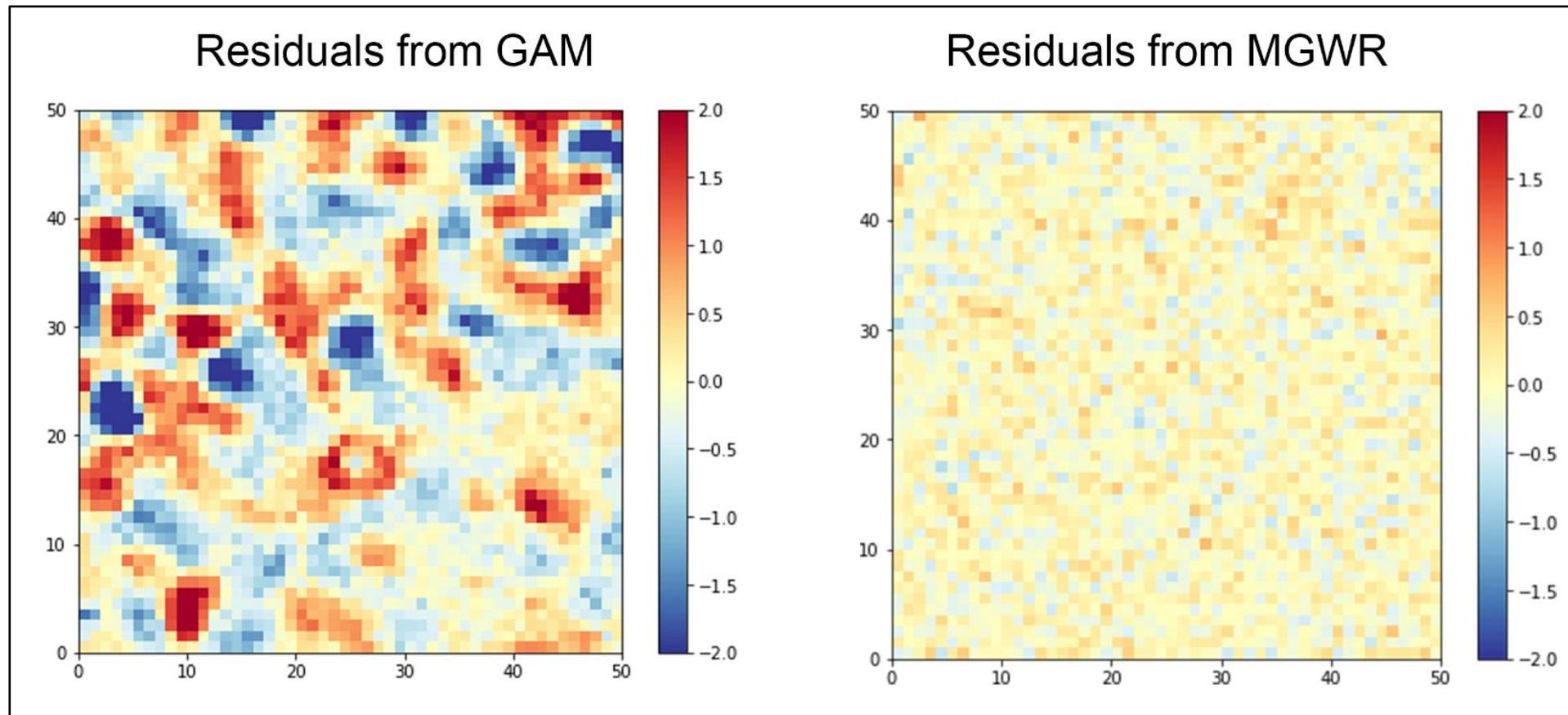
Nonlinearity Not Detected

Experiment I – GAM result plots



Experiment I – Diagnostic test: Detecting spatial nonstationarity when nonlinearity is assumed

Residuals from GAM and MGWR



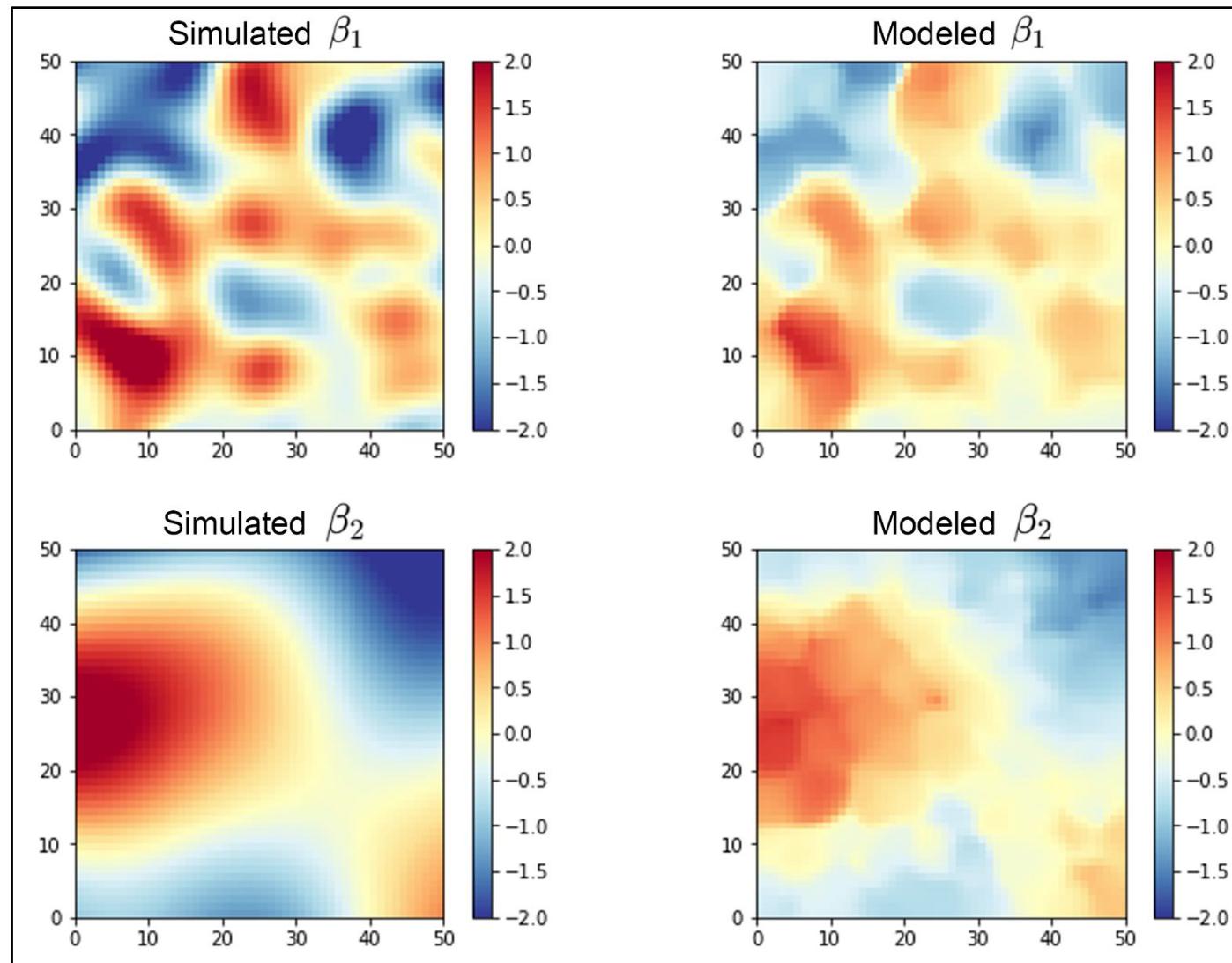
Moran's I value: +0.73

(p-value: 0.001)

Process Nonstationarity Detected

Experiment I - Further evidence of non-stationarity in processes

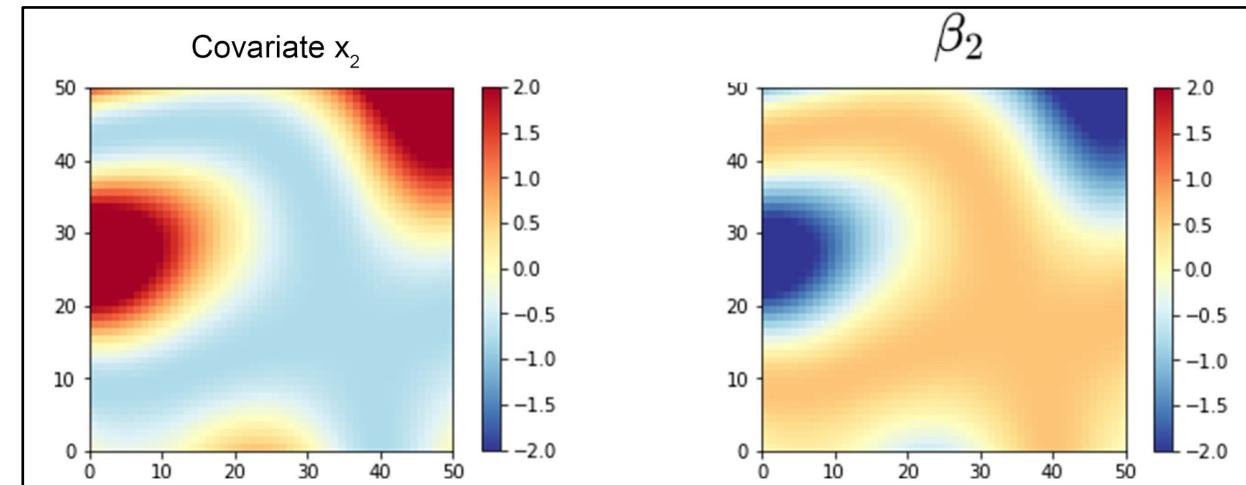
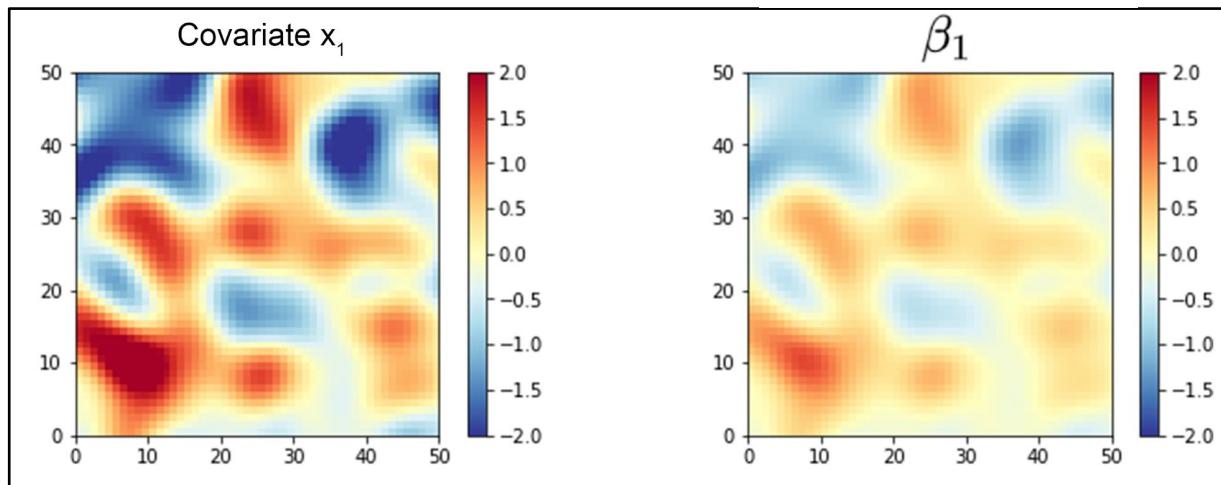
Simulated vs Modeled parameter estimates from MGWR



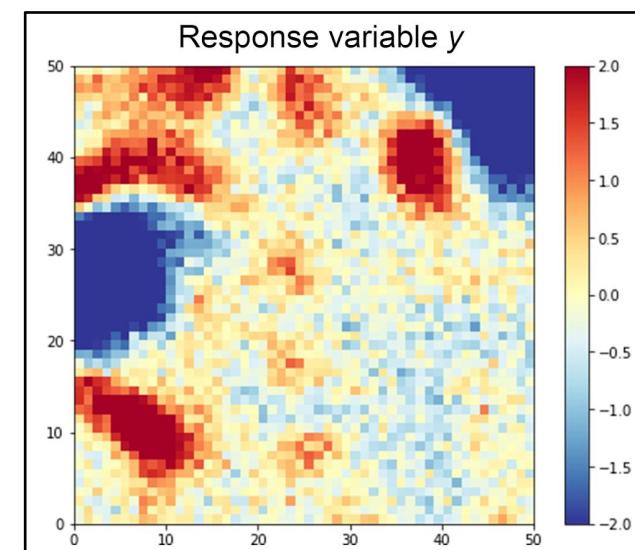
Experiment II – Complete non-linearity in relationships

$$\beta_1 = 0.5 * x_1$$

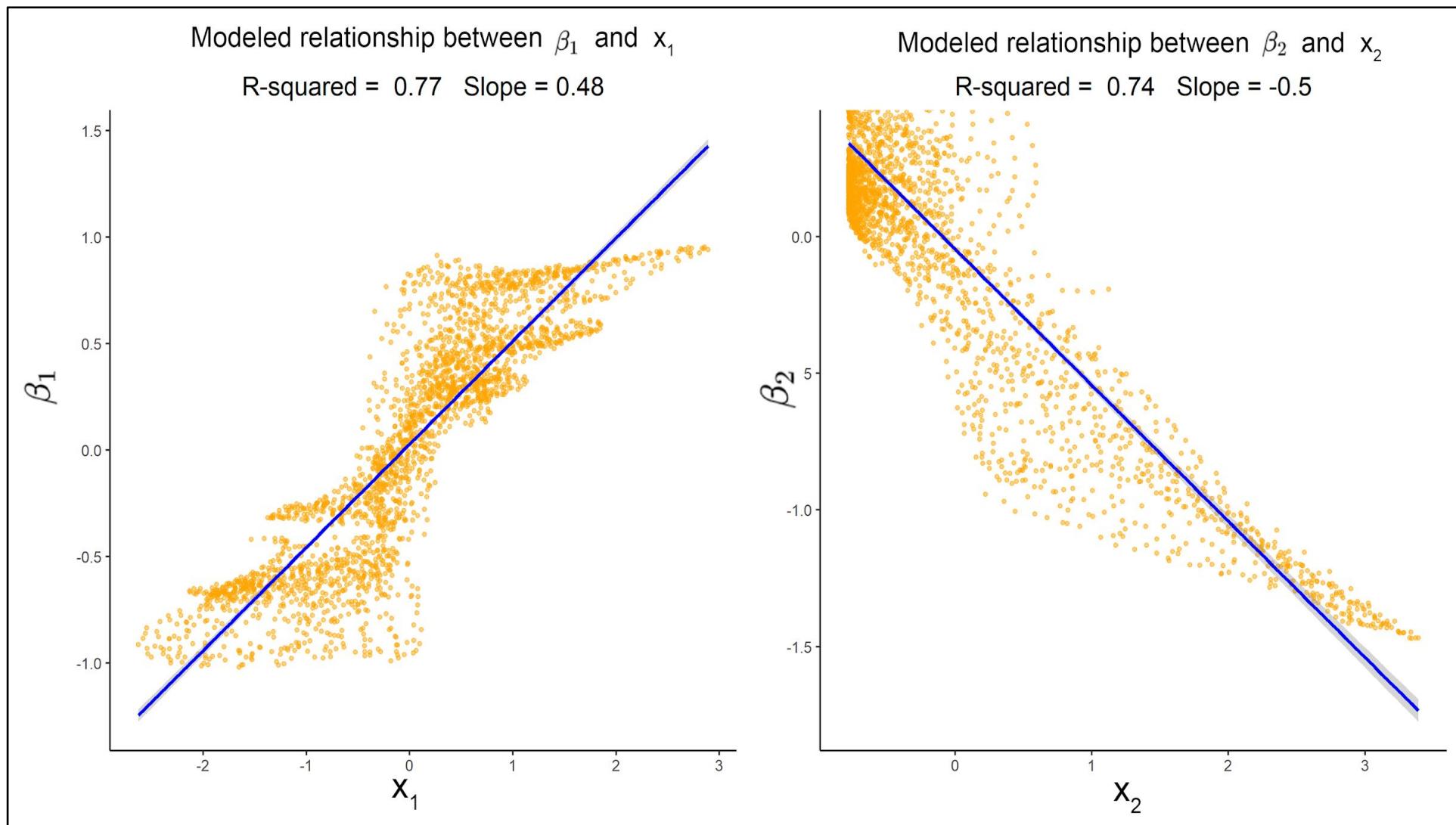
$$\beta_2 = -0.8 * x_2$$



$$y = 0.5 * x_1^2 - 0.8 * x_2^2 + \epsilon \quad \rightarrow$$



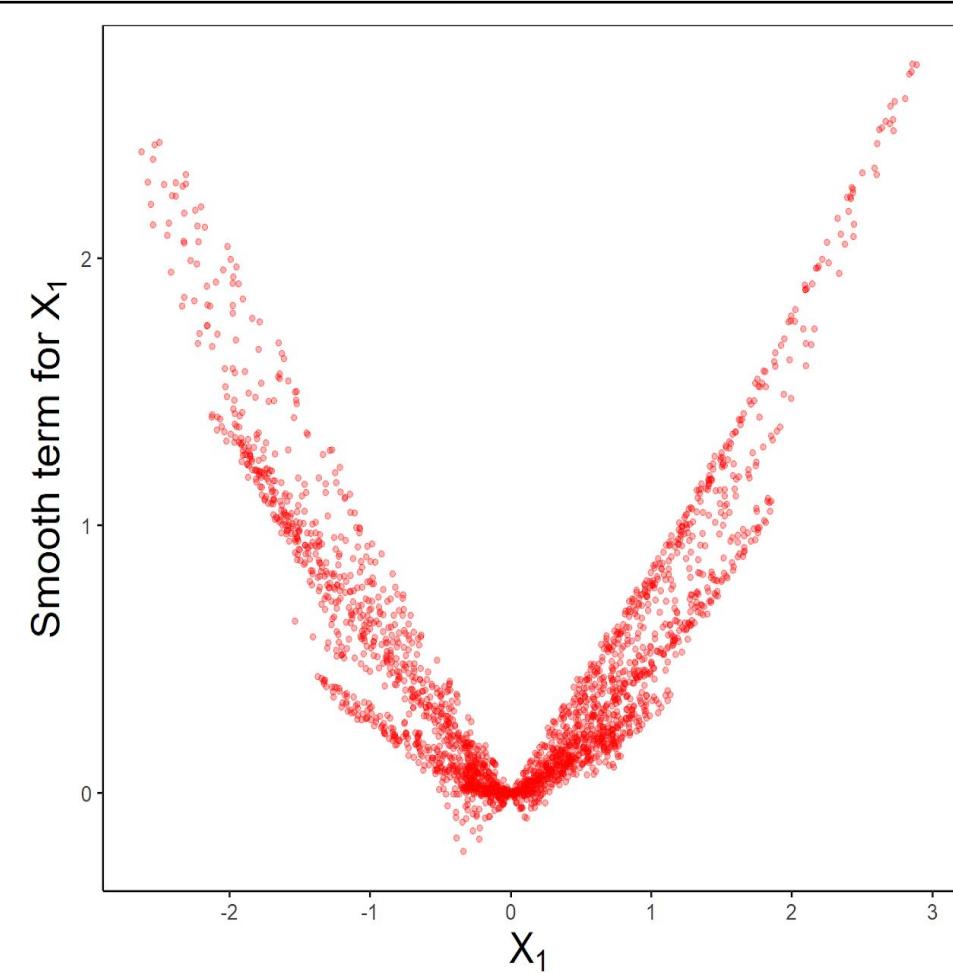
Experiment II – Diagnostic test: Detecting nonlinearity when spatial nonstationarity is assumed



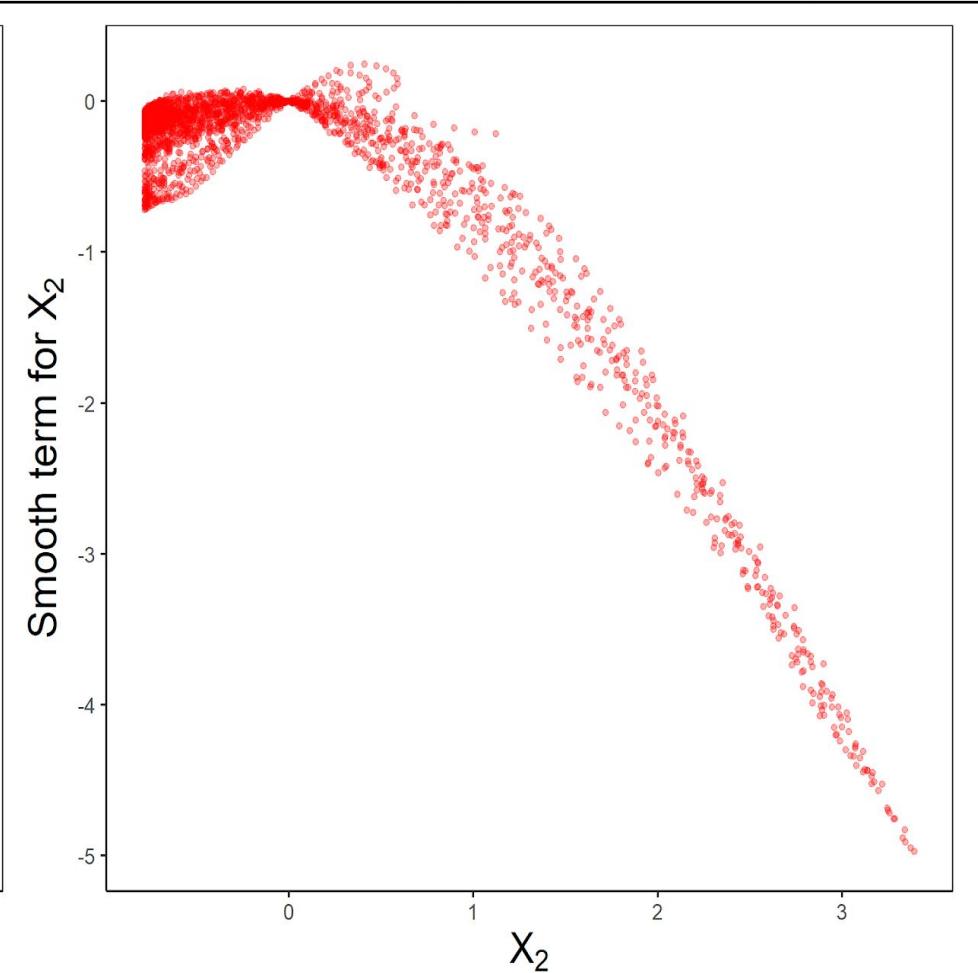
Nonlinearity Detected

Experiment II: Plot for further evidence on type of non-linearity

Modeled relationship between $\beta_1 * x_1$ and x_1

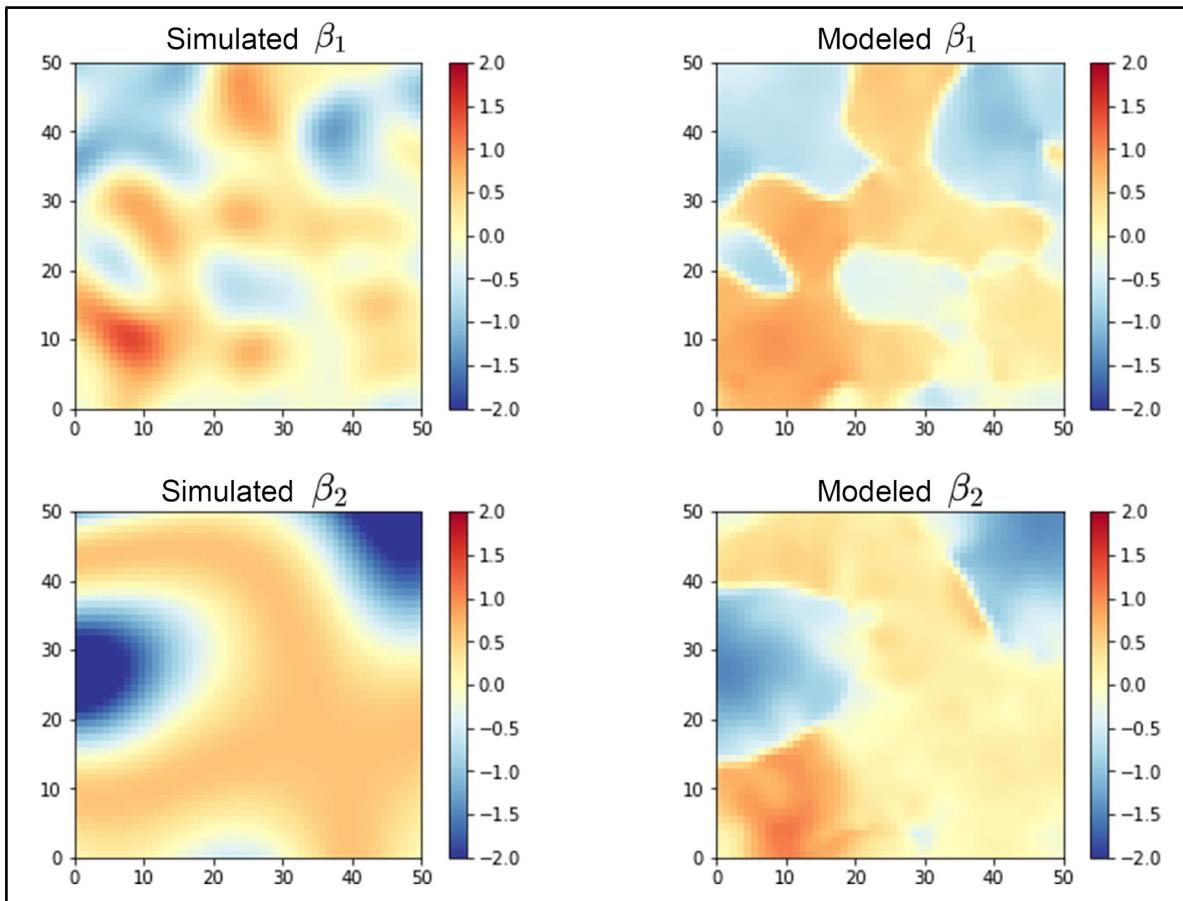


Modeled relationship between $\beta_2 * x_2$ and x_2

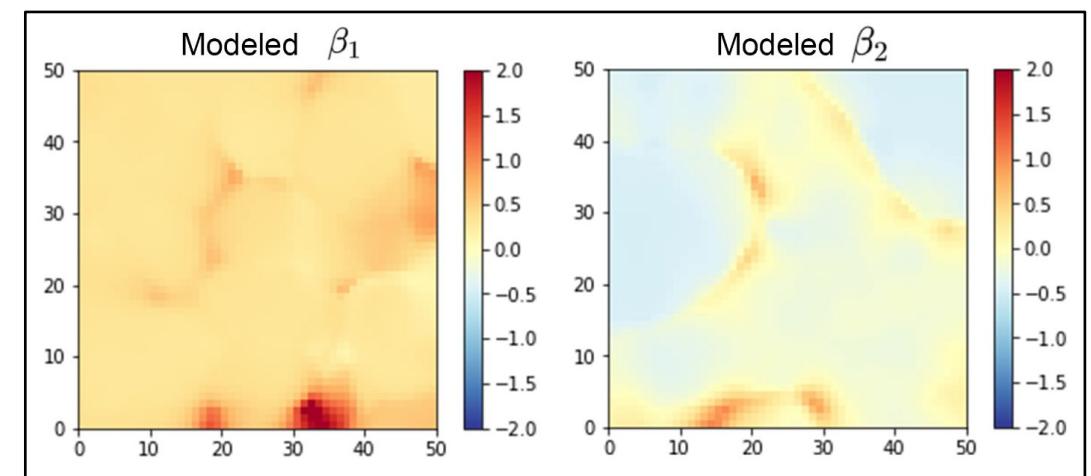


Experiment II: Parameter estimates from MGWR

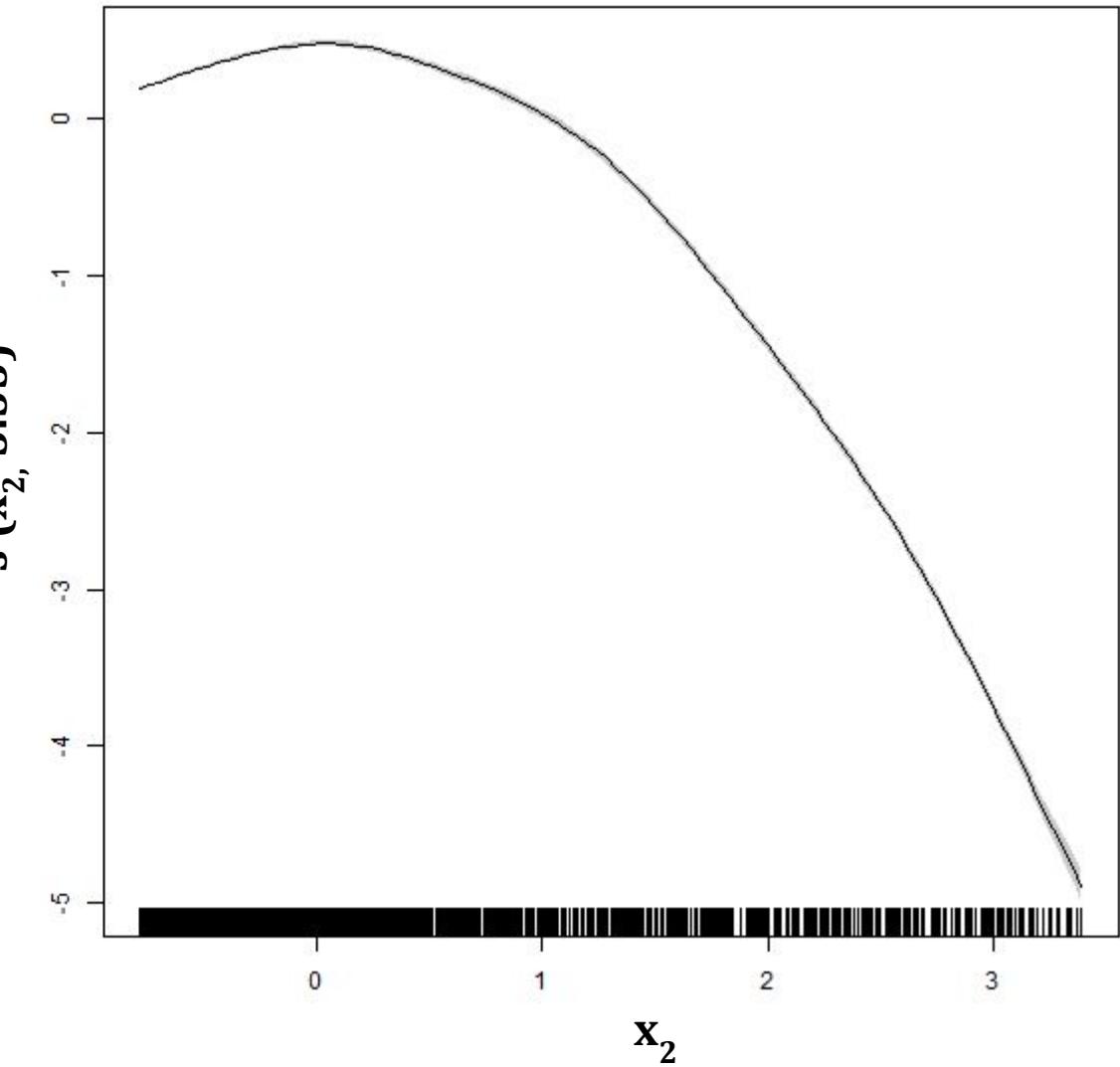
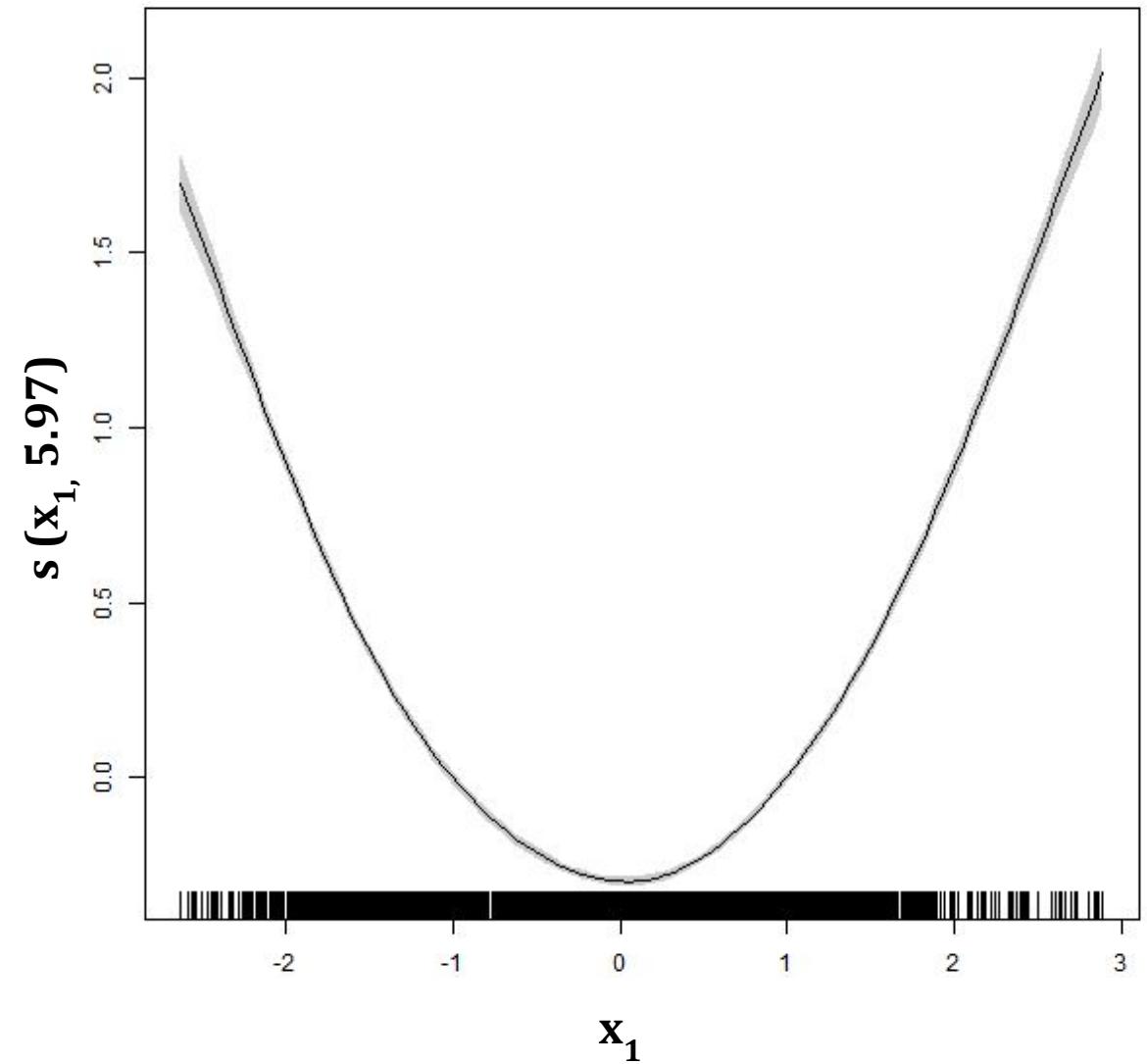
Using x_1 and x_2 as predictors



Using x_1^2 and x_2^2 as predictors

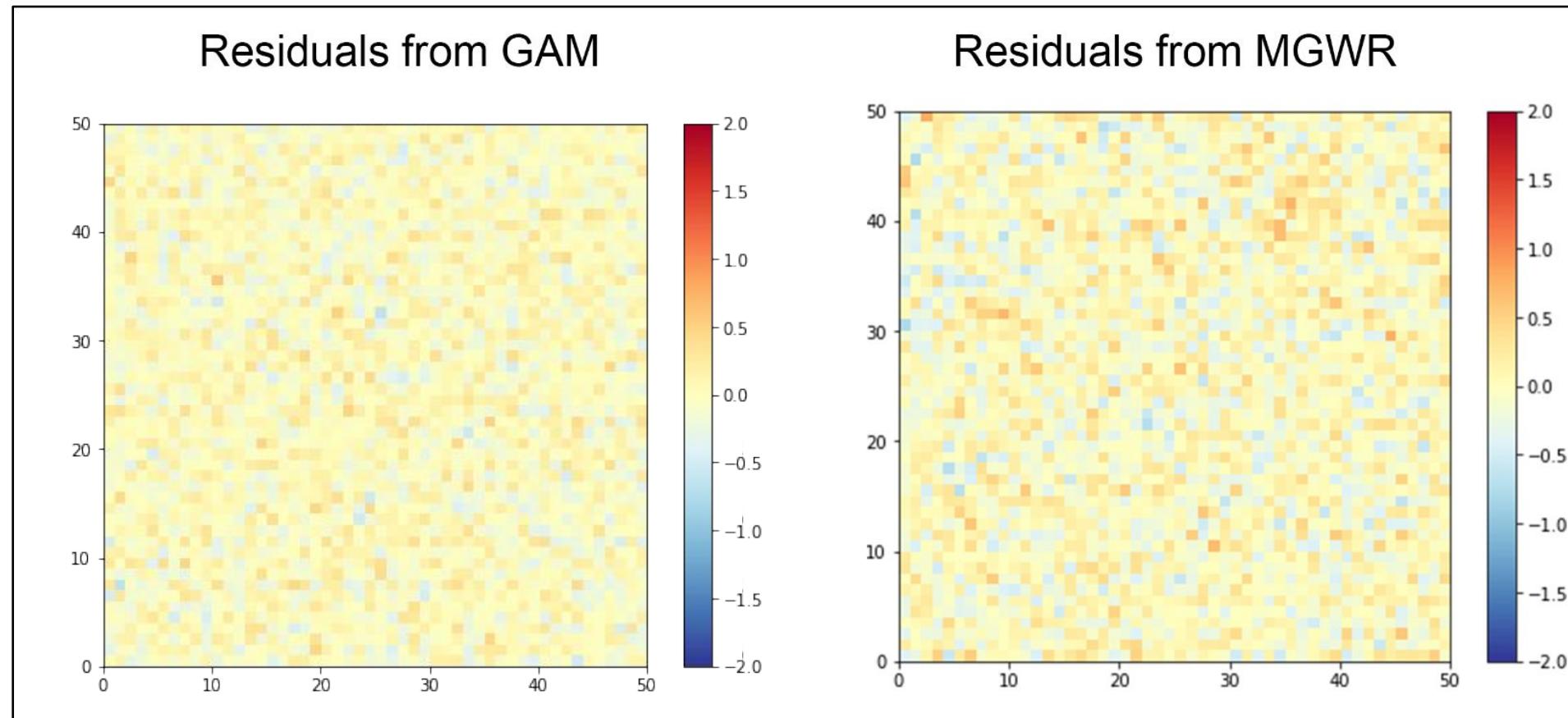


Experiment II - GAM result plots



Experiment II – Diagnostic test: Detecting spatial nonstationarity when nonlinearity is assumed

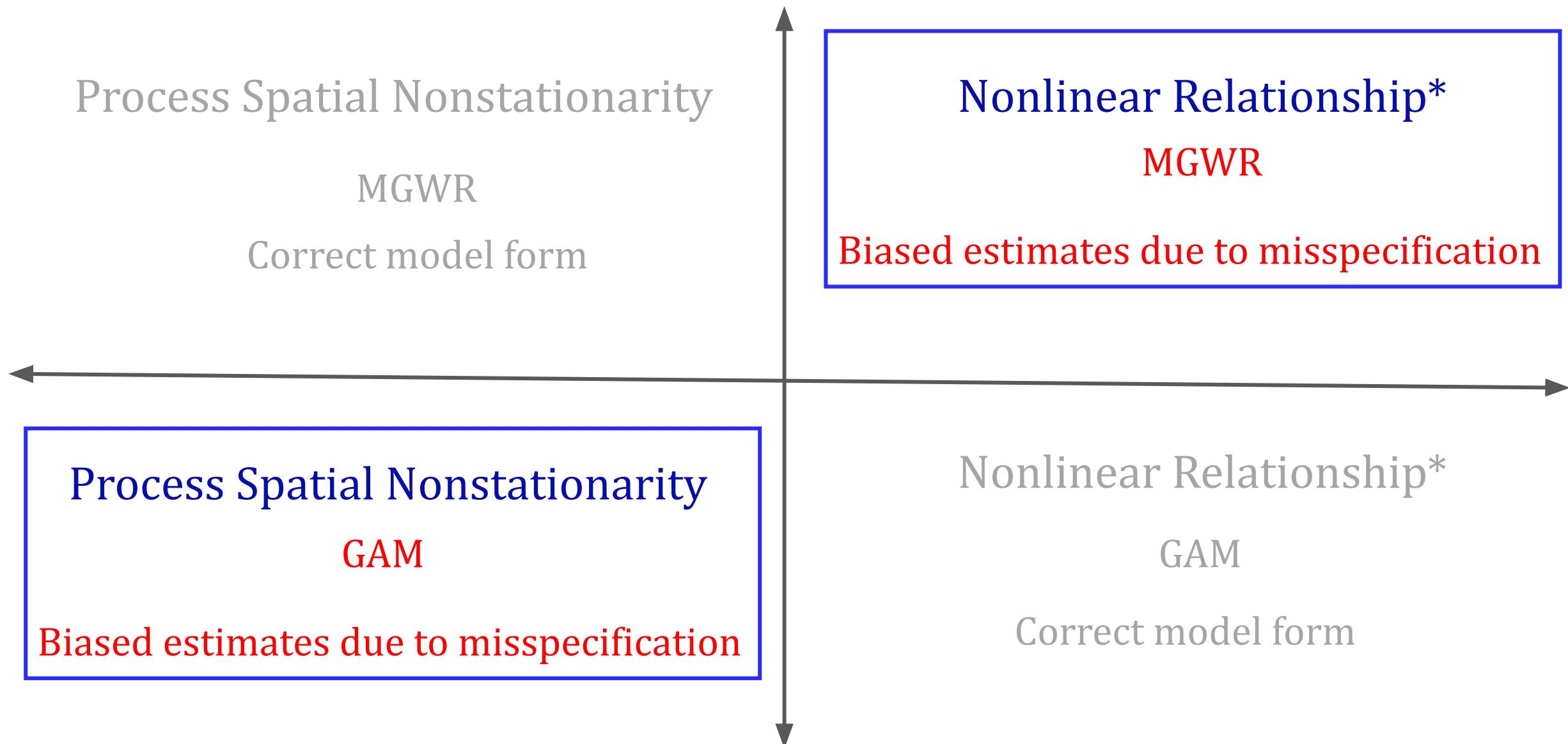
Residuals from GAM and MGWR



Moran's I value: Not significant

Process Nonstationarity Not Detected

Research describes **diagnostic tests** to detect these misspecifications



* Assuming the covariate is spatially varying

4. Intellectual merits and Broader impacts

- **Misinterpreting local estimates** due to misspecified covariate functional form plague the local analysis literature
- This **simple test must be a part of diagnostics** in local modeling
- **Expands the tools and tests within local spatial analysis** to enable stronger, more plausible interpretations

Thank you!

Any Questions?