

Can Conventional Measures Identify Geographically
Varying Mixed Regression Relationships? A
Simulation-based Analysis of Locally Weighted
Regression

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$$Y = \beta_0 + \beta_1 * X_1 + \beta_2 * X_2 + \epsilon$$

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Locally Weighted Regression (LWR)
to the Rescue?

OLS

$$\hat{\beta} = (X'X)^{-1}(X'Y)$$

LWR

$$\hat{\beta}(\textit{location}_i) = (X'W(\textit{location}_i)X)^{-1}(X'W(\textit{location}_i)Y)$$

OLS

$$\hat{\beta} = (X'X)^{-1}(X'Y)$$

LWR

$$\hat{\beta}(\text{location}_i) = (X'W(\text{location}_i)X)^{-1}(X'W(\text{location}_i)Y)$$

$$w_{jj} = \left[1 - \left(\frac{d_{ij}}{d_{ik}} \right)^2 \right]^2 \text{ if } d_{ij} < d_{ik}, \text{ otherwise } = 0,$$

Bandwidths are commonly selected with...

Leave One Out Cross Validation

Akaike Information Criterion

Generalized Cross Validation

Standardized Cross Validation

$$LOOCV = \frac{1}{N} \sqrt{\sum_{i=1}^N (y - \hat{y}_{\neq i})^2},$$

$$SCV_i(k) = \frac{(y_i - \hat{y}_{\neq i}(k))^2}{\sum_k (y_i - \hat{y}_{\neq i})^2}$$

$$SCV(k) = \sum_i SCV_i(k)$$

S Farber and A Páez. A systematic investigation of cross-validation in GWR model estimation: empirical analysis and Monte Carlo simulations. *Journal of Geographical Systems*, 9(4):371–396, 2007.

$$GCV = n * \sum_{i=1}^n \frac{(y_i - \hat{y}_i)^2}{(n - v_1)^2},$$

v_1 = “effective number of parameters”

$$AIC = 2 * n * \ln(\hat{\sigma}) + n * \ln(2 * \pi) + n * \frac{n + v_1}{n - 2 - v_1}$$

“Global” OLS

GGG

“Local” Regression

LLL

“Global” OLS

GGG

LGG

GLG

GGL

Mixed Models

LLG

LGL

GLL

“Local” Regression

LLL

The Researcher's Problem

Choose a bandwidth and a model

The Researcher's Problem

Choose a bandwidth and a model

Can we do both with conventional metrics?

Experiment Data Generation Process

$$Y_i = \beta_0(East_i, North_i) + \beta_1(East_i, North_i) * X_{1i} + \beta_2(East_i, North_i) * X_{2i} + \epsilon_i$$

$$n \in \{50, 100, 200, 400, 800\} \quad X_1 \sim u[0, 1]$$

$$\sigma^2 \in \{0.25, .5, 1, 2, 3\} \quad X_2 \sim u[0, 1]$$

$$East \sim u[0, 1]$$

$$North \sim u[0, 1]$$

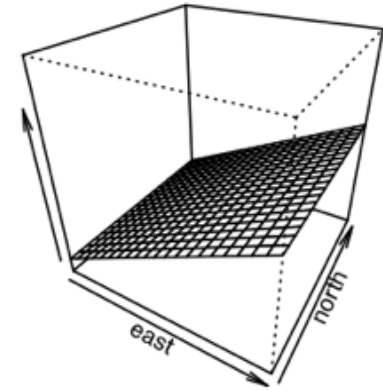
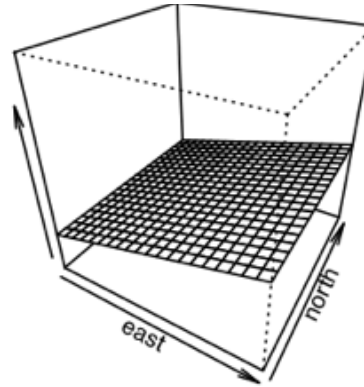
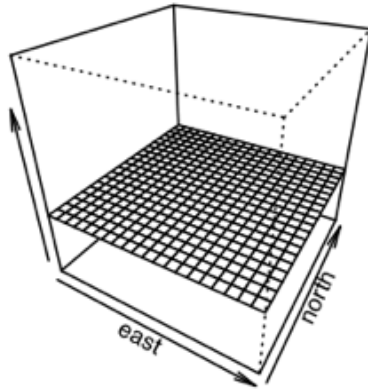
Coefficient Spatial Variation

Global

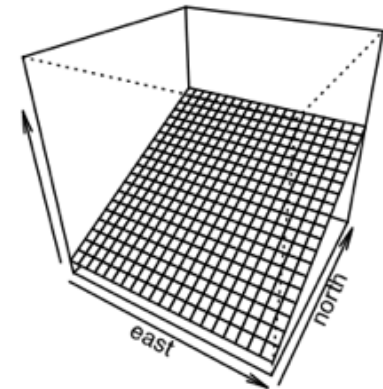
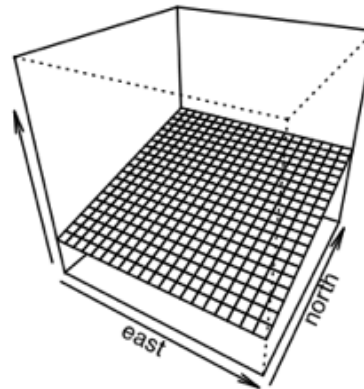
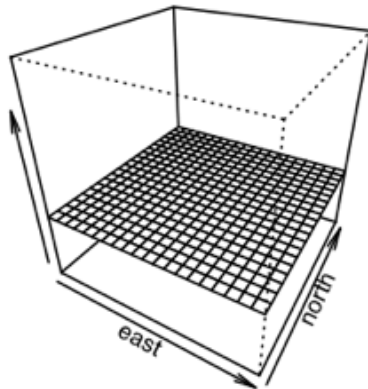
Local

(more) Local

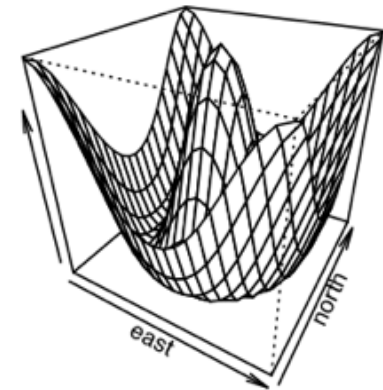
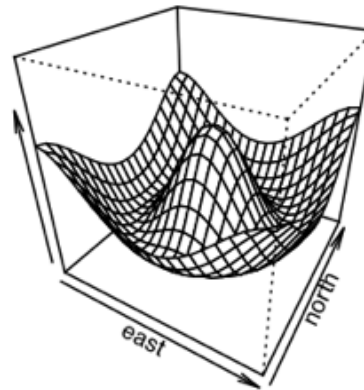
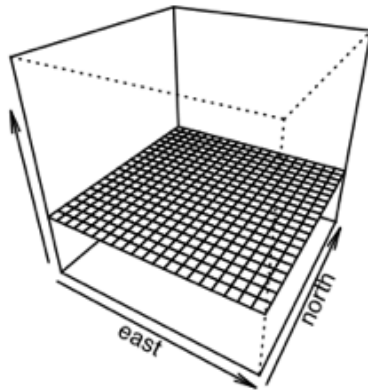
β_1



β_2



β_0



With Our Data...

$[Y, X_1, X_2, \textit{East}, \textit{North}]$

Estimate all models

GGG

LGG

GLG

GGL

LLG

LGL

GLL

LLL



7 bandwidths each

50 combinations total

Calculate the values of the
four metrics
(LOOCV, GCV, SCV, AIC)

Is the model with the optimized
metric value the correct model?

Start Simple... True Model: GGG

		Metric				
		LOOCV	GCV	SCV	AIC	
Model Selected	GGG	72	0	8	0	3/3 Correct
	LGG	7	28	29	28	2/3 Correct
	GLG	8	36	22	37	
	GGL	8	33	22	34	
	LLG	1	1	5	0	1/3 Correct
	LGL	2	1	5	1	
	GLL	1	1	8	0	
	LLL	0	0	1	0	0/3 Correct
		100	100	100	100	

True Model: GGG

metric: LOOCV

smallest — bandwidth size — biggest

model selected	GGG						72	
	LGG	1		1	2	1	2	
	GLG		1	1	2	2	2	1
	GGL		1	1	2	2	2	1
	LLG					1		
	LGL						1	
	GLL							
	LLL							

True Model: GGG

metric: GCV

smallest — bandwidth size — biggest

GGG							
LGG	1	1	2	2	2	4	17
GLG		1	2	2	3	5	23
GGL		1	1	2	3	4	21
LLG							
LGL							
GLL							
LLL							

metric: SCV

smallest — bandwidth size — biggest

GGG						8	
LGG		2	12	10	5	1	
GLG		2	6	7	5	2	
GGL		1	7	8	4	1	
LLG				2	2	1	
LGL				2	2	1	
GLL			1	3	3	1	
LLL					1		

metric: AIC

smallest — bandwidth size — biggest

GGG							
LGG		1	2	2	4	19	
GLG		1	2	3	6	25	
GGL		1	2	3	4	23	
LLG							
LGL							
GLL							
LLL							

True Model: GGG

Model Selected	Coefficient RMSE		
	β_0	β_1	β_2
GGG	6	6	7
LGG	3	23	24
GLG	22	4	24
GGL	23	23	3
LLG	5	3	33
LGL	5	36	2
GLL	32	4	4
LLL	4	2	3
	100	100	100

True Model: GGG

metric: B0RMSE

smallest — bandwidth size — biggest

GGG							6
LGG						1	3
GLG	9	3	2	2	2	2	1
GGL	9	3	3	3	2	1	1
LLG							4
LGL							4
GLL	14	5	4	4	3	2	1
LLL							4

metric: B1RMSE

smallest — bandwidth size — biggest

GGG							6
LGG	8	4	3	3	2	2	1
GLG							3
GGL	10	3	3	2	2	2	1
LLG							3
LGL	15	5	4	4	3	2	3
GLL							4
LLL							2

metric: B2RMSE

smallest — bandwidth size — biggest

GGG							7
LGG	9	4	3	2	2	2	2
GLG	10	4	2	1	2	2	2
GGL							3
LLG	12	5	4	4	3	3	2
LGL							2
GLL							4
LLL							3

True Model: GGG

		metric: B0RMSE									metric: B1RMSE									metric: B2RMSE						
		smallest — bandwidth size — biggest									smallest — bandwidth size — biggest									smallest — bandwidth size — biggest						
model selected	GGG							6	model selected	GGG							6	model selected	GGG							7
	LGG						1	3		LGG	8	4	3	3	2	2	1		LGG	9	4	3	2	2	2	2
	GLG	9	3	2	2	2	2	1		GLG							3		GLG	10	4	2	1	2	2	2
	GGL	9	3	3	3	2	1	1		GGL	10	3	3	2	2	2	1		GGL							3
	LLG							4		LLG							3		LLG	12	5	4	4	3	3	2
	LGL							4		LGL	15	5	4	4	3	2	3		LGL							2
	GLL	14	5	4	4	3	2	1		GLL							4		GLL							4
	LLL							4		LLL							2		LLL							3

Most accurate coefficient estimates:

- Tend not to be the correct model
- Allow other coefficients to vary

true model: LGG

model selection metric

		LOOCV	GCV	SCV	AIC	B0	B1	B2
model selected	GGG	4		1			4	4
	LGG	87	90	82	89	94	32	33
	GLG	3	4	6	5	1	3	24
	GGL	3	4	6	5	1	24	4
	LLG	1	1	2		1	1	33
	LGL	1		2		1	34	1
	GLL			1		1	1	1
	LLL						1	1

		true model: LGG						
		model selection metric						
model selected		LOOCV	GCV	SCV	AIC	B0	B1	B2
	GGG	4		1			4	4
	LGG	87	90	82	89	94	32	33
	GLG	3	4	6	5	1	3	24
	GGL	3	4	6	5	1	24	4
	LLG	1	1	2		1	1	33
	LGL	1		2		1	34	1
	GLL			1		1	1	1
	LLL						1	1

		true model: GLG						
		model selection metric						
model selected		LOOCV	GCV	SCV	AIC	B0	B1	B2
	GGG	11		2		12		11
	LGG	6	10	12	10		2	26
	GLG	68	78	49	80	26	70	25
	GGL	4	7	6	7	24	2	
	LLG	5	2	11	1	3	5	33
	LGL	1	1	2		1	4	1
	GLL	4	2	16	1	31	12	2
	LLL	1		1		3	4	1

		true model: GGL						
		model selection metric						
model selected		LOOCV	GCV	SCV	AIC	B0	B1	B2
	GGG	11		2		12	11	
	LGG	6	9	12	9	1	25	2
	GLG	5	8	5	8	24	1	2
	GGL	69	80	50	81	26	26	70
	LLG	1		2		1	2	3
	LGL	4	2	11	1	3	32	5
	GLL	4	2	16	1	31	2	12
	LLL	1		2		3	1	5

		true model: LLG						
		model selection metric						
model selected		LOOCV	GCV	SCV	AIC	B0	B1	B2
	GGG	3		1				5
	LGG	70	73	63	75	72	7	31
	GLG	6	6	11	8	1	32	25
	GGL	3	3	5	4	1	5	1
	LLG	17	17	15	13	21	14	35
	LGL	1		2		1	9	1
	GLL	1		4		2	21	2
	LLL					1	11	1

		true model: LGL						
		model selection metric						
model selected		LOOCV	GCV	SCV	AIC	B0	B1	B2
	GGG	3		1			5	
	LGG	71	73	65	75	73	31	8
	GLG	3	3	5	4	1	1	5
	GGL	6	6	10	8	1	26	34
	LLG	1		1		1	1	8
	LGL	17	17	15	13	20	34	13
	GLL	1		3		2	1	20
	LLL					1	1	11

		true model: GLL						
		model selection metric						
model selected		LOOCV	GCV	SCV	AIC	B0	B1	B2
	GGG	6		1		11	1	1
	LGG	6	9	10	9		2	2
	GLG	13	18	10	20	25	13	2
	GGL	13	18	10	20	26	2	13
	LLG	3	2	4	1	1	10	4
	LGL	3	2	5	1	1	4	11
	GLL	53	51	52	49	33	57	58
	LLL	3	1	8		3	10	10

true model: GGG
model selection metric

model selected		LOOCV	GCV	SCV	AIC	B0	B1	B2
	GGG	72		8		6	6	7
	LGG	7	28	29	28	3	23	24
	GLG	8	36	22	37	22	4	24
	GGL	8	33	22	34	23	23	3
	LLG	1	1	5		5	3	33
	LGL	2	1	5	1	5	36	2
	GLL	1	1	8		32	4	4
	LLL			1		4	2	3

true model: LGG
model selection metric

model selected		LOOCV	GCV	SCV	AIC	B0	B1	B2
	GGG	4		1			4	4
	LGG	87	90	82	89	94	32	33
	GLG	3	4	6	5	1	3	24
	GGL	3	4	6	5	1	24	4
	LLG	1	1	2		1	1	33
	LGL	1		2		1	34	1
	GLL			1		1	1	1
	LLL						1	1

true model: GLG
model selection metric

model selected	LOOCV	GCV	SCV	AIC	B0	B1	B2
GGG	11		2		12		11
LGG	6	10	12	10		2	26
GLG	68	78	49	80	26	70	25
GGL	4	7	6	7	24	2	
LLG	5	2	11	1	3	5	33
LGL	1	1	2		1	4	1
GGL	4	2	16	1	31	12	2
LLL	1		1		3	4	1

true model: LLG
model selection metric

model selected	LOOCV	GCV	SCV	AIC	B0	B1	B2
GGG	3		1				5
LGG	70	73	63	75	72	7	31
GLG	6	6	11	8	1	32	25
GGL	3	3	5	4	1	5	1
LLG	17	17	15	13	21	14	35
LGL	1		2		1	9	1
GGL	1		4		2	21	2
LLL					1	11	1

true model: GGL

model selection metric

	LOOCV	GCV	SCV	AIC	B0	B1	B2
GGG	11		2		12	11	
LGG	6	9	12	9	1	25	2
GLG	5	8	5	8	24	1	2
GGL	69	80	50	81	26	26	70
LLG	1		2		1	2	3
LGL	4	2	11	1	3	32	5
GLL	4	2	16	1	31	2	12
LLL	1		2		3	1	5

true model: LGL

model selection metric

	LOOCV	GCV	SCV	AIC	B0	B1	B2
GGG	3		1			5	
LGG	71	73	65	75	73	31	8
GLG	3	3	5	4	1	1	5
GGL	6	6	10	8	1	26	34
LLG	1		1		1	1	8
LGL	17	17	15	13	20	34	13
GLL	1		3		2	1	20
LLL					1	1	11

true model: GLL
model selection metric

model selected	LOOCV	GCV	SCV	AIC	B0	B1	B2
GGG	6		1		11	1	1
LGG	6	9	10	9		2	2
GLG	13	18	10	20	25	13	2
GGL	13	18	10	20	26	2	13
LLG	3	2	4	1	1	10	4
LGL	3	2	5	1	1	4	11
GLL	53	51	52	49	33	57	58
LLL	3	1	8		3	10	10

true model: LLL
model selection metric

model selected	LOOCV	GCV	SCV	AIC	B0	B1	B2
GGG	2				1	1	
LGG	63	65	55	69	61	8	8
GLG	5	5	7	7	2	16	6
GGL	4	5	7	6	2	5	17
LLG	8	8	8	5	11	14	10
LGL	8	8	8	6	11	9	14
GLL	1	1	8		2	31	30
LLL	9	9	6	6	12	15	15

Some Conclusions and Questions

LOOCV is pretty good!

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What happens, *ceteris paribus*, with greater error variance?