

PALESTRA



Estado da Arte nas Avaliações por Regressões Espaciais

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Promoção:



Realização:



1

Inferência Estatística

- Clássica
 - Regressão Tradicional
- Espacial
 - Regressão Espacial
 - Geoestatística



2

Efeitos Espaciais

- Dependência Espacial:

“As coisas mais próximas são mais parecidas”

Waldo TOBLER

- Heterogeneidade Espacial:

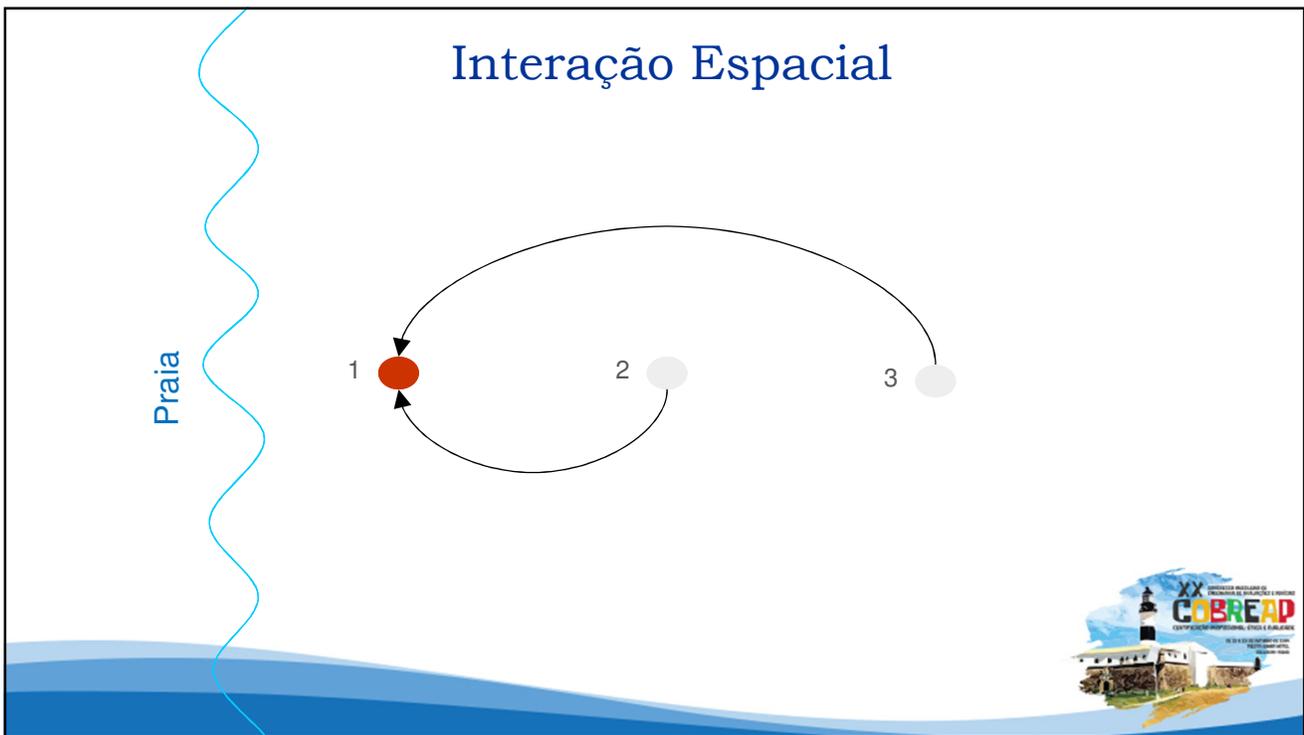
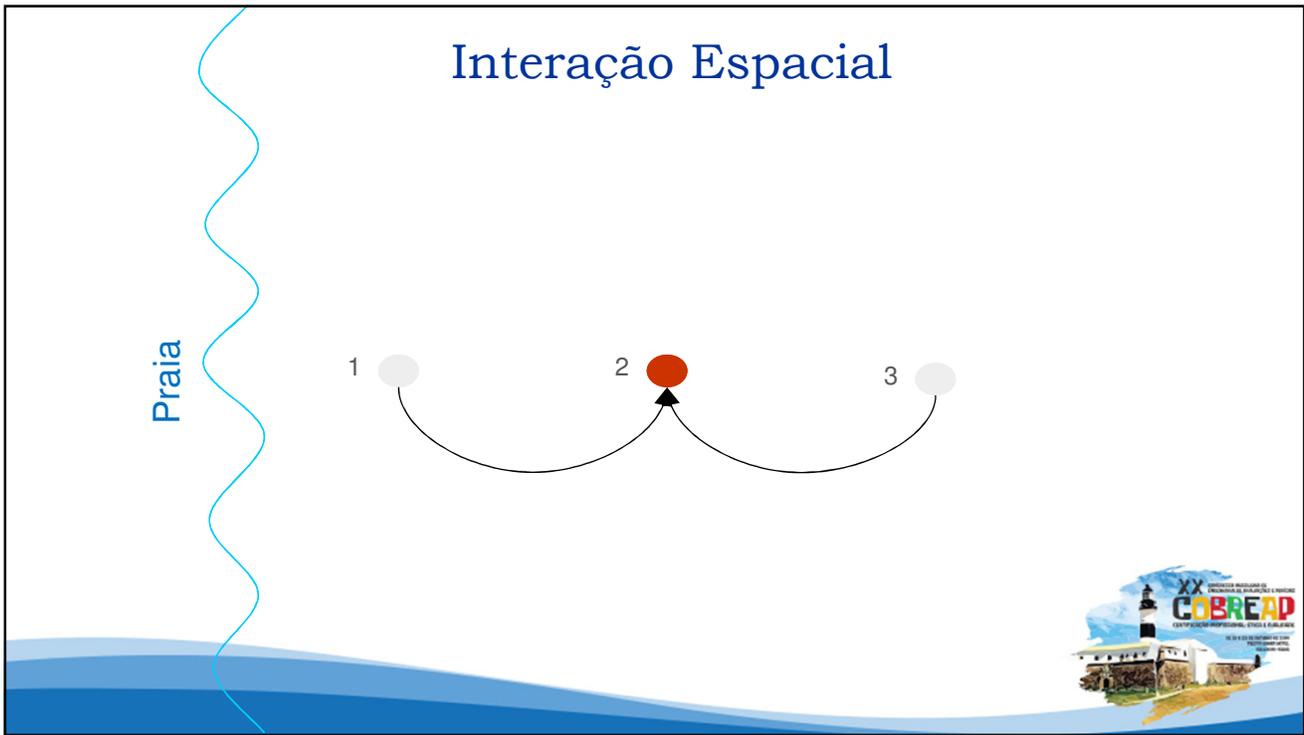
Instabilidade dos parâmetros do Modelo no espaço



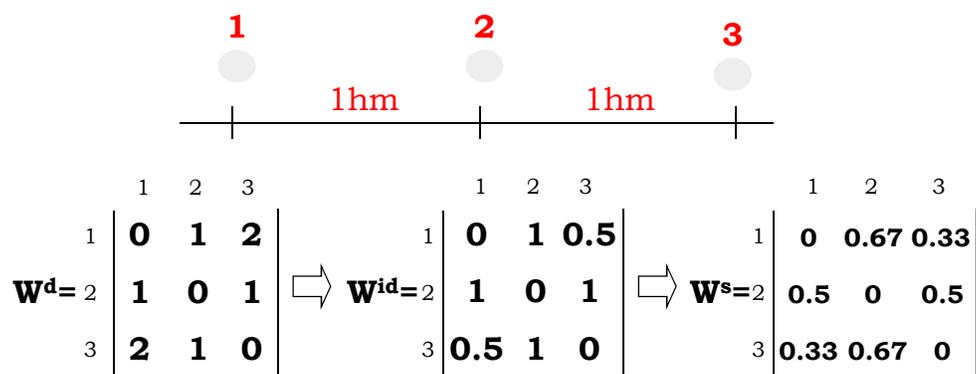
Interação Espacial

Praia





Matríz de Distancia



Variável Dependente Espacialmente Defasada

Diagram illustrating the spatially lagged dependent variable model with three points (1, 2, 3) on a line, with distances of 1hm between adjacent points. The dependent variable values are $P_1 = 100$, $P_2 = 80$, and $P_3 = 70$.

$$W^s \times P = \begin{matrix} & \begin{matrix} 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \end{matrix} & \begin{vmatrix} 0 & 0,67 & 0,33 \\ 0,5 & 0 & 0,5 \\ 0,33 & 0,67 & 0 \end{vmatrix} \end{matrix} \begin{vmatrix} 100 \\ 80 \\ 70 \end{vmatrix} = \begin{vmatrix} 76,70 \\ 85 \\ 86,6 \end{vmatrix}$$

- Modelo Clássico: $P = b_0 + b_1 * D$

- Modelo Defasagem Espacial: $P = b_0 + b_1 * D + b_3 * W * P$

- Modelo Erro Espacial: $P = b_0 + b_1 * D + b_3 * W * e$



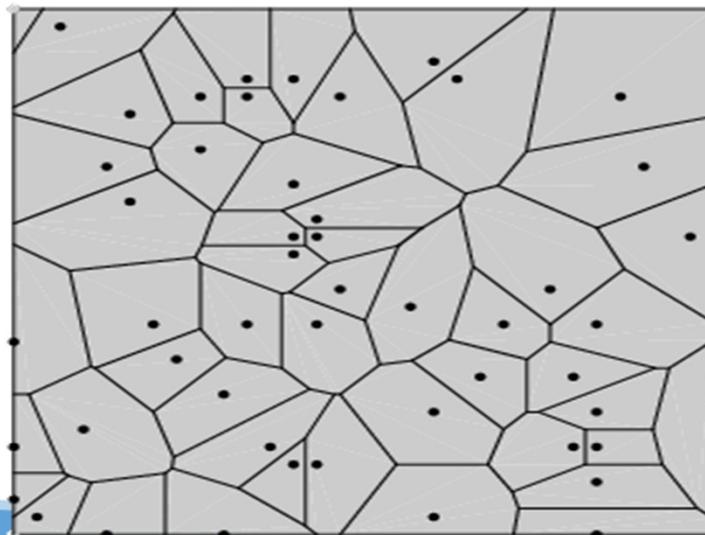
Dependência Espacial

- Distância (pontos)
- Contigüidade (polígonos)



9

Polígono de Voronoi



10

Matriz de Contiguidade W

Considera-se cada imóvel como uma unidade de vizinhança

- ⇒ $w_{ij} = 1$ se i e j são vizinhos;
- ⇒ $w_{ij} = 0$ caso contrario

A vizinhança pode ser do tipo:

- ⇒ Rook (Torre) : uma face comum
- ⇒ Bishop (Bispo) : uma quina comum
- ⇒ Queen (Rainha) : uma face ou quina comum



Matriz de Contiguidade tipo Rook para as Regiões do Brasil



$W =$

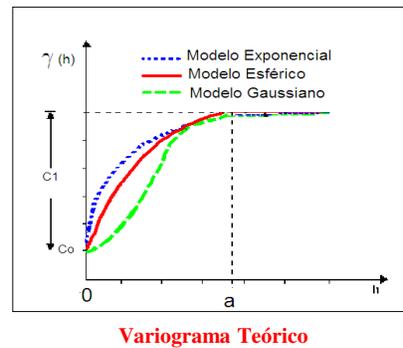
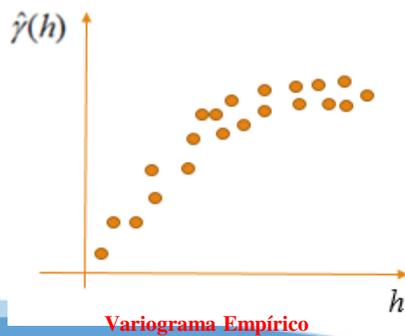
$$\begin{pmatrix} 0 & 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 & 0 \\ 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 1 & 0 \end{pmatrix}$$



Matriz de Covariância Espacial - Endógena

$$\hat{\gamma}(h) = \frac{1}{2n(h)} \sum_{i=1}^{n(h)} [z(s_i) - z(s_{i+h})]^2$$

$$C_{ij}(h) = \begin{cases} 0 & \text{for } i = j \text{ and } h = 0 \\ C_0 + C_1 - \gamma(h) & \text{for } i \neq j \text{ and } 0 < h < a \\ 0 & \text{for } i \neq j \text{ and } h > a \end{cases}$$



METODOLOGIA DE REGRESSÃO ESPACIAL

1º passo: Diagnóstico da Dependência Espacial

- Teste LM (erro)
- Teste LM (defasagem)

2º passo: Incorporação dos Efeitos Espaciais

- Modelo de Erro Espacial
- Modelo de Defasagem Espacial



Modelagem através da Regressão Espacial

Modelo de Erro Espacial

$$Y = X\beta + u,$$

onde $u = \lambda Wu + \epsilon$ e $y \approx N(0, \sigma^2 I)$

Sendo:

W - Matriz de Pesos Espaciais

λ - coeficiente de autocorrelação espacial do erro u

Modelo de Defasagem Espacial

$$Y = \rho WY + X\beta + \epsilon,$$

onde $y \approx N(0, \sigma^2 I)$

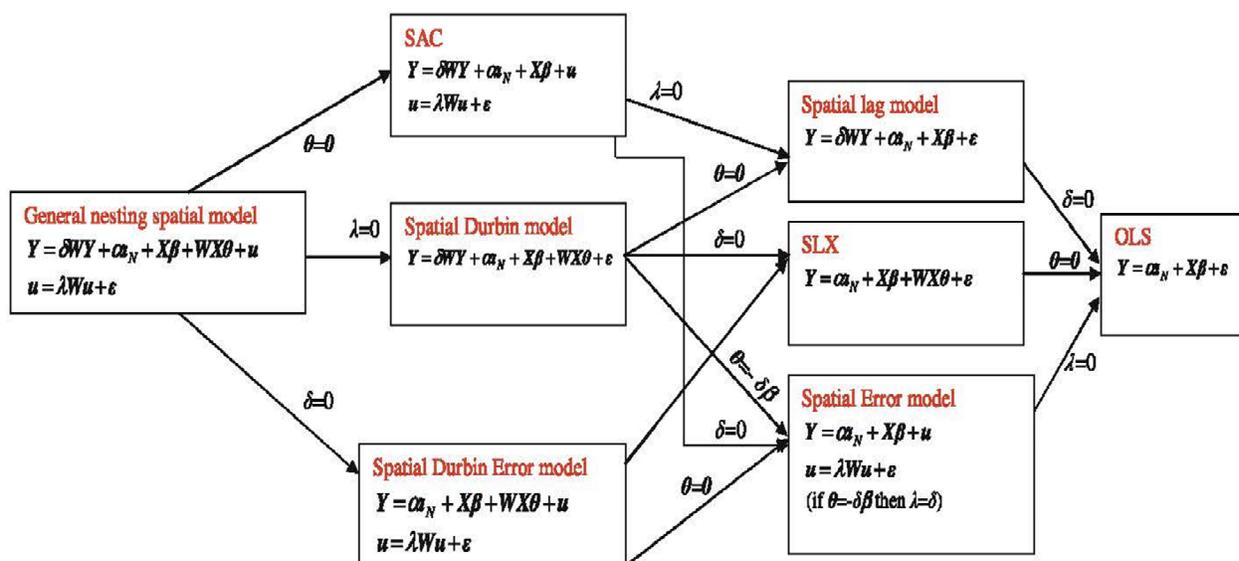
Sendo:

WY - variável dependente espacialmente defasada

ρ - coeficiente de autocorrelação espacial de WY



TIPOS DE MODELOS ESPACIAIS POSSÍVEIS



SAC = spatial autoregressive combined model, SLX = spatial lag of X model

Estimação dos Parâmetros: Método da Máxima Verossimilhança

- Funções de Log Verossimilhança: Defasagem Espacial

$$\ln L = -N/2 \ln 2\pi - N/2 \ln \sigma^2 + \ln |I - \rho W| - 1/2 \sigma^2 \{[(I - \rho W)y - X\beta]'[(I - \rho W)y - X\beta]\}$$

- Funções de Log Verossimilhança: Erro Espacial

$$\ln L(y/\lambda, \beta, \sigma^2) = -(N/2) \ln 2\pi - (N/2) \ln \sigma^2 + \ln |I - \lambda W| - 1/2 \sigma^2 \{[(I - \lambda W)(y - X\beta)]'[(I - \lambda W)(y - X\beta)]\}$$



Diagnósticos pelos Testes dos Multiplicadores de Lagrange - LM

$$\text{LM (erro)} = \frac{[e'W e / (e'e/N)]^2}{\text{tr}(W^2 + W'W)} \stackrel{a}{\approx} \chi^2_{(1)}$$

$$\text{LM (defasagem)} = \frac{[e'W y / (e'e/N)]^2}{\text{tr}(W^2 + W'W) + (X'X)^{-1} X' W y} \stackrel{a}{\approx} \chi^2_{(1)}$$



ESCOLHA DE MODELOS

I - Critério de informação de Akaike

$$AIC = -2L + 2k$$

II - Critério de Schwartz

$$SC = -2L + k * \ln(n)$$



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https://geodacenter.github.io

GeoDa

An Introduction to Spatial Data Analysis

Download View on GitHub Data Documentation Support 中文

Introducing GeoDa 1.12

GeoDa is a free and open source software tool that serves as an introduction to spatial data analysis. It is designed to facilitate new insights from data analysis by exploring and modeling spatial patterns.

GeoDa was developed by [Dr. Luc Anselin](#) and his team. The program provides a user-friendly and graphical interface to methods of exploratory spatial data analysis (ESDA), such as spatial autocorrelation statistics for aggregate data (several thousand records), and basic spatial regression analysis for point and polygon data (tens of thousands of records). To work with big data in GeoDa it should first be aggregated to areal units.

Since its initial release in February 2003, GeoDa's user numbers have increased exponentially to over 200,000 (June 2017). This includes lab users at universities such as Harvard, MIT, and Cornell. The

SUMMARY OF OUTPUT: ORDINARY LEAST SQUARES ESTIMATION

Data set : Columbus Crime
 Dependent Variable : hoval Number of Observations: 49
 Mean dependent var : 38.4362 Number of Variables : 3
 S.D. dependent var : 18.2767 Degrees of Freedom : 46

R-squared : 0.349514 F-statistic : 12.3582
 Adjusted R-squared : 0.321232 Prob(F-statistic) : 5.0637e-005
 Sum squared residual: 10647 Log likelihood : -201.368
 Sigma-square : 231.457 Akaike info criterion : 408.735
 S.E. of regression : 15.2137 Schwarz criterion : 414.411
 Sigma-square ML : 217.286
 S.E of regression ML: 14.7406

Variable	Coefficient	Std. Error	t-Statistic	Probability
CONSTANT	46.4282	13.1918	3.51948	0.00099
inc	0.628984	0.53591	1.17367	0.24657
crime	-0.484889	0.182673	-2.65441	0.01087

REGRESSION DIAGNOSTICS
 MULTICOLLINEARITY CONDITION NUMBER 12.537555
 TEST ON NORMALITY OF ERRORS

TEST	DF	VALUE	PROB
Jarque-Bera	2	39.7062	0.00000

DIAGNOSTICS FOR HETEROSKEDASTICITY
 RANDOM COEFFICIENTS

TEST	DF	VALUE	PROB
Breusch-Pagan test	2	5.7668	0.05594
Koenker-Bassett test	2	2.2700	0.32142

DIAGNOSTICS FOR SPATIAL DEPENDENCE
 FOR WEIGHT MATRIX : WColumbusRook
 (row-standardized weights)

TEST	MI/DF	VALUE	PROB
Moran's I (error)	0.2071	2.5040	0.01228
Lagrange Multiplier (lag)	1	1.8195	0.17737
Robust LM (lag)	1	0.6978	0.40351
Lagrange Multiplier (error)	1	3.9929	0.04569
Robust LM (error)	1	2.8713	0.09017

SUMMARY OF OUTPUT: SPATIAL LAG MODEL - MAXIMUM LIKELIHOOD ESTIMATION

Data set : Columbus Crime
 Spatial Weight : WColumbusRook
 Dependent Variable : hoval Number of Observations: 49
 Mean dependent var : 38.4362 Number of Variables : 4
 S.D. dependent var : 18.2767 Degrees of Freedom : 45
 Lag coeff. (Rho) : 0.23063

R-squared : 0.382337 Log likelihood : -200.439
 Sq. Correlation : - Akaike info criterion : 408.879
 Sigma-square : 206.322 Schwarz criterion : 416.446
 S.E of regression : 14.3639

Variable	Coefficient	Std. Error	z-value	Probability
W_hoval	0.23063	0.154956	1.48836	0.13666
CONSTANT	37.2681	13.9456	2.67239	0.00753
inc	0.563156	0.513164	1.09742	0.27246
crime	-0.45102	0.177697	-2.53815	0.01114

REGRESSION DIAGNOSTICS
 DIAGNOSTICS FOR HETEROSKEDASTICITY
 RANDOM COEFFICIENTS

TEST	DF	VALUE	PROB
Breusch-Pagan test	2	5.9099	0.05208

DIAGNOSTICS FOR SPATIAL DEPENDENCE
 SPATIAL LAG DEPENDENCE FOR WEIGHT MATRIX : WColumbusRook

TEST	DF	VALUE	PROB
Likelihood Ratio Test	1	1.8567	0.17300

```

SUMMARY OF OUTPUT: SPATIAL ERROR MODEL - MAXIMUM LIKELIHOOD ESTIMATION
Data set           : Columbus Crime
Spatial Weight     : WColumbusRook
Dependent Variable : hoval      Number of Observations: 49
Mean dependent var : 38.436224  Number of Variables   : 3
S.D. dependent var : 18.276669   Degrees of Freedom    : 46
Lag coeff. (Lambda) : 0.390498

R-squared         : 0.428509   R-squared (BUSE)      : -
Sq. Correlation   : -          Log likelihood         : -199.232493
Sigma-square      : 190.899   Akaike info criterion : 404.465
S.E of regression : 13.8166   Schwarz criterion     : 410.14

```

Variable	Coefficient	Std.Error	z-value	Probability
CONSTANT	48.0083	12.3126	3.89912	0.00010
inc	0.711532	0.497923	1.429	0.15300
crime	-0.559458	0.178773	-3.12943	0.00175
LAMBDA	0.390498	0.157587	2.47798	0.01321

REGRESSION DIAGNOSTICS

DIAGNOSTICS FOR HETEROSKEDASTICITY

RANDOM COEFFICIENTS

TEST	DF	VALUE	PROB
Breusch-Pagan test	2	5.6443	0.05948

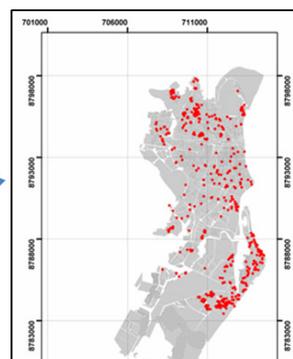
DIAGNOSTICS FOR SPATIAL DEPENDENCE

SPATIAL ERROR DEPENDENCE FOR WEIGHT MATRIX : WColumbusRook

TEST	DF	VALUE	PROB
Likelihood Ratio Test	1	4.2705	0.03878

Empirical Evidence using Spatial Covariance Matrix -WC

- ❖ Dataset used for illustration consists of 479 land lots transactions distributed along Atalaia Beach at the city of Aracaju.

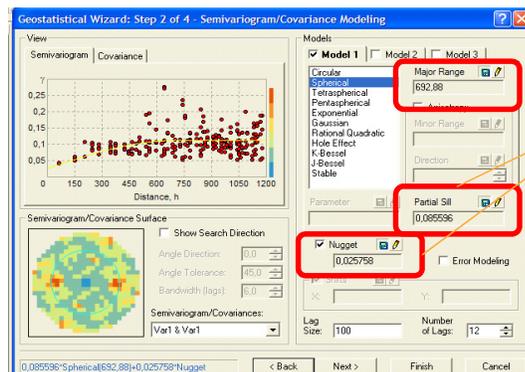


Available Information

- ❖ Land Lot Area (AR): Log-transformed;
- ❖ Sector (SE): Proxy Variable (average income of the head of household) Log-transformed;
- ❖ Topography (TO): Dummy indicating (plan)
- ❖ Location (SI): Dummy variable (corner)
- ❖ Paving: Dummy variable (pave)
- ❖ Frontage (FR): Quantitative variable Land-use Coefficient (CA): Variable interval that identifies the relationship between the land area and the area that can be constructed - Log-transformed;
- ❖ Axis (EI): Ordinal variable 3 - principal axis, 2 - secondary axis and 1 for the other cases;
- ❖ Nature of the event.



Variogram



$$(0.025758 / 0.111354) \% = 23 \%$$

- ❖ 23% of the total variability of the residuals is due to the random component;
- ❖ 77% are explained by the presence of the spatial autocorrelation in the residuals.



SPATIAL DEPENDENCE DIAGNOSE

	LMRobust Error		LMRobust lag	
WD	8.9330	0.0030	8.9110	0.0030
WG	6.0110	0.0140	109.7527	0.0000

	WD		WC	
	SAR	SEM	SAR	SEM
AIC	412,91	412,69	358,54	324,16
SC	452,62	458,58	414,42	370,05



Literatura Recente

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Obrigado

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