



**Disciplina SER 301 – Análise Espacial**  
**Professores: Eduardo G. Camargo e Antônio Miguel V. Monteiro**

# The Characteristic Analysis and Forecasting of Mid-Long Term Load Based on Spatial Autoregressive Model

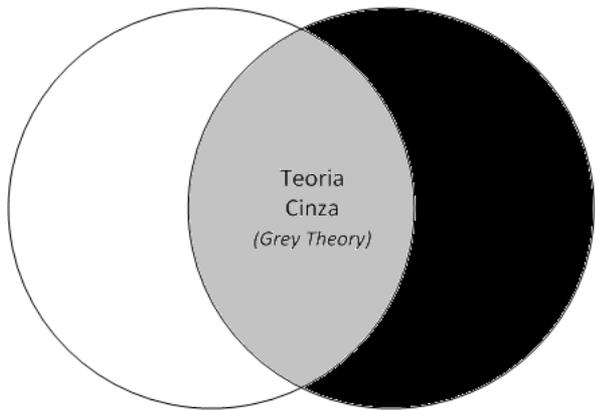
Guowei Cai, Deyou Yang, Ying Jiao, Chao  
Pan

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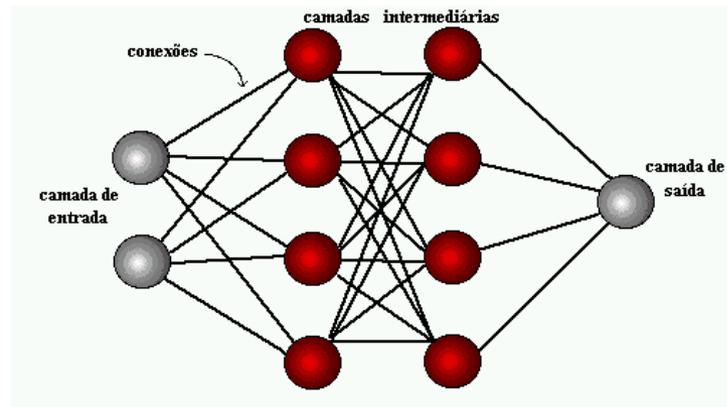
Novembro de 2011.

# Objetivo:

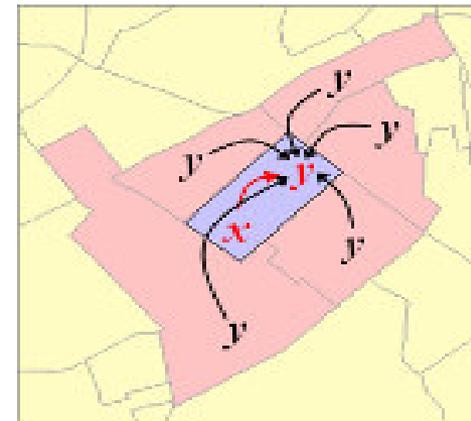
- Analisar a característica espacial entre o consumo de energia e PIB.
- Um modelo combinado é apresentado:



GM (1,1)

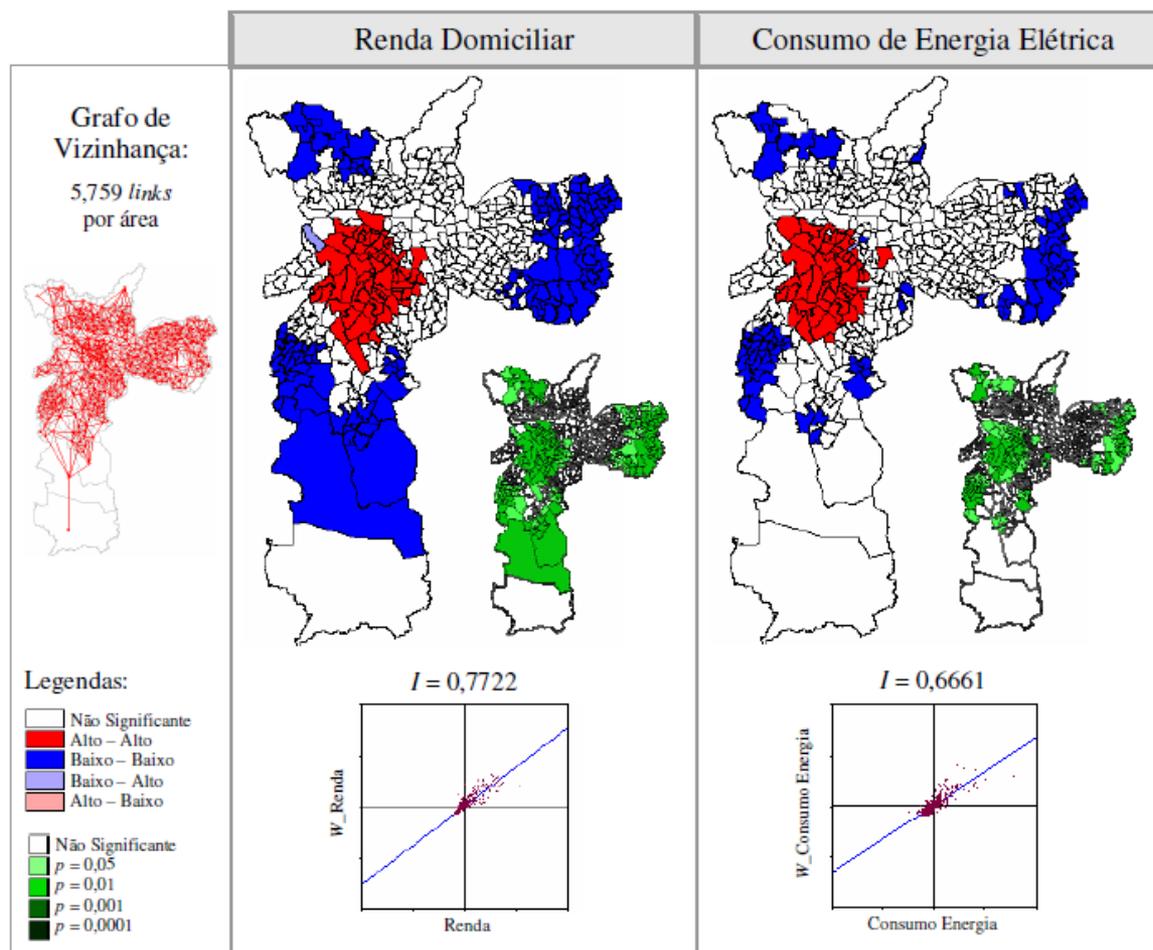


RN



SAR

# Porque é importante realizar um análise especial?:



# Modelo Cinza (GM) (1,1)

- Passo 1:  $X^0 = [x^0(1), x^0(2), \dots, x^0(n)]$  Na qual:  $x^0(i)$
- Passo 2:  $X^1 = [x^1(1), x^1(2), \dots, x^1(n)]$  Sendo que :  $x^1(k) = \sum_{m=1}^k x^0(m)$
- Passo 3:  $\frac{dx^1(k)}{dk} + ax^1(k) = u$
- A função de predição :  $\hat{x}^1(k) = (x^1(1) - \frac{u}{a})e^{-ak} + \frac{u}{a}$   
Para o tempo  $k+1$  :  $\hat{x}^0(k+1) = \hat{x}^1(k+1) - \hat{x}^1(k)$

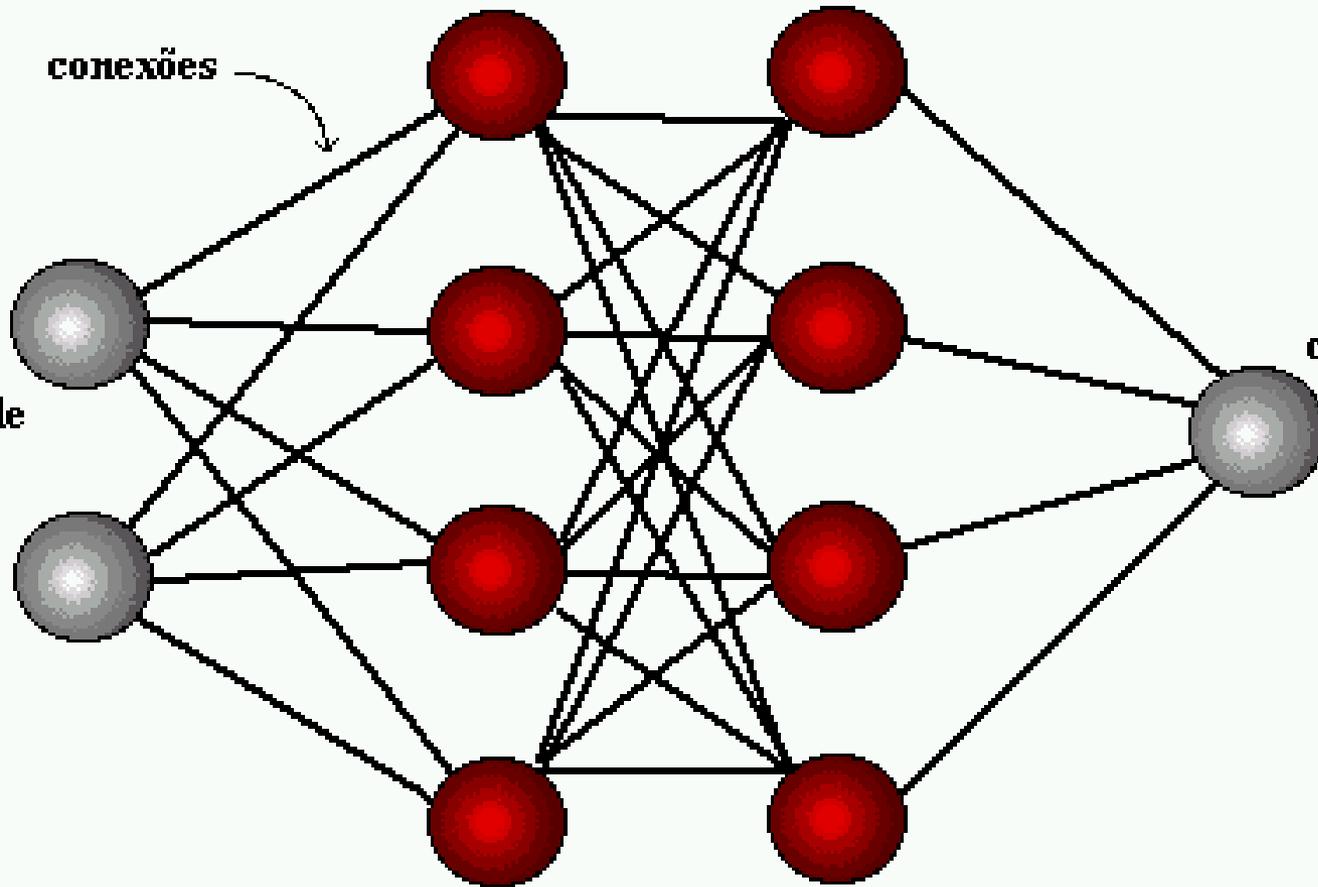
# Rede Neuronal

camadas intermediárias

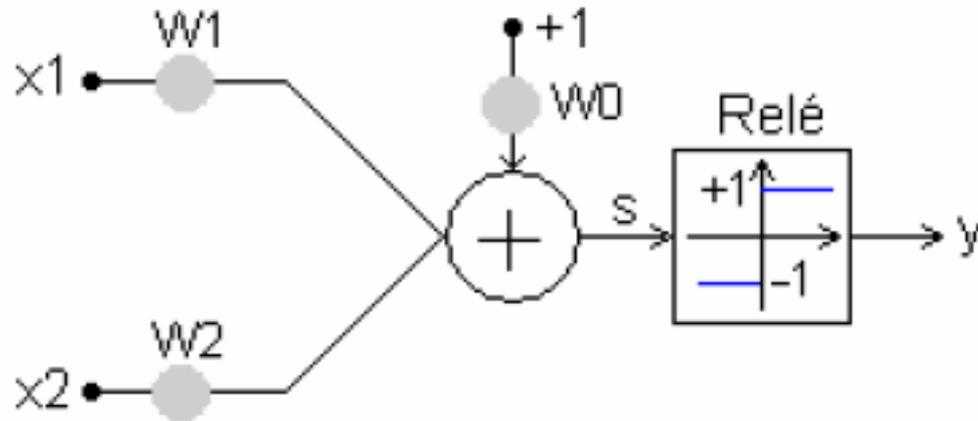
conexões

camada de entrada

camada de saída



# Rede Neuronal



$$s = W_1 x_1 + W_2 x_2 + W_0$$

# Modelo Espacial Autorregressivo :

$$y = \rho W_1 y + X\beta + \mu$$

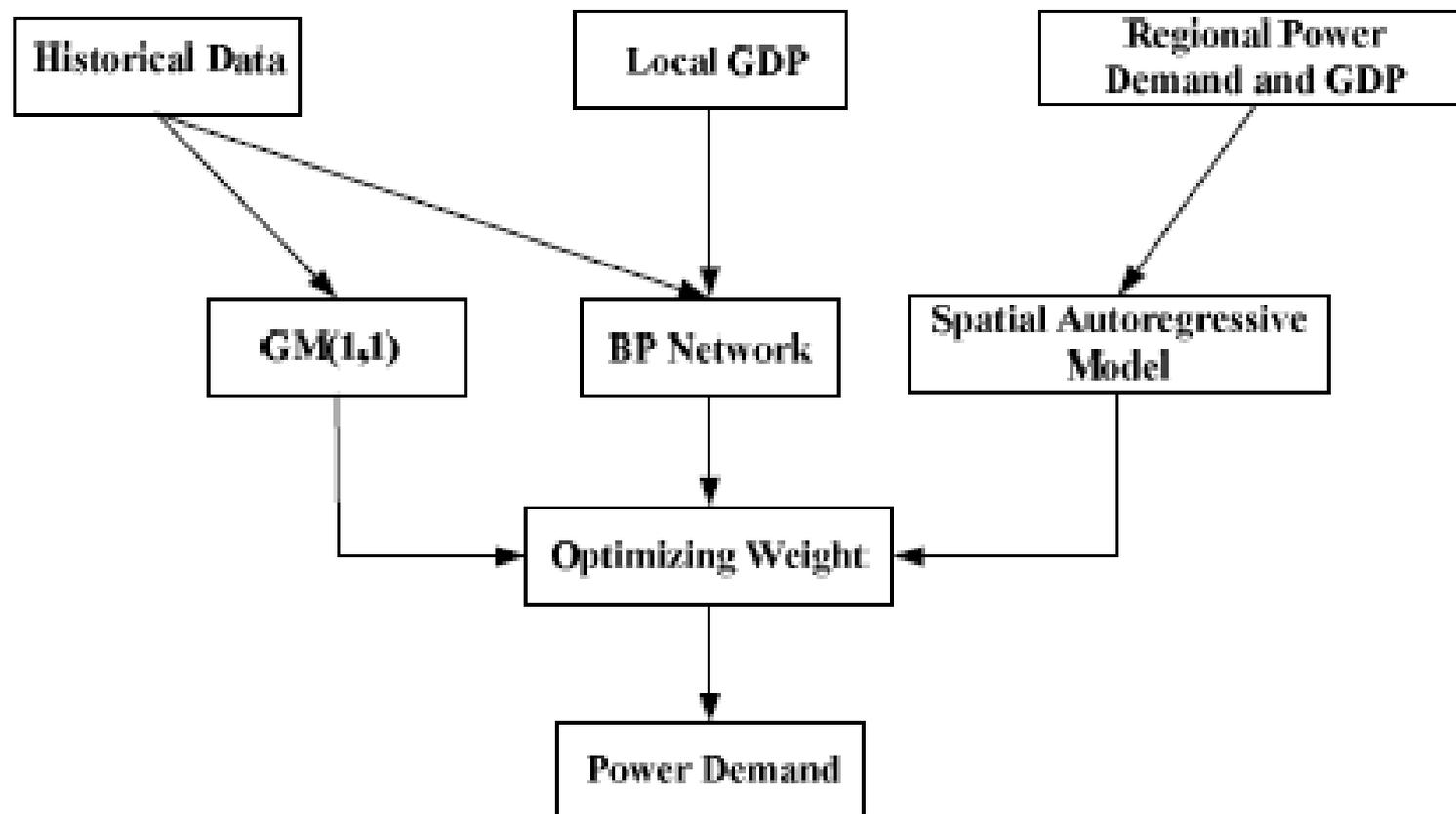
$$\mu = \lambda W_2 \mu + \varepsilon$$

$$\varepsilon \sim N(0, \sigma^2, I_n)$$

## Modelo de Otimização:

$$\left\{ \begin{array}{l} \min J = e \quad e = |y - \hat{y}| \\ \hat{y} = \sum_{i=1}^m \omega_i f_i \\ \sum_{i=1}^m \omega_i = 1 \\ \omega_i \geq 0 \quad i = 1, 2, \dots, m \end{array} \right.$$

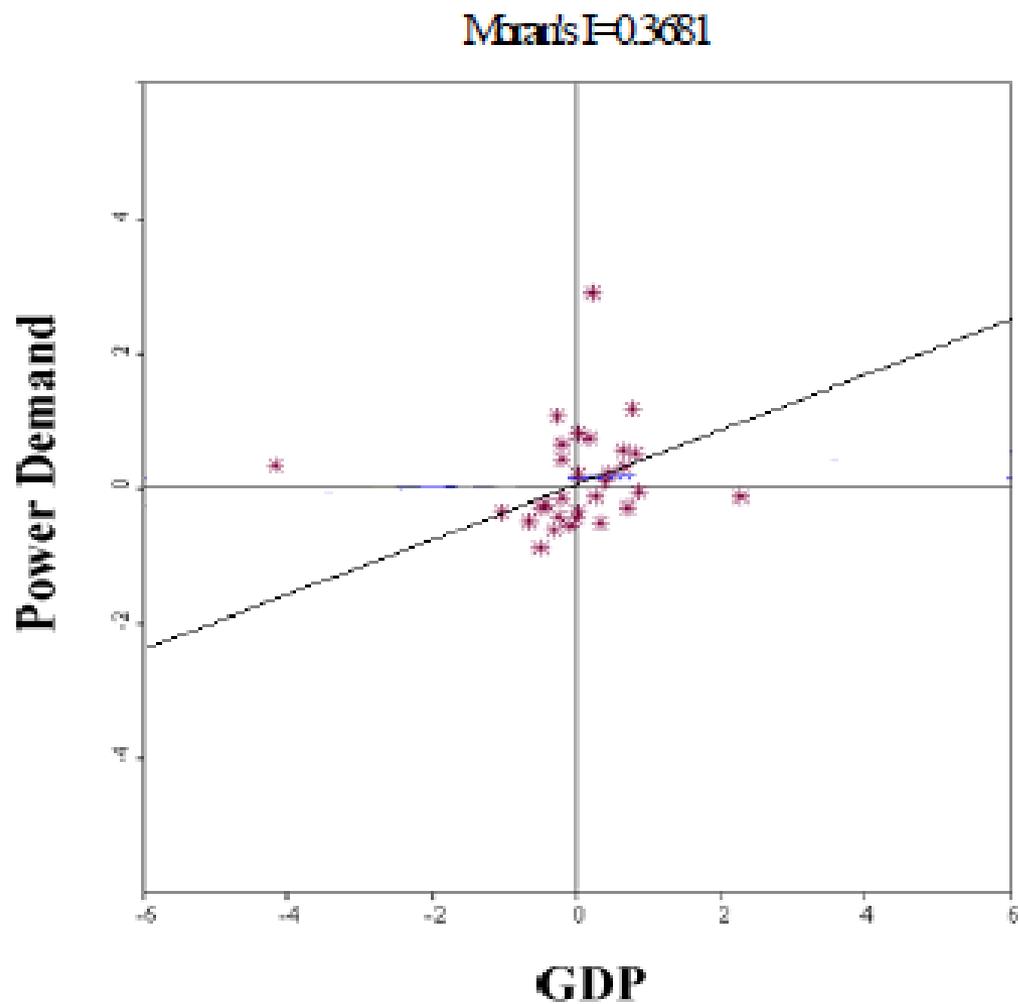
# Modelo Combinado:



# Resultados dos Modelos:

Model	Constant	$\beta$	$\rho$	$\lambda$	$R^2$
Regression Model	208.45 (0.001)	0.082 (0.008)			0.4571
Residual Spatially Autoregressive Model	236.21 (0.00)	0.061 (0.00)		0.6224 (0.00)	0.9112
Hybrid Spatially Autoregressive Model	176.32 (0.00)	0.053 (0.001)	0.4681 (0.00)		0.8252

# Moran Scatterplot:



# Previsão de Demanda:

Tab.2 Data for forecasting

Year	2001	2002	2003	2004	2005
Power Demand ( $10^9$ kwh)	295.08	306.27	338.7	371.8	378.22

Tab.3 Forecasting results

Method		Year	2006	2007
		Actual Value	412.26	466
Model I	Forecasted Facts	Forecasted Value	397.83	447.83
		Error (%)	3.5	3.9
	Actul Facts	Forecasted Value	401.13	451.09
		Error (%)	2.7	3.2
Model II	Forecasted Facts	Forecasted Value	433.7	448.84
		Error (%)	5.2	5.4
	Actul Facts	Forecasted Value	431.22	443.17
		Error (%)	4.6	4.9

# Comentários:

- Não se mencionaram características da área de estudo.
- Foi dada pouca informação dos modelos e as referências utilizadas são muito gerais para cada modelo.
- Deve fazer-se um comentário do poder preditivo dos modelos autorregressivos.
- Este artigo contribui com considerar modelos de regressão espacial dentro do estudo do consumo de energia nos sistemas elétricos.

# Referências:

- Cai, G.; Yang, D.; Jiao, Y. ; Pan, C. The Characteristic Analysis and Forecasting of Mid-Long Term Load Based on Spatial Autoregressive Model. In: INTERNATIONAL CONFERENCE ON SUSTAINABLE POWER GENERATION AND SUPPLY SUPERGEN, 9., 2009, Naijing. **Conference proceedings...** Naijing: IEEE, 2009. p. 1-6.
- Tieding, L.; Shijian, Z.; Wei, L.; Liting, Z. An improved algorithm of grey model-GM(1,1) based on total least squares and its application in deformation forecast. In: GREY SYSTEMS AND INTELLIGENT SERVICES, 2009. **Conference proceedings...** IEEE, 2009. p. 362 – 366.
- Saini, M.; Soni, K. Artificial neural network-based peak load forecasting using conjugate gradient methods. **IEEE Trans on Power Systems**, Vol.17, pp. 907-912, 2002

*Obrigado pela  
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