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Spatial AutoRegression (SAR) Model

Parameter Estimation Techniques



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Dedicated

to our advisors Prof. Shashi Shekhar and Prof. David Lilja.

Baris Kazar & Mete Celik

To my wife Neslihan, my teacher parents Mualla Hatice and Kemal, my brothers Ozgur and Caglar, my parent in-laws Fadime and Muzaffer, my brother in-law Emrullah, my sister in-law Emriye, and my whole family.

Baris Kazar

To my wife Filiz, my son Cagri Fatih, my daughter Elif Huma, my parents Fatma and Metin, and my whole family.

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