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# Spatial distribution dynamics: an application

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Spatial distribution  
dynamics:  
an application

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## Motivation:

- ▶ it is quite common in convergence analyses across spatial units (countries, regions) that data exhibit strong spatial dependence
  - ▶ neglecting spatial dependence may affect the results
- ⇒ employ SNP within the distribution dynamics approach for the analysis of cross-sectional convergence when data are spatially dependent

## Outline

- ▶ present data and discuss potential bias from cyclical dynamics
- ▶ analyze convergence among US states

## USA context

- ▶ 48 coterminous US states
- ▶ quarterly data on personal per capita income (1971:Q1-2010:Q4)
- ▶ orthodromic distance between state capitals

Allow for short-run, cyclical dynamics (Magrini, Gerolimetto and Duran, 2013; Gerolimetto and Magrini, 2014)

- ▶ the object of interest to convergence analysts is, essentially, the evolution of **potential** output
- ▶ measured output is a **noisy** indicator of potential output, contaminated by business cycle dynamics



In fact, policy makers need to discriminate between:

- ▶ a short-run component of the disparities (bound to vanish)
  - ▶ a long-run one (possibly requiring structural intervention)
- ⇒ the true object of interest is the evolution of (relative) **potential output**
- ▶ apply the Hodrick-Prescott filter to each regional time series to filter out short-run fluctuations
    - $\lambda$  (parameter controlling the degree of smoothness of the estimated trend) large enough to allow for all cyclical swings
  - ▶ choose any two points in time and study convergence on HP-filtered data
    - focus on the period between two major economic crises:  
1981:Q1-2007:Q2

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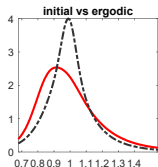
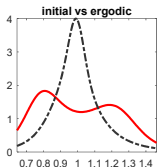
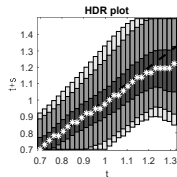
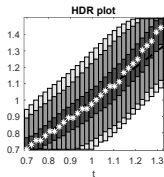
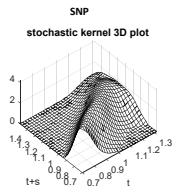
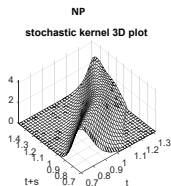
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# 1981:Q1-2007:Q2 with cross-validation minimization bandwidth

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Table: Results

	Moran's $I$	$p$ -value
observed initial	0.1993	0.0066
observed final	0.2877	0.0001
filtered initial	0.2206	0.0029
filtered final	0.3030	0.0001
residuals NP	0.4694	0.0000
residuals SNP	-0.0938	0.3627
	CV	IR
HP-filtered initial	0.1427	0.2144
HP-filtered final	0.1730	0.2382
ergodic NP	0.2189	0.4257
ergodic SNP	0.1819	0.2492

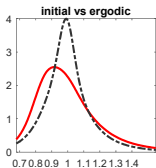
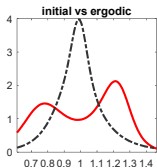
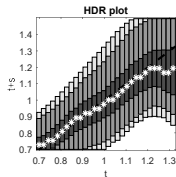
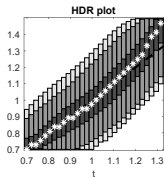
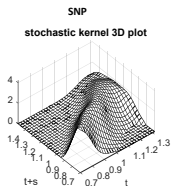
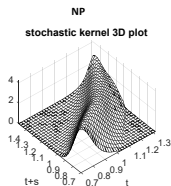
Table: Estimated half-life values

ergodic via NP	ergodic via SNP
7.3949	2.6620

# 1981:Q1-2007:Q2 with direct plug-in bandwidth

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Table: Results

	Moran's $I$	$p$ -value
observed initial	0.1993	0.0066
observed final	0.2877	0.0001
filtered initial	0.2206	0.0029
filtered final	0.3030	0.0001
residuals NP	0.4635	0.0000
residuals SNP	-0.0360	0.8528
	CV	IR
HP-filtered initial	0.1427	0.2144
HP-filtered final	0.1730	0.2382
ergodic NP	0.1817	0.2476
ergodic SNP	0.2186	0.4627

Table: Estimated half-life values

ergodic via NP	ergodic via SNP
9.5737	2.6622

Overall, we find

- ▶ evidence of a **very strong** process of **divergence** using NP to estimate the mean function
  - the ergodic distribution is clearly bimodal
- ▶ strong spatial dependence in data and NP regression residuals
- ▶ evidence of a **weak - moderate** process of **divergence** using SNP to estimate the mean function
  - no signs of bimodality in the ergodic distribution
- ▶ strong spatial dependence in data but no evidence of spatial dependence in SNP regression residuals
- ▶ results are robust to the choice of bandwidth

Specifically, results show that

- ▶ neglecting spatial dependence might affect the results
- ▶ the bias is particularly strong in the analysed period (1981:Q1-2007:Q2) stretching between two major crises

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