Thesis Progress Meeting

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Achieved Work (by projects)

- Biblio/Meetings/Organisation [0.7w]
- Conference [0.7w]
- Reading Records (Synergetics) [0.2w]
- Monitorat [1,3w]
- Cybergeo Project [1w]
- Correlated Synthetic data [3w]
- Theory construction [0.2w]
- BP Case Study / Spatial Econometrics [0,3w]

Context

Def.: Synthetic Data are output of generative models (and possibly inputs of models using them).

Methodology used in various fields, e.g. therapeutic evaluation [Abadie et al., 20 territorial systems analysis [Moeckel et al., 2003, Pritchard and Miller, 2009], machine learning [Bolón-Canedo et al., 2013] or bio-informatics [Van den Bulcke

Few examples at the second order: specific examples as [Ye, 2011] for discrete choices; methods that can be interpreted this way: generation of complex networks [Newman, 2003].

Generic Method

 \vec{X}_I multidimensional stochastic process, $\mathbf{X} = (X_{i,j})$ realizations.

Aim : Generate a statistical population $\mathbf{\tilde{X}} = \tilde{X}_{i,j}$ such that:

- proximity to data : given a precision ε and an indicator f, $\|f(\mathbf{X}) f(\tilde{\mathbf{X}})\| < \varepsilon$
- ② control of the estimated correlation structure : $\hat{Var}\left[(\tilde{X}_i)\right] = \Sigma R$ with R fixed.

Geographical data: Context and Objective

- In geography, generation of synthetic populations for agent-based models [Pritchard and Miller, 2009].
- Generation of spatial synthetic configuration not used (Geo. Weighted Regression [Brunsdon et al., 1998] can be interpreted this way); however crucial for abstract models [Schmitt, 2014]
- [Cottineau et al., 2015] recently proposed to estimate the sensitivity of spatial models of simulation to initial configuration (application to Schelling model).
- Case study: city-transportation interactions, complex to understood quantitatively [Offner, 1993, Bretagnolle, 2009] → simple model of population density and transportation network morphogenesis.

Modle

Simple coupling between

- Iterative generation of a density grid by preferential attachment/diffusion [Raimbault, 2016] calibrated on morphological objectives on european density grid.
- Heuristic network generation conditional to density :
 - Distribution of a fixed number of centers preferentially following density
 - Deterministic percolation between closest neighbors
 - Breaking of interaction potentials

$$V_{ij}(d) = \left[(1 - k_h) + k_h \cdot \left(\frac{P_i P_j}{P^2} \right)^{\gamma} \right] \cdot \exp\left(-\frac{d}{r_g(1 + d/d_0)} \right)$$

for a fixed number of couples N_L such that $V_{ij}(d_N)/V_{ij}(d_{ij})$ is minimal among $K \cdot N_L$ strongest euclidian potentials (K = 5 fixed)

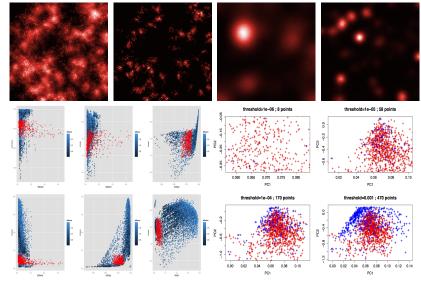
Planarization

Indicators: morphology [Le Néchet, 2015] (Moran, mean distance, entropy, hierarchy) and network (centrality, mean width, speed, diameter).

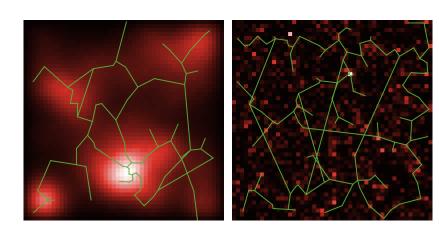
Implementation and Exploration

- \rightarrow Formal and Operational coupling : modular implementation (scala/NetLogo encapsulated by OpenMole [Reuillon et al., 2013]
- \rightarrow Exploration by intensive computation on grid via OpenMole : calibration of density model alone ($\sim 1.5 \cdot 10^6$ runs) ; brutal exploration by LHS sampling for feasible correlations ($\sim 5 \cdot 10^4$ runs)

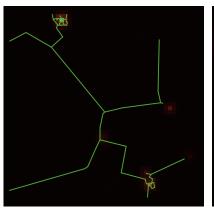
Results: Density Model alone

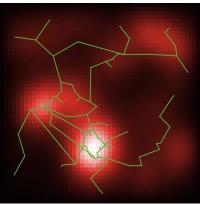


Results: examples of configurations

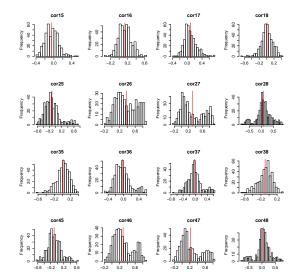


Results: examples of configurations



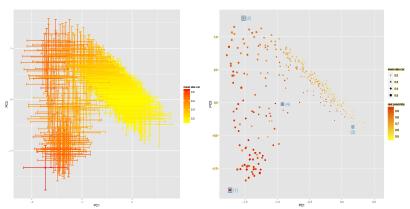


Results: cross-correlations

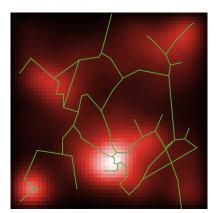


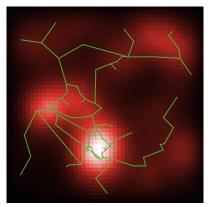
Results: feasible correlations

Mean matrices in a principal plan



Results: exemples of correlations





 $ho[ar{d},ar{c}]\simeq 0.34$ $ho[ar{d},ar{c}]\simeq -0.41$ ightarrow gravity hierarchy more important in (1) $\gamma=3.9,k_h=0.7$ against $\gamma=1.07,k_h=0.25$ for (2)

Applications

- Calibration of the coupled model, street network data(edge effects !) → generation of correlated synthetic data corresponding to a given urban system → intrinsic correlations to be compared to estimated correlations between different states : non-ergodicity of urban systems [Pumain, 2012]).
- Oynamical correlations in a strongly coupled model / spatio-temporal correlations in a strong spatial coupling.

Context

On Accessibility

Statistical Analysis

P. Bourgine framework for Complex Adaptive Systems

Next steps (until February 15th 2016)

- Theory exemplification, paper finalization [1w]
- Spatial Econometrics / Case study [0.5w]
- Cybergeo [0.5w]
- Wrap everything within a 1-year Memoire [1w]

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