A Macro-scale Model of Co-evolution for Cities and Transportation Networks

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Systems of Cities and Transportation Networks

(Left) Hong-Kong-Zhuhai-Macao Bridge; (Right) Near XiaoLan station





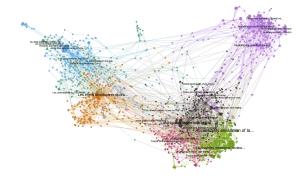
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Co-evolution

- ightarrow Co-evolution between networks and territories shown empirically [Bretagnolle, 2009]
- \rightarrow Models endogeneizing it are very rare ; possible explanation : interdisciplinary question and compartmentalization of disciplines [Raimbault, 2015]
- \rightarrow Diverse precursor approaches: LUTI models [Wegener and Fürst, 2004]; Network Growth (Geometrical [Courtat et al., 2011], Economical [Xie and Levinson, 2009], Biological [Tero et al., 2010])



Multidisciplinary citation network for studies of relations between networks and territories

Towards Models of Co-evolution

Why model it? Insights into dynamical processes in System of Cities; Perspective of Operational Models

Several possible Scales and Ontologies

- Micro-scale: mostly chaotic regimes, too precise for reasonable models (shown for traffic by [Raimbault, 2017a])
- Meso-scale: Urban Form, Accessibility and Network Shape [Raimbault, 2017b]; Transportation Governance in Mega-city Regions [Le Néchet and Raimbault, 2015]
- Macro-scale: SimpopNet model [Schmitt, 2014]

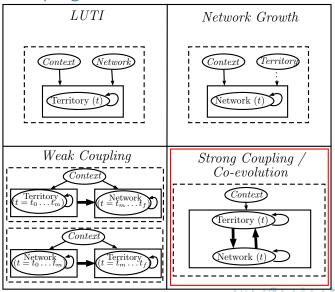
Research Objective

ightarrow Introduce a parsimonious but modular model of co-evolution of cities and networks at the scale of a System of Cities.





Submodel coupling







Model: Rationale

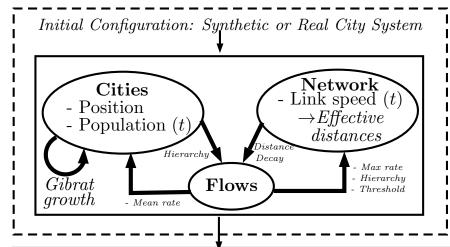
Interaction between cities at different orders are main drivers of their growth

- Cities represented by their population follow deterministic growth based on self growth (Gibrat) and interactions with other cities (similar to [Favaro and Pumain, 2011], extension of [Raimbault, 2016b]); approach of the Evolutive Urban Theory [Pumain, 1997]
- Drivers of network growth are interaction flow demands
- Adjustable network growth scale and stochasticity level
- \rightarrow Generic for any city system with dynamical population and network data; tested on synthetic city systems (following Rank-size Zipf's law).





Generic Model



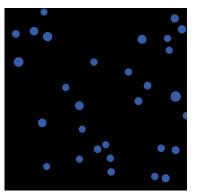
Indicators: Hierarchy, Entropy, Correlations, Trajectories diversity and complexity, Real Data fit

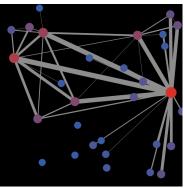




Model Specification: Abstract Network

Complete virtual network between cities, initialized with euclidian distances ; thresholded hierarchical growth of speeds as a function of flows.





Exemple of run ($t_f = 30$). Level of red gives overall growth and link width flows.



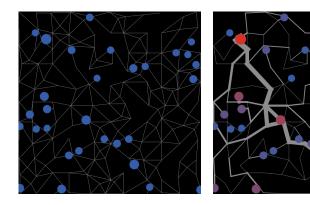


Results





Model: Physical Network









Further Extensions and Applications

Further Work and Extensions

- Targeted explorations using Genetic Algorithms (e.g. output diversity with PSE [Chérel et al., 2015])
- Potential breakdown / investment network heuristics

Application Case Studies

- French Urban System : Dynamical Railway Network 1850-2000 ;
 Dynamical Freeways Network (Database in construction)
- Chinese Urban System : High-speed Railway Network 2005-2015
- Comparisons: implication of governance context and planning level on interactions between networks and territories
- \rightarrow Does adding co-evolution improve fit on cities populations (correcting for additional parameters) ? How are produced network shapes realistic ?





Conclusion

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 All code and data available at https://github.com/JusteRaimbault/CityNetwork/tree/master/ Models/MacroCoevol





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Reserve Slides





Evolutive Urban Theory

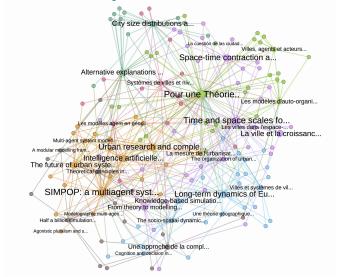
Definition: A Geographical Theory aiming at gathering most of known stylized facts on cities and their organisation within territories, in a non-static and out-of-equilibrium perspective, by following them on long time periods and putting an emphasis on structural factors and bifurcations. [Interview with D. Pumain, 03/2017]

- ightarrow Seminal works : theoretical manifest [Pumain, 1997] and modeling [Sanders et al.,
- \rightarrow Reciprocal relationship with computer scientists [Interview with R. Reuillon, 04/2017] : OpenMole [Reuillon et al., 2013] et Meta-heuristics [Chérel et al., 2015]
- \rightarrow Diverse and deep fieldworks ([Swerts, 2013] [Baffi, 2016]), integrated modeling ([Cottineau, 2014] [Schmitt, 2014]), epistemology [Rey-Coyrehourcq, 2015]
- \rightarrow Theoretical Frame and Methods to renew Ontologies : e.g. Definition of the city [Interview with C. Cottineau, 05/2017]





Evolutive Urban Theory



Citation Network of core references of the Evolutive Urban Theory





Model Formalization: Interactions

ightarrow Work under Gibrat independence assumptions, i.e. $\operatorname{Cov}[P_i(t),P_j(t)]=0$. If $\vec{P}(t+1)=\mathbf{R}\cdot\vec{P}(t)$ where \mathbf{R} is also independent, then $\mathbb{E}\Big[\vec{P}(t+1)\Big]=\mathbb{E}[\mathbf{R}]\cdot\mathbb{E}\Big[\vec{P}\Big](t)$. Consider expectancies only (higher moments computable similarly)

ightarrow With $ec{\mu}(t)=\mathbb{E}\Big[ec{P}(t)\Big]$, we generalize this approach by taking $ec{\mu}(t+1)=f(ec{\mu}(t))$





Model Formalization: Interactions

Let $\vec{\mu}(t) = \mathbb{E}\Big[\vec{P}(t)\Big]$ cities population and (d_{ij}) distance matrix Model specified by

$$f(\vec{\mu}) = r_0 \cdot \mathsf{Id} \cdot \vec{\mu} + \mathsf{G} \cdot \mathsf{1} + \mathsf{N}$$

with

- $G_{ij} = w_G \cdot \frac{V_{ij}}{\langle V_{ij} \rangle}$ and $V_{ij} = \left(\frac{\mu_i \mu_j}{\sum \mu_k^2}\right)^{\gamma_G} \exp\left(-d_{ij}/d_G\right)$
- $N_i = w_N \cdot \sum_{kl} \left(\frac{\mu_k \mu_l}{\sum \mu} \right)^{\gamma_N} \exp\left(-d_{kl,i}\right) / d_N$ where $d_{kl,i}$ is distance to shortest path between k,l computed with slope impedance $(Z = (1 + \alpha/\alpha_0)^{n_0})$ with $\alpha_0 \simeq 3$





Model Formalization: Network Growth





Model Formalization: Indicators

- Initial-final rank correlation (changes in the hierarchy) for variable X: $\rho\left[X_i(t=0),X_i(t=t_f)\right]$
- Trajectory diversity for variable X : with $\tilde{X}_i(t) \in [0;1]$ rescaled trajectories,

$$\frac{2}{N \cdot (N-1)} \sum_{i < j} \left(\frac{1}{T} \int_{t} \left(\tilde{X}_{i}(t) - \tilde{X}_{j}(t) \right)^{2} \right)^{\frac{1}{2}}$$

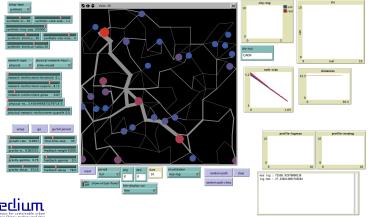
Average trajectory complexity (number of inflexion points)





Model Implementation and Exploration

- \rightarrow Model implemented in NetLogo (heterogeneous coupling, very diverse sub-models); explored with OpenMole [Reuillon et al., 2013]
- ightarrow Synthetic City-systems : follow Zipf's law with $lpha \in \{1.0, 1.5\}$ and N=30 cities ; relaxed Central Place Theory (no influence of other's initial size on localization).



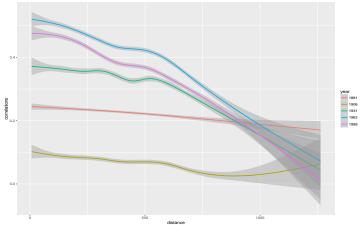




Macro Stylized facts

Population data for French-cities (Pumain-INED database: 1831-1999)

Non-stationarity of log-returns correlations function of distance

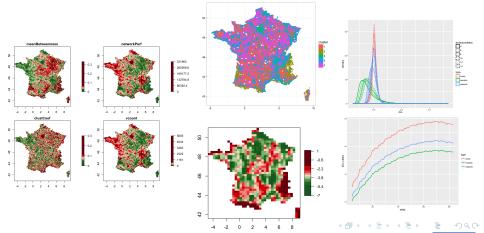






Meso Stylized facts

[Raimbault, 2016a]: Data, tools and methods showing the spatial non-stationarity and multi-scalar nature of correlations between urban form and network topology

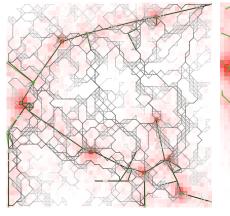


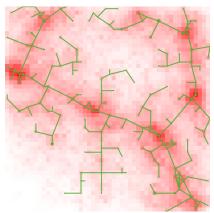




Modeling co-evolution at the Meso-scale

Multi-modeling of co-evolution at the meso-scale [Raimbault, 2017b]



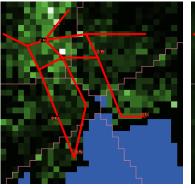






Transportation Governance in Mega-city Regions

Lutecia ([Le Néchet and Raimbault, 2015]): a model of co-evolution that includes governance processes of transportation network extension; application to Pearl River Delta Mega-city Region









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