

From individual spatio-temporal trajectories to spatial networks

Maxime Lenormand

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From individual spatio-temporal trajectories to spatial networks

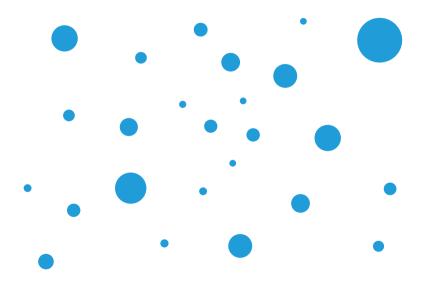


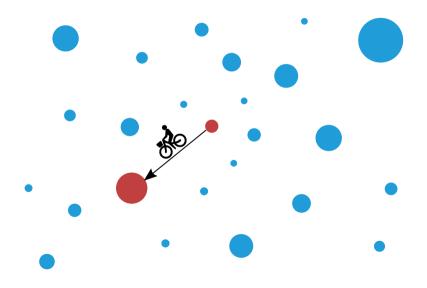
Maxime Lenormand UMR TETIS, Irstea, France

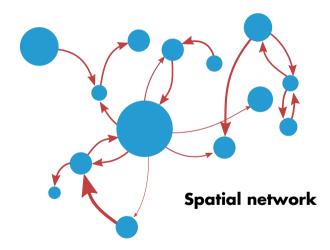
XTerM2019 | Le Havre, France

June 27, 2019









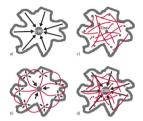
Tii

-,					-		
-	1	2	3	4	5		
1	0	1	8	4	0	13	
2	2	0	2	35	3	42	
3	13	1	0	9	4	26	O _i
4	1	23	2	0	1	27	
5	5	34	8	2	0	49	

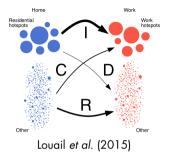
20 59	20	50	8
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Dj

Origin-Destination matrix



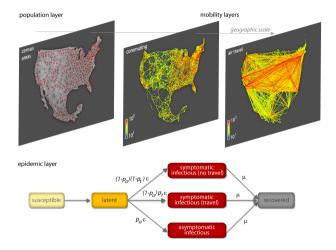
Bertaud & Malpezzi (2003)



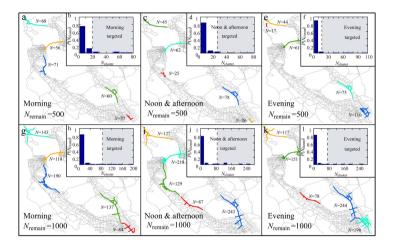


Lenormand et al. (2015)

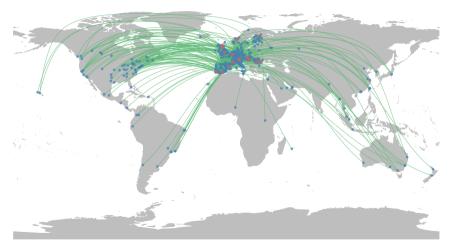
Louf et al. (2015)



Balcan et al. (2009) Seasonal transmission potential and activity peaks of the new influenza. BMC Medecine 7, 15052015.



Wang et al. (2014) Encapsulating urban traffic rhythms into road networks. *Scientific Reports* 4, 4141.



Lenormand et al. (2018) Multiscale socio-ecological networks in the age of information. *PLoS ONE 13, e0206672.*

Census & survey

- Census & survey
- Spatial interaction models

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- Spatial interaction models
- Individual geolocalized data

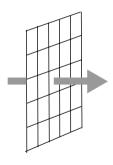
- Census & survey
- Spatial interaction models
- Individual geolocalized data



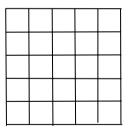
Sampling framework

"Reality"

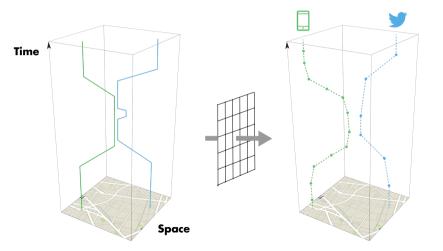




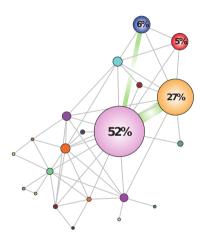




Sampling framework



Most frequented locations



Song et al. (2010) Limits of predictability in human mobility. Science 327, 1018-1021.

Home

Most frequented location between 7pm and 7am

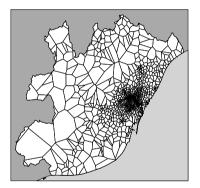
Work

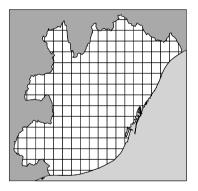
Most frequented location between 8am and 5pm on weekdays

Origin-Destination Matrix

Tij: number of individuals living in cell **i** and working in cell **j**

Spatial discretization





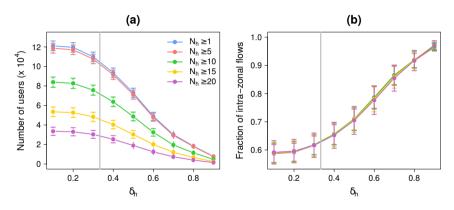
Location?

Extracting most frequented locations

- The hours of activity are divided into two groups, daytime hours and nighttime hours. Only days of the week from Monday to Thursday are taken into account.
- First filter: Consider only individuals "actives" during at least N_h hours spread over at least N_d days.
- For each hour of activity, the most frequently visited zone during this hour is identified.
- ▶ For both groups of hours (daytime and nighttime), we identify the zone in which the user has been localized the highest number of hours.
- Second filter: Select only users whose fraction of hours spent at "home" and "work" are larger than a fraction δ_h of the total number of locations visited during nighttime and daytime, respectively.

Lenormand et al. (2016) Is spatial information in ICT data reliable? In proceedings of the 2016 Spatial Accuracy Conference, 9-17, Montpellier, France. https://gitlab.com/maximelenormand/Most-frequented-locations

Extracting most frequented locations



Lenormand et al. (2016) Is spatial information in ICT data reliable? In proceedings of the 2016 Spatial Accuracy Conference, 9-17, Montpellier, France.

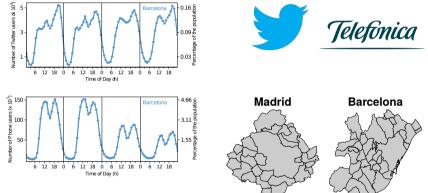
Cross-checking different sources of mobility information

Lenormand et al. 2014

Data

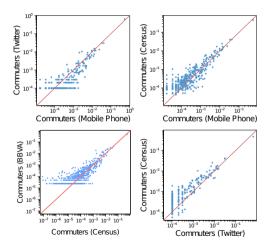
BBVA





Lenormand *et al.* (2014) Cross-checking different sources of mobility information.*PlosOne*,9(8):e105407. Louail *et al.* (2017) Crowdsourcing the Robin Hood effect in cities. Applied Network Science 2, 11.

Pairwise OD comparison

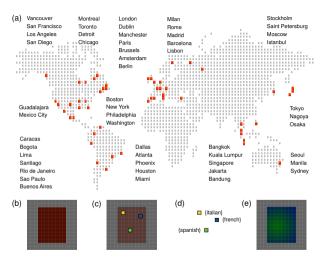


Lenormand *et al.* (2014) Cross-checking different sources of mobility information.*PlosOne*,9(8):e105407. Louail *et al.* (2017) Crowdsourcing the Robin Hood effect in cities. Applied Network Science 2, 11.

Immigrant community integration in world cities

Lamanna et al. 2018

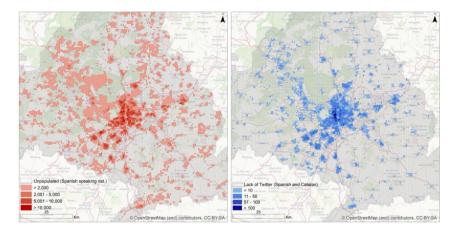
Data



Lamanna et al. (2018) Immigrant community integration in world cities. *Plos One 13, e0191612.*



Spatial distribution of residence

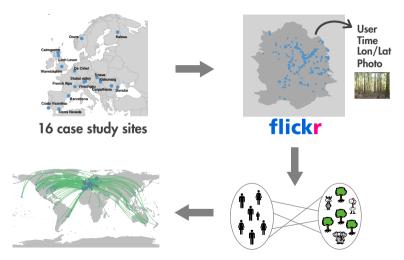


Lamanna et al. (2018) Immigrant community integration in world cities. *Plos One 13, e0191612.*

Multiscale socio-ecological networks in the age of information

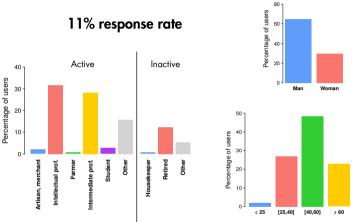
Lenormand et al. 2018

Data



Lenormand et al. (2018) Multiscale socio-ecological networks in the age of information. *PLoS ONE 13, e0206672.*





90% of accuracy in the users' place of residence detection!

Lenormand et al. (2018) Multiscale socio-ecological networks in the age of information. *PLoS ONE 13, e0206672.*

Is spatial information in ICT data reliable? Lenormand *et al.* 2016

Data



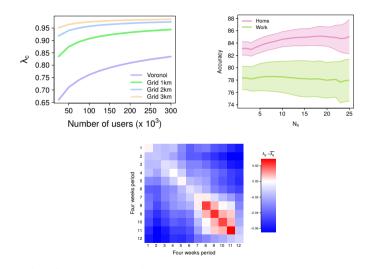
300,000 mobile phone users' trajectories x 25 two-week periods



Identifying **home-work locations** from mobile phone activity

Lenormand et al. (2016) Is spatial information in ICT data reliable? In proceedings of the 2016 Spatial Accuracy Conference, 9-17, Montpellier, France.

OD comparison



Lenormand et al. (2016) Is spatial information in ICT data reliable? In proceedings of the 2016 Spatial Accuracy Conference, 9-17, Montpellier, France.

Take home messages...

Good agreement between the different data sources.

- Uncertainty & accuracy are highly dependent of the spatial resolution and sample size.
- More studies in this spirit need to be done to assess the biases and uncertainty associated with ICT data.
- It could be interesting to involve (more strongly and widely) the indiviual ICT data providers.

References

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- Iqbal et al. (2014) Development of origin-destination matrices using mobile phone call data, Transp. Res. Part C:EmergingTechnol. 40, 63–74.
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- Toole et al. (2015) The path most traveled: Travel demand estimation using big data resources Transp. Res. PartC: Emerging Technol. 58, PartB, 162–177.
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www.maximelenormand.com

Acknowledgement



























