



**Cite this article:** Lotero L, Hurtado RG, Floría LM, Gómez-Gardeñes J. 2016 Rich do not rise early: spatio-temporal patterns in the mobility networks of different socio-economic classes. *R. Soc. open sci.* 3: 150654. <http://dx.doi.org/10.1098/rsos.150654>

Received: 1 December 2015

Accepted: 9 September 2016

**Subject Category:**

Physics

**Subject Areas:**

complexity/statistical physics

**Keywords:**

urban mobility networks, spatio-temporal patterns, socio-economic status

**Author for correspondence:**

Jesús Gómez-Gardeñes

e-mail: [gardenes@gmail.com](mailto:gardenes@gmail.com)

# Rich do not rise early: spatio-temporal patterns in the mobility networks of different socio-economic classes

Laura Lotero<sup>1,2</sup>, Rafael G. Hurtado<sup>3</sup>, Luis Mario Floría<sup>4,5</sup> and Jesús Gómez-Gardeñes<sup>4,5</sup>

<sup>1</sup>Facultad de Ingeniería Industrial, Universidad Pontificia Bolivariana, Medellín, Colombia

<sup>2</sup>Departamento de Ciencias de la Computación y de la Decisión, Universidad Nacional de Colombia, Medellín, Colombia

<sup>3</sup>Departamento de Física, Universidad Nacional de Colombia, Bogotá, Colombia

<sup>4</sup>Departamento de Física de la Materia Condensada, Universidad de Zaragoza, Zaragoza 50009, Spain

<sup>5</sup>Instituto de Biocomputación y Física de Sistemas Complejos, Universidad de Zaragoza, Zaragoza 50018, Spain

LL, 0000-0002-6537-3276; JGG, 0000-0001-5204-1937

We analyse the urban mobility in the cities of Medellín and Manizales (Colombia). Each city is represented by six mobility networks, each one encoding the origin-destination trips performed by a subset of the population corresponding to a particular socio-economic status. The nodes of each network are the different urban locations whereas links account for the existence of a trip between two different areas of the city. We study the main structural properties of these mobility networks by focusing on their spatio-temporal patterns. Our goal is to relate these patterns with the partition into six socio-economic compartments of these two societies. Our results show that spatial and temporal patterns vary across these socio-economic groups. In particular, the two datasets show that as wealth increases the early-morning activity is delayed, the midday peak becomes smoother and the spatial distribution of trips becomes more localized.

## 1. Introduction

Understanding and modelling urban mobility is crucial for urban planning and decision-making and has been a topic of great interest for sociologists, urban planners, engineers, physicists,

16. Balcan D, Gonçalves B, Hu H, Ramasco JJ, Colizza V, Vespignani A. 2010 Modeling the spatial spread of infectious diseases: the Global Epidemic and Mobility computational model. *J. Comput. Sci.* **1**, 132–145. ([doi:10.1016/j.jocs.2010.07.002](https://doi.org/10.1016/j.jocs.2010.07.002))
17. Colizza V *et al.* 2009 Estimate of novel influenza A/H1N1 cases in Mexico at the early stage of the pandemic with a spatially structured epidemic model. *PLoS Curr.* **1**, RRN1129. ([doi:10.1371/currents.RRN1129](https://doi.org/10.1371/currents.RRN1129))
18. Van den Broeck W, Gioannini C, Gonçalves B, Quaggiotto M, Colizza V, Vespignani A. 2011 The GLEaMviz computational tool, a publicly available software to explore realistic epidemic spreading scenarios at the global scale. *BMC Infect. Dis.* **11**, 37. ([doi:10.1186/1471-2334-11-37](https://doi.org/10.1186/1471-2334-11-37))
19. Cardillo A, Zanin M, Gómez-Gardeñes J, Romance M, García del Amo AJ, Boccaletti S. 2013 Modeling the multi-layer nature of the European Air Transport Network: resilience and passengers re-scheduling under random failures. *Eur. Phys. J. Spec. Top.* **215**, 23–33. ([doi:10.1140/epjst/e2013-01712-8](https://doi.org/10.1140/epjst/e2013-01712-8))
20. Caschili S, de Montis A. 2013 Accessibility and complex network analysis of the U.S. commuting system. *Cities* **30**, 4–17. ([doi:10.1016/j.cities.2012.04.007](https://doi.org/10.1016/j.cities.2012.04.007))
21. De Montis A, Caschili S, Chessa A. 2013 Commuter networks and community detection: a method for planning sub regional areas. *Eur. Phys. J. Spec. Top.* **215**, 75–91. ([doi:10.1140/epjst/e2013-01716-4](https://doi.org/10.1140/epjst/e2013-01716-4))
22. Wilson A. 2010 Entropy in urban and regional modelling: retrospect and prospect. *Geogr. Anal.* **42**, 364–394. ([doi:10.1111/j.1538-4632.2010.00799.x](https://doi.org/10.1111/j.1538-4632.2010.00799.x))
23. Duranton G. 2005 Delineating metropolitan areas: measuring spatial labour market networks through commuting patterns. In *The economics of interfirm networks* ch. 6 (eds T Watanabe, I Uesugi, A Ono), pp. 107–134. Advances in Japanese Business and Economics, vol. 4. Berlin, Germany: Springer.
24. De Montis A, Barthélémy M, Chessa A, Vespignani A. 2007 The structure of interurban traffic: a weighted network analysis. *Environ. Plan. B: Plan. Des.* **34**, 905–924. ([doi:10.1068/b32128](https://doi.org/10.1068/b32128))
25. Ramasco JJ, Colizza V, Panzarasa P. 2009 Using the weighted rich-club coefficient to explore traffic organization in mobility networks. In *Lecture Notes of the Institute for Computer Sciences, Social-Informatics and Telecommunications Engineering*, vol. 4, pp. 680–692. ([doi:10.1007/978-3-642-02466-5](https://doi.org/10.1007/978-3-642-02466-5))
26. Goetz SJ, Han Y, Findeis JL, Brasier KJ. 2010 U.S. commuting networks and economic growth: measurement and implications for spatial policy. *Growth Change* **41**, 276–302. ([doi:10.1111/j.1468-2257.2010.00527.x](https://doi.org/10.1111/j.1468-2257.2010.00527.x))
27. Chowell G, Hyman J, Eubank S, Castillo-Chávez C. 2003 Scaling laws for the movement of people between locations in a large city. *Phys. Rev. E* **68**, 066102. ([doi:10.1103/PhysRevE.68.066102](https://doi.org/10.1103/PhysRevE.68.066102))
28. Brockmann D, Hufnagel L, Geisel T. 2006 The scaling laws of human travel. *Nature* **439**, 462–465. ([doi:10.1038/nature04292](https://doi.org/10.1038/nature04292))
29. Lenormand M, Louail T, Cantu-Ros OG, Picornell M, Herranz R, Arias JM, Barthélémy M, San Miguel M, Ramasco JJ. 2015 Influence of sociodemographic characteristics on human mobility. *Sci. Rep.* **5**, 10075. ([doi:10.1038/srep10075](https://doi.org/10.1038/srep10075))
30. González M, Hidalgo C, Barabási A-L. 2008 Understanding individual human mobility patterns. *Nature* **453**, 779–782. ([doi:10.1038/nature06958](https://doi.org/10.1038/nature06958))
31. Louail T, Lenormand M, Cantu-Ros OG, Picornell M, Herranz R, Frias-Martinez E, Ramasco JJ, Barthélémy M. 2014 From mobile phone data to the spatial structure of cities. *Sci. Rep.* **4**, 5276. ([doi:10.1038/srep05276](https://doi.org/10.1038/srep05276))
32. Candia J, González M, Wang P, Schoenharl T, Madey G, Barabási A-L. 2008 Uncovering individual and collective human dynamics from mobile phone records. *J. Phys. A Math. Theor.* **41**, 224015. ([doi:10.1088/1751-8113/41/22/224015](https://doi.org/10.1088/1751-8113/41/22/224015))
33. Cáceres N, Wideberg JP, Benítez FG. 2007 Deriving origin-destination data from a mobile phone network. *IET Intell. Transport Syst.* **1**, 15. ([doi:10.1049/iet-its:20060020](https://doi.org/10.1049/iet-its:20060020))
34. Calabrese F, Di Lorenzo G, Liu L, Ratti C. 2011 Estimating origin-destination flows using mobile phone location data. *IEEE Pervasive Comput.* **10**, 36–44. ([doi:10.1109/MPRV.2011.41](https://doi.org/10.1109/MPRV.2011.41))
35. Iqbal MS, Choudhury CF, Wang P, González M. 2014 Development of origin-destination matrices using mobile phone call data. *Transportation Res. Part C: Emerging Technol.* **40**, 63–74. ([doi:10.1016/j.trc.2014.01.002](https://doi.org/10.1016/j.trc.2014.01.002))
36. Mellegard E, Moritz S, Zahoor M. 2011 Origin/destination-estimation using cellular network data. In *2011 IEEE 11th International Conference on Data Mining Workshops*, pp. 891–896. IEEE. ([doi:10.1109/ICDMW.2011.132](https://doi.org/10.1109/ICDMW.2011.132))
37. Coscia M, Hausmann R. 2015 Evidence that calls-based and mobility networks are isomorphic. *PLoS ONE* **10**, e0145091. ([doi:10.1371/journal.pone.0145091](https://doi.org/10.1371/journal.pone.0145091))
38. Jia T, Jiang B, Carling K, Bolin M, Ban Y. 2012 An empirical study on human mobility and its agent-based modeling. *J. Stat. Mech. Theory E* **2012**, P11024. ([doi:10.1088/1742-5468/2012/11/P11024](https://doi.org/10.1088/1742-5468/2012/11/P11024))
39. Liu Y, Sui Z, Kang C, Gao Y. 2014 Uncovering patterns of inter-urban trip and spatial interaction from social media check-in data. *PLoS ONE* **9**, e86026. ([doi:10.1371/journal.pone.0086026](https://doi.org/10.1371/journal.pone.0086026))
40. Liu Y, Kang C, Gao S, Xiao Y, Tian Y. 2012 Understanding intra-urban trip patterns from taxi trajectory data. *J. Geogr. Syst.* **14**, 463–483. ([doi:10.1007/s10109-012-0166-z](https://doi.org/10.1007/s10109-012-0166-z))
41. Wang W, Pan L, Yuan N, Zhang S, Liu D. 2015 A comparative analysis of intra-city human mobility by taxi. *Phys. A* **420**, 134–147. ([doi:10.1016/j.physa.2014.10.085](https://doi.org/10.1016/j.physa.2014.10.085))
42. Hasan S, Schneider CM, Ukkusuri SV, González M. 2012 Spatiotemporal patterns of urban human mobility. *J. Stat. Phys.* **151**, 304–318. ([doi:10.1007/s10955-012-0645-0](https://doi.org/10.1007/s10955-012-0645-0))
43. AREA Metropolitana del Valle de Aburrá. 2006 Chapter 2: Diagnóstico. Formulación del Plan Maestro de Movilidad para la Región Metropolitana del Valle de Aburrá. *Informe Final*, pp. 102–108. Oficina Asesora de Comunicaciones del Área Metropolitana del Valle de Aburrá.
44. Universidad Nacional de Colombia and AREA Metropolitana del Valle de Aburrá. 2006 Encuesta origen destino de viajes 2005 del Valle de Aburrá, estudios de tránsito complementarios y validación. Technical Report.
45. Alcaldía de Manizales. 2005 Plan de movilidad para el municipio de Manizales. Informe del estudio origen-destino. Technical Report.
46. Lotero L, Hurtado RG, Flória LM, Gómez-Gardeñes J. 2016 Data from: Rich don't rise early: spatio-temporal patterns in the mobility networks of different socio-economic classes. Dryad Digital Repository. (<http://dx.doi.org/10.5061/dryad.hjt44>)
47. Peterson EB, Hamburg JR. 1986 Travel surveys: current options. *Transportation Research Record*, (1097).
48. Thériault M, Vandersmissen MH, Lee-Gosselin M, Leroux D. 1999 Modelling commuter trip length and duration within GIS: application to an OD survey. *J. Geogr. Inform. Decis. Anal.* **3**, 41–55.
49. Zahabi SAH, Miranda-Moreno LF, Patterson Z, Barla P. 2012 Evaluating the effects of land use and strategies for parking and transit supply on mode choice of downtown commuters. *J. Transport Land Use* **5**, 103–119. ([doi:10.5198/jtlu.v5i2.260](https://doi.org/10.5198/jtlu.v5i2.260))
50. Sider T, Alam A, Zukari M, Dugum H, Goldstein N, Eluru N, Hatzopoulou M. 2013 Land-use and socio-economics as determinants of traffic emissions and individual exposure to air pollution. *J. Transport Geogr.* **33**, 230–239. ([doi:10.1016/j.jtrangeo.2013.08.006](https://doi.org/10.1016/j.jtrangeo.2013.08.006))
51. Medina C, Morales L, Bernal R, Torero M. 2007 Stratification and public utility services in Colombia: subsidies to households or distortion of housing prices? *Economía* **7**, 41–99. ([doi:10.1353/eco.2007.0013](https://doi.org/10.1353/eco.2007.0013))
52. Lotero L, Cardillo A, Hurtado R, Gómez-Gardeñes J. 2016 Several multiplexes in the same city: the role of socioeconomic differences in urban mobility. In *Interconnected Networks*, ch. 9 (ed. Garas), pp. 149–164. Series on Understanding Complex Systems. Berlin, Germany: Springer.
53. Boccaletti S, Bianconi G, Criado R, DelGenio CI, Gómez-Gardeñes J, Romance M, Sendiña-Nadal I, Wang Z, Zanin M. 2014 The structure and dynamics of multilayer networks. *Phys. Rep.* **544**, 1–122. ([doi:10.1016/j.physrep.2014.07.001](https://doi.org/10.1016/j.physrep.2014.07.001))