

#### **Big Data for the Social Sciences**

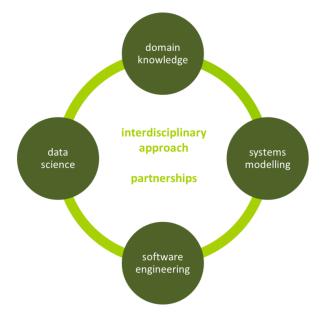
The Example of Mobile Phone Records

1<sup>st</sup> September 2016

#### About us

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- Technology company based in Madrid
- Decision support tools for the planning and management of transport systems





- Joint company Nommon Luis Willumsen
- Analysis of geolocated data
- Sectors: transport, tourism, geomarketing...
- Projects in Europe & LatAm

# **Activity-mobility statistics**

Different types of institutions collect and analyse activity-mobility data in order to produce studies that are then made available to third parties or publicly available: national/regional/local statistical offices, research services, academic institutions, etc.

**Traditional data collection methods (e.g., household travel surveys) are expensive and time-consuming**, so many institutions cannot access reliable and frequently updated activity-mobility information

Modern ICTs allow the automatic collection of vast amounts of spatio-temporal data and open **new opportunities for complementing and/or replacing traditional statistics and data collection methods** 

## **Information needs**

# Distribution of population



#### Distribution of workplaces

Urban mobility





#### Interurban mobility



#### Impact of events



**Tourism flows** 



# Data collection: traditional approach

- Surveys:
  - Official statistics
  - Ad-hoc surveys (e.g., mobility surveys)
- Administrative registers
  - Trend towards replacing surveys by administrative registers, where possible



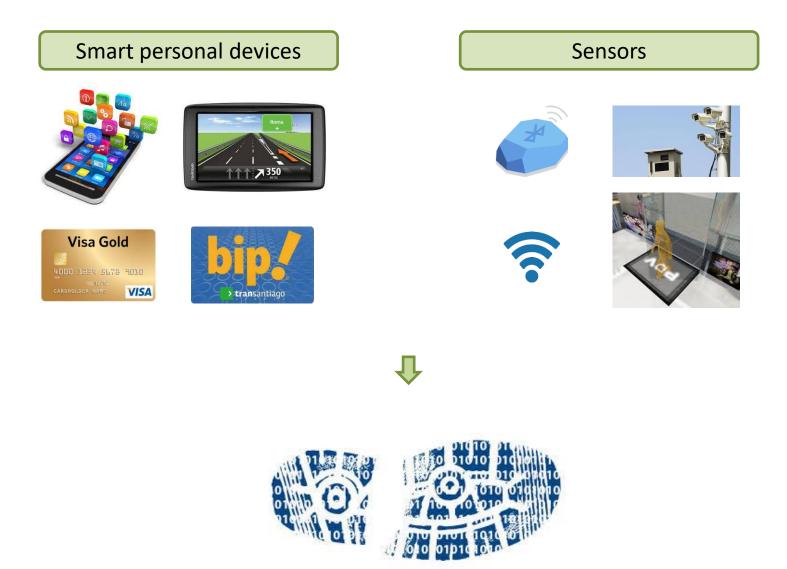


#### **Problems**

- Information not fully adapted to the problem
- Outdated information
- High cost
- Time consuming
- Small samples
- Quality of information



#### **Opportunities**



## What we do

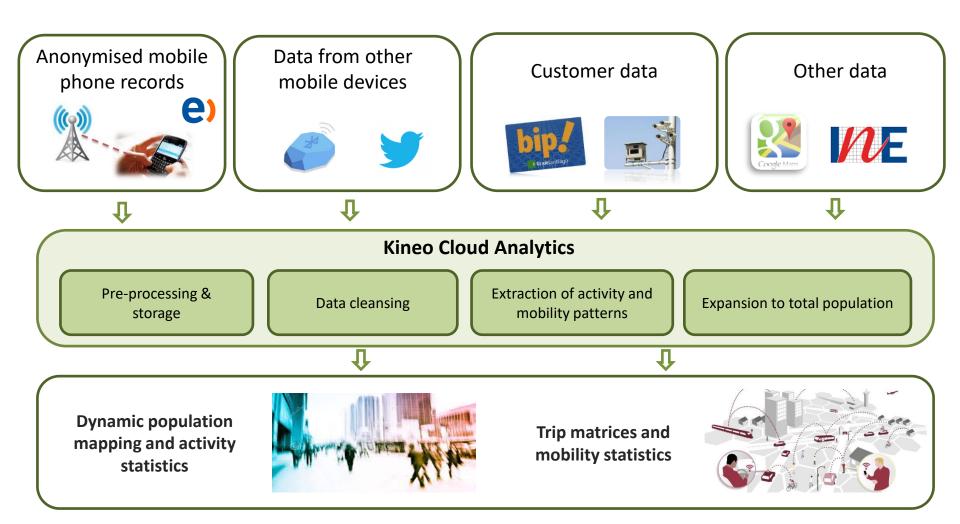
We mine and analyse anonymised geolocation data from mobile devices and blend it with other data provided by our customers or available from public databases **to provide continuously updated and rich activity-travel information in a fraction of the time and cost required by traditional methods** 







#### What we do



#### Where are the people? What are they doing?

#### How do they move? Why?





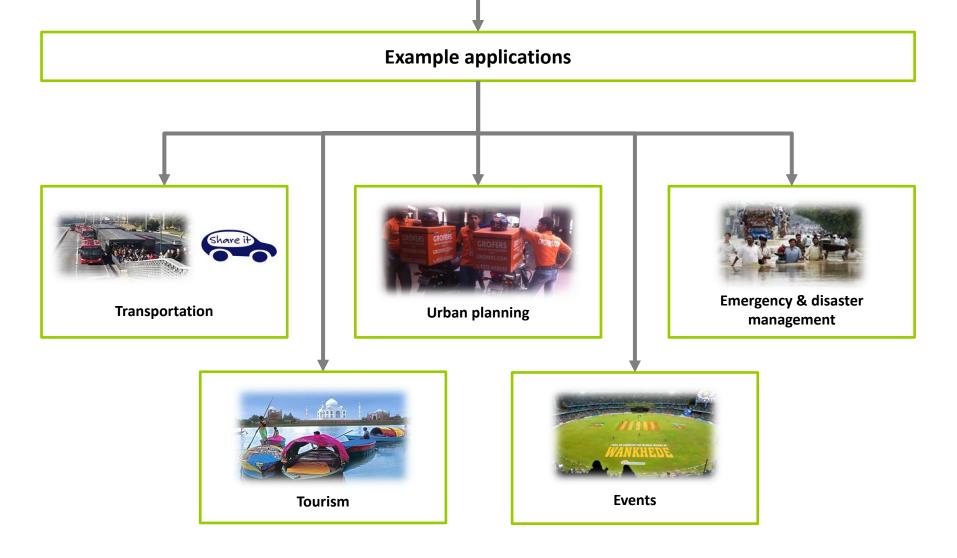


#### Origin-Destination Matrices and Mobility Statistics



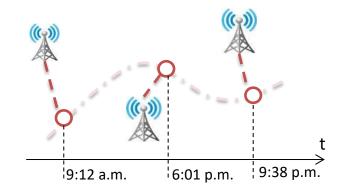


#### Dynamic Population Mapping and Activity Statistics



#### Mobile phone data: what do we mean exactly?

- Typically anonymised call detail records (CDRs) or data from network probes
  - Spatio-temporal data: time and user position every time an event occurs
  - Position typically at cell level (but some MNOs are beginning to store triangulated positions)
- MNO's clients + roamers that connect to its network
- Sociodemographic data for clients (age and gender)



#### So how does it work?

#### What we do (more or less):

- Sample construction
- Identification of users' home
- Identification of other activities (work, study...)
- Generation of activity-travel diaries
- Expansion to the total population
- Calculation of indicators

#### Why mobile phone records?

Great potential for the analysis of human activity and mobility:

- Passive data collection eliminates many of the intrinsic limitations of surveys, such as incorrect and imprecise answers
- Opportunistic data collection no need for specific infrastructure
- Reasonably good temporal and spatial granularity
- Large samples
- Longitudinal data
- Opportunities for "natural experiments"
- Drastic cost and time reduction with respect to traditional methods

#### But...

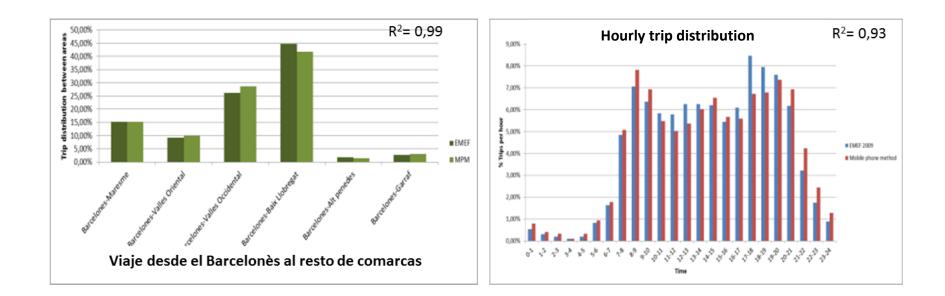
These data also come with a number of challenges:

- Noise and errors
- Biases
- Less explanatory power
- Computational challenges
- $\Rightarrow$  Mobile phone data do not provide the full picture: need for data fusion
- $\Rightarrow$  Algorithms matter: devil's in the details
- $\Rightarrow$  Rigorous validation is essential

# Example applications

#### **OD Matrices for Barcelona Metropolitan Area**

 Calculation of origin-destination matrices in the Barcelona Metropolitan Area and comparison with the Mobility Survey on Working Days (EMEF)

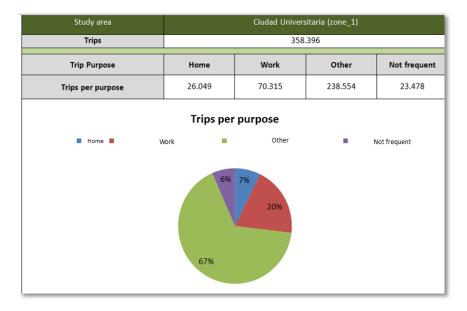


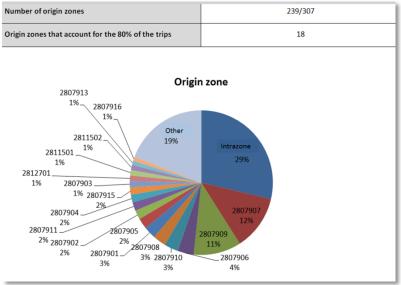




#### **Trip Attraction Areas in Madrid**

• Trip matrices for several attraction areas in the Region of Madrid







#### **Analysis of Commuting Patterns in Spain**

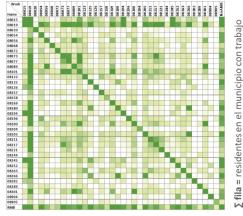
• Commuting Patterns in Spain: pilot study for next Spanish Census

| Ma KALLA   | Sample | Sample %                                | ID /    | Population              |
|--|--------|---|---------|-------------------------|
|  | 111    | 7.602739726027398                       | 0104001 | 1460.000000000000000000 |
|  | 80     | 5.925925925925927                       | 0104002 | 1350.00000000000000000  |
|  | 2834   | 11.626666666666667                      | 0104003 | 24375.00000000000000000 |
|  | 23     | 3.357664233576643                       | 0104005 | 685.0000000000000000    |
|  | 1114   | 10.257826887661142                      | 0104006 | 10860.0000000000000000  |
|  | 36     | 3.956043956043956                       | 0104007 | 910.0000000000000000    |
|  | 35     | 5.600000000000001                       | 0104008 | 625.0000000000000000    |
|  | 14     | 2.043795620437956                       | 0104010 | 685.0000000000000000    |
|  | 160    | 4.221635883905013                       | 0104011 | 3790.0000000000000000   |
| and the second | 23007  | 12.174948404508653                      | 0104013 | 188970.000000000000     |
| 15-10-00-00-00-00-00-00-00-00-00-00-00-00-   | 33     | 17.368421052631579                      | 0104014 | 190.00000000000000000   |
|  | 365    | 11.196319018404909                      | 0104016 | 3260.0000000000000000   |
|  | 166    | 3.660418963616317                       | 0104017 | 4535.00000000000000000  |
|  | 33     | 9.565217391304348                       | 0104018 | 345.0000000000000000    |
|  | 16     | 4.266666666666667                       | 0104020 | 375.0000000000000000    |
|  | 22     | 9.77777777777777777                     | 0104021 | 225.0000000000000000    |
|  | 12     | 1.276595744680851                       | 0104022 | 940.0000000000000000    |
|  | 536    | 12.807646356033453                      | 0104024 | 4185.00000000000000000  |
|  | 23     | 8.36363636363636363                     | 0104028 | 275.0000000000000000    |
|  | 1710   | 11.350813143046798                      | 0104029 | 15065.0000000000000000  |
|  | 243    | 16.933797909407666                      | 0104030 | 1435.00000000000000000  |
|  | 406    | 10.545454545454545                      | 0104031 | 3850.0000000000000000   |
|  | 967    | 12.279365079365078                      | 0104032 | 7875.00000000000000000  |
|  | 883    | 6.740458015267175                       | 0104035 | 13100.00000000000000000 |
|  | 21     | 7.0000000000000000000000000000000000000 | 0104036 | 300.00000000000000000   |
|  | 70     | 4 202022471010112                       | 0104027 | 1790.00000000000000000  |

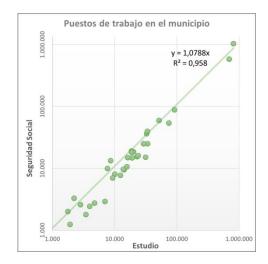


#### **Analysis of Workplaces Distribution in Barcelona**

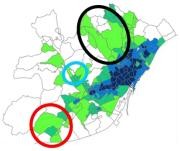
• Home-Work matrices for the Metropolitan Area of Barcelona



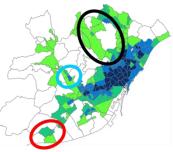
 $\Sigma$  col = puestos de trabajado en el municipio







Asignación sin usos del suelo



Asignación con usos del suelo

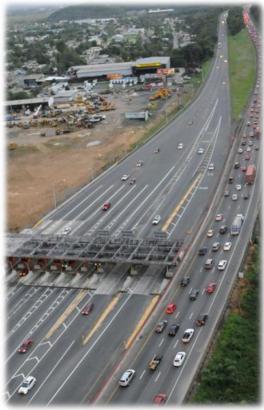


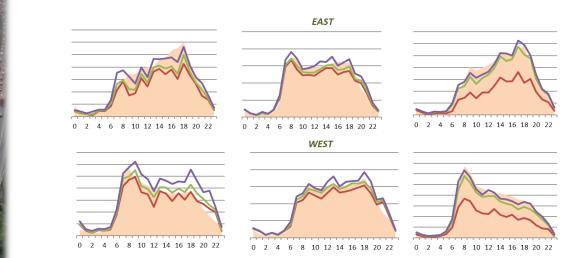
BARCELONA REGIONAL

AGÈNCIA DESENVOLUPAMENT URBÀ

#### **Analysis of Toll Road Potential Demand**

• Estimation of potential demand for a toll road and modelling of different revenue optimisation strategies

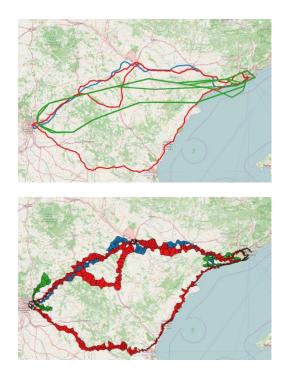


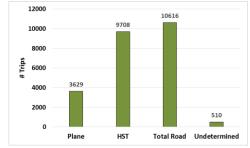


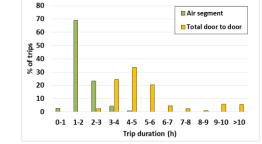


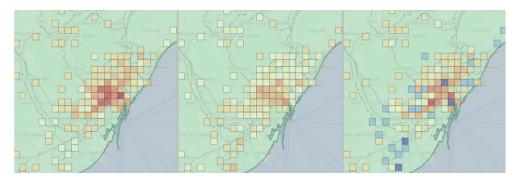
#### **Analysis of Madrid-Barcelona Transport Corridor**

• Analysis of modal split, airport and HST catchment areas, and door-to-door travel times









No Train trips

No Air trips

[No Train trips - No Air trips]

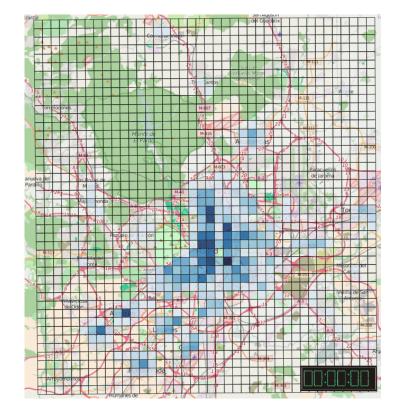


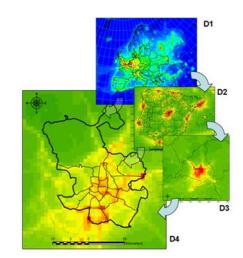


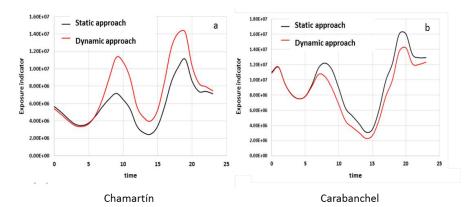


#### **Population Exposure to Air Pollution**

 Exposure of population to NOx emissions in Madrid based on dynamic population mapping







MADRID! ÁREA DE GOBIERNO DE MEDIO AMBIENTE Y MOVILIDAD

# Some conclusions...

## The promise of Big Data

Big data streams generated by smart personal devices offer a great potential for analysing human activity-mobility patterns and producing sociodemographic statistics

#### Data is not (always) a substitute for theory

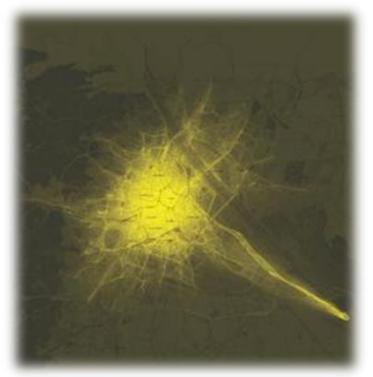
#### "Models without data are just stories - data without models are just numbers"

Neither data nor algorithms are "neutral": lack of understanding of underlying principles and limitations may lead to a false sense of objectivity and democratisation of information

Critical interpretation of the results in terms of the initial problem is essential, and we (researchers, consultants...) must do so responsibly

Big Data is usually noisy, unstructured and biased ("all data are biased, but some are useful"), so producing actionable information requires a lot of effort for data cleaning, data fusion, and algorithm validation...

...but this is the hard part, not really cool, so many people prefer to focus on fancy (and often useless) visualisation



# Big Data also comes with a number of regulatory, organisational and business challenges

(We won't discuss this today, but it should not be forgotten)

# We have to live with

- Those overselling the benefits of Big Data as the cure for all diseases: more common in industry, but also happens in public administration and academia
  - But we can and must be critical and claims to be rigorously supported
- Those unable to see the added value: equally common in industry, in public administration and in academia, but probably more reproachable in the latter case ("I'm happy with my good old crappy data, I still get my papers published")
  - Nothing much to do here probably the same people who 20 years ago didn't see the added value of Internet over teletext

"A new scientific truth does not triumph by convincing its opponents and making them see the light, but rather because its opponents eventually die, and a new generation grows up that is familiar with it"

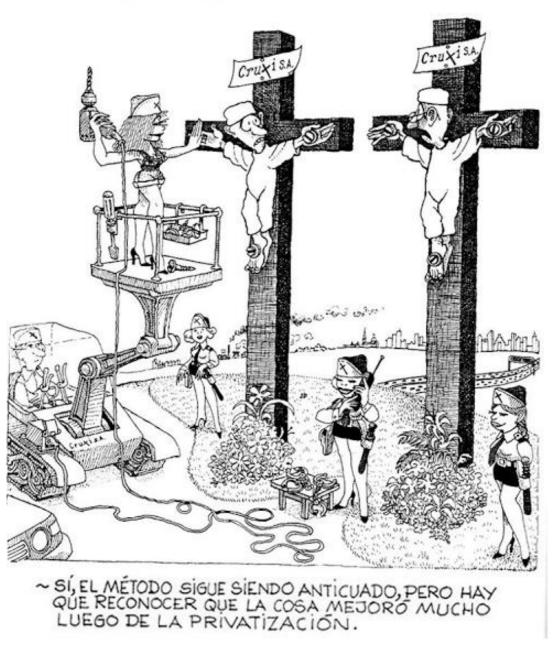
Max Planck

# ...and one fundamental open question

#### At present

Big Data has already yielded important practical advances in fields like transportation planning, but so far this is mostly about better/cheaper ways of measuring the same indicators and use them to feed the same models





# In the future

For sure, Big Data will produce new indicators that couldn't be measured before, and will help us make the most of state-of-the-art theoretical frameworks, but...



.. just as the telescope revolutionised astronomy or the microscope revolutionised biology, can Big Data inspire the development of new theories and revolutionise the social sciences?

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