# **Urban Mobility Flows from Mobile Phone Data**

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**INSA Lyon INRIA UrbaNet** 

**Urban Modelling Symposium** Lyon







#### **About Me**

Associate professor with the Telecom department at INSA Lyon

**Researcher at CITI lab, a joint INSA-INRIA research unit** 

Member of the INRIA UrbaNet team, focused on wireless networks in urban environments

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#### **Context**

- **Datasets**
- Methodology
- Mobility Flows

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#### **Our motivation**

Global mobile data traffic grew 81% in 2013\*

**526M** mobile devices and connections were added in 2013\*

Global mobile data traffic is expected to increase nearly 11-fold by 2018\*

\* Cisco Visual Networking Index: Global Mobile Data Traffic Forecast Update, 2013-2018

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#### **Our motivation**

Global mobile data traffic grew 81% in 2013\*

□ 526M mobile devices and connections were added in 2013\*

Global mobile data traffic is expected to increase nearly 11-fold by 2018\*

A need to enhance current wireless infrastructure

\* Cisco Visual Networking Index: Global Mobile Data Traffic Forecast Update, 2013-2018

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## **Our approach**

**Understand human dynamics** 

**Understand customers' demand** 

□ Adapt networking services

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**Understand human dynamics** 

**Understand customers' demand** 

Adapt networking services

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Analyse mobile phone data

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#### Data

**Orange Data for Development Challenge 2013** 

Anonymized Call Detail Records (CDR) of Orange customers in Ivory Coast

**Data interval: 05/12/2011 - 22/04/2012** 

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#### Datasets

**D1:** Hourly antenna-to-antenna aggregated calls

**D2: Small subset of individual trajectories with a high spatial resolution** 

**D3:** Large subset of individual trajectories with a reduced spatial resolution

**D4:** Individual communication subgraph

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□ Goal: Understand how people move over time in a typical day in an urban environment

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Goal: Understand how people move over time in a typical day in an urban environment - Abidjan



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□ Goal: Understand how people move over time in a typical day in an urban environment – Dataset D2

Consecutive calls from the same base station are filtered out



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**Goal: Understand how people move over time in a typical day in an urban** environment – Dataset D2

- Consecutive calls from the same base station are filtered out
- Hourly O/D matrices
- A movement duration is variable
- Each movement is assigned a weight of  $1/\Delta t$ , where  $\Delta t$  is the movement duration

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Goal: Understand how people move over time in a typical day in an urban environment

- Reduced number of movements in dataset D2
- The only period with all antennas present: 2 weeks in April
- Idea: aggregate D2 movements from multiple similar days
- Questions: what is "similar"? what is "typical"?

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#### Typical behavior

□ The notion of snapshot: representation of the load generated by mobile users on the access network during a certain time period



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## **Typical behavior**

□ This allows us to calculate distances between snapshots to detect similar user distributions



\* D. Naboulsi, R. Stanica, M. Fiore – "Classifying Call Profiles in Large-Scale Mobile Traffic Datasets", Proc. Infocom 2014

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### **Typical behavior**

#### U We obtain clusters of snapshots, and can distinguish typical and outlying **behaviors**



#### **Mobility flows**

We aggregate data from D2 to increase the mobility information
Smallest geographic area: the cell covered by a base station



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#### **Mobility flows**

#### U We also aggregate spatially (per region), to filter some of the noise



#### To summarize

Promising use of Call Detail Records

□ Mobile phone data gets richer and richer (your smartphone connects without your knowledge)

□ Not (yet) capable of providing a "standardized" O/D matrix

A different reasoning, in terms of flows, might be better suited

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