# Analysis of Human mobility to develop pervasive urban applications

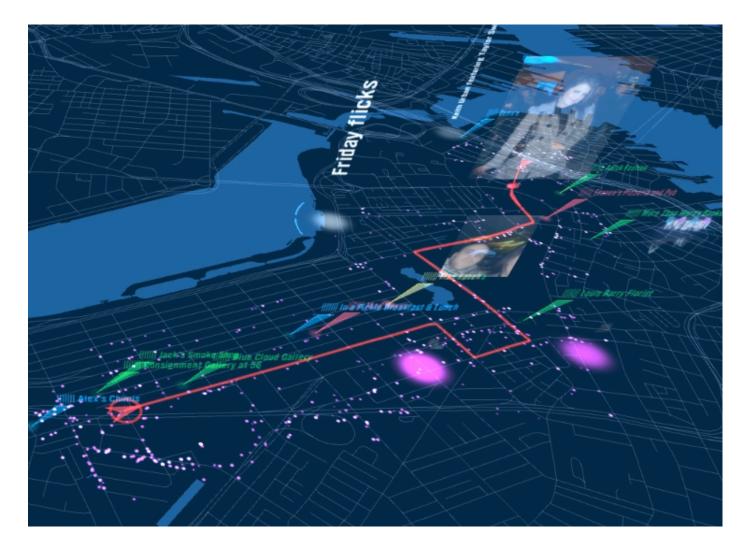
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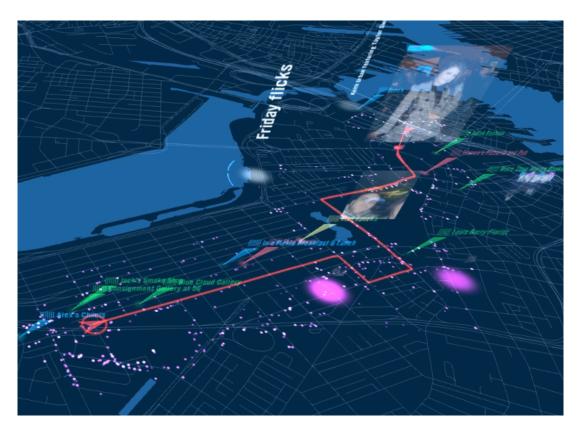


# Why people move?



## AIDA - Activity detection and prediction module

Solution: driver mobility model



- 1) Which are the places where the user goes?
- 2) Which kind of activities he/she likes?
- 3) In which sequence he/she does some activities?

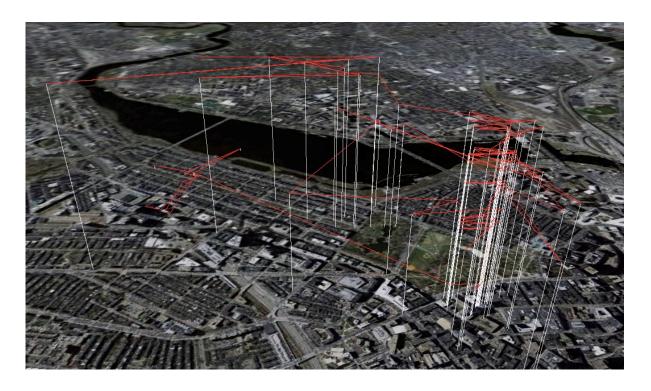
G. Di Lorenzo, S. Phithakkitnukoon, C. Ratti, Human Context-Aware Navigation: Improving urban living experience with predictive navigation system, Proceedings of UBI Challenge Workshop - Real World Urban Computing, Ubicomp, 2010



# Activity based modeling of human mobility

In activity-based models, travel demand is derived from the activities that individuals need/wish to perform

Goal: Understand urban mobility from individual telecommunication usage in space



## Mobile phone data set description

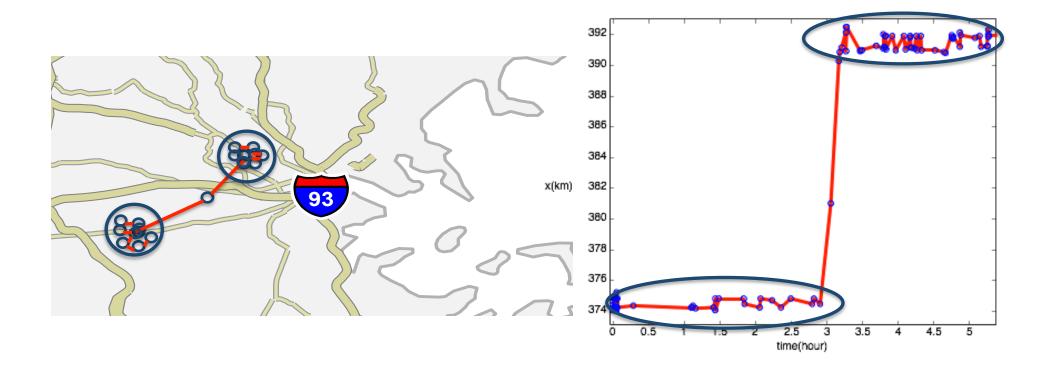
#### Sample:

- 130 million anonymous location estimations from approx. 1 million devices during 2 months
- 20% market share
- Location estimations when:
  - making or receiving a call
  - exchanging SMS messages
  - accessing internet
- Average localization error of 320 meters (median 220)

F. Calabrese, G. Di Lorenzo, and C. Ratti, Human Mobility Prediction based on Individual and Collective Geographical Preferences, Proceedings of 13th International IEEE Annual Conference on Intelligent Transportation Systems, ITCS, 2010.



#### Mobile phone trajectory



# Individual mobility model

#### Routine/non-routine

- Routine: home, work, market, ...
  - Historical data can model the sequence of stops
- Non-routine: concert, restaurant, a new city, ...
  - Historical data is not useful, but user preferences
- Individual/collective behavior
  - User preferences of the individual or of a group of people (e.g. people going to a concert)

$$P_{NEXT} = (1 - \alpha)P_{I} + \alpha P_{C}$$



- For each user
- 1. we extract the sequence of locations that he/she has visited during a day with a sampling rate of 1hour.
- 2. we connect each location to a geography of human activity
- 3. we then employ a process to identify and classify recurring patterns of users
- 4. we define an individual mobility model

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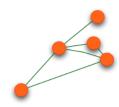


- For each user
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- 1. Stops
- 2. Sequence of trips

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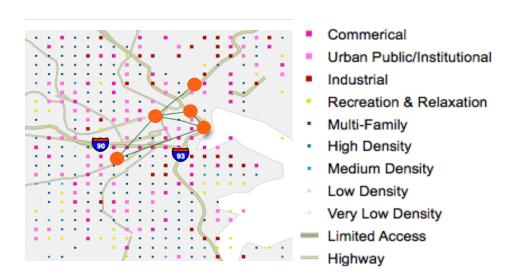


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- Land use dataset
- 2. Point of interest dataset



Giusy Di Lorenzo



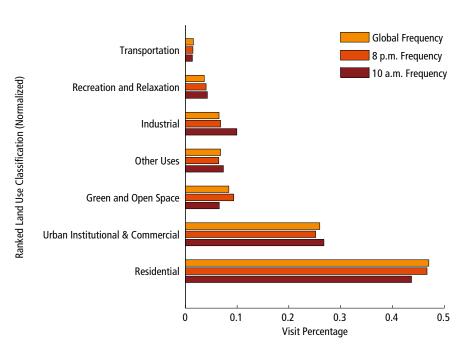
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- Land use and Point of interest visit 1. percentage distribution 2.
  - Trip length distribution

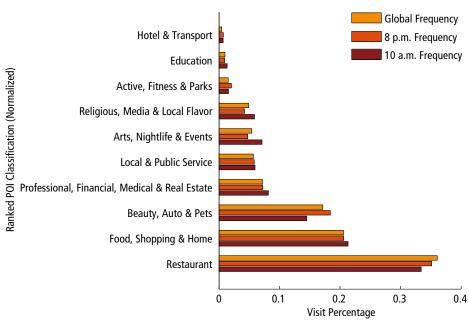


#### Land use and points of interest visit percentage distribution

Are geographical preferences useful?

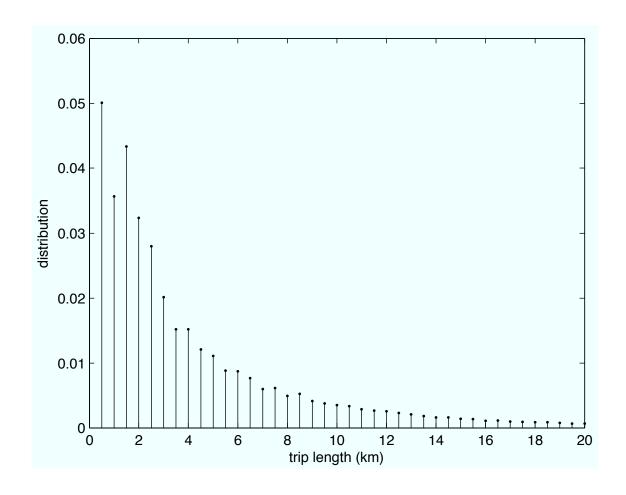


Distribution of visited land use categories



Distribution of nearby point of interest categories

# Trip length distribution



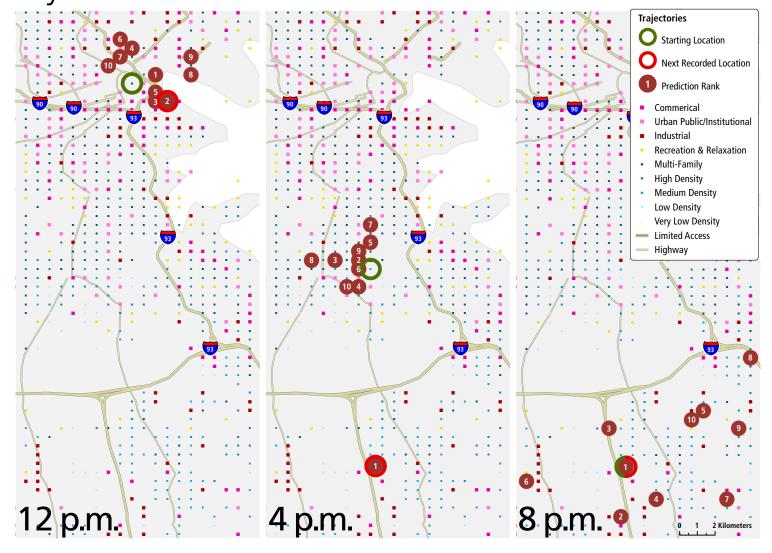
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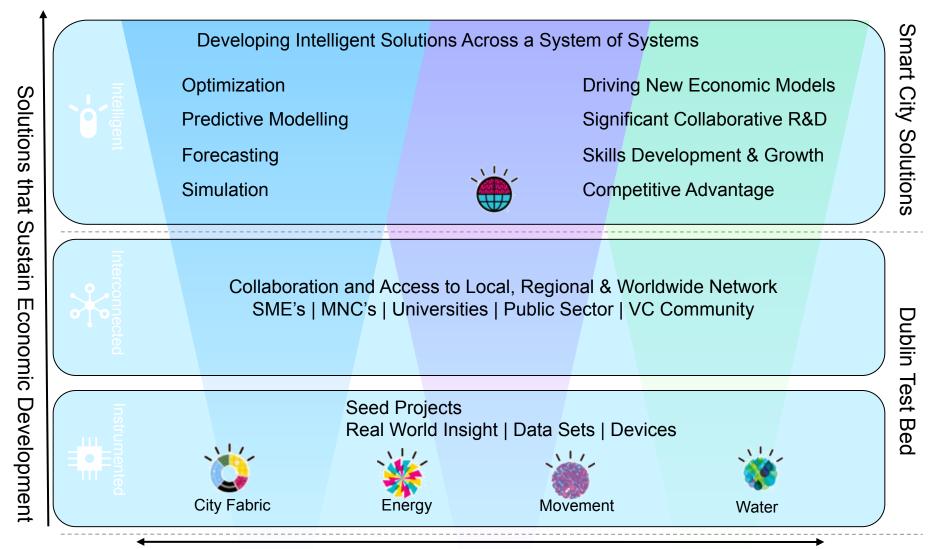
Individual and Collective behavior



Comparison of predictions & actual behavior of a randomly-selected user



#### The Smarter Cities Technology Centre merges Collaborative Research & Smarter Cities opportunities







# Thanks Giusy Di Lorenzo giusydil@ie.ibm.com