Commuting Dynamics For Change NetMob 2013 May 1 - 3, 2013 MIT / Data for Development [D4D] challenge

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INTRODUCTION

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GIS based technologies have been sufficiently proved in some sustainable development projects carried out in Africa related to international cooperation as well as in local development where the cross-sectorial work is critical. Our team used the geolocation data from call detail records extracted from Orange's customer base in order to know in which areas the customers have been moving around, to help us



HYPOTHESIS

Our hypothesis is that we can automatically identify the different stages of the commuting process from the data provided, creating a commuting profile of the studied area. The main idea behind this reasoning is that people usually

discover the morning and evening rush hours: the time when users were commuting between their place of residence and their place of work.

CONCLUSIONS

- Knowing people dynamics and, specially, commuting patterns, constitutes a great tool for improving infrastructure, logistics, and communications in general.
- A simply anonymized dataset does not contain name, home address, phone number or other obvious identifier. Yet, if individual's patterns are unique enough, additional information can be used to link the data back to an individual.
- Mobile devices are a great tool for real time sensing. Being almost everywhere, they can be of great help in acquiring data for generating more accurate predictive models.

METHOD

In our model, the temporal component is useful for providing a tool to move backward and

forward in order to get a more detailed understanding of the dynamics, not only moving across the timeline, but by creating temporal windows to group events. The geographical component provides a more high-level understanding related to the human mobility across the space at different levels, and connecting it to some other spatial features.









Using the antenna's position and the traffic intensity, a kernel function has been used to calculate a weight density in a neighbourhood around those points. Our kernel function is based on the quadratic kernel function. Conceptually, a smoothly curved

surface is fitted over each point. The surface value is higher at the location of the point and diminishes with increasing distance from the point, reaching zero at the 'search radius'



FUTURE WORK

- Apply our model to automatically identify the commuting profile for different places.
- Create geographical and time-based prediction models to better manage logistics and infrastructures.
- Reinforce prediction models with sentiment analysis extracted knowledge.
- Use the commuting profile of a region for outliers detection, anticipating possible critical events or

distance from the point. The volume under the surface equals the 'population field' value for the point. The density at each output cell is calculated by adding the values of all the kernel surfaces where they overlay the center of the cell.















