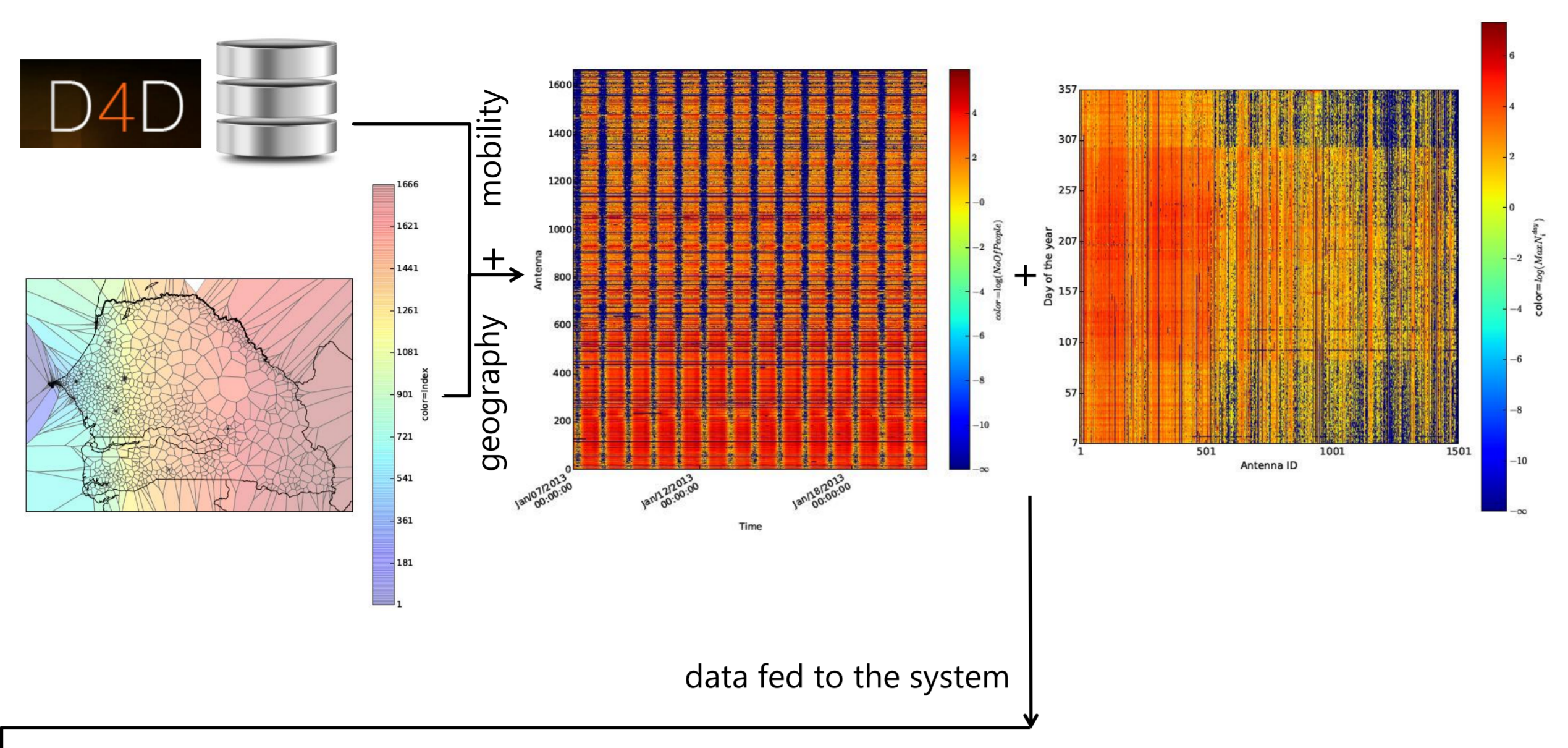
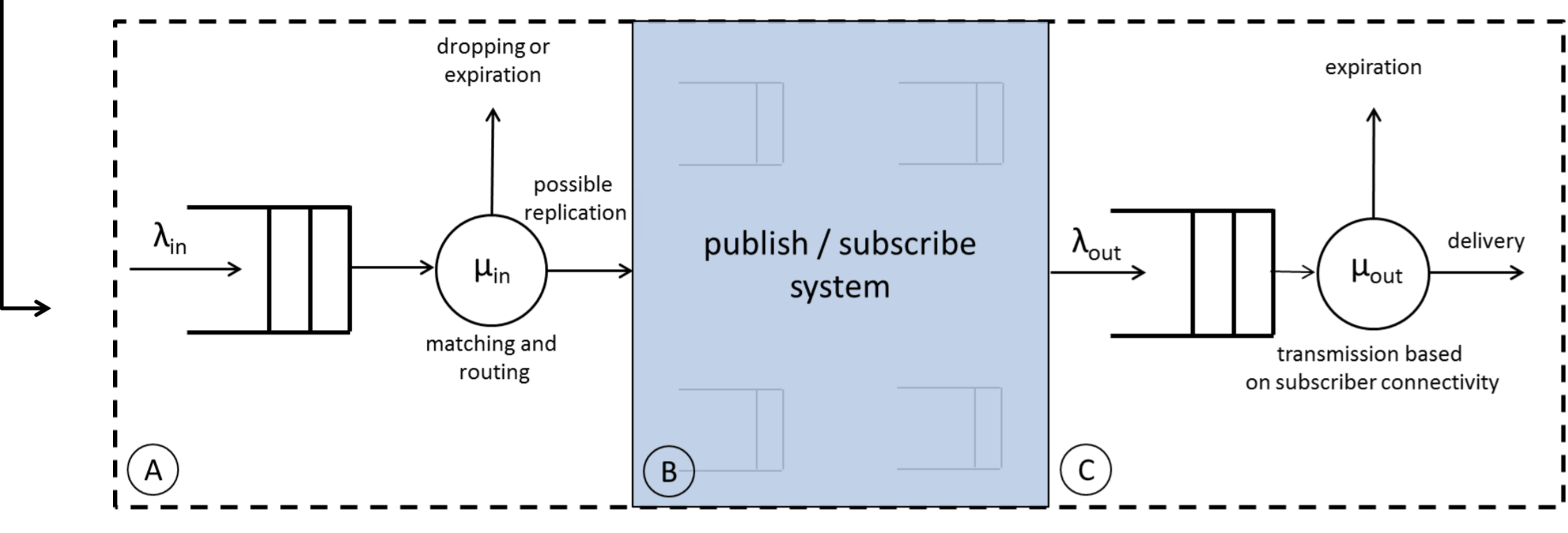


B Towards Mobile Social Crowd-Sensing for Transport Info. Management

Health	Transport Urban	National Statistics	Other
Agriculture	Energy	DataViz	Network



data fed to the system



- Georgantas, Nikoloas, Sr. Res. Scientist (lead Author)
- Bouloukakis, Georgios, Phd. Student
- Pathak, Animesh, Res. Scientist
- Issarny, Valerie, Res. Director ▪ Agarwal, Rachit, Post Doc

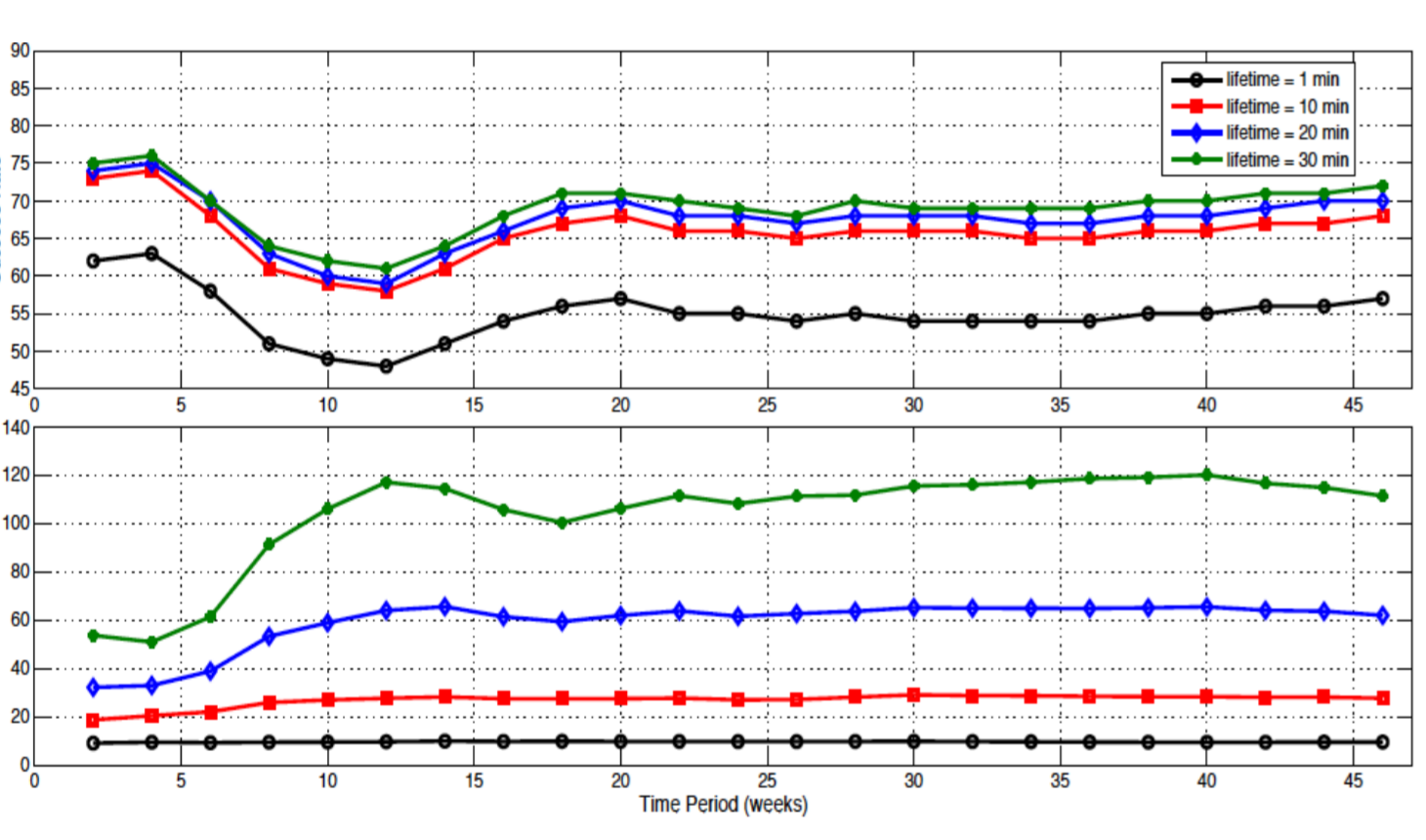


Project Summary:

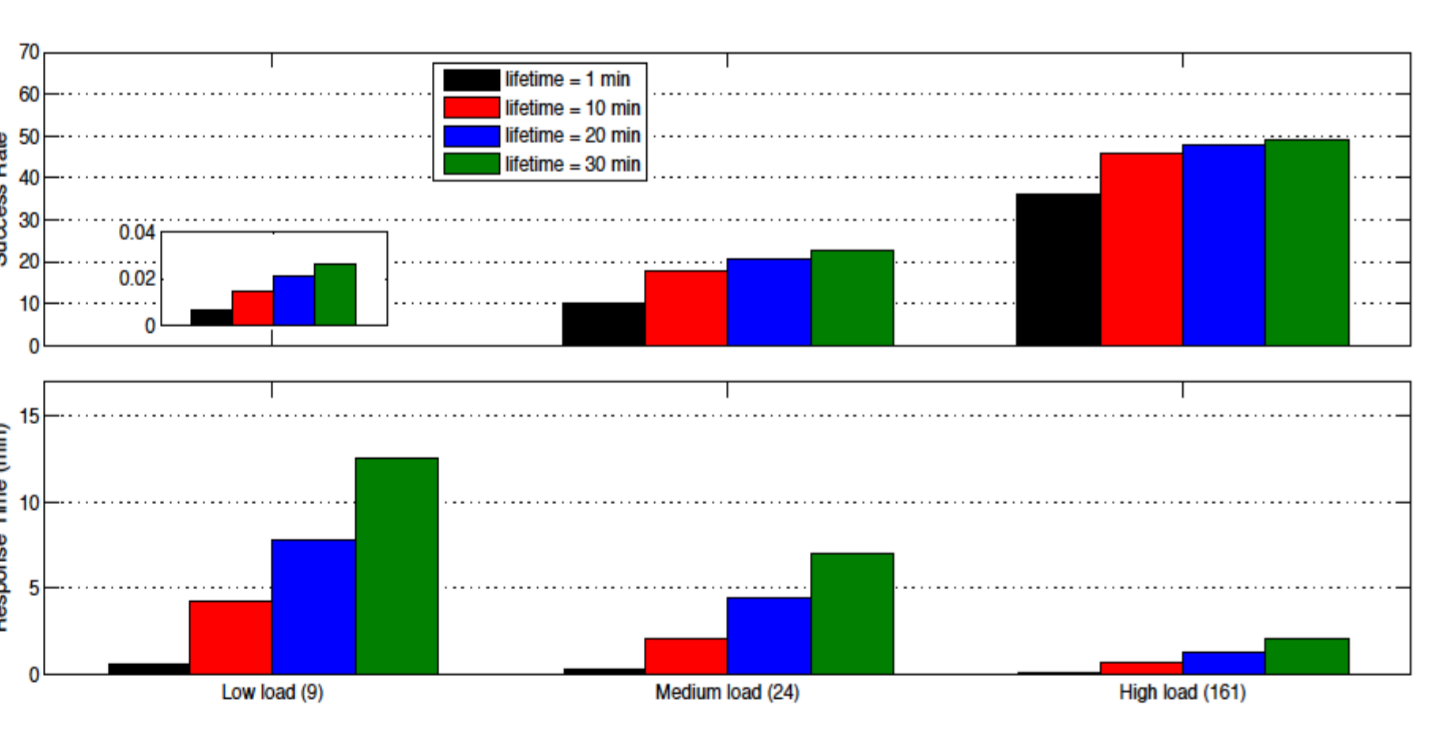
Transport in Senegal is chaotic and large, especially in main cities. Additionally, although most people have mobile phones, large part them still rely on SMS. Considering this, we propose the development of an application platform for large-scale transport information management relying on 'mobile social crowd-sensing'. To support this platform, we model a large-scale mobile publish/subscribe system using queuing theory. We developed the MobileJINQS simulator with the realistic load for the analysis.

Possible use for development:

The project provides telecom providers inputs to better tune the communication backbone, and application developers a platform for transportation application.



Success rates and response times for network traffic from low load Antenna 9 to high load Antenna 161 with varying message lifetime periods



Mean end-to-end transaction success rates and response times

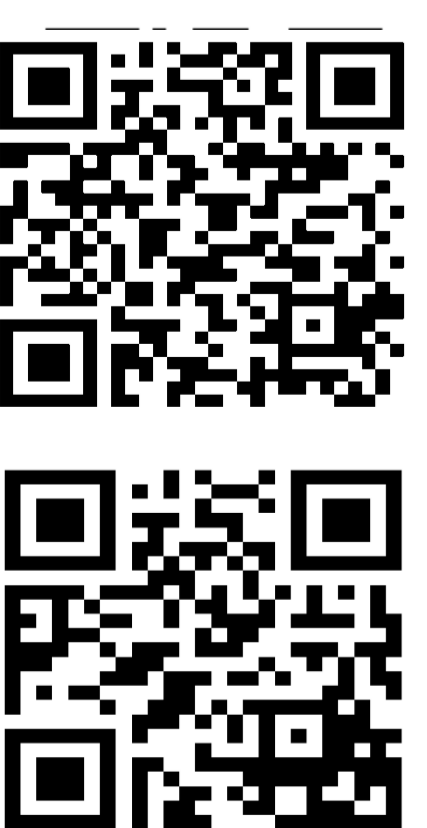
Main results:

- High load observed in antennas near Dakar.
- Varying incoming loads and service delays has a significant effect on response time.
- Success rate and response time are directly proportional to Message lifetime with proportionality constant greater for response time.
- Response time is dependent on subscribers behavior.

By properly setting event lifetime spans, system or application designers can best deal with the tradeoff between freshness of information and information delivery success rates. Still, both of these properties are highly dependent on the dynamic correlation of the event input flow and delivery flow processes, which are intrinsically decoupled.

Method:

- Let N_i^t be the number of people in an antenna i at a given time t over the period of the trace (50 weeks)
 - Let λ_{in} be the input process at the input access point associated to the antenna i , then λ_{in} is a non-homogeneous Poisson process with rate $\lambda(t) = N_i^t / |t|$. Similarly μ_{out} is a non-homogeneous Poisson process with rate $\mu(t) = N_j^t / |t|$ at the output access point associated to the antenna j .
 - μ_{out} is equivalent to service time that follows an exponential distribution with mean equal to $1/\mu(t)$.



Full paper is here:
<http://xsb.inria.fr/docs/d4d2015.pdf>

DataViz or video are here:
<http://xsb.inria.fr/d4d#visualization>

- Data sources used for this project:
- D4D data set 1, communication between antennas
 - D4D data set 2, high resolution movement routes
 - D4D data set 3, low resolution movement routes
 - D4D synthetic data set
- Other data sets used in this project:
- None

- Main Tools used:
- XSB
 - MobileJINQS
 - Queueing Theory
 - Python mpl_toolkit
- Open Code available:
- Yes
 - No