B. Ducourthial

Team

Introduction Architecture for sensing Pollution Summary

Vlethodology Reference mag Region of Interest Kriging

Results

Comparing maps Robustness against errors

Conclusion



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> IWCMC August 2015

heudiasyc





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Team Institutions

Mobile Pollution Sensing

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Université de Technologie de Compiègne ~4500 students, master degree (engineer diploma), PhD http://www.utc.fr

- One of the first French engineering school for computer science
- Close to Paris and Charles de Gaulle airport



Heudiasyc lab from the UTC & CNRS http://www.hds.utc.fr Equipex Robotex, Labex MS2T





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- Point of view
 - Dynamic networks are different!
- Methodology
 - Real applications
 - 2 Designing new algorithms
 - Proof of concept Performances issues Analytic proofs
 Proof of concept Tests or network emulation Distributed algorithms
- Tools
 - Airplug Software Distribution
 - Communicating embedded disposals
 - On-Board-Units, Road-Side-Units in Compiègne

https://www.hds.utc.fr/airplug





Team

Approach

Team Projects

Team

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- Inter-vehicles cooperative perception for road safety National project 2008-2011
- Distributed system for vehicle dynamic evaluation Regional project 2008-2011
- Data gathering from VANET to infrastructure Industrial project Orange lab
 2008-2010
- Distributed applications for dynamic networks Regional project 2007-2010
- SafeSPOT European IP project 2006-2010
- Network services for com. between mobiles objects Industrial project Orange lab 2004-2008
- Road anticipating

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Regional project 2004-2007



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Team Contributions

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| • Experiments with dist. data fusion | I [VNC | 2014] |
|--------------------------------------------------|----------|-------|
| • Experiments with sensors | [WiSARN | 2014] |
| I2V experiments | [ITSC | 2014] |
| V2I experiments | [IWCMC | 2014] |
| V2V unicast communication | [WCNC | 2014] |
| Distributed data fusion | [SSS | 2012] |
| Data collection on the road | [IV | 2012] |
| • Performances in a convoy of vehic | les [VTC | 2011] |
| V2I architecture | [Mobiwac | 2010] |
| • Distributed dynamic group service | [SPAA | 2010] |
| Vehicular networks emulation | [ICCCN | 2010] |
| • Simulation of vehicular networks | [VTC | 2010] |
| Experimenting on the road | [VTC | 2009] |
| Messages forwarding | IEEE TVT | 2007] |
| | | |

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- POSIX OS
- Core program
 - user-space process

Airplug framework 1/2

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- networking
- Applications
 - user-space process
 - read on stdin
 - write on stdout
 - API close to IEEE WSMP
- Ensure tasks and OS independence for robustness
- Open to any programming language

utc

Team Airplug framework 2/2

Team

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• New protocols developed in user space processes

- Open to new networking solutions
- Cross-layer solutions facilitated





Team Complete platform

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• Airplug-term \rightarrow rapid prototyping

- Airplug-emu \sim study by emulation
- Airplug-live → real experiments (vehicles, UAV)
- + remote, notk...

(airplup) app=005 ident=10

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Architecture for sensing Measuring pollution in cities Summary





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Agenda

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| Architecture | for | sensing |
|--------------|-------|--------------|
| Horizo | ontal | architecture |



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| CITS | | |
| heudiasyc | | utc Université de Te Complègny |

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Mobile

Pollution Sensing

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• Context

- Excessive accumulation of the pollutants can occur in parts of cities
- Detailed maps help at notifying people
- But
 - Pollution mapping relies on very precise sensors
 - Very high cost → limited number
- Mobile cheaper sensors to improve the maps?





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Mobile sensing Objectives: Ozone case study

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Good ozone

- In the stratosphere 15 to 50 km above the Earth
- Protects the life from the sun's harmful UV-b
- ${\sim}80\%$ of 0_3 but the layer is thin...
- Bad ozone
 - In the troposphere 0 to 15 km above the Earth
 - Air pollutant damaging human health, vegetation...
 - Ground ozone created by chemical reactions between oxides of nitrogen (NOx) and volatile organic compounds (VOC) with sunlight
- Consequences
 - Even low level of 0_3 can cause health effect
 - Eg. inflame the lining of the lungs Especially children because their lungs are developping



Mobile sensing Objectives: Compiègne case study

Mobile Pollution Sensing

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Mobile sensing Objectives: Compiègne case study



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| Parameters | Station 1 | Station 2 | Station 3 | |
|----------------------------------|-------------------------|---------------------|--------------------------------|--|
| station_european_code | FR18014 | FR18015 | FR18032 | |
| station_local_code | FR18014 | FR18015 | FR18032 | |
| country iso code | FR | FR | FR | |
| country_name | France | France | France | |
| station_name | A. Thierry COMPIEGNE | Mairie COMPIEGNE | Desbordes Compiegne | |
| station_start_date | 01/01/79 | 01/01/79 | 05/11/97 | |
| station_end_date | | | | |
| type_of_station | Industrial | Background | Background | |
| station_ozone_classific ation | | | urban | |
| station_type_of_area | suburban | urban | suburban | |
| station_subcat_rural_ba ck | | | | |
| street_type | | | | |
| station_longitude_deg | 2.838.058 | 2.827.789 | 2.818.055 | |
| station_latitude_deg | 49.425.556 | 49.418.613 | 49.402.500 | |
| station_altitude | 35 | 53 | 57 | |
| station_city | | | | |
| lau_level1_code | 6097 | 6097 | 6097 | |
| lau level2 code | | | | |
| lau_level2_name | Compiègne | Compiègne | Compiègne | |
| EMEP_station | no | no | no | |
| Measured indexes | SO2 | SO2 | SO2, PM10, O3, NO2, NOX, NO | |



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Mobile sensing Objectives: summary and methodology

- Objectives
 - Comparing static and mobile sensing
 - To estimate the pollutant levels in a Rol
 - while varying the number of samples
 - and taking sensor errors into account.
- Methodology
 - Reference map
 - 2 Extrapolating measures to compute a map
 - Comparing the reference and the computed maps



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Methodology Reference map Region of Interest Kriging









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3 Methodology Reference map

Region of Interest Kriging







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- Adapting spatial patterns from city of Badajoz [Moral-García et al. 2010]
- Using the June map
 - Appears to be more difficult to estimate

•
$$f(x) = ae^{-\frac{1}{2}(\frac{x-x_0}{\sigma})^2} + d$$

• Determining the parameters





Methodology Reference map 2/2

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- Extending the function to two dimensions
- Determining hot spots
- → Reference map for Compiègne



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Methodology Region of Interest (Rol)

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- All bus stops
- Line 5 bus stops
- Line 5 route





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- Extrapolating the measures to compute the map → kriging
- Require to estimate the

co-variance

Mean variation as the distance from a sample increases

• Depends on the samples set All bus stops vs. line 5



Methodology

Extrapolating the measures





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Introduction Architecture for sensing Pollution Summary

1.87%

Methodology

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Using all bus stops: average value of the relative absolute error =



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Results

4000 5000

Reference map vs. computed map 1/3



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• Using only bus stops of Line 5

average value of the relative absolute error = 5.33% (in the Rol)



Results

Reference map vs. computed map $2/3\,$





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Using the measures on the moving bus: average value of the relative absolute error = 2.27% in the Region Of Interest





Results

Reference map vs. computed map 3/3

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Comparing maps

Robustness against errors

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Results Robustness against errors 1/2

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- Robustness against errors



- All bus stops
 - Measure errors up to 5% → minimal impact
 - Only 60% of bus stops ~> minimal impact



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Robustness against errors 2/2

Results

- Line 5 bus stops
 - Measure errors has a greater impact (linear)
- Moving sensor on the Bus 5
 - Less impact with the moving sensor
 - Measure error of $10\% \rightarrow$ result error of 6.7%





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Mobile Pollution Sensing

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• Problem

- Detailed pollutant maps are mostly unavailable
- What about using cheaper less precise but mobile sensors instead?

• Method

- Generating a reference map
- Pick up some exact measures All bus stations / Line 5 only / Bus 5 only
- Extrapolating using kriging
- Measuring errors
- Results
 - A single equipped bus:
 - \sim 40% of all bus stations
 - 2% better than only Line 5 stations
- Mobile sensing
 - Promizing
 - Still many things to confirm



