

Young Researchers Seminar 2009

Torino, Italy, 3 to 5 June 2009

Integrating the effects of adverse weather conditions on traffic:

Methodology, empirical analysis and simulation

Romain Billot



Outline

- I. Problem Statement
- II. Methodology
- III. Results from real world data
- IV. Online traffic state estimation
- V. Concluding remarks



I. Problem Statement



- Proactive real-time traffic management systems involve a comprehensive knowledge of all elements impacting traffic conditions.
- Adverse weather conditions are well recognized as one important event which can severely impact traffic in terms of traffic operations and safety
- What are adverse weather conditions ?
The presented work focuses only on the effects of **rain** on traffic.



How does adverse weather impact traffic ?

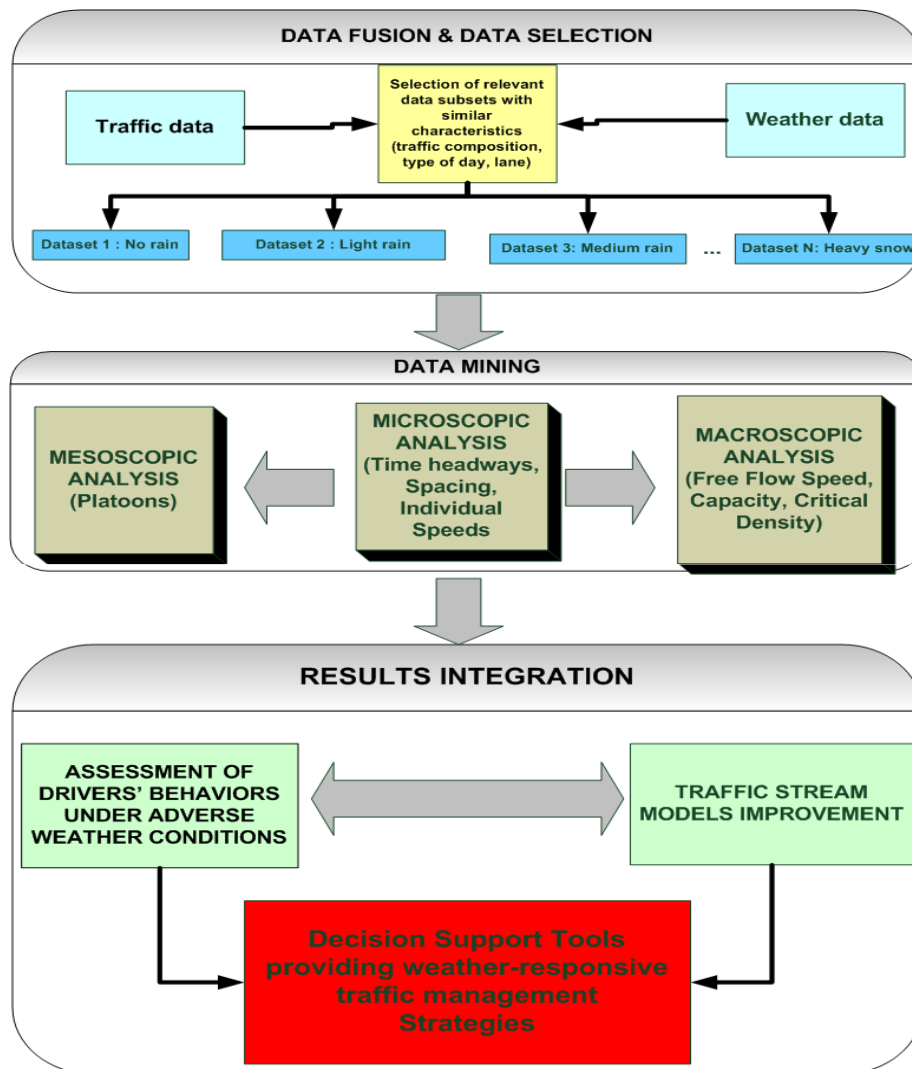
DRIVERS' BEHAVIOUR	Speed and acceleration	↘
	Time Headways	↗
	Spacing	↗
	Lane Changing	
	Platooning ?	
TRAFFIC OPERATIONS	Capacity	↘
	Traffic Volume	↘
	Speed	↘
	Speed Variation	↗
	Congestion Severity	↗

How does adverse weather impact safety ?



- Precipitation impacts the driver's safety by degrading the state of the pavement, reducing the visibility as well as light,
- Increase of the crash frequencies and above all **crash severity**,
- Lagged effect of precipitations across days : the effect of rain is higher if many days have passed since the last precipitation.

II. Methodology



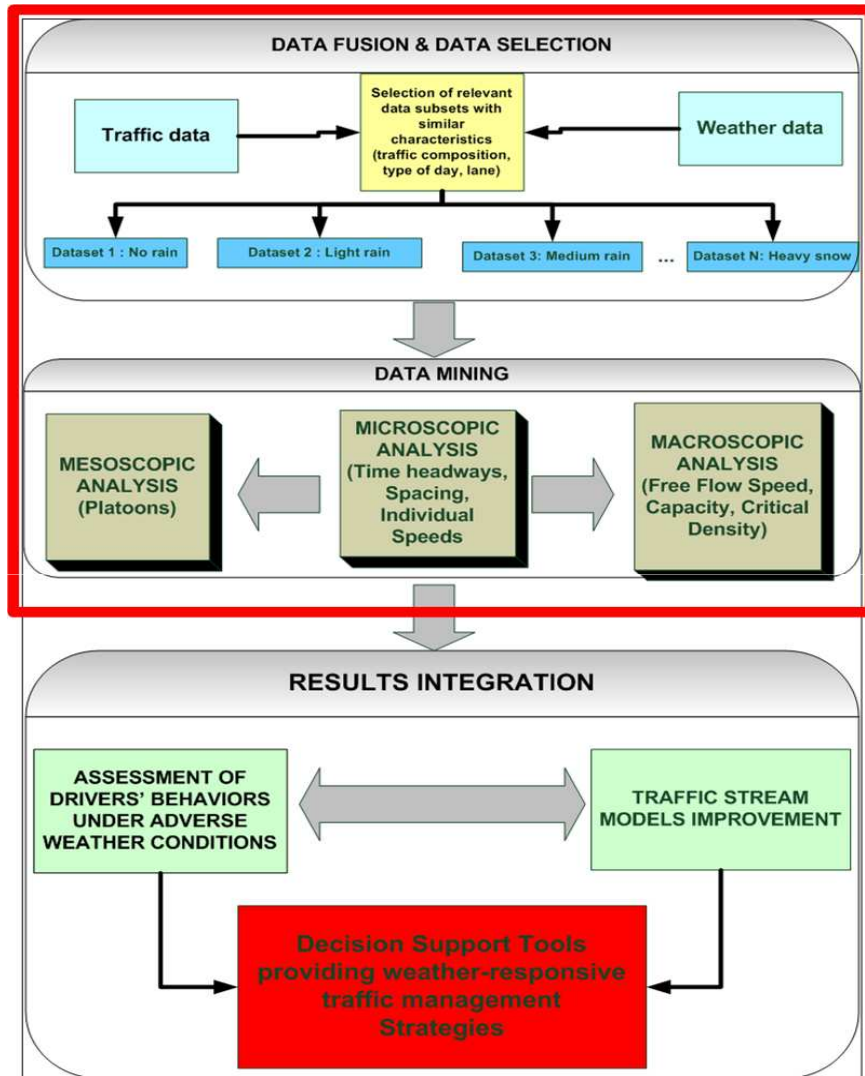
- Relevant datasets
- Multi-level assessment
- Integration of weather effects into traffic models
- Online weather-responsive decision support systems (DSS)

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Methodology



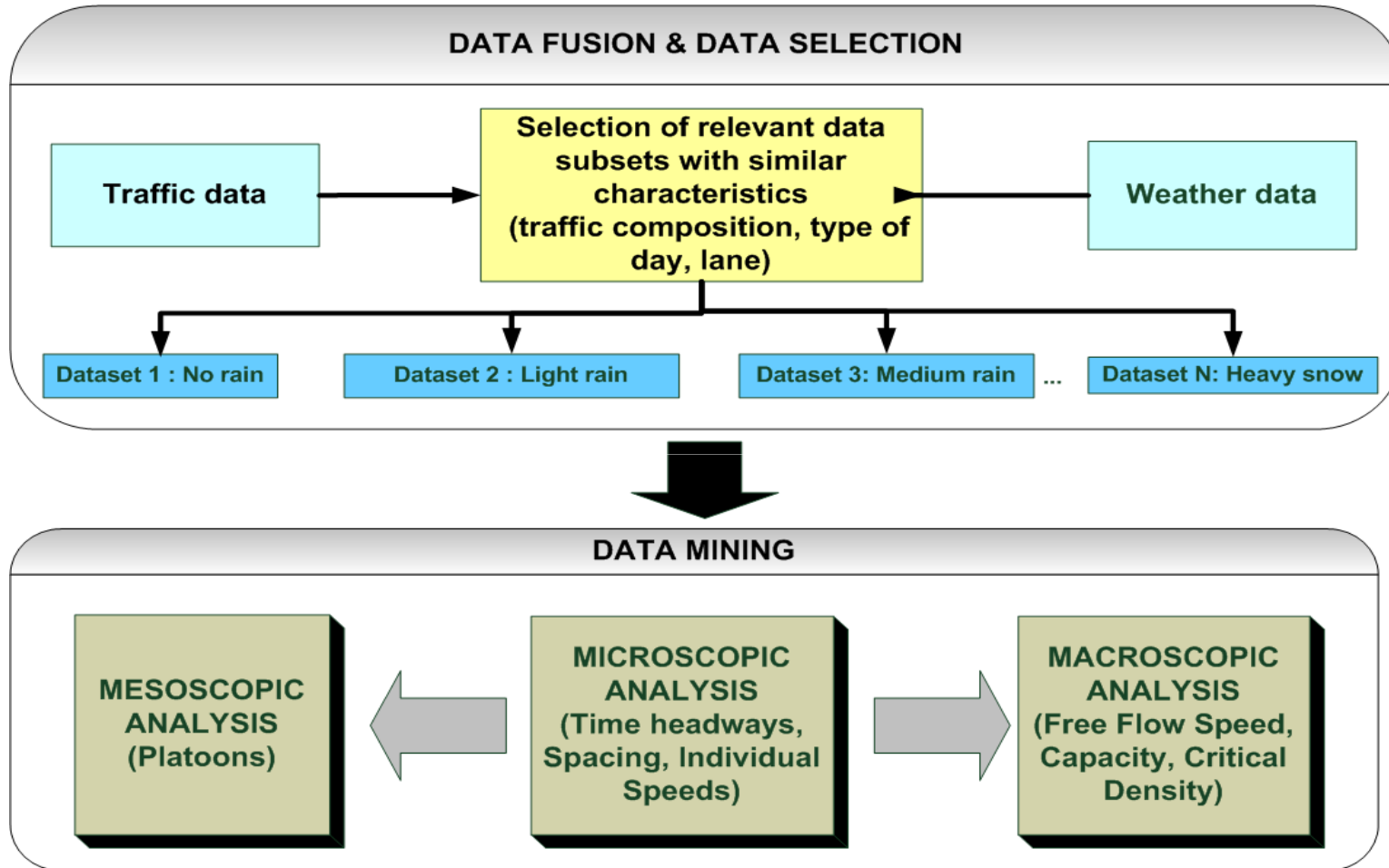
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Methodology (II): data selection & data mining



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III. Results from real world data: 118 national road



- Two-lane freeway section near Paris with high traffic volume
- Traffic data collected from 2005 to 2007 from 9 double-trap loop sensors.
- First traffic data selection from 3 out of 9 sensors located on a homogeneous section (same topography)

Datasets construction

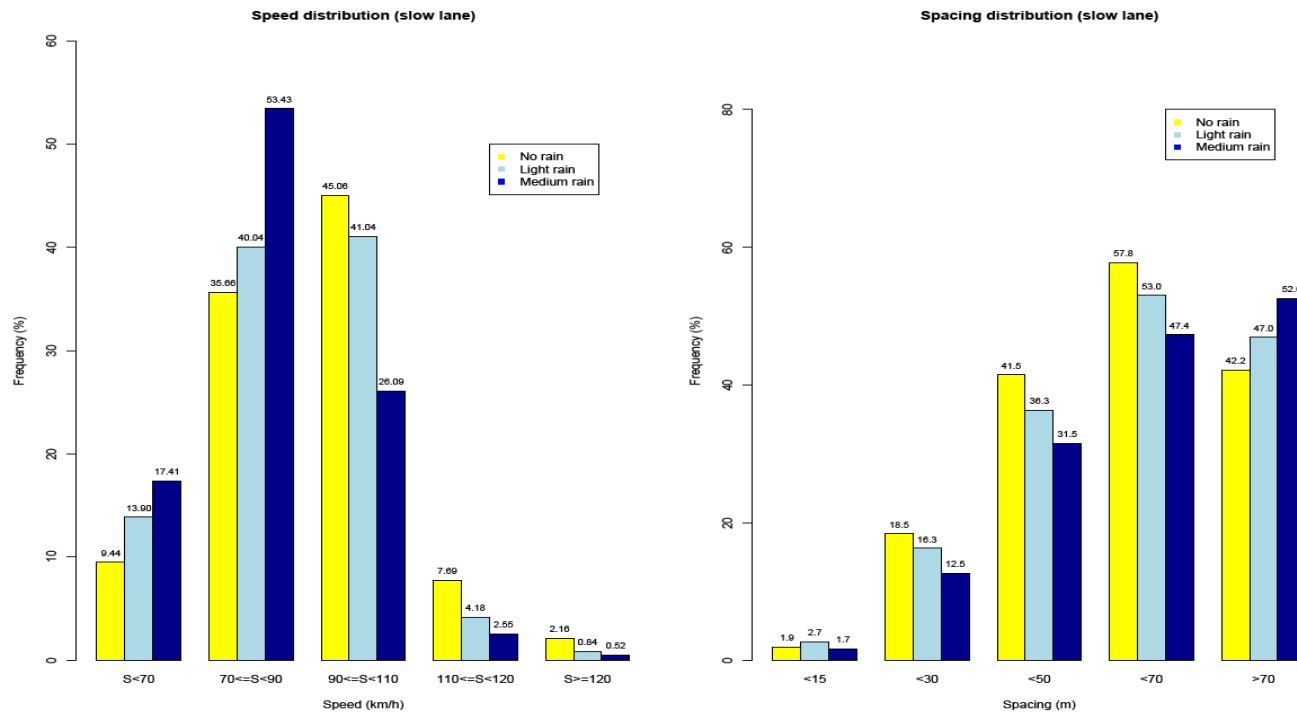
TRAFFIC DATA

- Same day category (regular weekday) and same time periods (morning/evening peak hours)
- Similar traffic composition & significant amount of vehicles (>10 000 vehicles per lane).
- Speed limit = 110 km/h (68.3 mph). Only the southbound direction towards Paris is considered.

WEATHER DATA

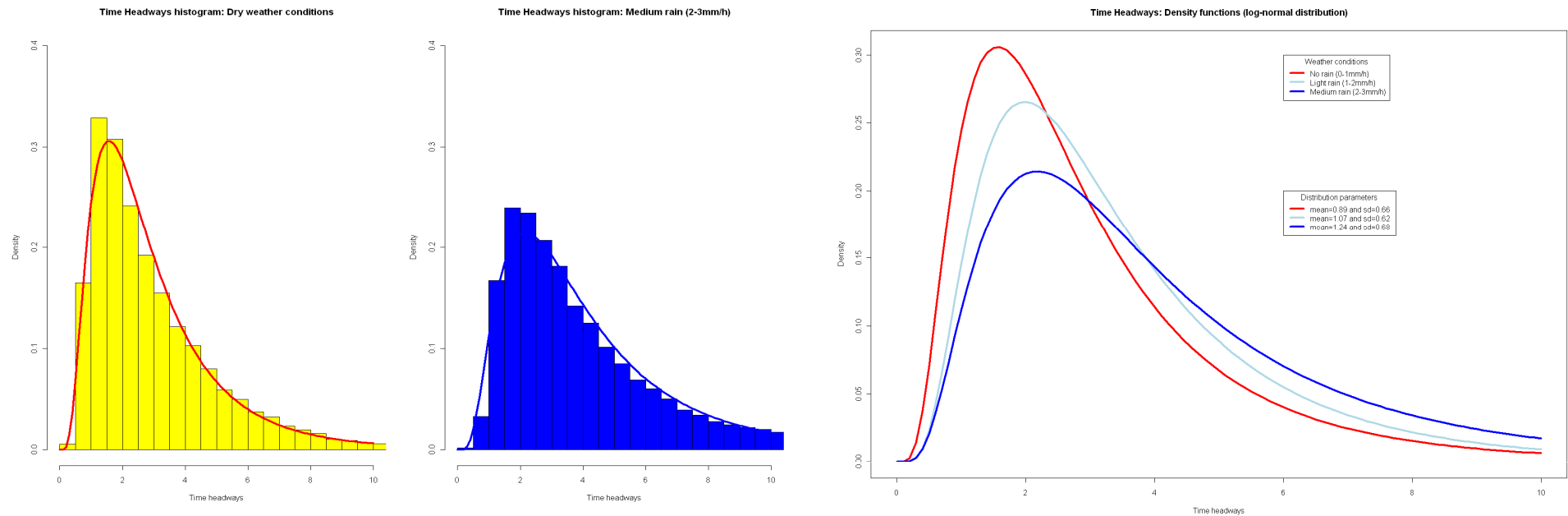
- Hourly weather data provided by a consistent weather station located near the section.
- Then, traffic data were divided into 3 datasets according to the rain intensity:
 - **No rain:** rainfall= 0 mm/h
 - **Light:** up to 2 mm/h
 - **Medium:** from 2 to 3 mm/h

Results: speed and spacing distributions



- Clear decrease for the slow lane of the frequencies of speeds > 90 km/h under rainy conditions. Primary adaptation to rain = speed reduction.
- Frequency of short spacing decreases during inclement weather conditions -> Drop of about 10% of the spacing < 50m under medium rain conditions.

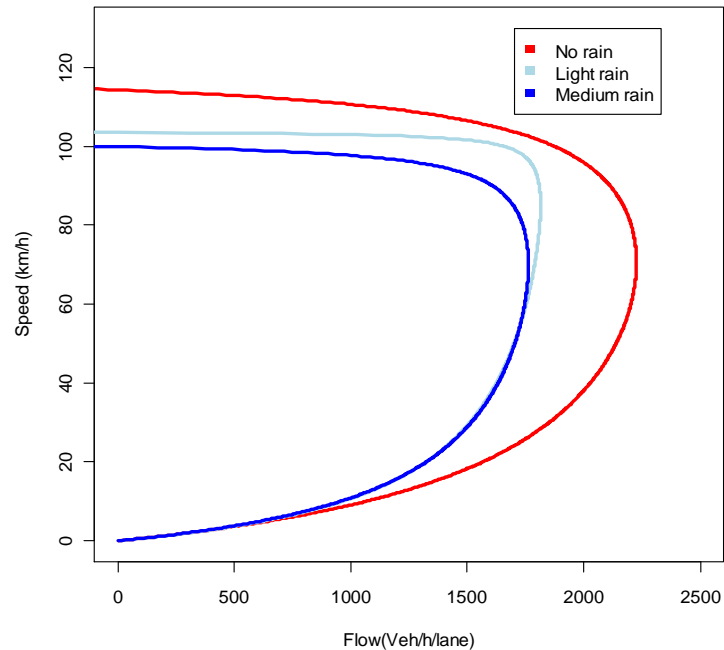
Results: Time headways (TH) distributions



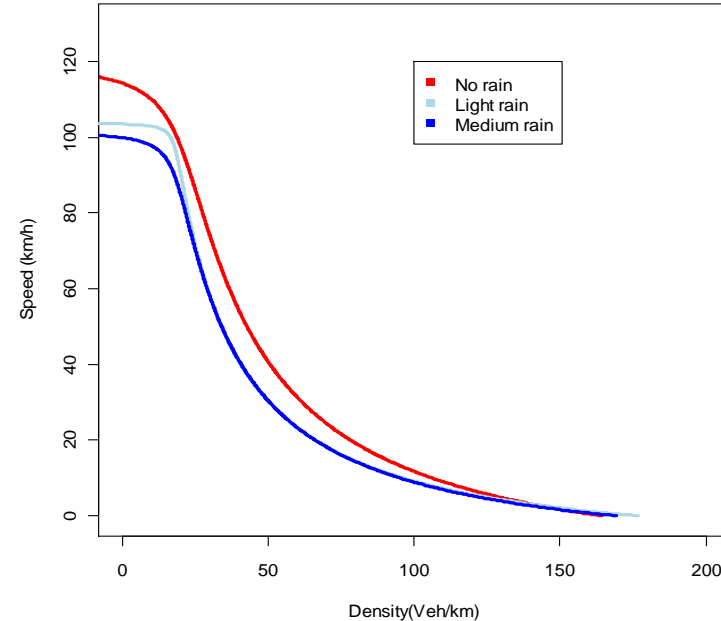
- Frequency of TH <3s higher under dry conditions.
- Good fit of the log normal distribution with our TH distribution.
- The higher the intensity of rain is, the lower the density of short TH is.

Results: Macroscopic level

Speed-Flow Relationship (Van Aerde Model)



Speed-Density Relationship (Van Aerde Model)



- Calibration of Van Aerde Model.
- Drop in free-flow speed: 8% under light rain conditions and of about 12.6% under medium rain conditions.
- Roadway capacity decreases by 18.5% vs 21% under light & medium rain conditions.

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Consistency with other studies

Light rain conditions: reductions coefficient of two key parameters on freeways:
free-flow speed and capacity:

Country	Road type	Free-flow speed	Capacity	Reference
France	interurban area	8% – 12.5%	18.5% – 21%	Previous study
France	urban ring road	9%	15.5%	Billot et al. (2008)
Canada	urban freeway	10%		Andrey et al. (2006)
USA	3 metropolitan areas	6% – 9%	10% – 11%	Rakha et al. (2008)
Japan	metropolitan expressway	5%	6 – 9%	Chung et al. (2006)

- Results are context-dependent (urban/interurban area).
- Regarding other world-wide studies, traffic parameters reductions are slightly higher.
- Regional differences need to be analyzed.

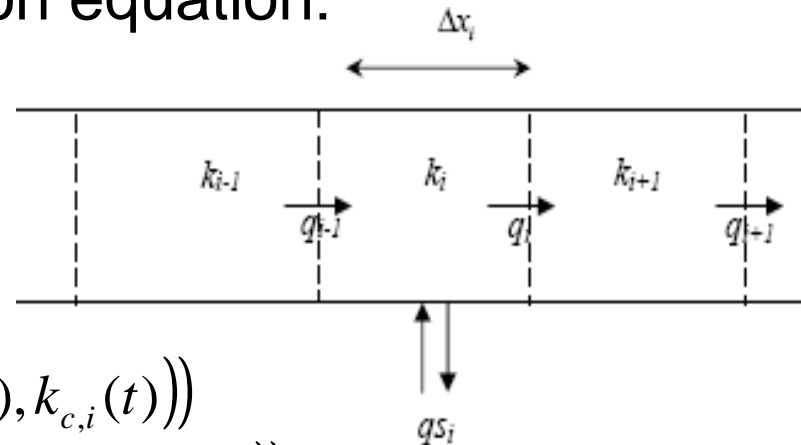
IV. Toward online weather-responsive estimation

- Goal: to show how the new knowledge about the rain effects would be useful from a simulation point of view.
- Tools: Bayesian framework with the use of Monte Carlo observer-based traffic flow estimations.
- Simulation case from real world data: eastern part of Lyon's ring road (urban motorway)

Macroscopic traffic model

- Discretized version of conservation equation:

$$k_i(t + \Delta t_N) = k_i(t) + \frac{\Delta t_N}{\Delta x_i} (q_{i-1}(t) - q_i(t))$$



- Gudunov balance equation:

$$\left\{ \begin{array}{l} \text{Cell } i \text{ demand: } \Gamma_i(t) = Q_{e,i,t}(\min(k_i(t), k_{c,i}(t))) \\ \text{Cell } i+1 \text{ supply: } \Omega_{i+1}(t) = Q_{e,i+1,t}(\max(k_{i+1}(t), k_{c,i+1}(t))) \\ \text{Resulting flow: } q_i(t) = \min(\Gamma_i(t), \Omega_{i+1}(t)) \end{array} \right.$$

- State Vector: $x_t = (k_1(t), k_2(t), \dots, k_n(t), q_0(t), q_1(t), \dots, q_n(t))^T$

$$\longleftarrow \xrightarrow{2n+1}$$

$$x_{t+1} = f(x_t)$$

Observation equation and state estimation

- The measured quantities (vector y_t) are flows and densities in certain cells, i.e. components of the state vector x_t :

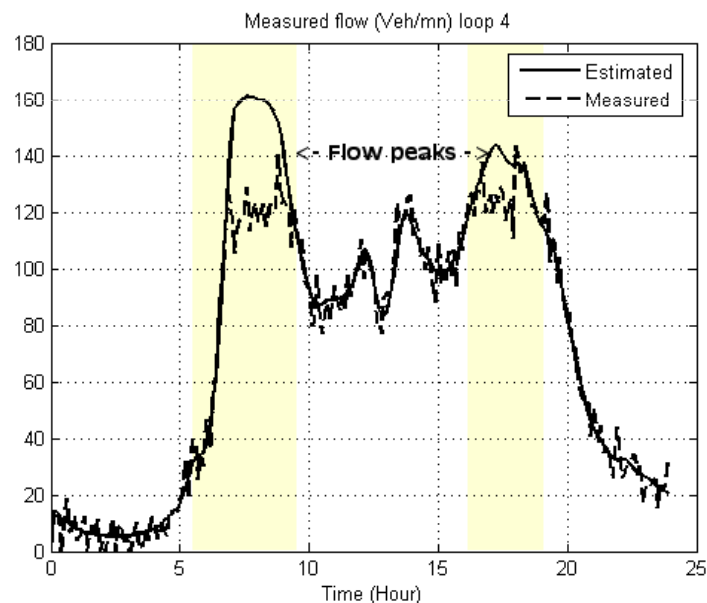
$$y_t = Cx_t$$

- Bayesian framework: the dynamic state estimation is carried out through the construction of the posterior probability density of the state.
- Tools: sequential Monte Carlo (SMC) State Estimation

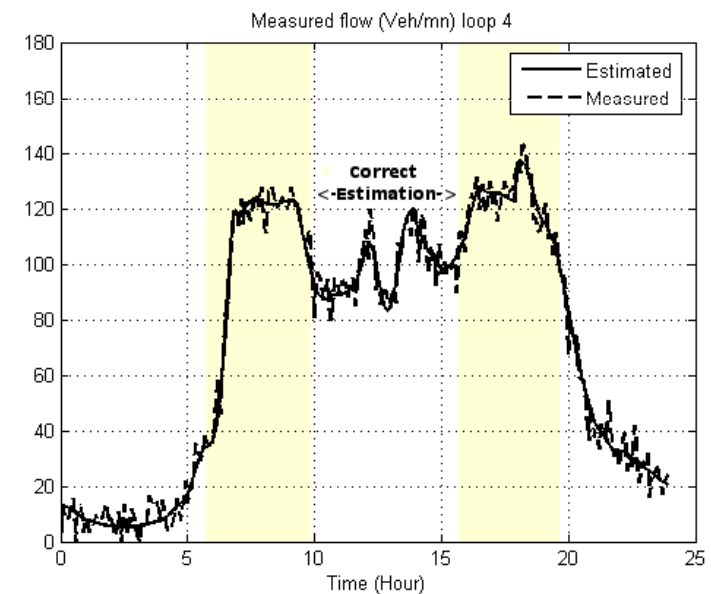
deriving an estimate of: $P(x_t | x_{t-1}, u_{t-1}, y_t)$

Integration of weather in a observer-based model

- A day-long scenario under adverse weather conditions is created.
- Without any knowledge about rain effects => overestimation of traffic volume during peaks hours
- By introducing the new knowledge about rain impacts into the fundamental diagram => new simulation leads to **correct estimations**



Switch to an
adverse weather
fundamental diagram



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Perspectives & current works

- Analysis of weather impact on platooning: does a platooning phenomenon raise more often under adverse weather conditions ?
- Wider range of rain and snow intensities -> analysis of swiss and japanese data enabling also comparison of regional differences.
- Effect of weather on travel times: french regional project TPTEO with road operator AREA
- **Adaptative traffic modelling according to prevailing weather conditions.**
- **Vlasov Fokker Planck traffic modelling taking into account platooning and the weather parameter**

Conclusion

- Standardized methodology for weather impact integration into DSS.
- Work undertaken within an european project enabling data and knowledge sharing:
COST ACTION TU0702 (<http://tu0702.inrets.fr>)
- Goal: Provide the road operators with weather-sensitive management tools (DSS)

Thank you for your attention !

**Contact:
romain.billot@inrets.fr**

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