

Winter risk estimations through infrared cameras an principal component analysis

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Ressources, territoires et habitats
Énergie et climat
Prévention des risques
Développement durable
Infrastructures, transports et mer

**Présent
pour
l'avenir**



UNIVERSITÉ
— PARIS-EST

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Ministère de l'Écologie, de l'Énergie, du Développement Durable, des Transports et du Logement

1- Objectives

Context:

- Instrument on a vehicle to establish road susceptibility to ice
- Measurements on one lane at a time \Rightarrow tests too long
- Measurements on a small spot (*radiometer FOV 20°*)
- No appreciation of road radiative environment
- No dynamic appreciation with season of ice occurrence susceptibility

Objectives:

- improve the winter risk determination of road network with an infrared camera

(several lanes, environment contribution evaluation, ...),

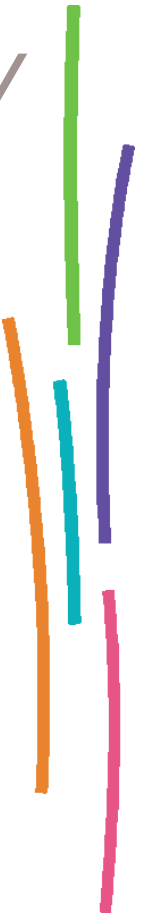
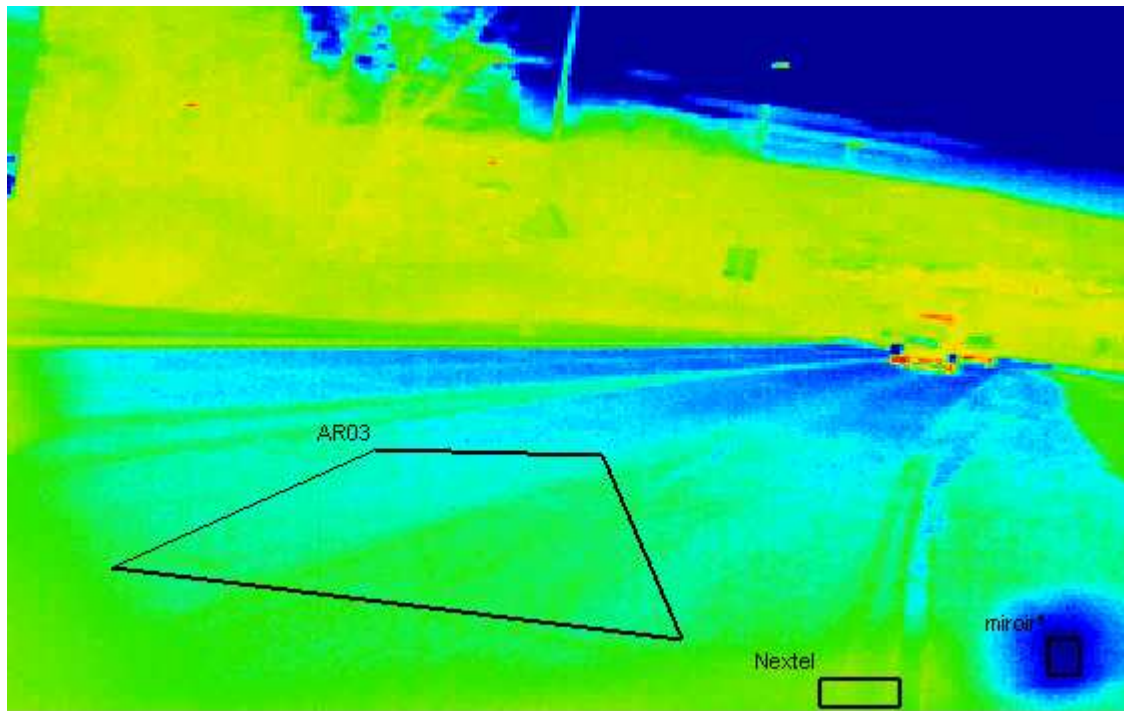
- extend measurements season

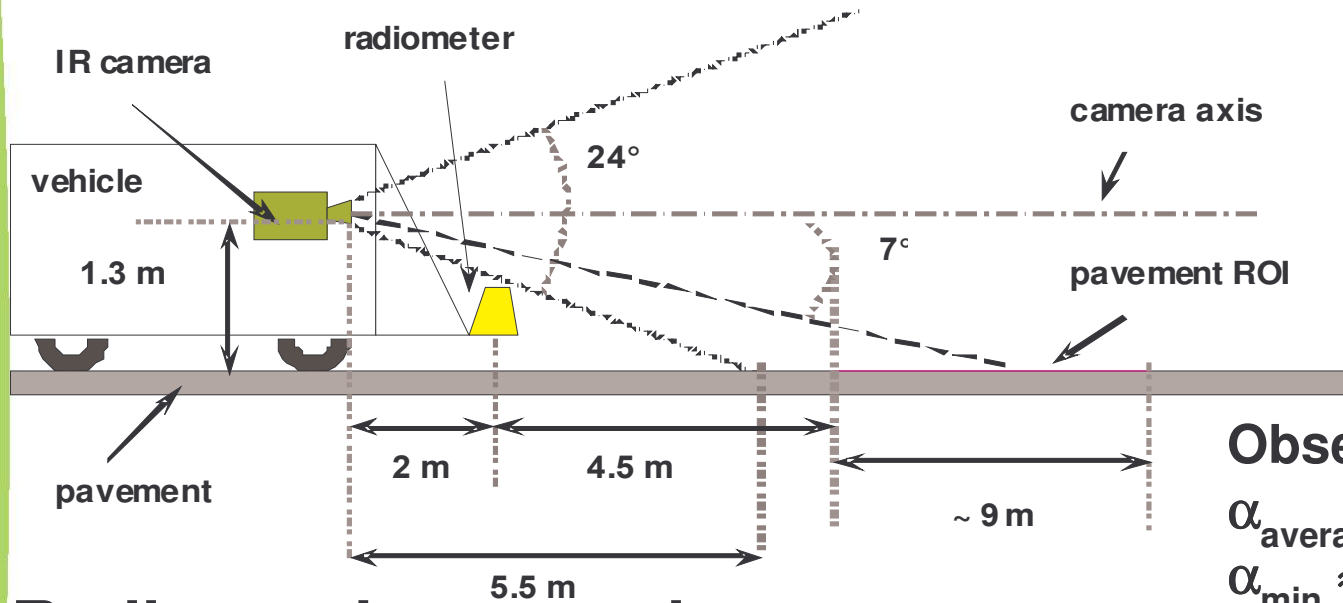
(spring and fall)

- build itinérary thermal fingerprints forecasts (*Principal Component Analysis*).

2- Winter risk index and thermography

Thermal images sequence of the Itinerary

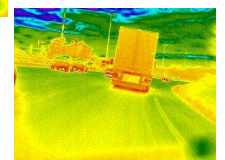
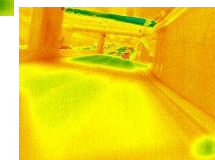
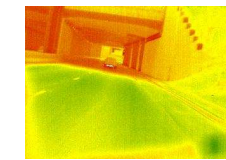




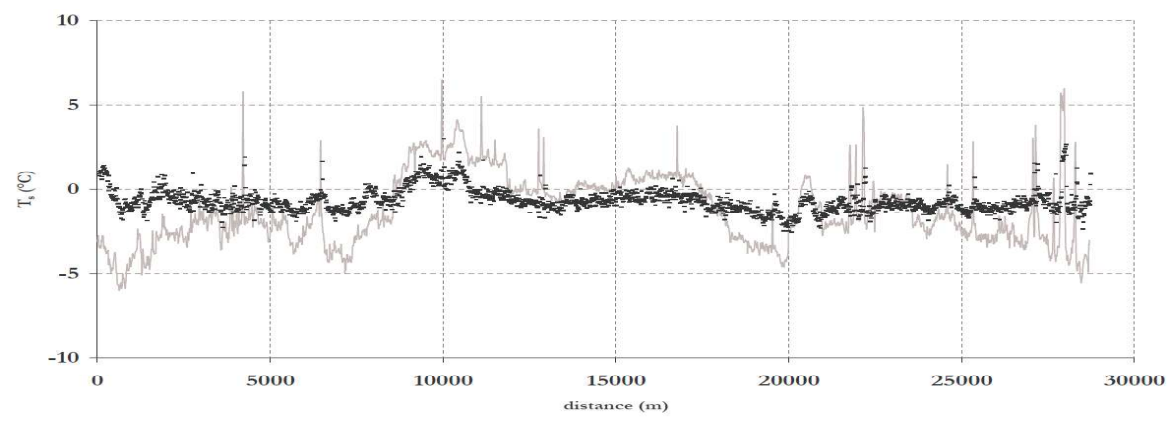
Observation angle:
 $\alpha_{\text{average}} \approx 7^\circ$
 $\alpha_{\text{min}} \approx 4^\circ$
 $\alpha_{\text{max}} \approx 11^\circ$

Radiometric corrections

- low distance and clear atmosphere: $\tau_{\text{atmosphere}} \sim 1$
 - no specularity: Stefan law $L = \sigma T^4$
 - emissivity correction over the pavement ROIs
- ⇒ emissivity distribution over pavement ROI



$$T_{\text{measured}}^4 = \epsilon_{\text{pavement}} \cdot T_{\text{pavement}}^4 + (1 - \epsilon_{\text{pavement}}) \cdot T_{\text{environment}}^4$$



- Ts PRT5 (°C) — Ts FLIR corrigée

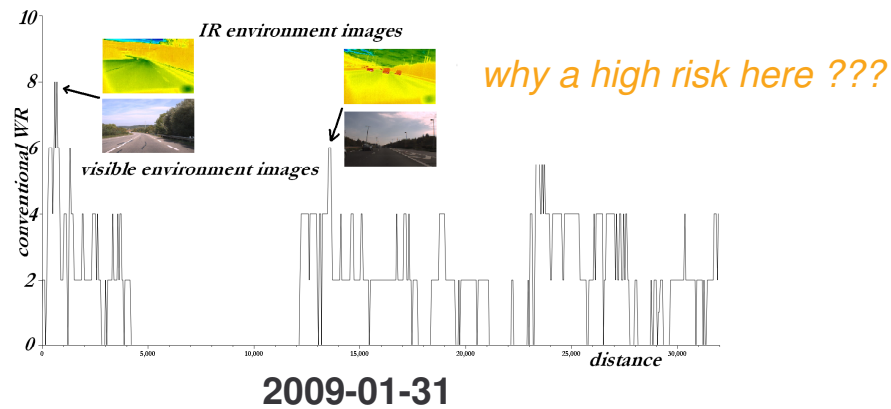
Consistency with seasons and infrastructure

Approach based on a moving average

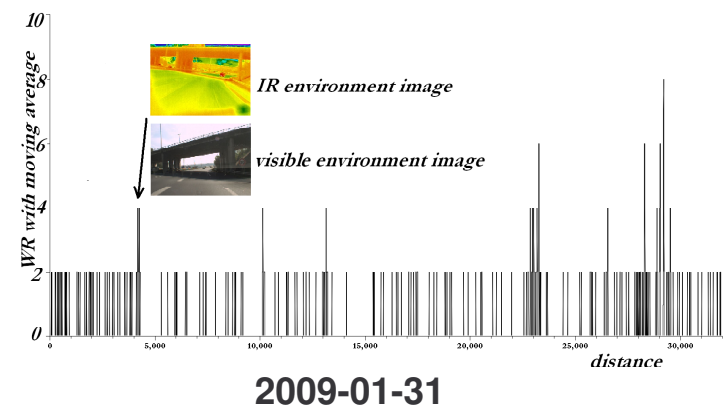
$$WR = 2 \cdot WR(T_s) + WR(T_d),$$

with $WR(T_s) = 0$ if $0^\circ\text{C} \leq T_{s, \text{moving average}} - T_s < 0.5^\circ\text{C}$; 1 if $0.5^\circ\text{C} \leq T_{s, \text{moving average}} - T_s < 1^\circ\text{C}$; ...

and $WR(T_d) = 0$ if $0^\circ\text{C} \leq T_{d, \text{moving average}} - T_d < 0.5^\circ\text{C}$; 1 if $0.5^\circ\text{C} \leq T_{d, \text{moving average}} - T_d < 1^\circ\text{C}$; ...



2009-01-31
former conventional WR as a function of distance (in m)

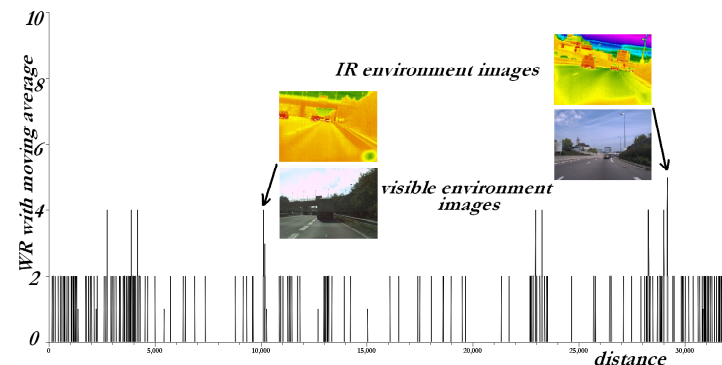


2009-01-31
moving average WR as a function of distance (in m)

greater consistency with the infrastructure ...

... and with seasons

2009-08-19
moving average WR as a function of distance (in m)



3- Winter risk index and PCA

Principal Component Analysis (PCA) = statistical sensitivity analysis method to deal with a large set of data.

Among a set of descriptive techniques (mainly mathematical matrix algebra) without guessing any probabilistic model.

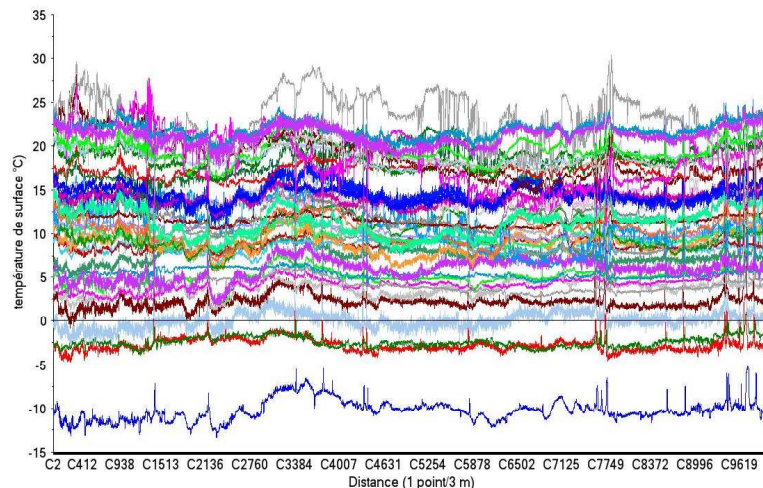
Statistical tool = correlations matrix, or the variance-covariance matrix

Research of the correlations within the dataset.

Related variables considered as a single entity, to produce an overall result taking the relationship among the variable into account.

Linear transformations of a group of correlated variables obtained in such a way that certain optimal conditions are reached.

Most important conditions: transformed variables are uncorrelated.

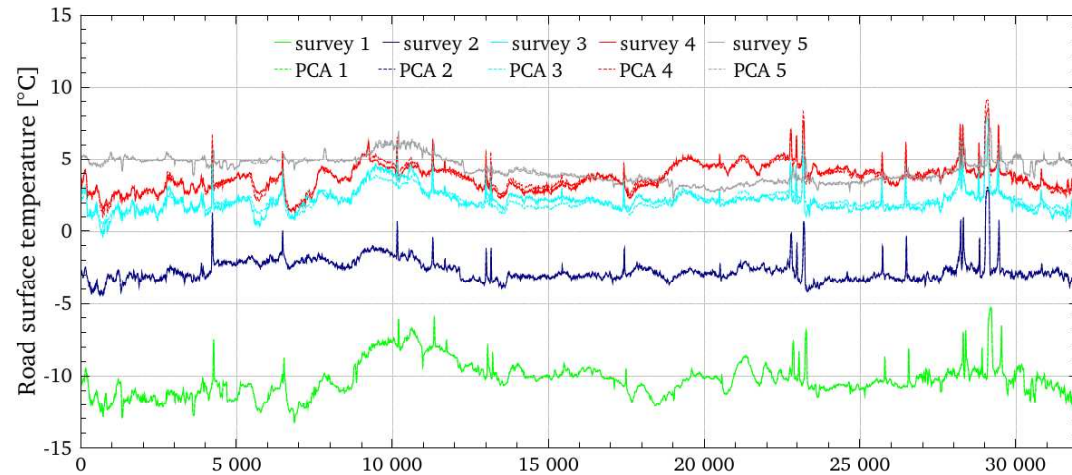


Existence of a generic profile to be adapted
- with seasons,
- with weather conditions

Case study	Case 1 All measurements (53)	Case 2 All measurements under 5°C (8)	Case 3 5 selected measurements under 5°C
Number of principal components (PC) used	10	6	3
Percentage of explained variance (with 1 st PC)	98%	99%	99%
Outliers detected (number of data points)	1000	91	94

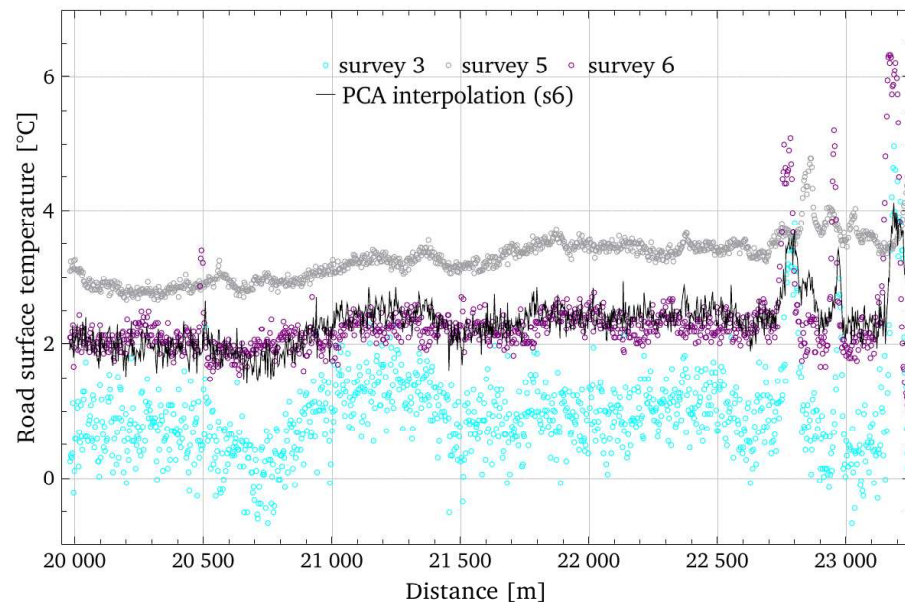
calculations on the whole itinerary

excellent match between field measurements and thermal fingerprints re-calculated with PCA data
(mean matrix + (scores matrix)x/loading matrix)



calculations on specific itinerary section + interpolations between PCA results

Excellent match with field measurements
 Possibility to generate winter risk index based on measurements/forecast of atmospheric parameters



Thanks for your attention

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