

Integrating fault detection methods in monitoring wastewater quality

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Summary

- Introduction
- Objectives
- Materials and methods
- Results
- Discussions
- Conclusions and future work

Introduction

- Monitoring water quality:
 - Application of sensors in WRRFs or other systems (rivers, sewers).



River Dommel in Eindhoven



Station pilEAUte at Université Laval

Introduction

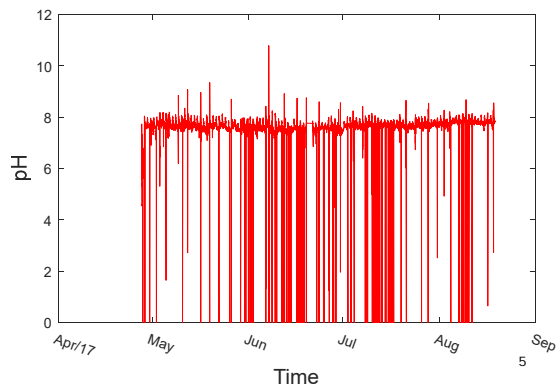
- Monitoring water quality:
 - Application of sensors in WRRFs or other systems (rivers, sewers).
 - Measurement of several parameters.

$\text{NH}_4\text{-N}$ $\text{NO}_3\text{-N}$ COD
 O_2 T°
PH Turbidity $\text{NO}_2\text{-N}$ TSS
Flow Conductivity

Introduction

- Monitoring water quality:
 - Application of sensors in WRRFs or other systems (rivers, sewers).
 - Measurement of several parameters.
- Collection of long time series.
 - Typical loss of data due to problems: 5 – 60 % (Alferes et al., 2013)

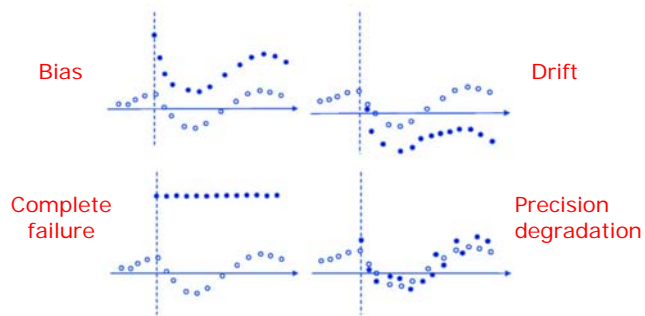
# datapoints	1 923 054
Sampling frequency	2 seconds



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Introduction

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 - Application of sensors in WRRFs or other systems (rivers, sewers).
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- Typical sensors faults:



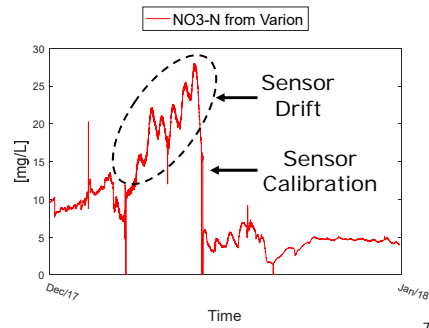
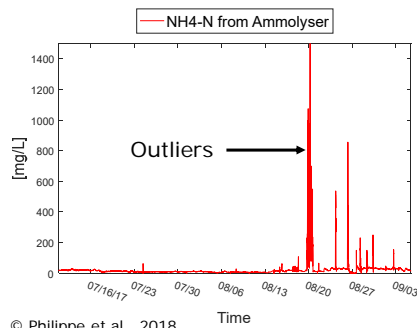
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(Yoo et al., 2008)

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Introduction

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Introduction

- Data management complexity:
 - Many sensors and many variables measured.
 - Collection of long time series.
 - Typical loss of data due to problems: 5 – 60 % (Alferes et al., 2013).
 - Difficulties to identify sensor faults.

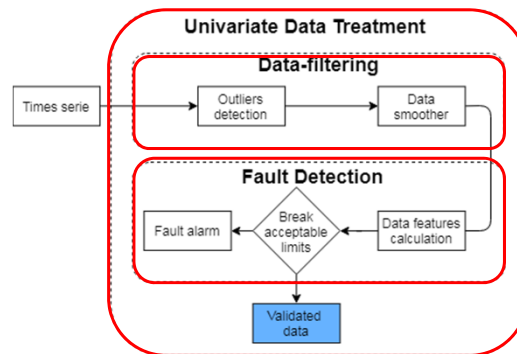
Our solution: A framework to detect outliers and isolate faults to improve data quality.

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Materials and Methods

- General and modular Framework.
- Two main steps: Data-filtering and Fault detection.

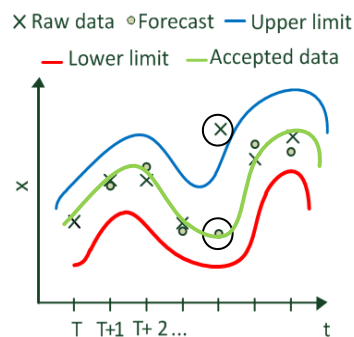
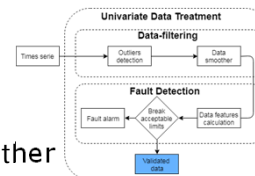


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Materials and Methods

- Data-filtering: Outlier detection and Data smoother
 - 11 parameters (4 - 5 parameters to tune).
 - Outlier detection:
 - For each data point:
 - Estimate the average of the signal
 - Compare the observed value to the average
 - Decide whether or not to keep it
 - Data smoother:
 - Moving Average over n points



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Materials and Methods

- Fault detection:

- Scores

- Sign run-test: Check randomness of error
 $sign(\text{accepted data} - \text{smoothed data})$

- Slope: Change realistic.

$$\frac{dx}{dt} \cong \frac{(x_{i+1} - x_i)}{dt}$$

- Standard deviation: Noise of data.

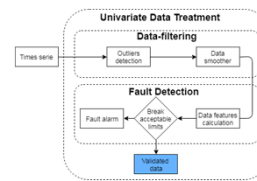
$$std = \text{accepted data} - \text{smoothed data}$$

- Range [Min – Max]: Realistic values.

$$x_i > Max \text{ or } x_i < Min \rightarrow x_i \text{ is a fault}$$

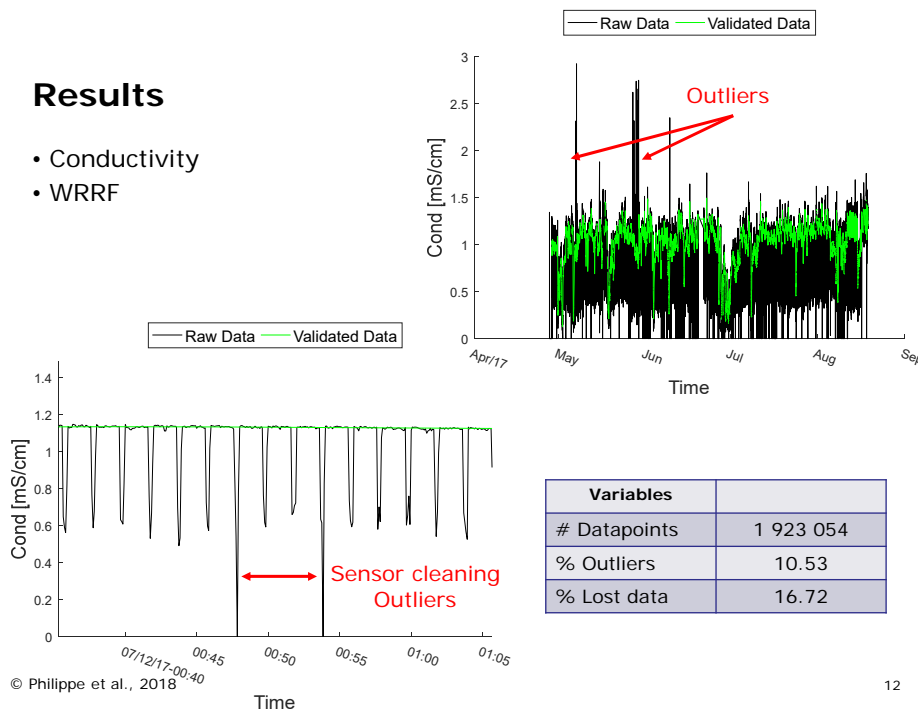
- Data validation

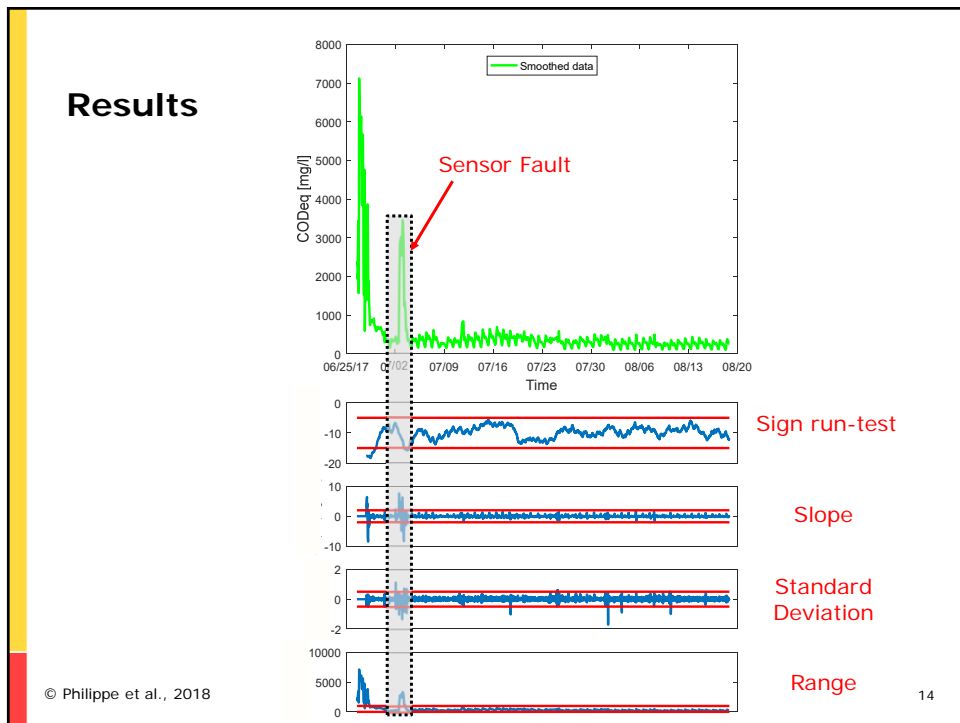
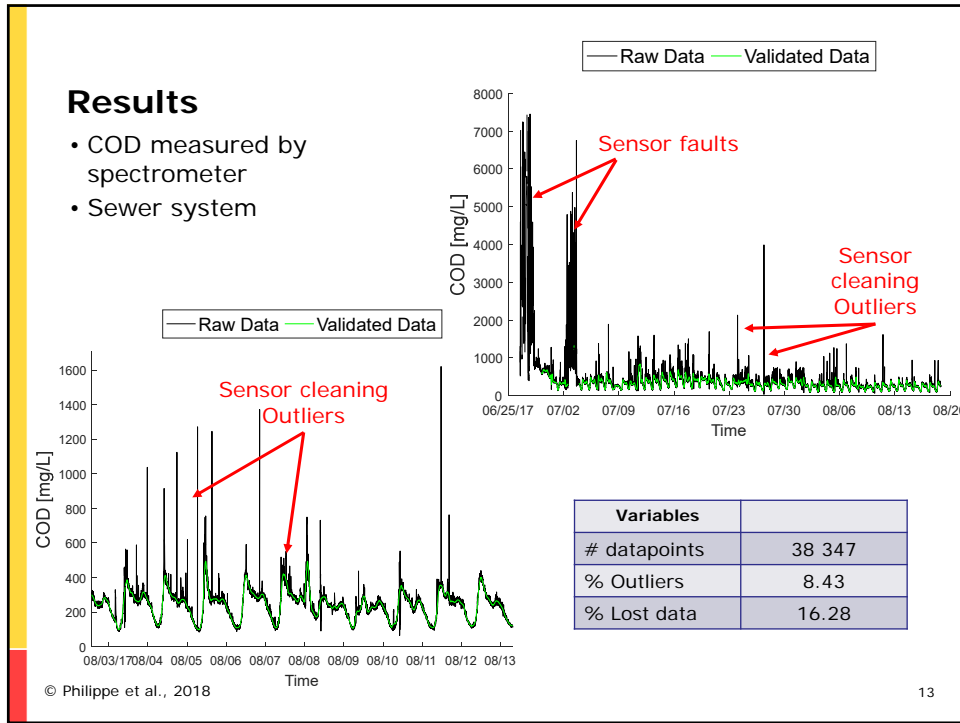
- 8 parameters to tune.



Results

- Conductivity
- WRRF





Discussion

- Method: **Modular** and **automated** data filtration and validation.
 - Outlier detection: EWMA, Neural network, Custom-made method.
 - Data smoother: Moving average, Custom-made method.
 - Fault detection: Slope, range, standard deviation, sign run-test, Custom-made test.
- **Good data** :
 - Improve the maintenance (Sensor cleaning): Pro-active maintenance
 - Indicate when the sensor have to be cleaned

Conclusions and future work

- Method: **Modular** and **automated** data filtration and validation
- Parameters to be tuned:
 - Outliers: 11 parameters (4-5 important parameters).
 - Data validation : 8 parameters.
- **Good data** for next stage:
 - Modelling,
 - Control,
 - Pro-active maintenance,
 - Etc.
- Future work:
 - Multivariate data validation (i.e. PCA),
 - Gap filling in time series.

Thank you for your attention

Questions ?



References

Alferes, J., Tik, S., Copp, J., & Vanrolleghem, P. A. (2013). Advanced monitoring of water systems using in situ measurement stations: data validation and fault detection. *Water Science and Technology*, 68(5), 1022–1030.

Yoo, C. K., Villez, K., Van Hulle, S. W. H., & Vanrolleghem, P. A. (2008). Enhanced process monitoring for wastewater treatment systems. *Environmetrics*, 19(6), 602–617.