



# Use of control to improve nutrient removal

## Perspectives (LCA and fault-tolerant control)

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Neptune workshop: Technical Solutions for Nutrient and Micropollutants Removal in WWTPs  
Université Laval, Québec, March 25-26, 2010



## Overview

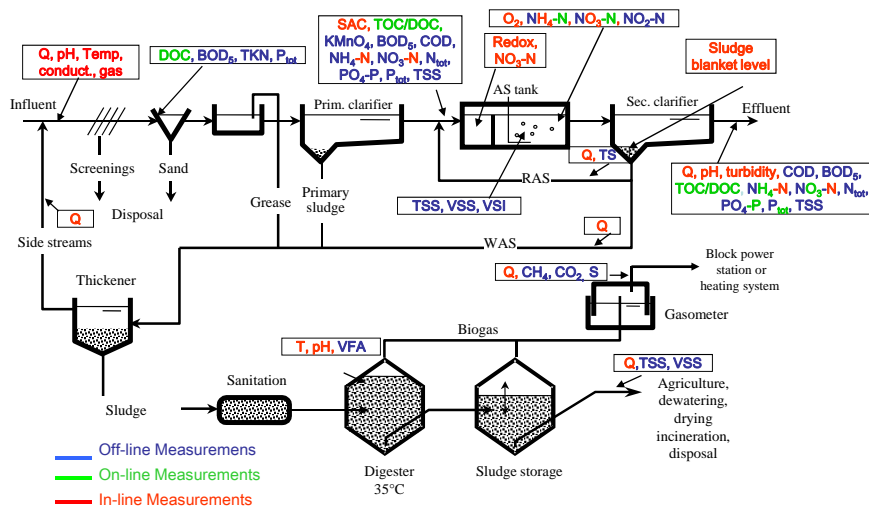
1. Introduction
2. LCA to evaluate control strategies
3. Fault-detection
4. Conclusions

# 1. Introduction

- ✓ Sensors are installed in WWTPs for monitoring and control purposes



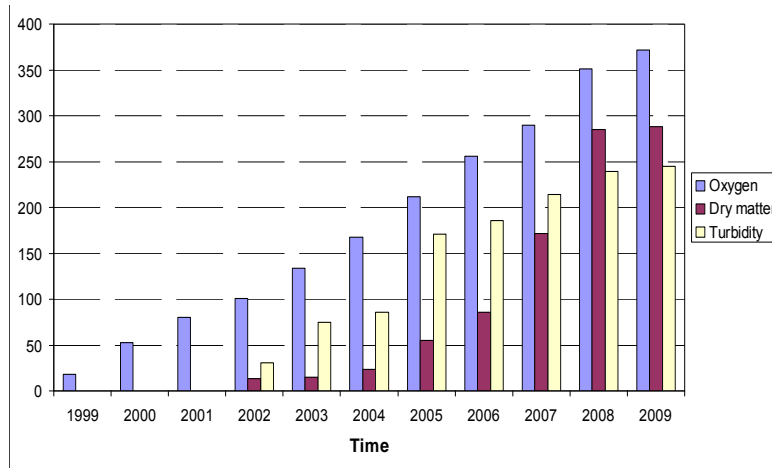
# 1. Introduction



Source: Hansruedi Siegrist

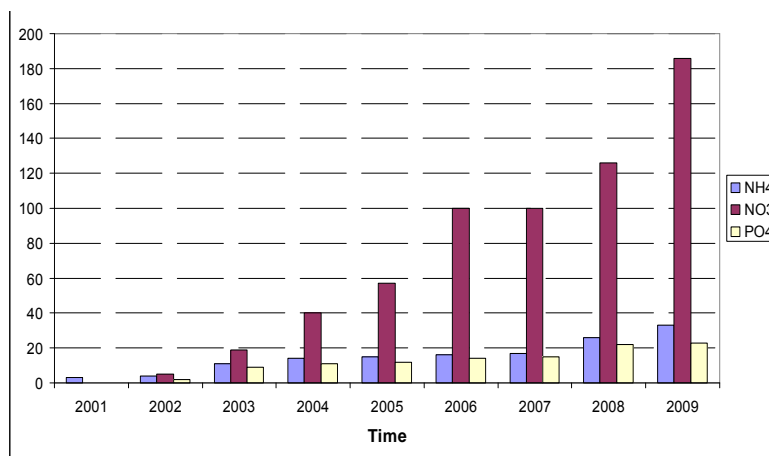
# 1. Introduction

Evolution of number of oxygen, dry matter and turbidity sensors at Aquafin plants (about 220 plants) *Source: Aquafin*



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Evolution of the number of nutrient sensors at Aquafin plants (about 220 plants) *Source: Aquafin*

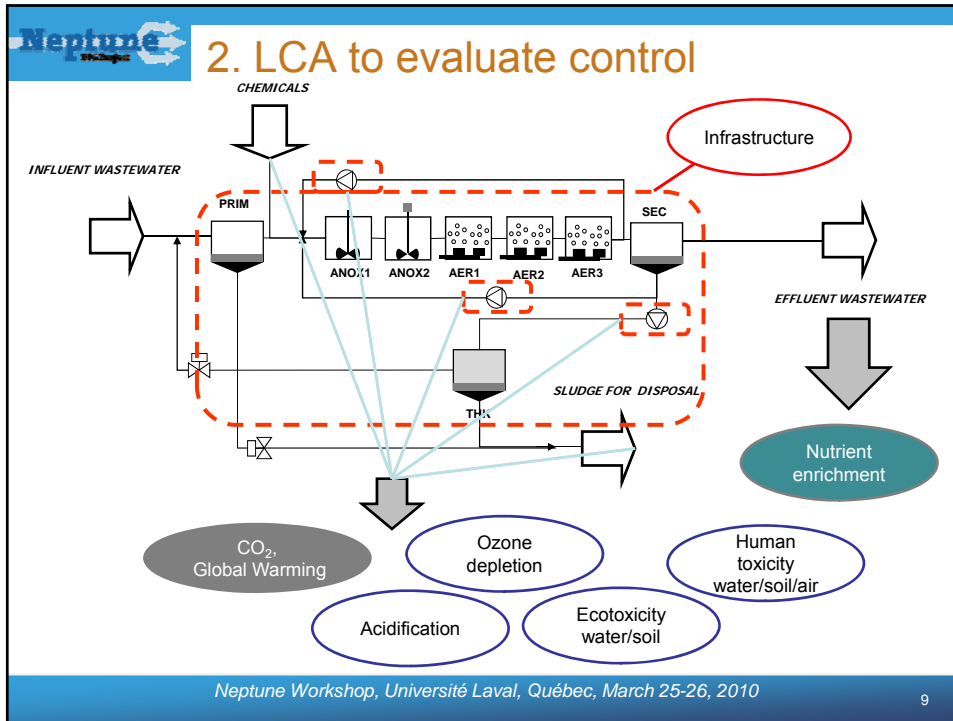


## 1. Introduction

- ✓ Driving force: Water Policies
  - ✓ Sustainable development → need for tools to estimate GHG emissions and perform Life Cycle Analysis (LCA)
  - ✓ Increasing demands on treatment efficiency (new technologies/optimization and control)
    - Draw-back with control: equipment failures (sensors and actuators) can cause severe effluent limit violations → Fault-tolerant control*

## Overview

1. Introduction
- 2. LCA to evaluate control strategies**
3. Fault-detection
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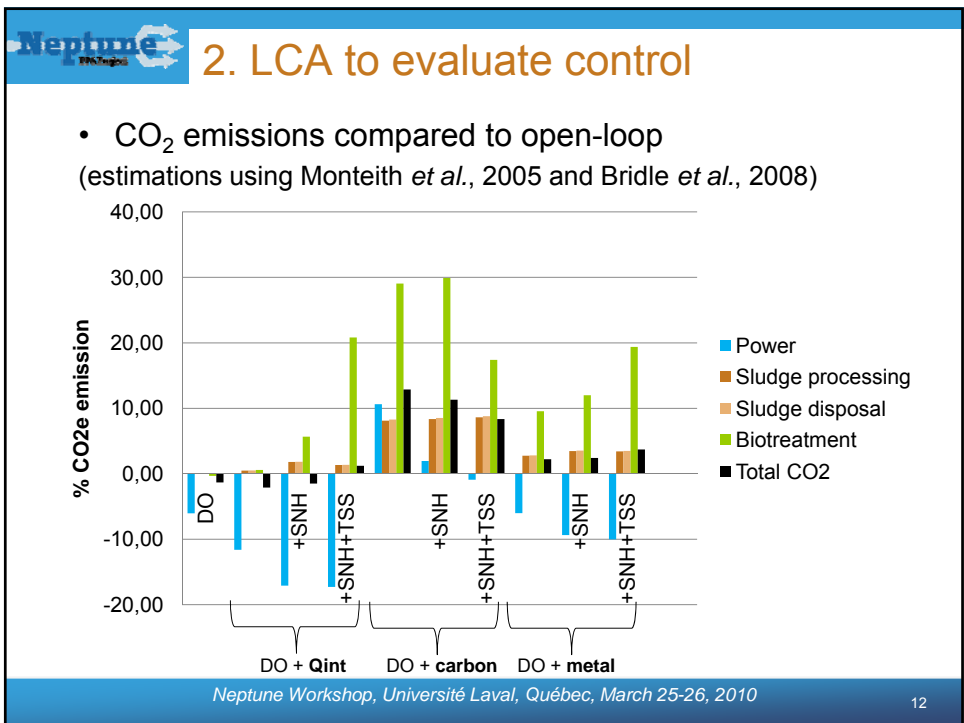
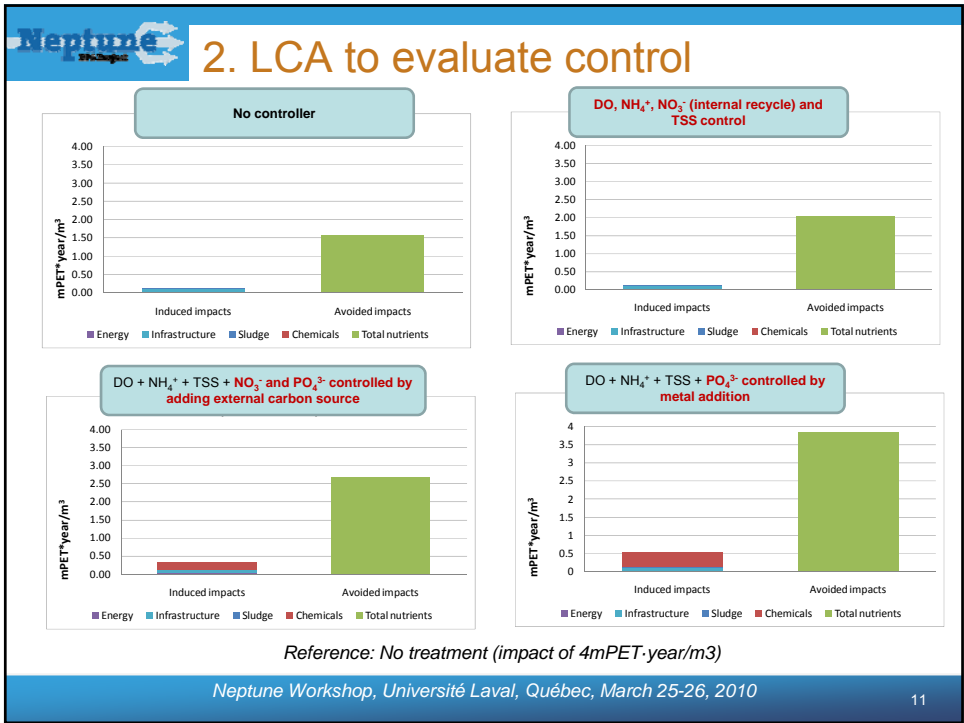
**Neptune** 2. LCA to evaluate control

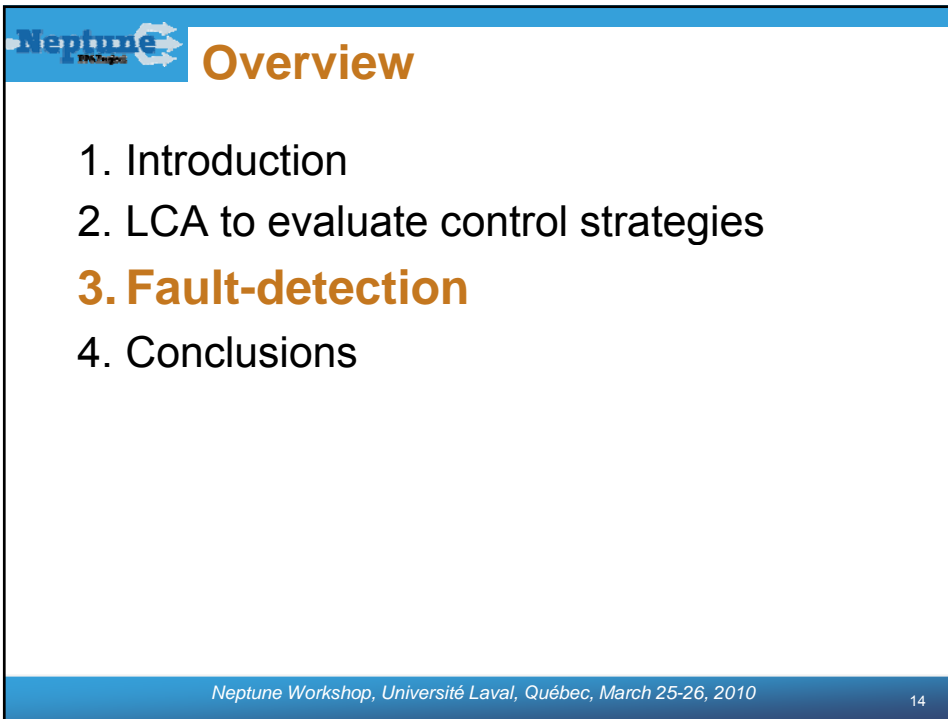
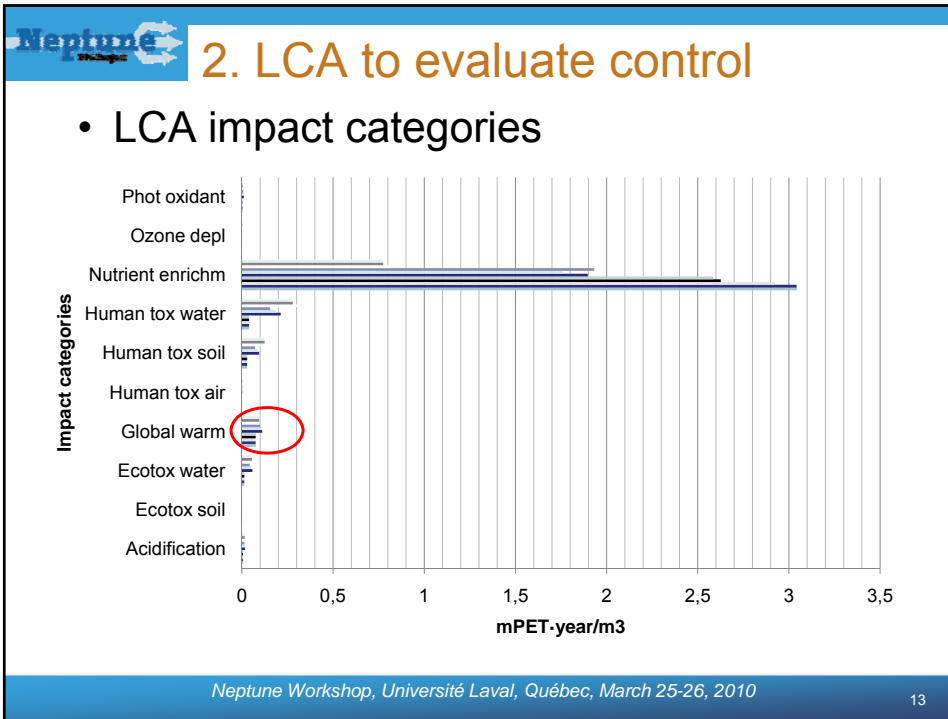
- Variables and impact factors

Variables (var)	Impact factors (mPET*year/unit) WF=1
Nitrogen (kg N)	37.23
Phosphorus (kg P)	269.2
Electricity consumption (kWh)	0.12324
Sludge production (kg sludge, 63% water)	0.1
Infrastructure (m <sup>3</sup> influent treated)	0.127
External carbon source (acetate)	3.8781
Metal (FeCl <sub>3</sub> , 40%)	2.6110

- Functional unit (1 m<sup>3</sup> of treated wastewater)
- Calculation
 
$$NIP_{(var)} = \frac{Value_{(var)} \times IF}{m^3 \text{ treated WW}} \quad [mPET * year / m^3]$$
- Presentation of the results
  - Avoided impact: Influent – effluent nutrient impact
  - Induced impact: Effluent nutrient + Electricity + Sludge + Infr + chemicals

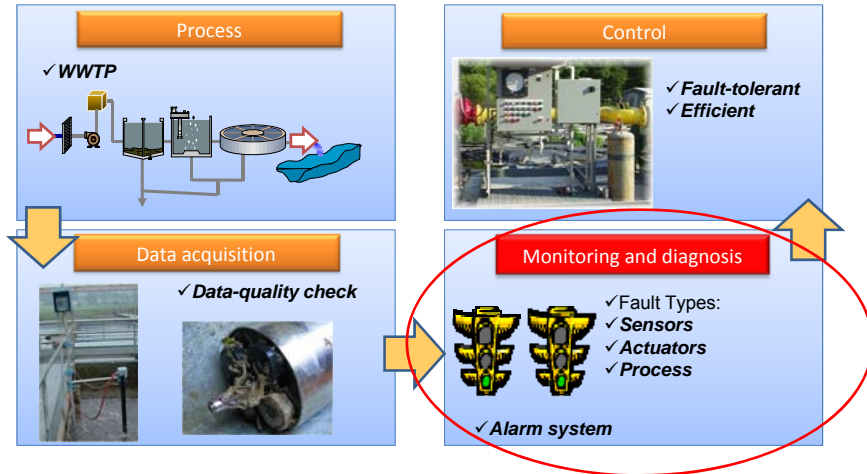
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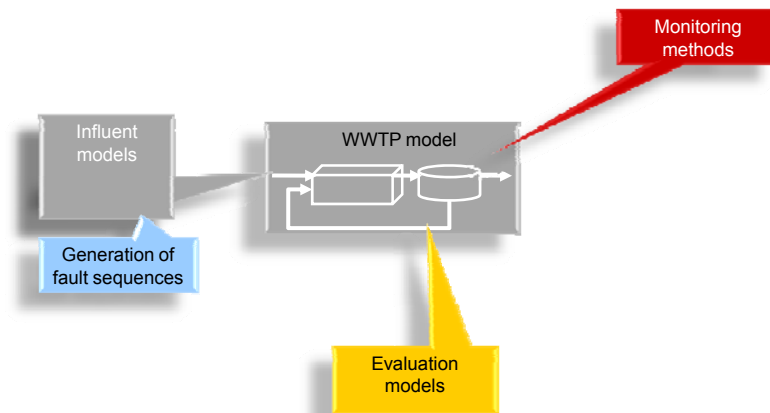
### 3. Fault-detection

- Fault-tolerant control



### 3. Fault-detection

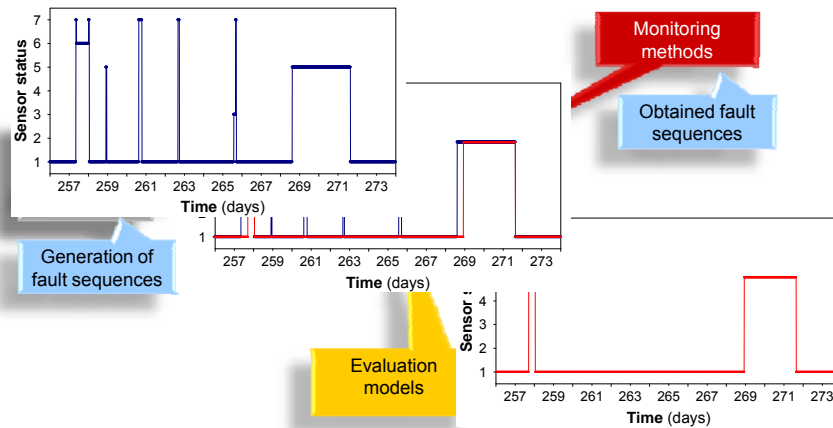
- Goal: Compare performance of fault-detection methods





### 3. Fault-detection

- Goal: Compare performance of fault-detection methods

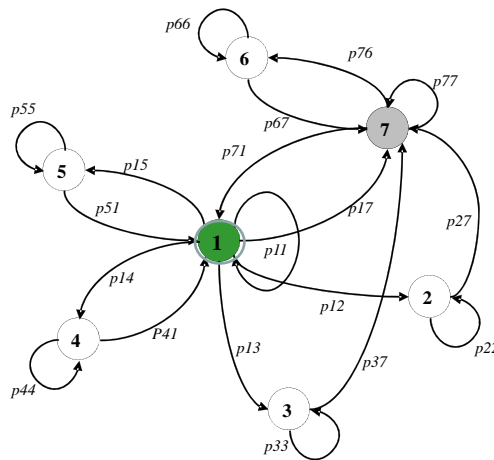



### 3. Fault-detection

- Sequences of faults using Markov Chains

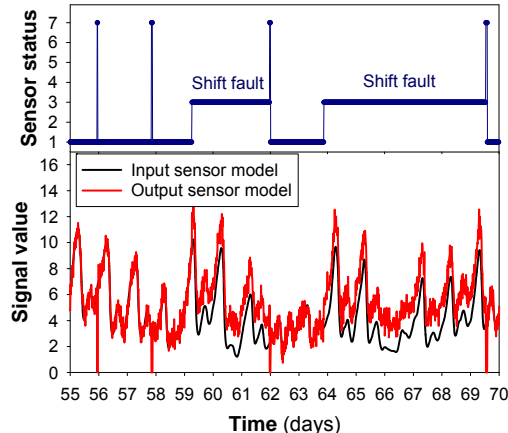
**Sensor status:**

1. Correct functioning
2. Excessive drift
3. Shift
4. Fixed value
5. Complete failure
6. Error gain
7. Calibration




Neptune  3. Fault-detection

- Fault models (phenomenology)

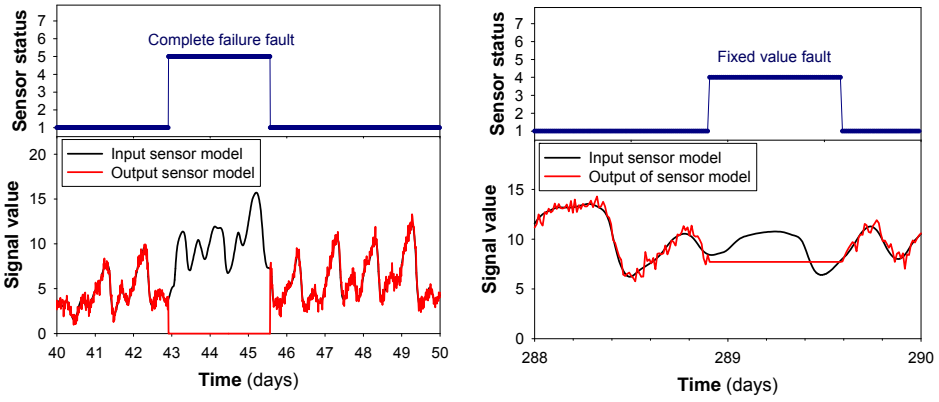


The graph displays two stacked plots. The top plot shows 'Sensor status' on the y-axis (ranging from 1 to 7) against 'Time (days)' on the x-axis (ranging from 55 to 70). A blue step function indicates two 'Shift fault' events, where the sensor status changes from 1 to 3. The bottom plot shows 'Signal value' on the y-axis (ranging from 0 to 16) against 'Time (days)'. It contains two lines: a black line for the 'Input sensor model' and a red line for the 'Output sensor model'. Both lines show a noisy signal that shifts its baseline level during the fault periods.

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Neptune  3. Fault-detection

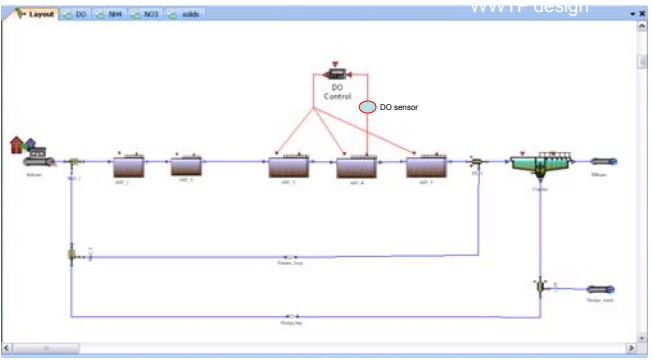
- Fault models (phenomenology)



The left graph shows a 'Complete failure fault'. The top plot has 'Sensor status' (1-7) vs 'Time (days)' (40-50). The status is 1 until day 43, jumps to 5, and returns to 1. The bottom plot shows 'Signal value' (0-20) vs 'Time (days)'. The black 'Input sensor model' line continues to fluctuate, while the red 'Output sensor model' line drops to zero at day 43. The right graph shows a 'Fixed value fault'. The top plot has 'Sensor status' (1-7) vs 'Time (days)' (288-290). The status is 1 until day 289, jumps to 4, and returns to 1. The bottom plot shows 'Signal value' (0-15) vs 'Time (days)'. The black 'Input sensor model' line fluctuates around a mean of 10, while the red 'Output of sensor model' line is fixed at a value of approximately 8 during the fault period.

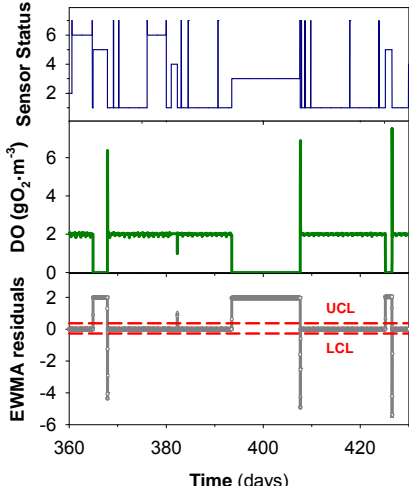
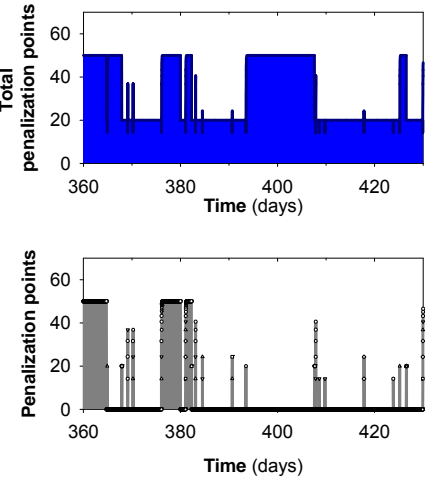
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**Neptune** 3. Fault-detection

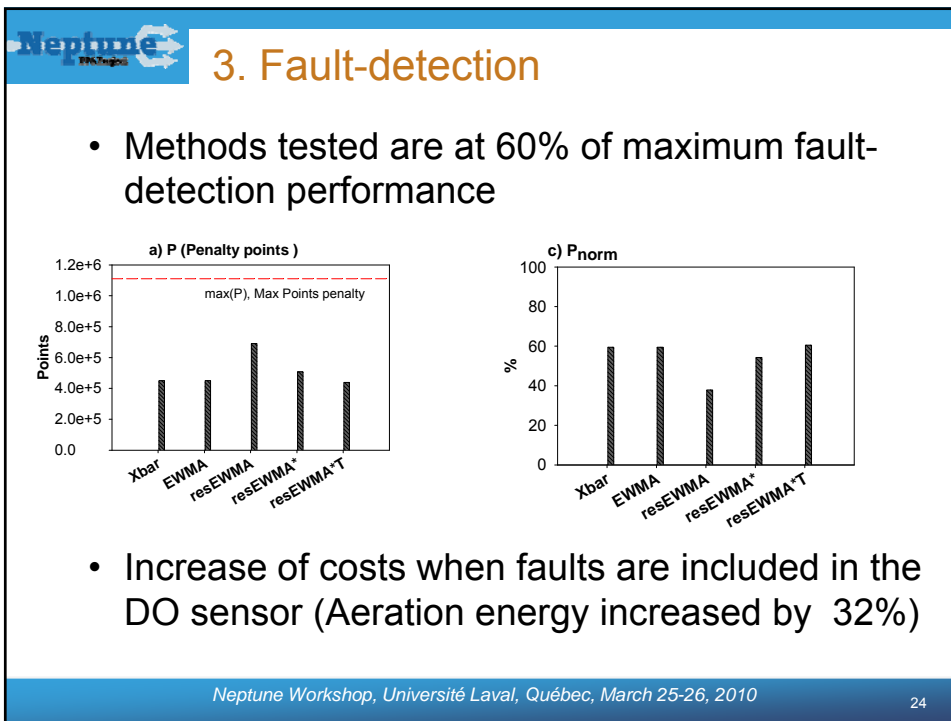
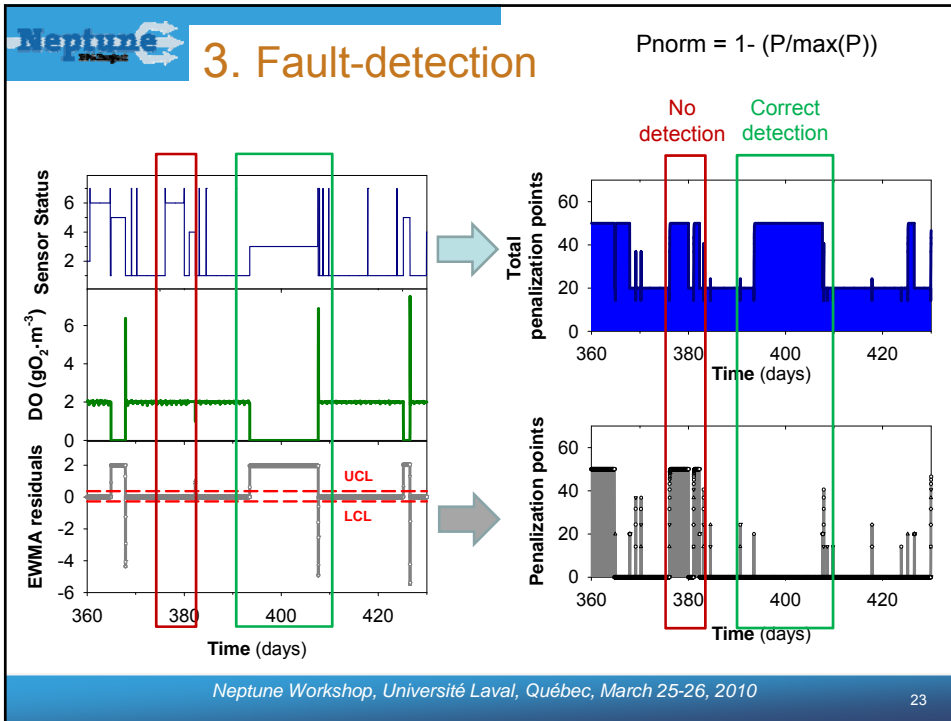
- Case-study:
 
- Methods:
  - Shewhart (DO)
  - EWMA (DO)
  - Residuals on EWMA (DO,  $k_L a$ )

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**Neptune** 3. Fault-detection

$$y = P_{sat} \times (1 - e^{-t/\tau})$$



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- The implementation of control leads to an increase of the avoided impact and a decrease in the induced impact
- The most environmentally friendly strategies are:
  - Metal addition: effluent phosphorous ↓
  - Carbon addition: effluent nitrate ↓ but GHG ↑
- In LCA nutrient removal gets more attention than global warming
- Fault-tolerant control is necessary since equipment failures can cause severe effluent limit violations
- Fault-detection methods tested so far are at 60% of maximum performance. Further research is warranted.



## Acknowledgements

This research is supported by the Canada Research Chair in Water Quality Modeling and a NSERC Special Research Opportunities grant as part of the Canadian contribution to the European Union 6th framework project NEPTUNE. This study was part of the EU Neptune project (Contract No 036845, SUSTDEV-2005-3.II.3.2), which is financially supported by grants obtained from the EU Commission within the Energy, Global Change and Ecosystems Program (FP6-2005-Global-4).



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