



# Monitoring the Performance of Flow Meters Through Advanced Modelling Techniques

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Introduction

Case Study 1 and Case Study 2: Predict the Cause of Drifts in Ultrasonic Flow Meters

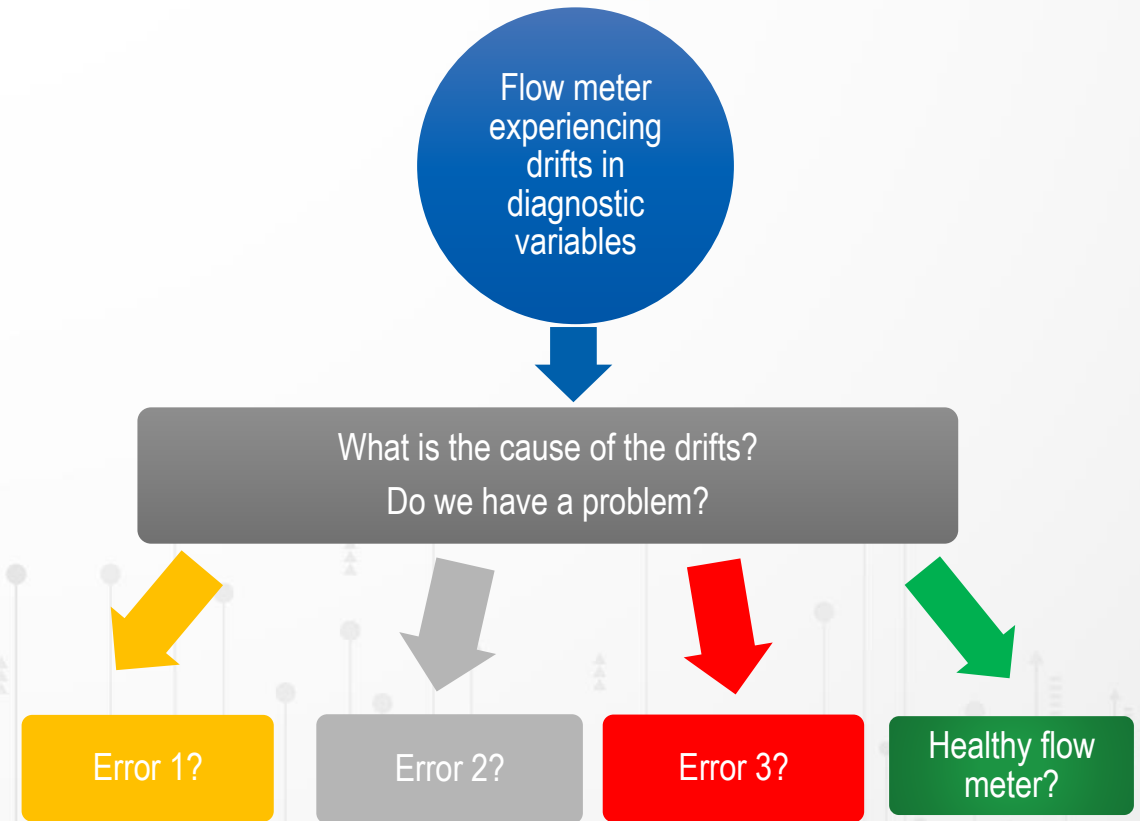
Case Study 3: Anomaly Detection in Ultrasonic Flow Meters

Case Study 4: Predict the Remaining Useful Life of Coriolis and Turbine Meters

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# Introduction

- Potential measurement errors can be detected by drifts in both individual and combinations of digital process values.
- Challenge of being able to distinguish between errors as some errors can incur the same types of drifts in the same variables.
- Without using modelling techniques, it is difficult to fully understand the correlation between different variables.
- Data driven models can be constructed to analyse the relationship between process variables under different conditions.
- Future predictions can be made using those models to aid decision making and improve our efficiency.



# Predict the Cause of Drifts in USMs

## Case Study 1: Installation Error

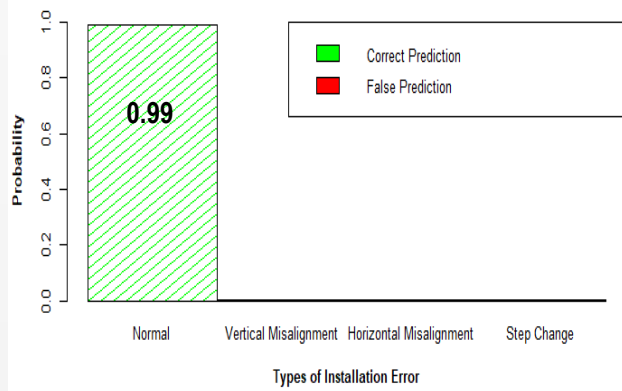
- We consider 4 conditions:
  - Normal Ideal Set up
  - Horizontal Misalignment
  - Vertical Misalignment
  - Step Change
- **Aim:** Use machine learning models to predict the types of installation error based on drifts and patterns seen in diagnostic variables.

## Case Study 2: Wax deposition

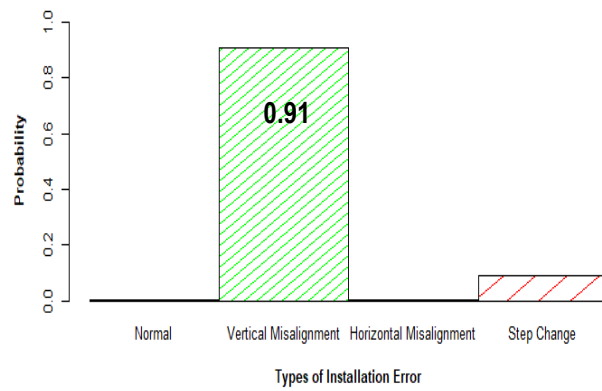
- We consider 4 different conditions:
  - No wax present
  - Wax in 1 downstream port
  - Wax in all downstream ports
  - Wax in all ports
- **Aim:** Use machine learning models to predict the location of wax build up based on drifts and patterns seen in diagnostic variables.

# Prediction Results: Case Studies 1 and 2

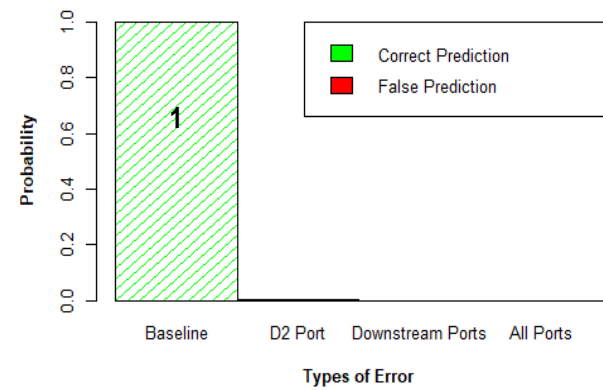
Prediction On Unseen Data from Test 1 Meter A



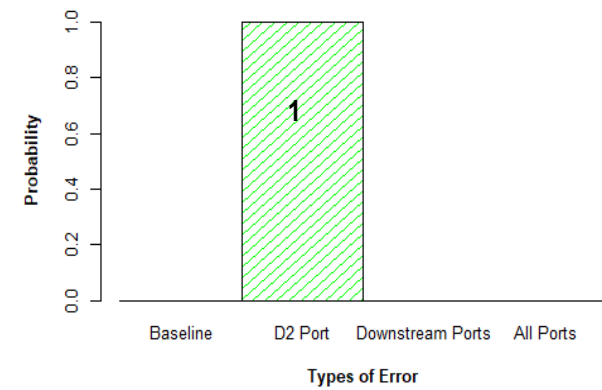
Prediction On Unseen Data from Test 2 Meter A



Prediction On Unseen Wax Data from Test 9 (Baseline)



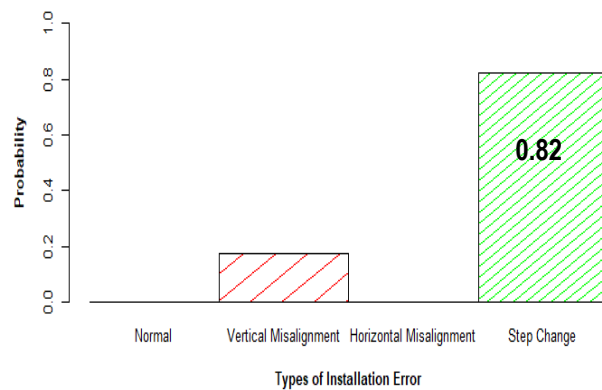
Prediction On Unseen Wax Data from Test 6 (D2 Ports)



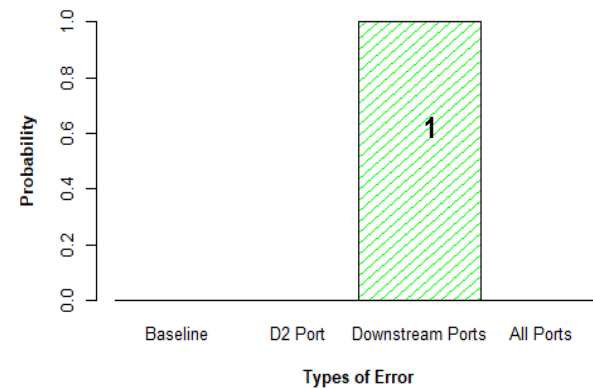
Prediction On Unseen Data from Test 3 Meter A



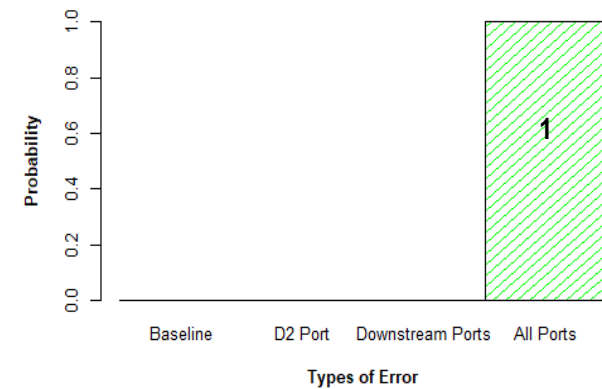
Prediction On Unseen Data from Test 4 Meter A



Prediction On Unseen Wax Data from Test 7 (Downstream Ports)

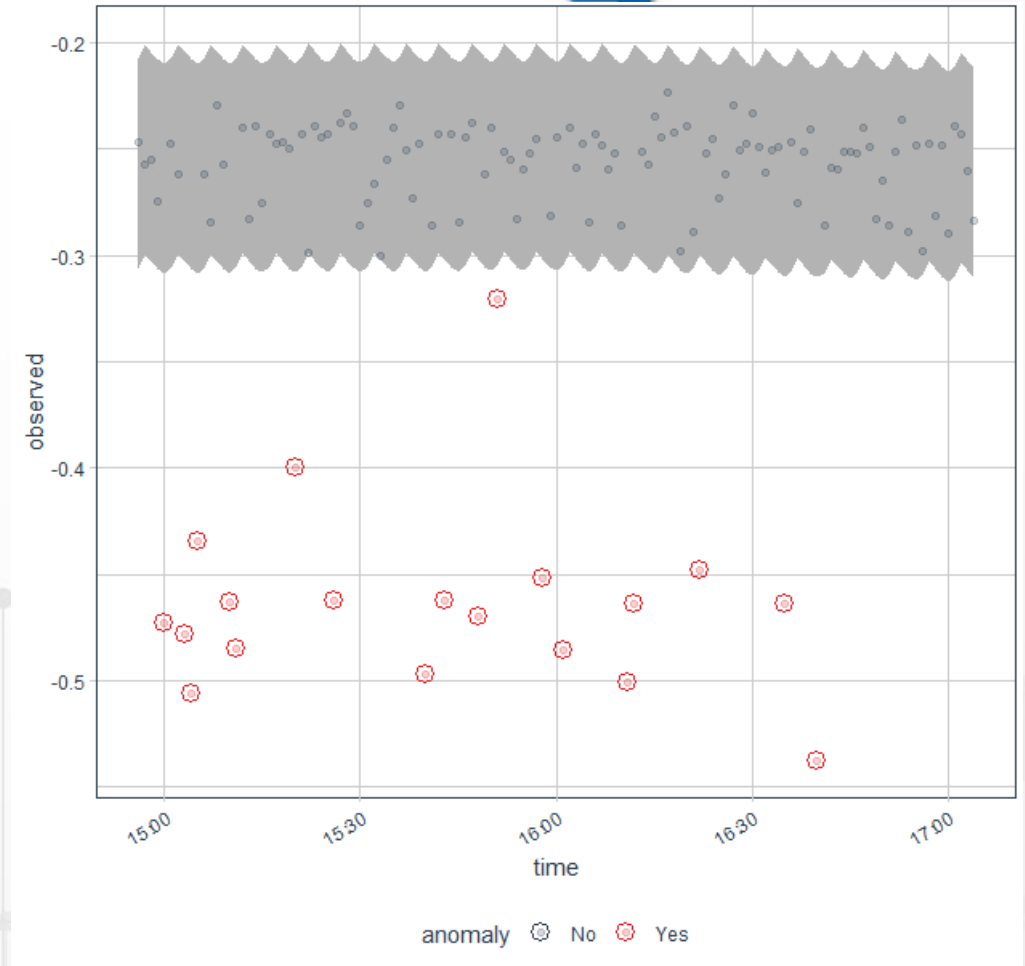


Prediction On Unseen Wax Data from Test 8 (All Ports)



# Case Study 3: Anomaly Detection Model on USM Data

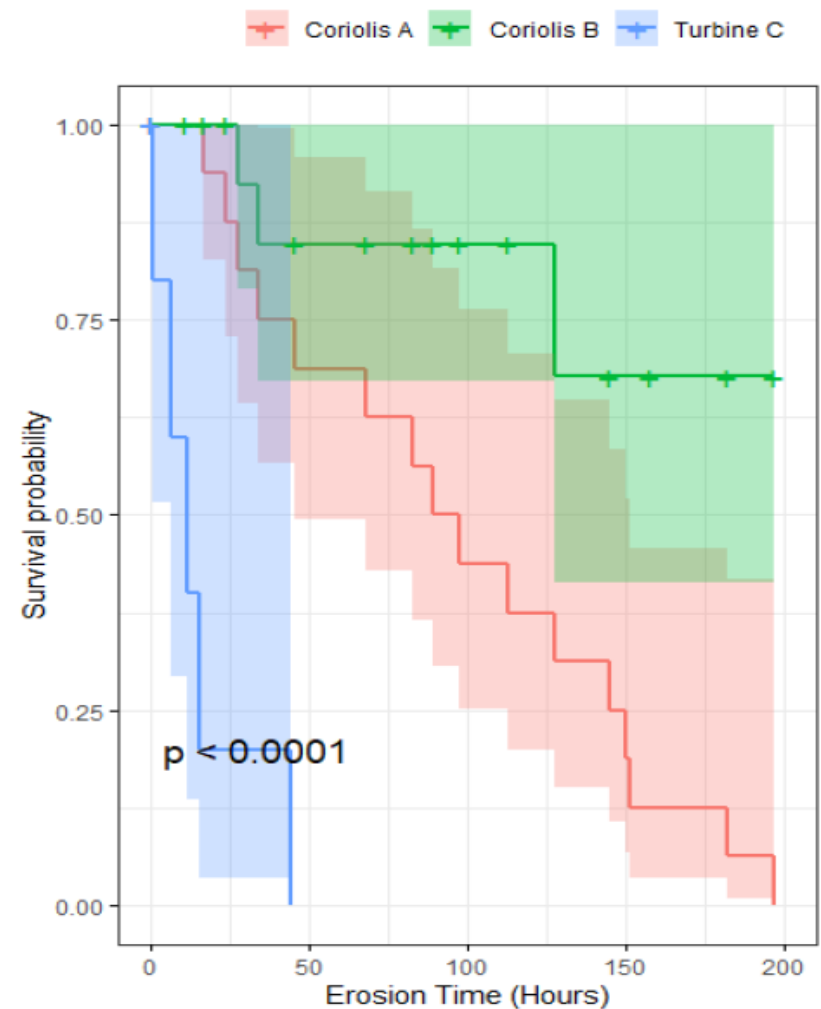
- Examine the trend of data and alert end-users to the presence of **anomalies** which could indicate potential problems in flow meters.
- **Red dots**: anomalies
- **Black dots**: healthy data
- **Grey region**: anomalies threshold



# Case Study 4: Predict the Remaining Useful Life of Flow Meters

- In hydrocarbon recovery, the presence of sand within fluid is inevitable.
- Predict the **failure rates** of flow meters at certain time period through advanced probability analysis when exposed to sand-water with 5% silica sand (180-250 microns) by weight.
- Aid decision-making process by estimating the **remaining useful life** of flow meters.
- **Aim:** Predicting the remaining useful life of Coriolis meters and turbine meter when exposed to sand erosion.

# Case Study 4: Remaining Useful Life of Flow Meters



- Coriolis Meter B was more resistant towards sand erosion.
- Turbine Meter C was the worst affected by sand erosion.
- After 120 hours of sand erosion:
  - Meter A : 63% chance would fail.
  - Meter B: 16% chance would fail.
  - Meter C: Failed



# Conclusion

- Additional insights can be extracted from data using advanced data-driven models
  - ✓ Enable condition-based monitoring and predictive maintenance
  - ✓ Improve fault diagnosis process, reduce unexpected downtime.
  - ✓ Errors can be rectified promptly.
  - ✓ Reduce operating costs.
- 4 case studies were mentioned:
  - **High accuracy prediction** achieved in determining the exact cause of drifts. Overcome ambiguity problem in drifts.
  - **Anomalies detected** in data when end-users might not expect any anomalies.
  - **Remaining useful life of flow meters was predicted** when exposed to erosive flow. Better decision can be made in selecting the most suitable meter and when to replace meter.

***Collecting and storing data are not enough. Valuable insight and opportunities can only be unlocked if appropriate modelling strategies are used to extract the underlying values in data.***



## Contact Us

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