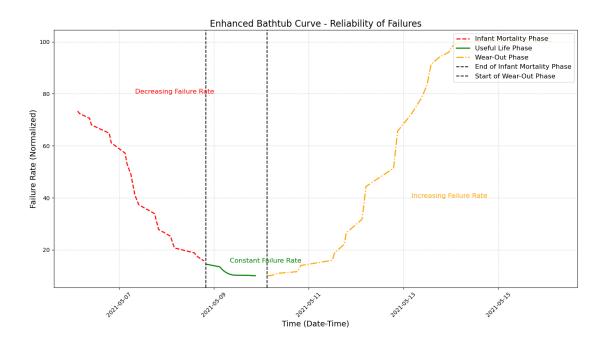
# **PREDICTIVE** MAINTENANCE UNEOSTATS **SOLUTION**

# **OBJECTIVE**

- Develop a **Predictive Maintenance Solution Model** to anticipate system failures and minimize downtime.
- Analyze trends in sensor data to identify early failure indicators.
- Map alarms and source readings for a comprehensive failure prediction strategy.

# LIFECYCLE OF SYSTEM RELIABILITY: THE BATHTUB CURVE



• The Enhanced Bathtub Curve illustrates system reliability and failure rates over its lifecycle:

### Infant Mortality Phase:

- High failure rate initially
- Failures decrease as issues are resolved.

#### Useful Life Phase:

 Stable operation with random, constant failure rates.

#### Wear-Out Phase:

- Failures increase due to aging and component wear.
- Preventive maintenance is critical here.

#### Transition Points:

- End of Infant Mortality Phase.
- Start of Wear-Out Phase.

# MODEL FOR PREDICTIVE MAINTENANCE

- Analysis: Captures trends, seasonality, and noise in sensor readings over time.
- Exogenous Variables (X): Incorporates external factors to improve prediction accuracy.
- **Seasonality Support:** Effectively models recurring patterns in sensor data, crucial for periodic maintenance needs.
- Implementation:
- 1. Data Preprocessing:
  - Cleaned and aggregated data from sensor readings and alarms.
  - o Addressed missing values and ensured data stationarity.

### 2. Model Design:

- Selected parameters based on trend, seasonality, and residual analysis.
- o Included relevant exogenous factors like pressure and temperature etc.
- 3. Training and Evaluation:
  - o Trained the model on historical data for compressors.
  - Validated predictions using actual failure and alarm events.

# TARGET PREDICTIONS

#### **Focus on Failure Prediction:**

• Historical trends and sensor data were analyzed to accurately anticipate potential failures, enabling proactive maintenance.

Next 7 days FailureStartTime forecast:

FailureStartTime	mean	mean_se	mean_ci_lower	mean_ci_upper
2021-08-31 17:00:00	0.0	0.000010	-0.000020	0.000020
2021-09-01 17:00:00	0.0	0.000014	-0.000028	0.000028
2021-09-02 17:00:00	0.0	0.000017	-0.000034	0.000034
2021-09-03 17:00:00	0.0	0.000020	-0.000039	0.000039
2021-09-04 17:00:00	0.0	0.000022	-0.000044	0.000044
2021-09-05 17:00:00	0.0	0.000024	-0.000048	0.000048
2021-09-06 17:00:00	0.0	0.000026	-0.000052	0.000052

### **Approach for Alarms Prediction:**

- Model was utilized to predict alarms and their corresponding indications.
- Alarms serve as early warning signals, helping operators take preventive actions to avoid downtime and reduce risks.

Machineld	ClassName	Name	Timestamp	Confidence
Comp_A	Oil Temperature	Fault Low	2024-10-20 00:00:00	75.51829701
Comp_A	Oil Pressure Alarm	<b>Warning Low</b>	2024-10-20 00:30:00	96.30872277
Comp_A	Oil Pressure Alarm	Warning Low	2024-10-20 01:00:00	13.53501756
Comp_A	Inlet Temperature	Warning Low	2024-10-20 01:30:00	36.82704076
Comp_A	<b>Control Gas Pressure</b>	Fault Low	2024-10-20 02:00:00	67.16496258
Comp_A	Inlet Pressure	<b>Warning Low</b>	2024-10-20 02:30:00	41.875202
Comp_B	<b>Control Gas Pressure</b>	Fault Low	2024-10-20 00:00:00	95.26575289
Comp_B	<b>Control Gas Pressure</b>	Warning Low	2024-10-20 00:30:00	59.13587304
Comp_B	Inlet Pressure	Fault Low	2024-10-20 01:00:00	33.60762302
Comp_B	Inlet Pressure	Fault High	2024-10-20 01:30:00	14.55694853
Comp_B	Inlet Pressure	Fault Low	2024-10-20 02:00:00	21.00937759
Comp_B	<b>Control Gas Pressure</b>	Fault Low	2024-10-20 02:30:00	57.83793756
Comp_C	Oil Pressure Alarm	Warning Low	2024-10-20 00:00:00	94.90801093
Comp_C	Oil Pressure Alarm	Fault Low	2024-10-20 00:30:00	16.74480042
Comp_C	Inlet Pressure	Fault High	2024-10-20 01:00:00	31.31773213
Comp_C	Inlet Pressure	Warning High	2024-10-20 01:30:00	17.58302935

# **WORKFLOW**

## AZURE SQL SERVER

Raw data for compressors and alarms stored in Azure SQL Server.

### DATA PIPELINE

ETL Process: Data is extracted, transformed, and loaded into the Data Lakehouse.

### **DATA LAKEHOUSE**

Centralized data repository enabling scalable and efficient data storage for analysis.

### **ML MODEL**

Performance monitoring and insights for model improvement.

### **NOTEBOOK**

Data analysis, feature engineering, and model building performed here.