



Predicting Failure in Wind Turbines

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Duke Energy Sustainable Solutions**

Commercial Portfolio



- Own more than 4.5 GW of wind & solar projects
- Operate & maintain 600+ MW of 3rd party wind and solar projects
- On track to grow Renewables Fleet 3x by 2030



Duke Energy Renewables



Duke Energy Renewables
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DUKE ENERGY®
BUILDING A SMARTER ENERGY FUTURE®

Wind:

- **23 Sites**
- **1,316 Turbines**
- **3,091 MW**

Solar:

- 61 Sites
- 1,565 MW

Third Party:

- 8 Wind Sites
- 609 MW



Predicting Failure in Wind Turbines



CHALLENGE

Wind turbine component failures are a leading cause of downtime, lost revenue, and expensive repairs across the wind industry



SOLUTION

Detecting signs of impending failure give an operator the capability to inspect and make repairs on their own timeline

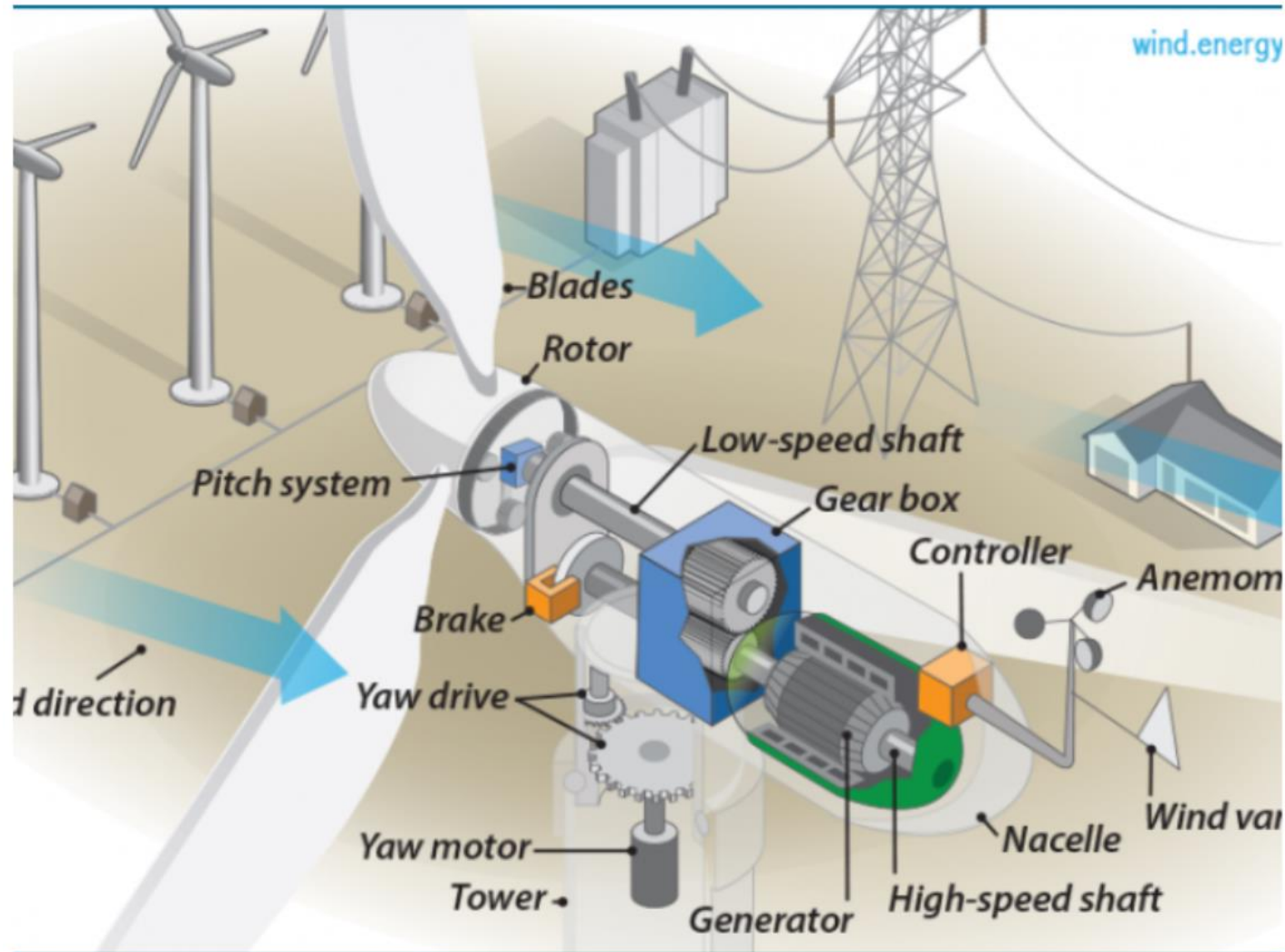


RESULTS

Duke Energy is enhancing visibility across systems and automating alerts, enabling technicians to keep our fleet running better and longer



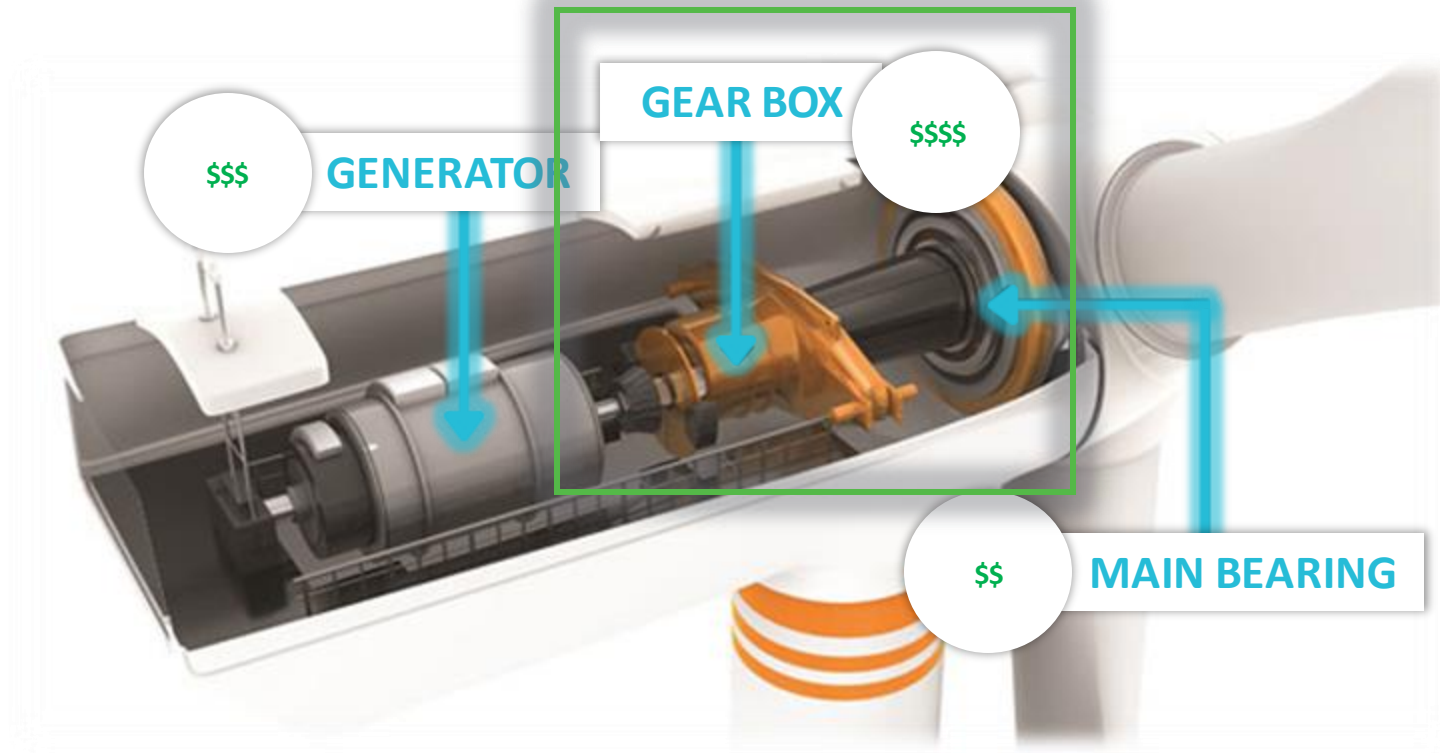
What is the Problem?



Major Components of Wind Turbines

- Blades
- Rotor
- Pitch
- Yaw
- Gearbox
- Generator
- Main Bearing

What is the Problem?



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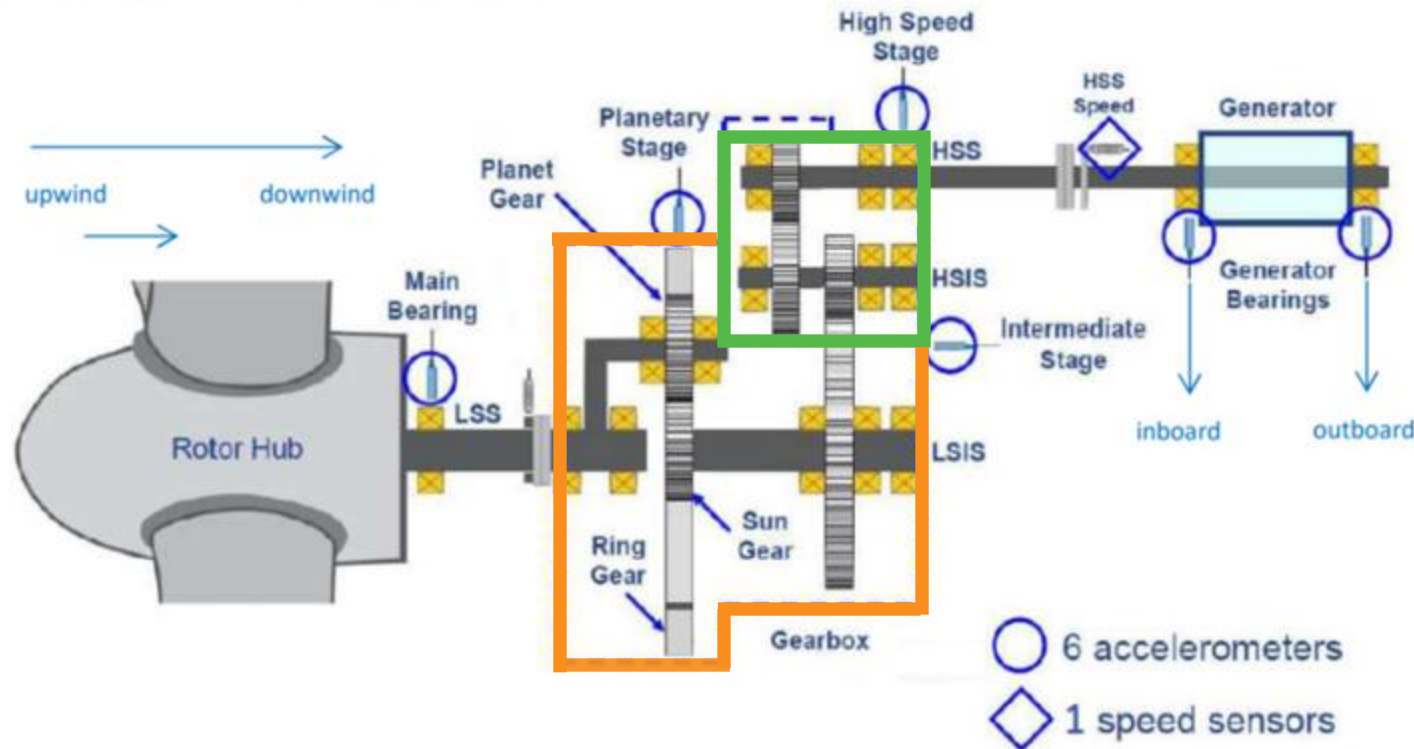
Optimize

- Gearbox
- Generator
- Main Bearing

Predict

What is the Problem? Benefits of Early Detection

Nomenclature & Diagrams:



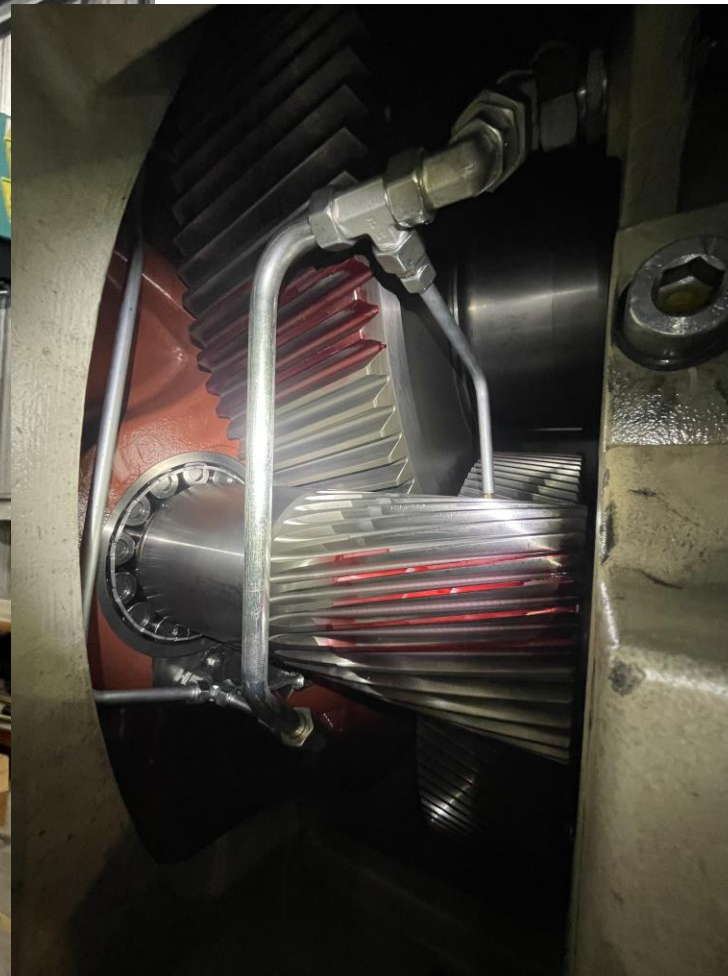
High Speed Stage:

- Uptower replacement
- Avoid full gearbox replacement - \$300-500K
- HSIS depends on technology though 85% is repairable uptower

Low Speed Stage:

- Full replacement
- Minimizing risk of gearbox casing split - \$30k
- Mitigate potential environmental impacts of oil spill

Gearbox and High Speed Shaft



Analytical Approach – Model Determination

KEY SIGNALS

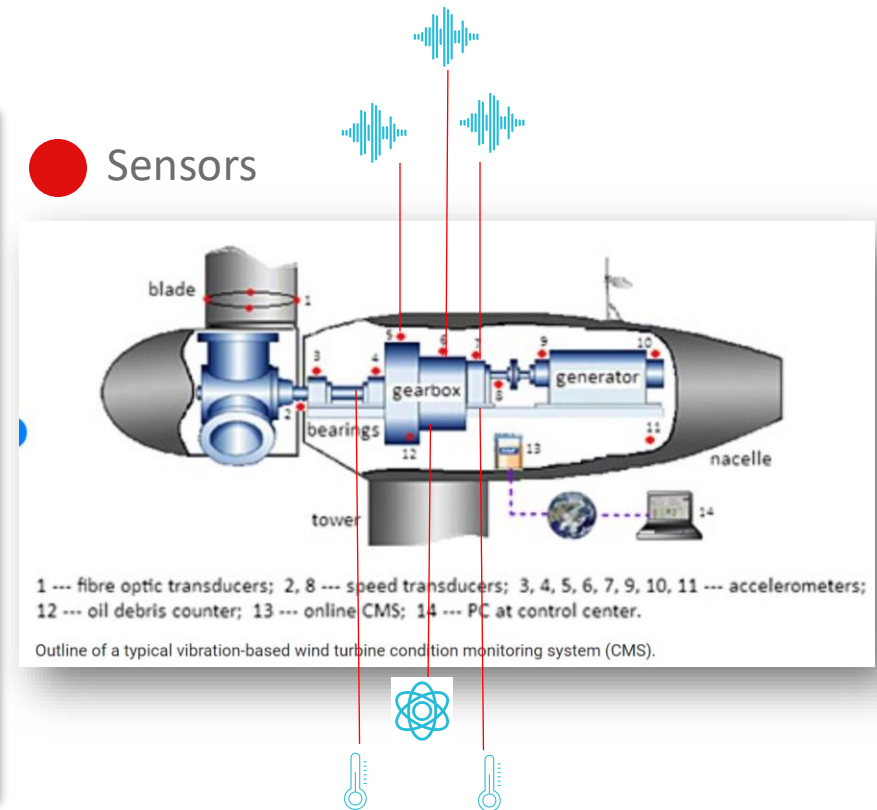
VIBRATION



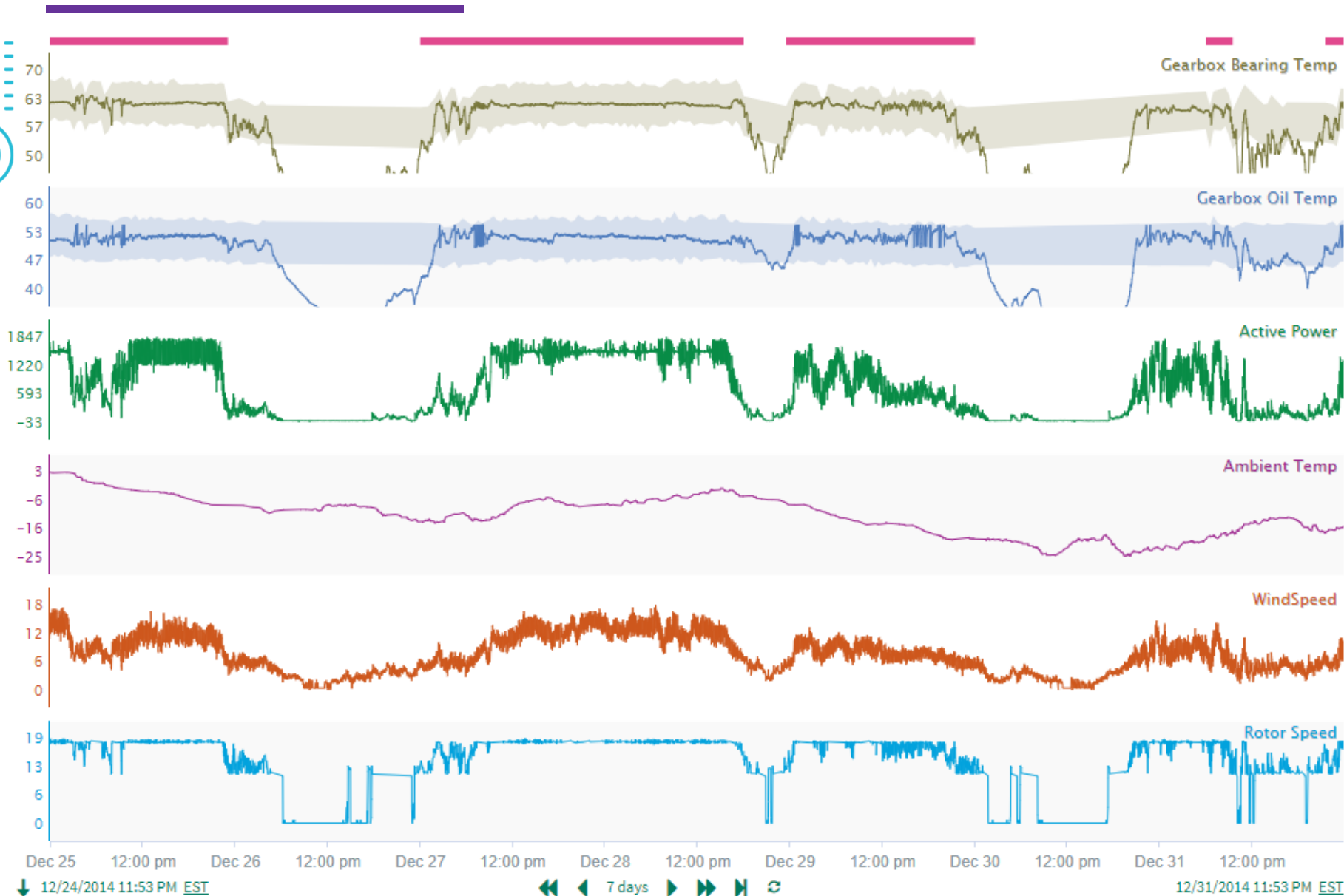
PARTICLE
COUNT



TEMPERATURE



Analytical Approach – Selecting Model Inputs



Training Window

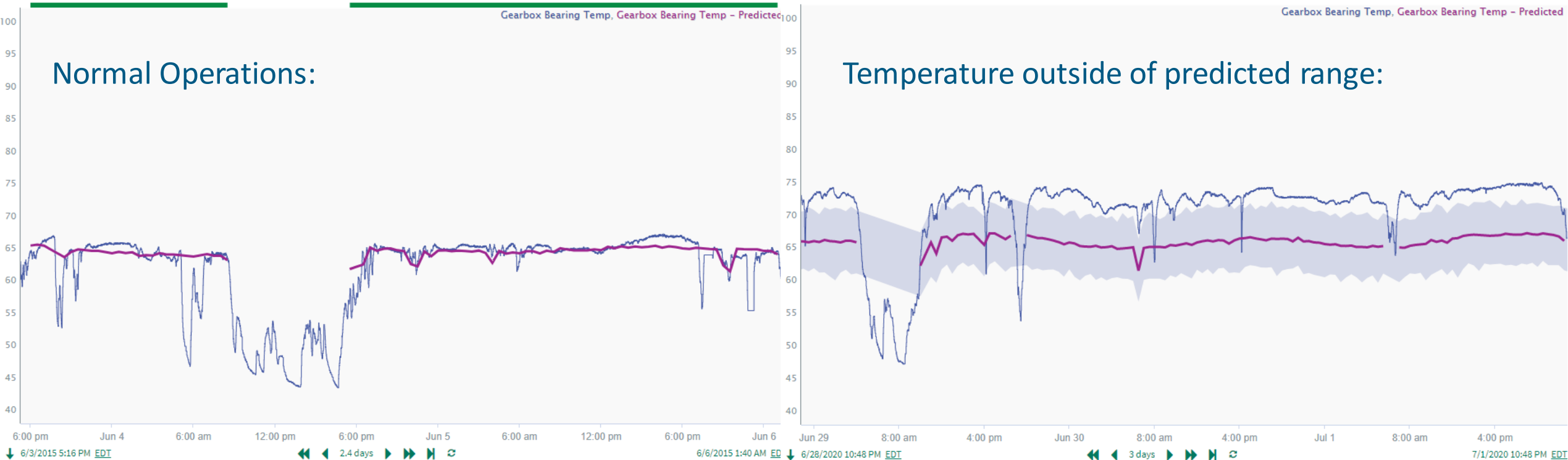
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Targets

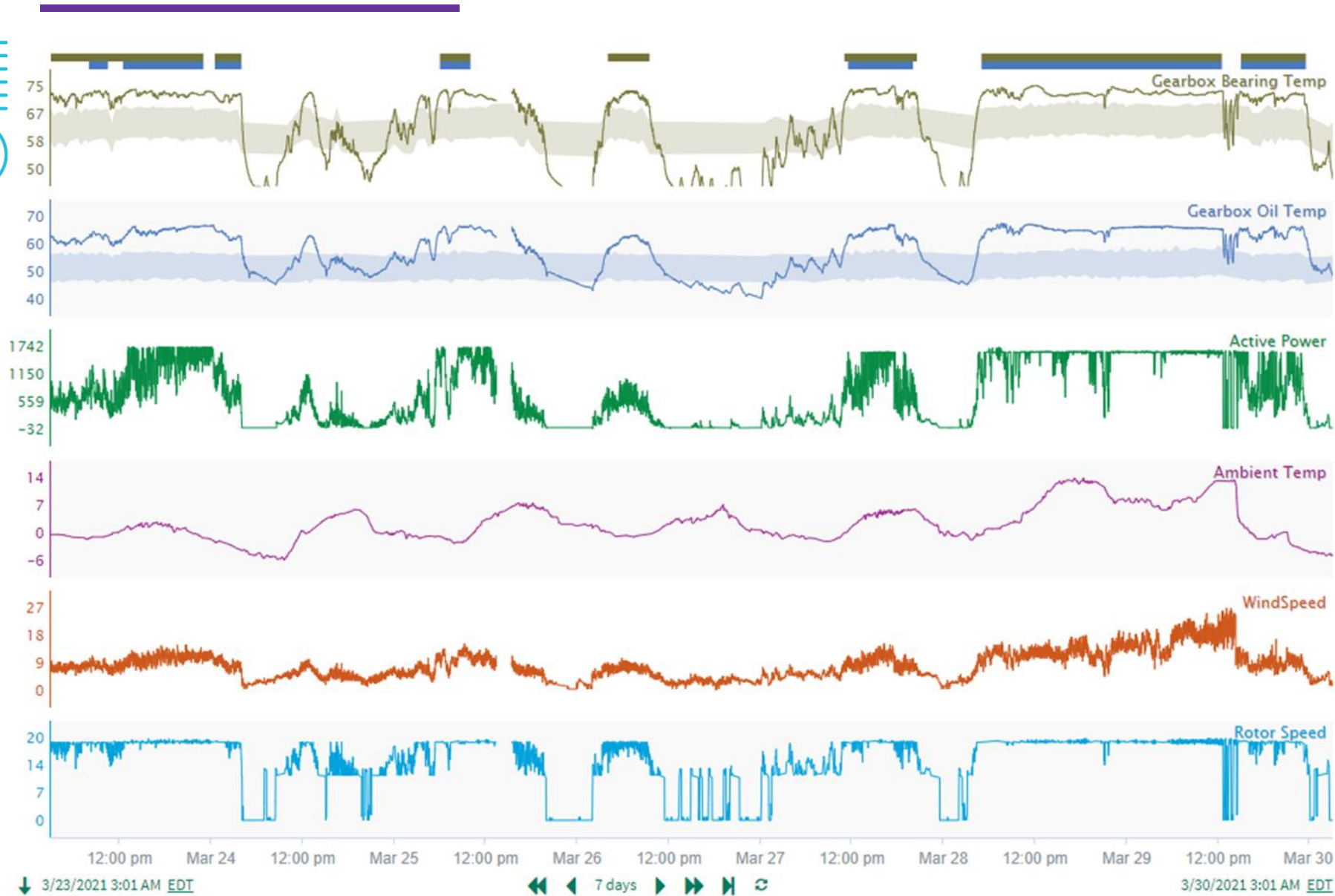
1

Inputs

Analytical Approach – Predicting Bearing Temperature



Analytical Approach – Continuous Anomaly Detection



Anomaly Detected

2

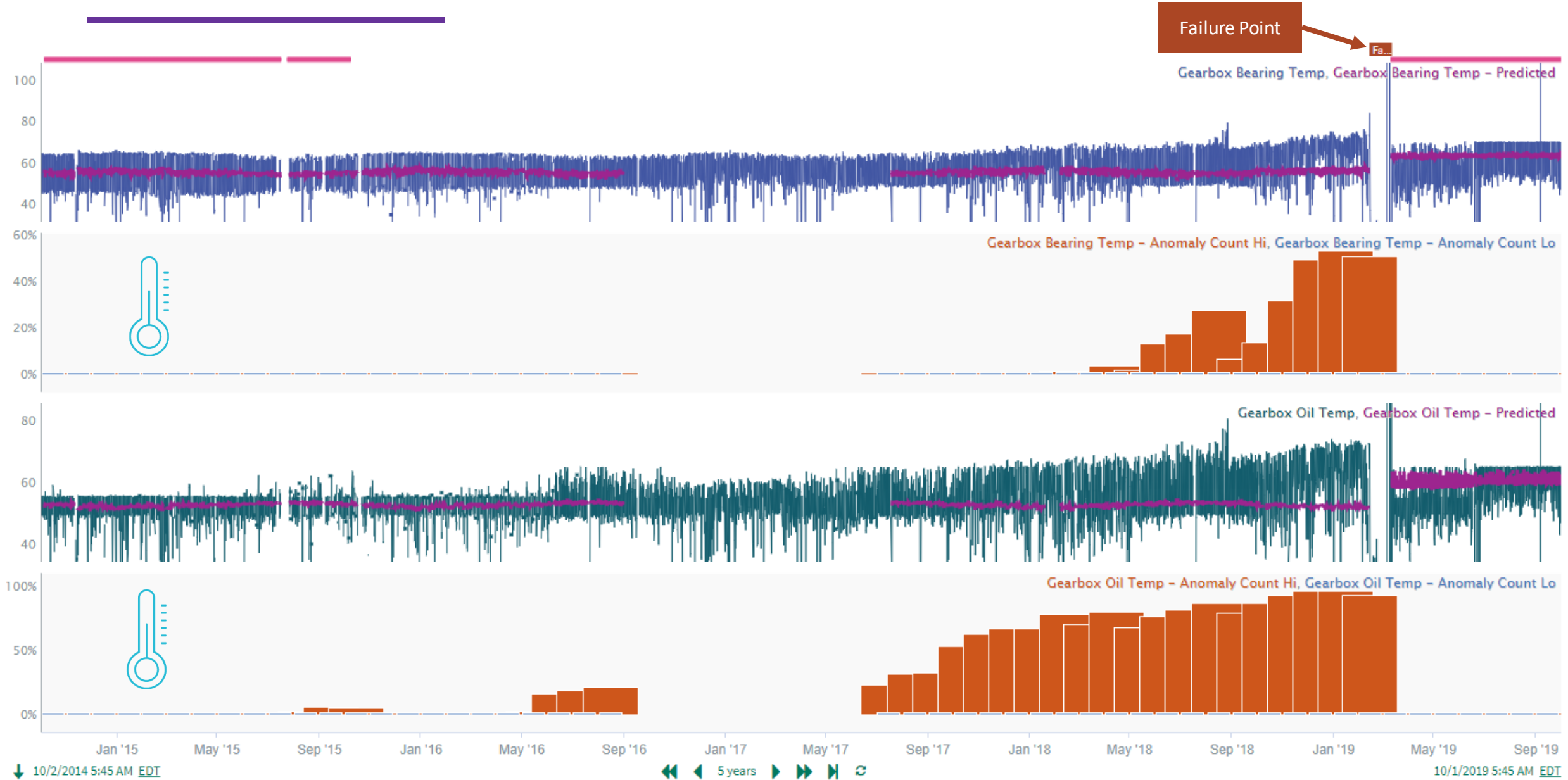
Results

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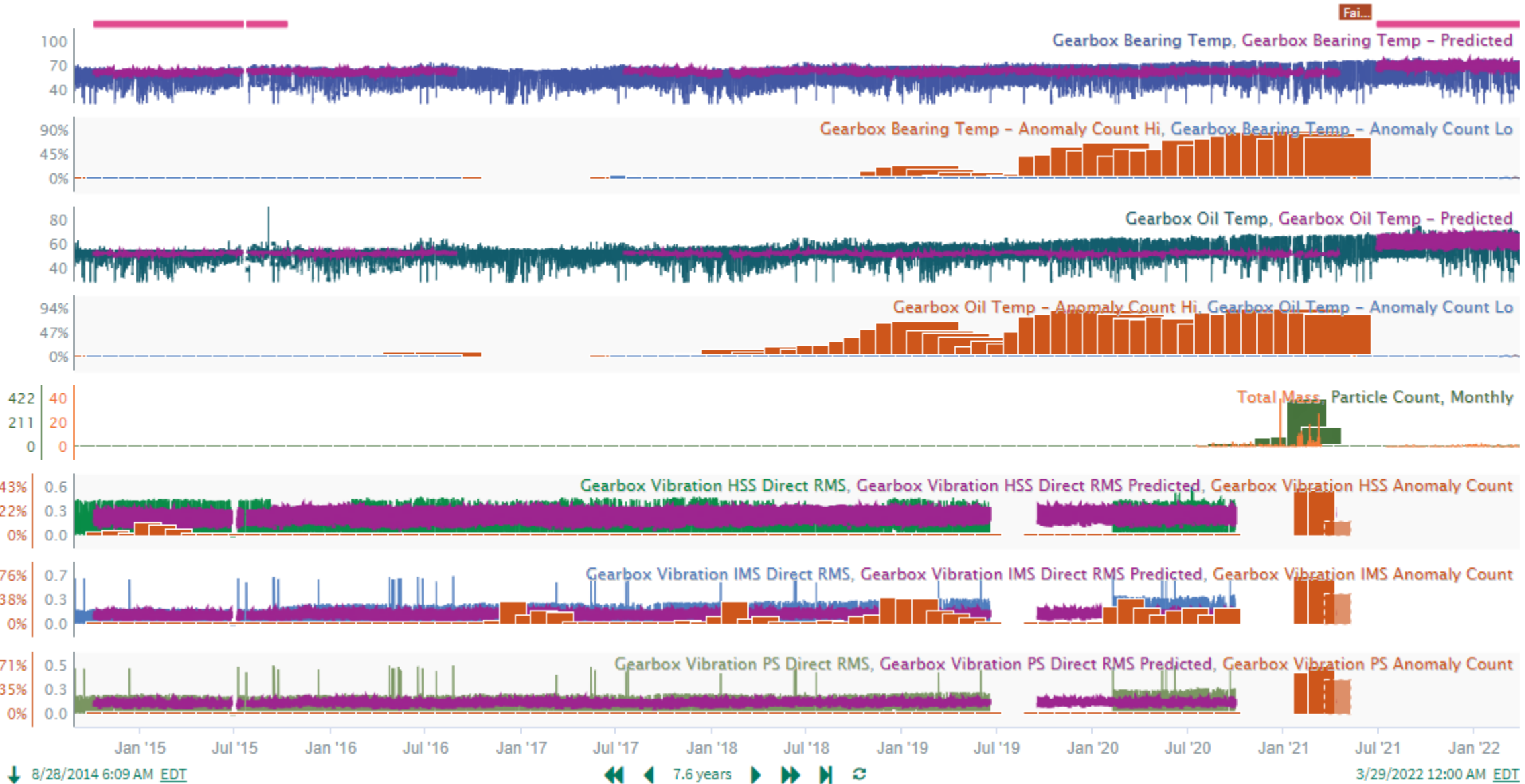
Inputs



Analytical Findings: Increasing Failure Indicators Over Time



Wholistic Health Monitoring: Consolidated View



Operational Results

Turbine	Bearing Temp	Oil Temp	Particle	Vibration HSS	Vibration IMS	Vibration PS	Notes
Turbine A			☉	✓	✓		HSS & IMS Vib warnings preceding failure
Turbine B			☉	✓	✓	✓	Warnings from all vibrations, none from temp, particle not available
Turbine C			☉				No warnings preceding failure, particle not yet active
Turbine D	✓	✓	☉		✓		Temp & IMS Vib warnings preceding failure, particle not available
Turbine E	✓	✓	☉	✓	✓		Temp & HSS/IMS Vib warnings preceding failure
Turbine F			✓				Particle warnings peaked 1y prior to failure
Turbine G	✓	✓		☉	☉	☉	Temp warnings only
Turbine H	✓	✓	✓	☉	✓	☉	Temp, particle, & IMS vib warnings preceding failure
Turbine I	✓	✓	✓	✓	✓	✓	All warnings issued preceding failure
Turbine J	✓	✓	✓	✓	✓	✓	All warnings issued preceding failure

Warning present preceding failure

Data not available preceding failure

Wind Predictive – Working Smarter

Meeting: Work Week Screening



VS



Monthly CBM Report: PDF
through email

- × Static
- × Manual
- × Disparate



- ✓ Aggregated Data
- ✓ One User Interface
- ✓ Accessible Anywhere
- ✓ Real-Time



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