

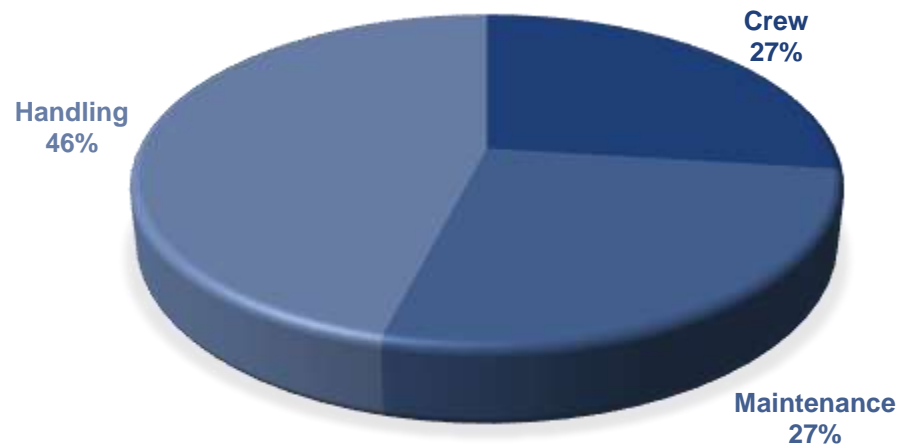


# Domain Adaptation for One-Class Classification: Monitoring the Health of Critical Systems Under Limited Information

Michau Gabriel

# Maintenance of Critical Systems

E.g. costs in Aircraft Operation

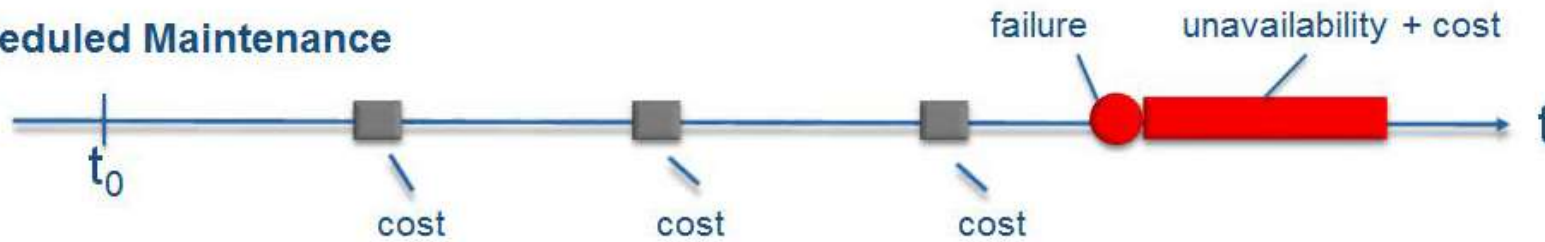


# When to perform Maintenance Operations?

## Corrective Maintenance



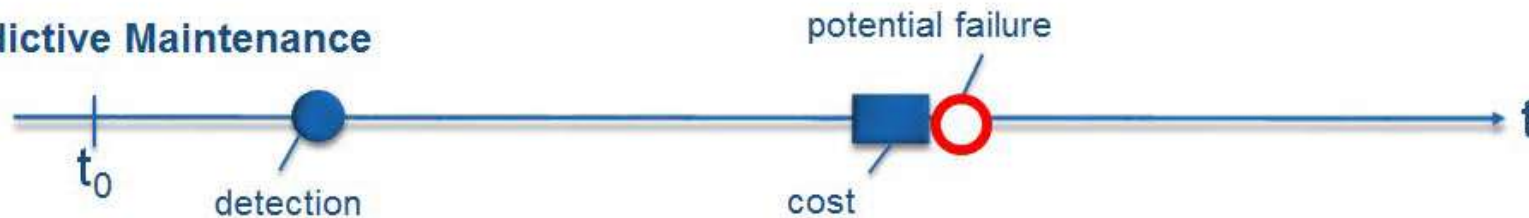
## Scheduled Maintenance



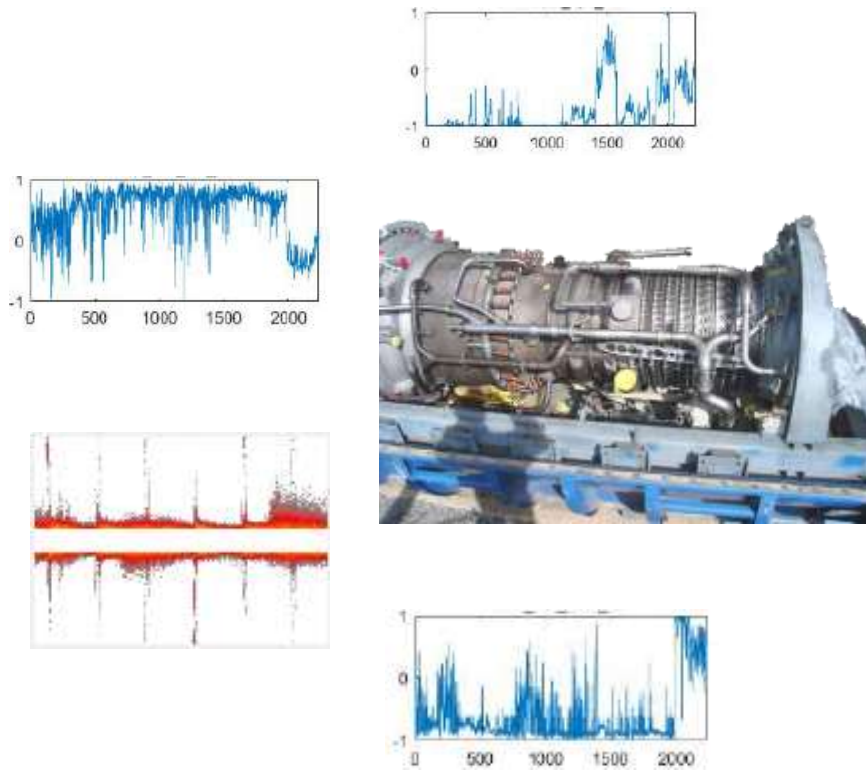
## Condition-Based Maintenance



## Predictive Maintenance



# Challenges in the Monitoring of Complex Industrial system



Large number and large variety of sensors  
→ Heterogeneous condition monitoring data

Rare Faults:

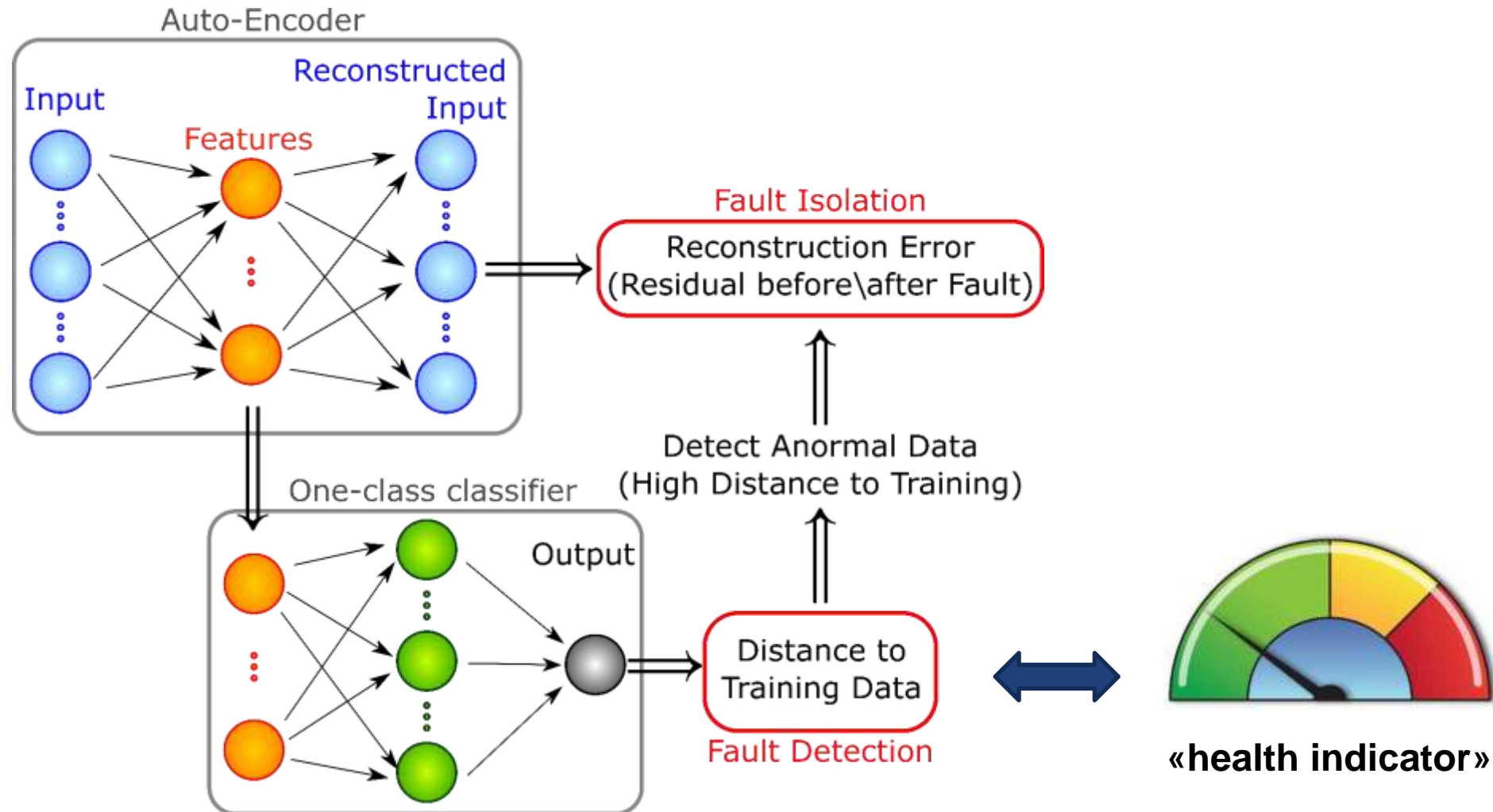
Many possible faults  
Preventive maintenance  
reliable systems

→ Not sufficient examples to learn from

Diagnostics from classification results impossible  
→ No clear fault signatures

Recorded data not representative  
(varying operating conditions)  
→ Not enough data from one system

# Proposition: Hierarchical Neural Networks – ELM<sup>1</sup>



1. Michau, G., Thomas, P., & Fink, O. (2017). Deep feature learning network for fault detection and isolation. In *PHM 2017, St. Petersburg, USA, 2-5 October 2017* (pp. 108-118). PHM Society

## Example on a Generator (PHM Society 2017)

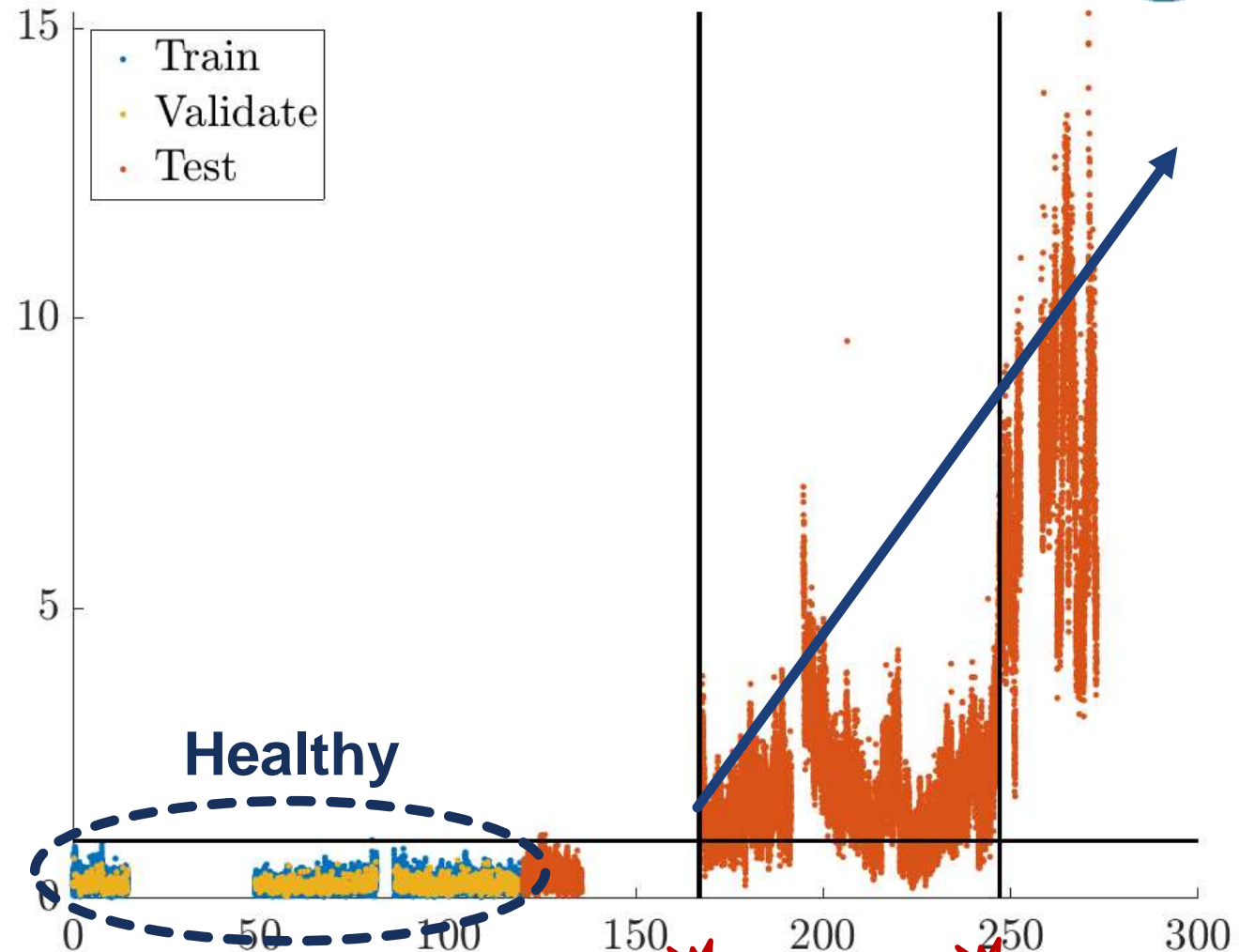


275 days  
60 000 points  
1 fault

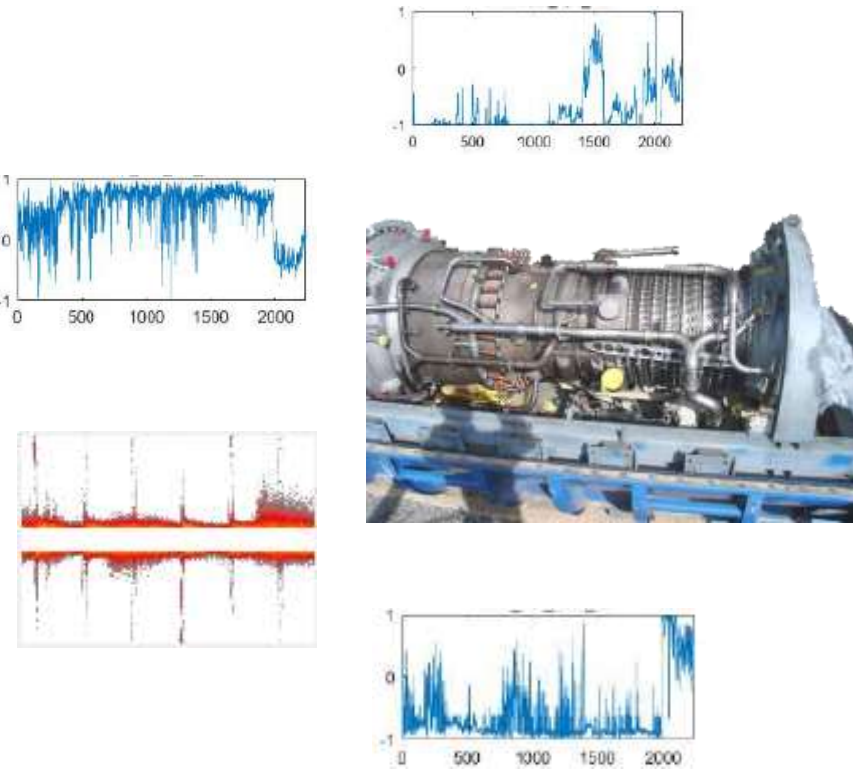
320 monitoring sensors:

- Partial discharge
- Rotor shaft voltage
- Rotor flux
- Stator end winding vibration
- Stator Water Temperature

**Abnormal behavior 100 days before!**



# Challenges in the Monitoring of Complex Industrial system



Large number and large variety of sensors  
→ Heterogeneous condition monitoring

Rare Faults:

Many possible failure modes  
Preventive maintenance  
reliable systems

→ Not enough examples to learn from

Diagnostics from classification results impossible  
→ No clear fault signatures

**Unsupervised Anomaly Detection**

Recorded data not representative  
(varying operating conditions)  
→ Not enough data from one system



# New Challenges

What if, I don't have so much training data?

- What if my system is **new** (or has been refurbished)?
- What if I am expecting operating conditions to **change**?

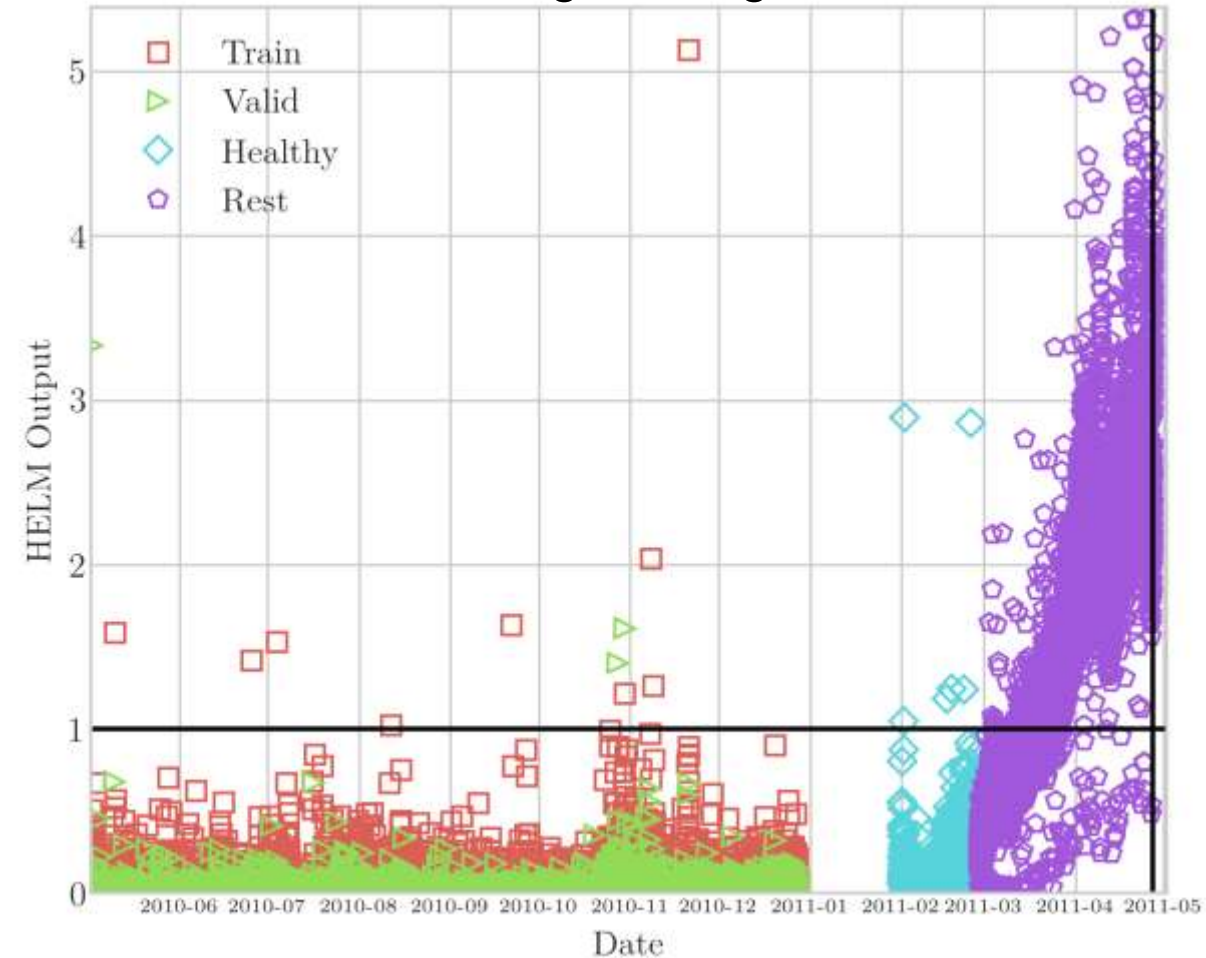
⇒ Use data from other similar systems

**Fleet approach** to PHM (from manufacturer perspective)

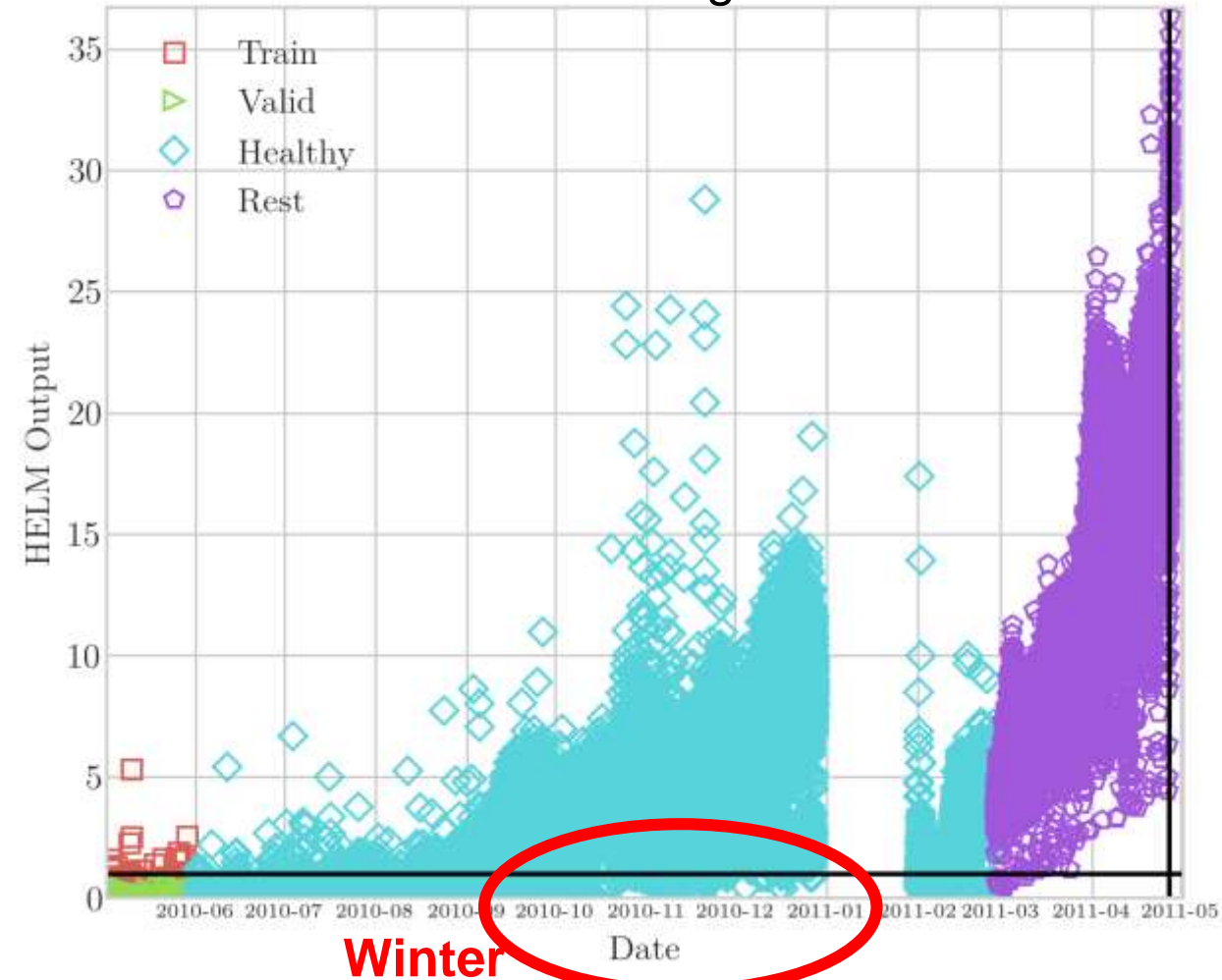


# Example: Stator Data

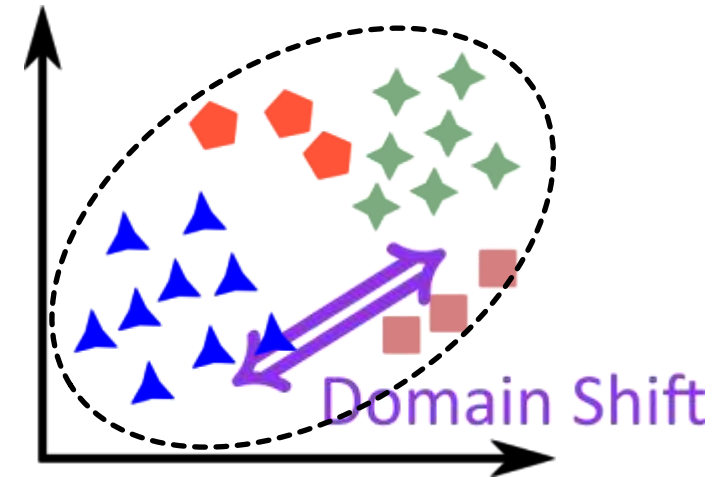
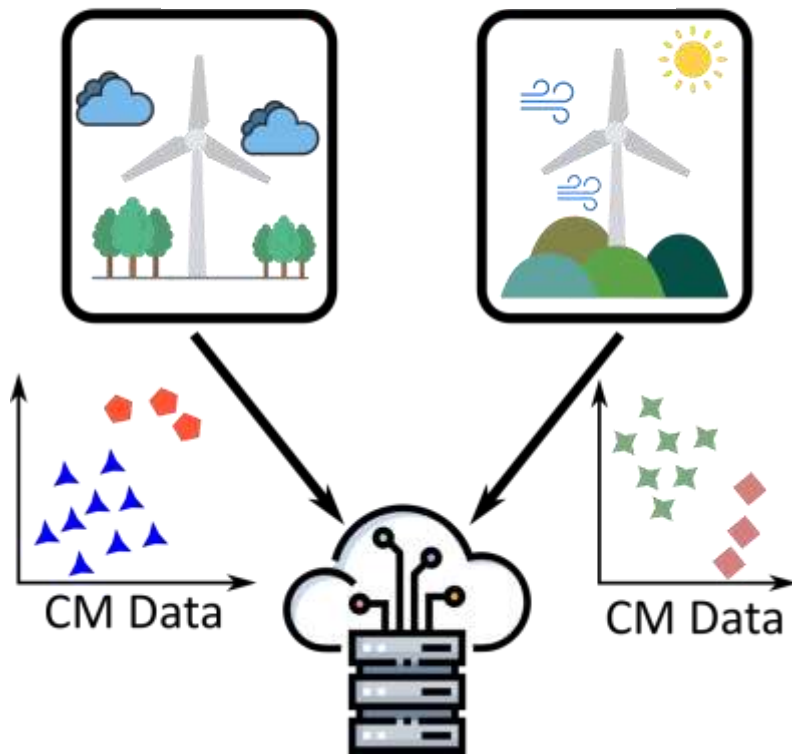
## Long training



## Short training



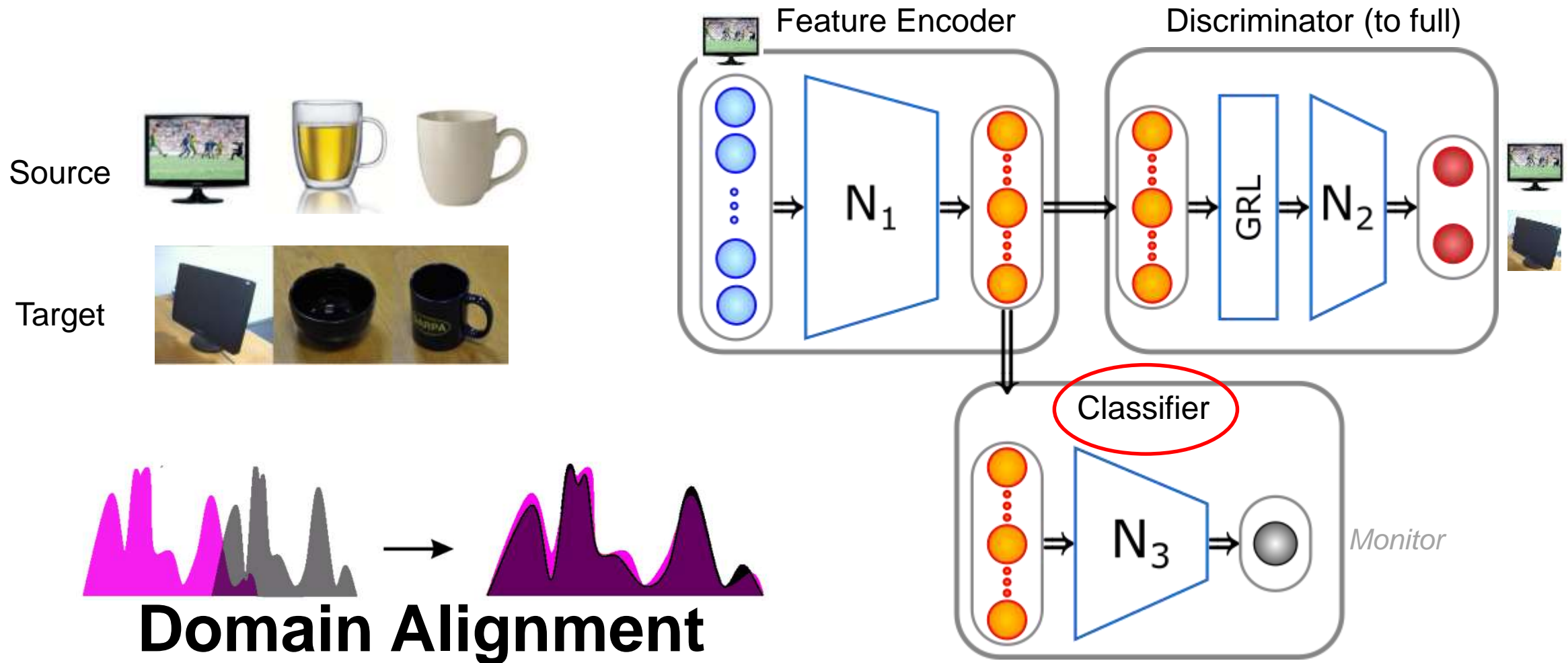
# Solution: Learn from other units



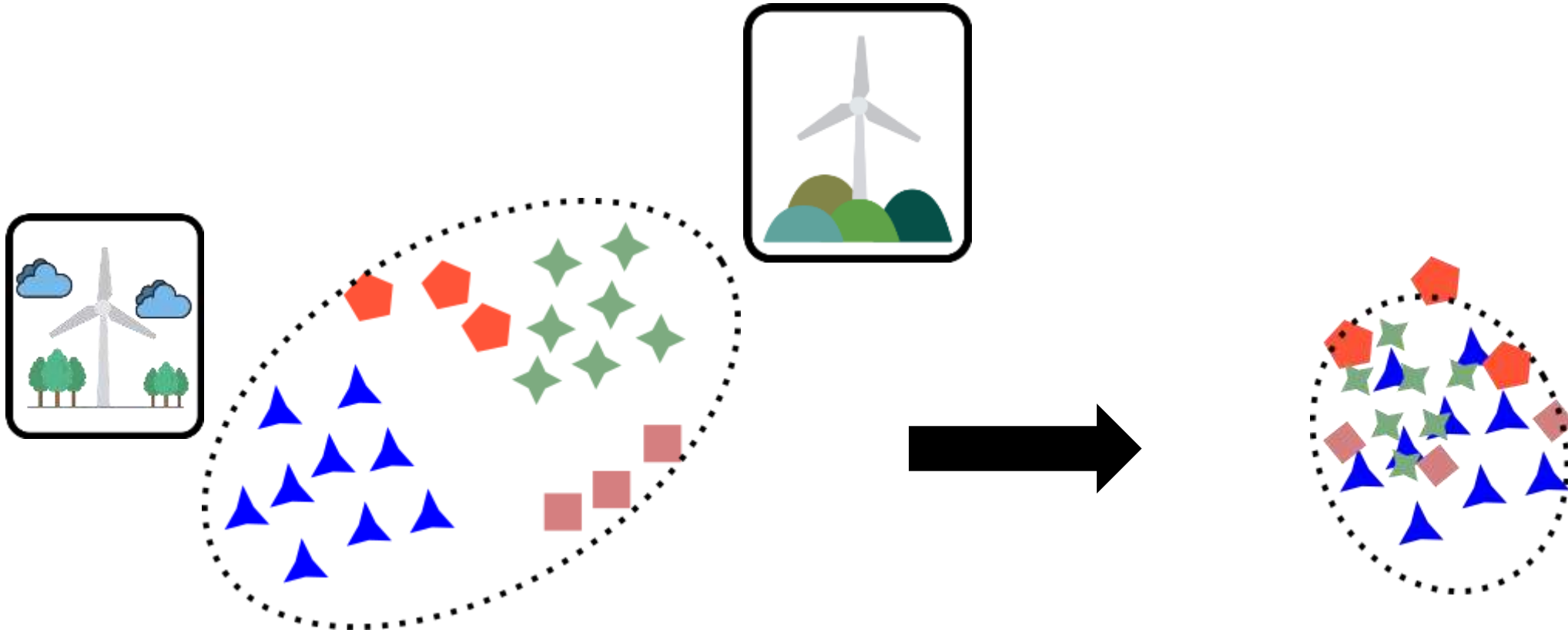
Identify similar units and operating conditions

- ✗ Not always possible
- ✗ Multi-dimensional time series comparison

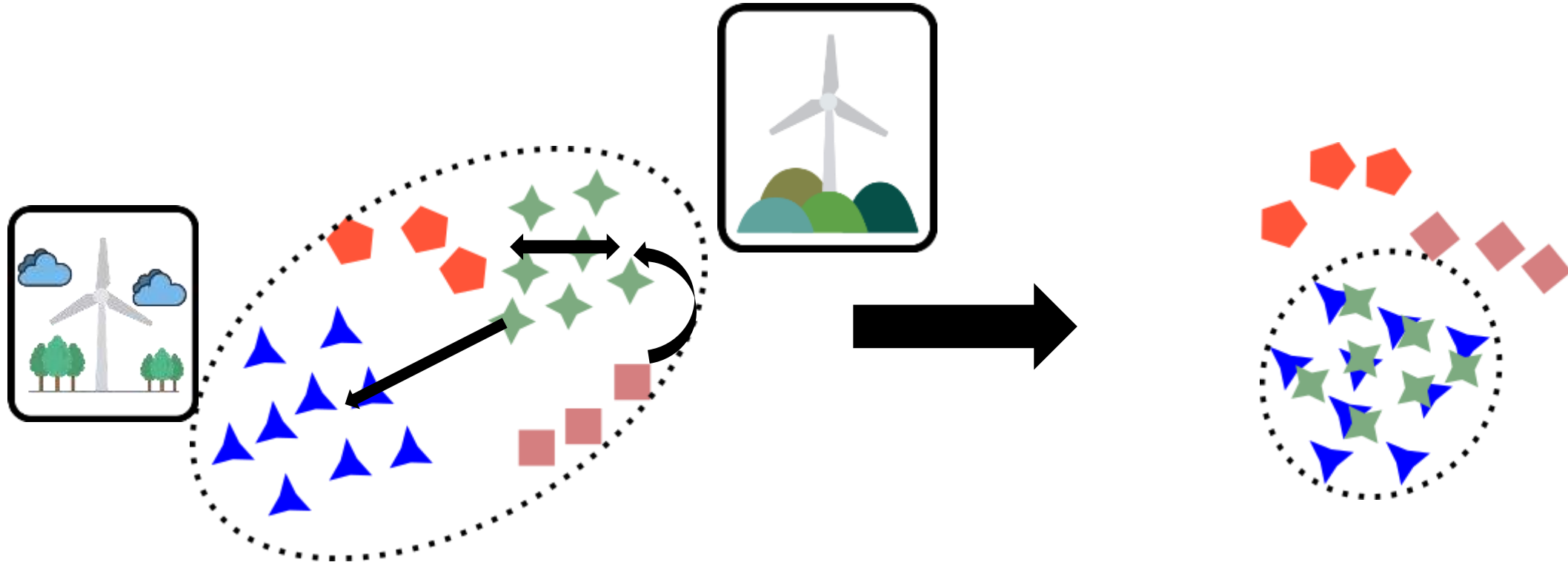
# Domain Adaptation Neural Network



# Risk in Unsupervised Alignment



# Keep inter-sample relationships while doing the overlap



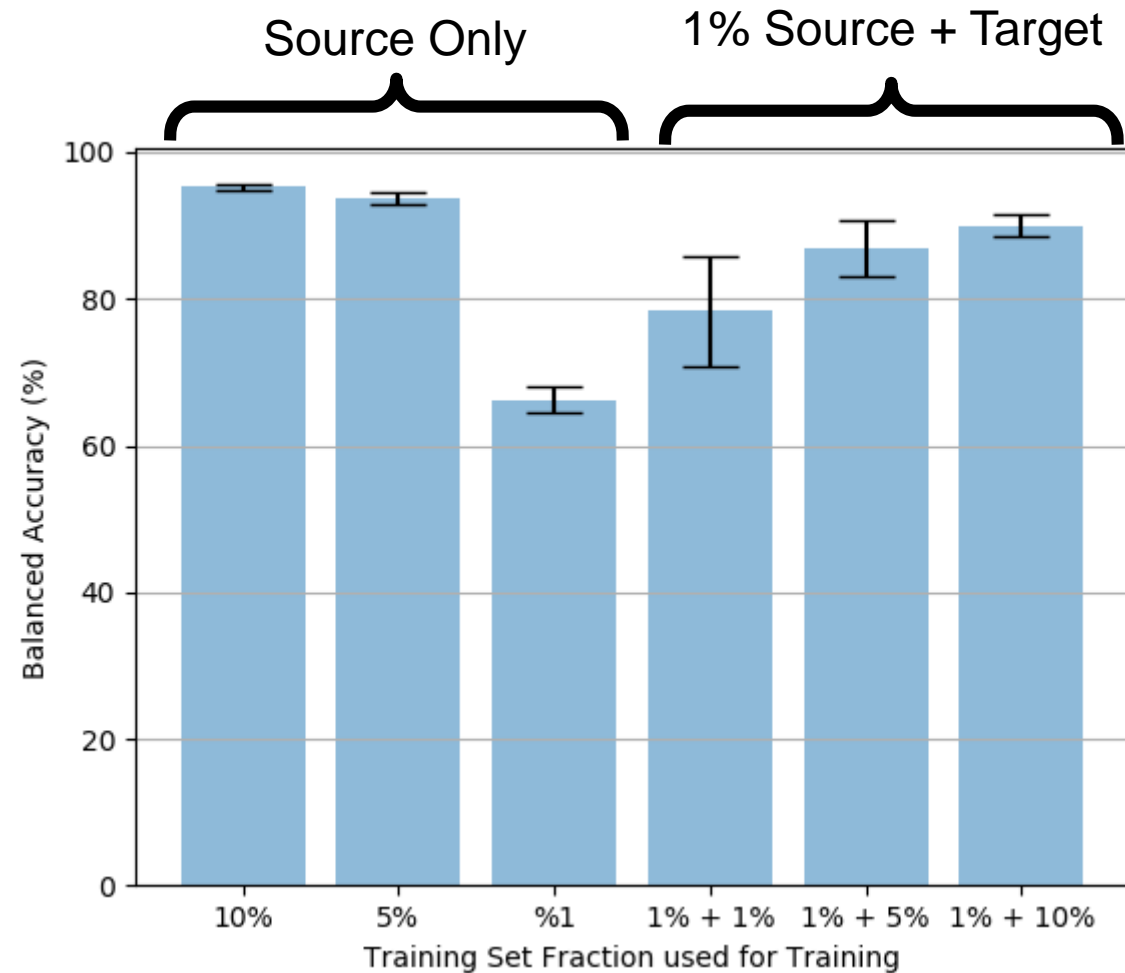
## Dimensionality Reduction Tool (Multi-dimensional Scaling)

Add this as an additional loss to minimize (instead of classification loss)

# Case Study – Anomalous Digit Detection on MNIST – MNIST-M



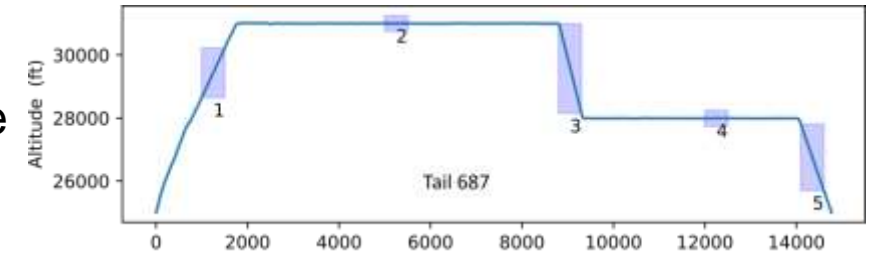
Main Class: 0  
 Anomalies: 1-9



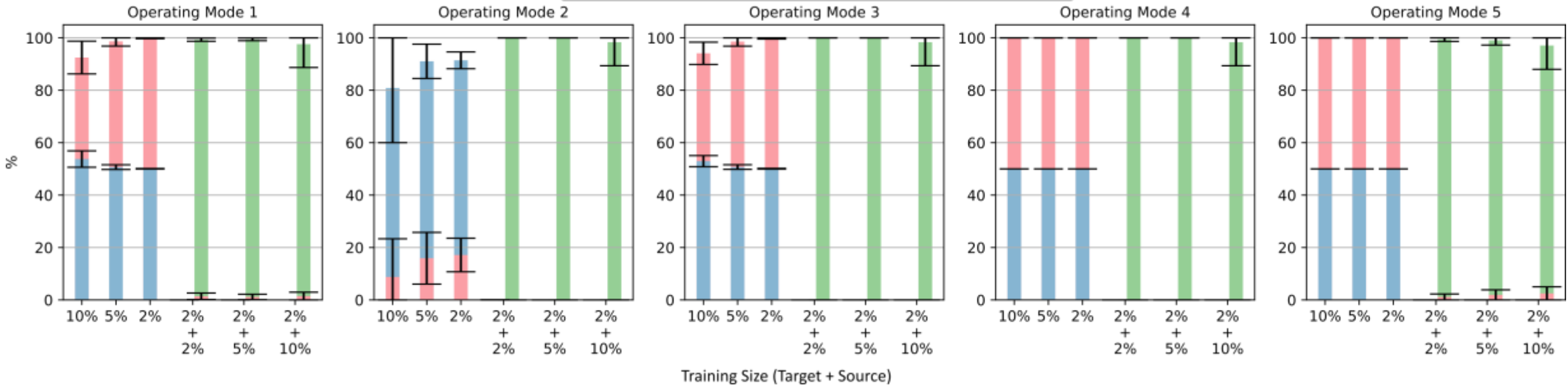
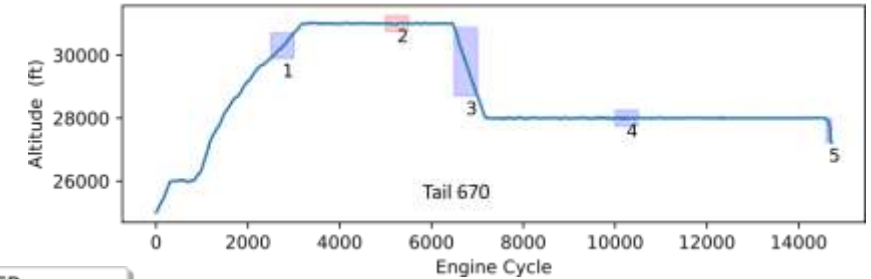


# Flight Engine Data Monitoring

Source



Target



Thanks for your attention!

