# Predictive Machine Maintenance Project (systemRx)

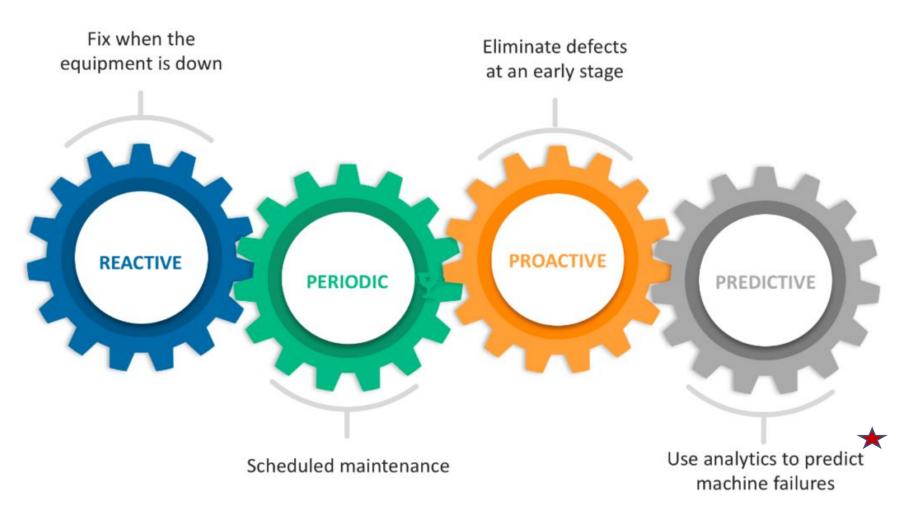


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Presenter: Sheneka Allen

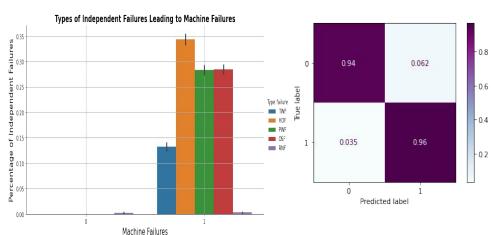
**June 2022** 

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## **Objective**

- Develop a modeling tool that accurately predicts the percent likelihood of machine failure. If possible, identify common contributors to machine failure to minimize equipment downtime and maintenance costs.
- Prediction performance goals: >90% correct positive predictions, < 5% for false negative predictions</li>
- Technology & tools: predictive maintenance (pDm) dataset (synthetic) of 10K rows, data analytics

## **Visualizations**



# Model 'Problem-Solving' Value

### Strengths:

- Helps to easily identify machine failure trends and patterns
- Responds well to new machine maintenance data and returns high quality predictions

#### **Limitations:**

- Limited public availability for predictive maintenance data to re-train model
- No date-time values in dataset to calculate machine average runtime or availability
- Must collect a good mix of machine data to ensure quality model performance (e.g., balanced dataset improved performance by 31%!!)

## Recommendations

- 1) Focus initial maintenance tasks on <u>highest</u> <u>percentage causes</u> for machine failures
- 2) Migrate from scheduled maintenance culture to formal condition monitoring program
- Identify & collect machine data on <u>critical assets</u> (e.g., safety and revenue loss focus)
- Determine <u>what typically goes wrong</u> on machines that cause failures (e.g., 34% heat dissipation (HDF), 28% power failures (PWF) for this dataset)
- Consider altering testing frequency between degraded machine operation and breakdown

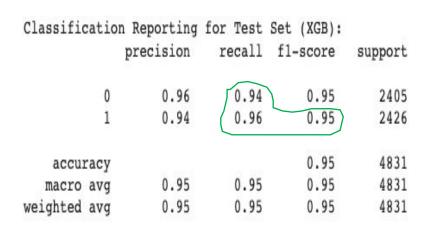
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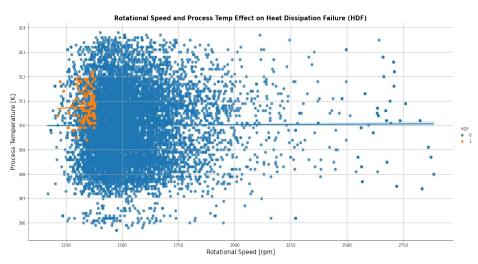
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# **Backup - More systemRx Visualizations**

### Classification Metrics for Confusion Matrix (Slide 2)

## **Heat Dissipation Failure - Feature Analysis**





### **Imbalanced Data & Model Performance**

Imbalanced datasets cause model predictions to be <u>skewed</u> in favor of the majority class versus the minority class (target). The prediction accuracy is 99% but of ZERO analysis value! You MUST correct the data imbalance and re-train the model to improve performance quality and ensure its value to stakeholders.

