The Future of Maintenance: Al-Driven Failure Prediction & Anomaly Detection

Lionel Low, Capstone Project



A section of power at burned throug Tuas West Extensio re the primary cause affected three MRT . The incident lasted 123,000 commuters. Many were stuck in tra

a faulty circuit breaker llong the massive disruption on Oct 14 that e than three hours, and affected r hours when the power went out.

MAD A TO

¿ stalleu bad pulled into Joo Koon station y were r system. Passengers on the

Investigations into the incident found that a s Bishan and Braddell MRT stations – design was likely close to full before the inciden √ater pi maintenance, sludge and debris had ac e tunnel between ollect a water pit, which could have affected In out rainwater _ s also f at due to a lapse in ated in ver compartment or the eration abilities of pumps and float

East-West Line MRT disruption. How a faulty train left a trail of destruction

Train services along a stretch of the East-West Line (EWL) have been disrupted since Sept 25, affecting more than 1.3 million passengers. What exactly happened?



\$1,300,000 / h



Objectives

How to predict machine failures?

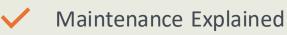
How to detect anomalies in machine?

Stakeholders





Agenda





Failure Predictions

Detecting Anomalies

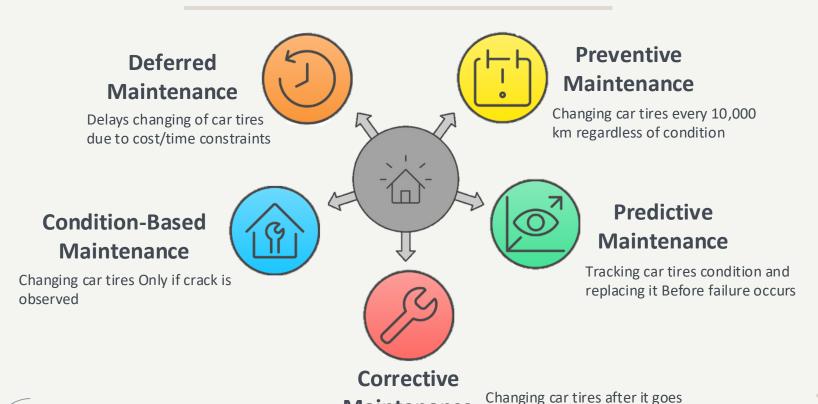
Business recommendations

What is Maintenance?



Maintenance involves the regular actions taken to keep equipment or systems in working condition, prevent failures, and optimize performance, ensuring longevity and reliability. It aims to reduce downtime and improve operational efficiency.

What are the different types of Maintenance?



Maintenance

flat

Data Summary

Ensure data is curated for machine learning

Identify relevant sources such as fact and dimension tables.
Validate data to ensure data integrity

Data Preparation

Failure Predict using Random Forest

Predicting failure with supervised modeling with actual deployment

Unsupervised modelling

Data Sourcing

(from Kaggle, synthetic)

Supervised prediction

Prepare data so that it meets modeling specifications

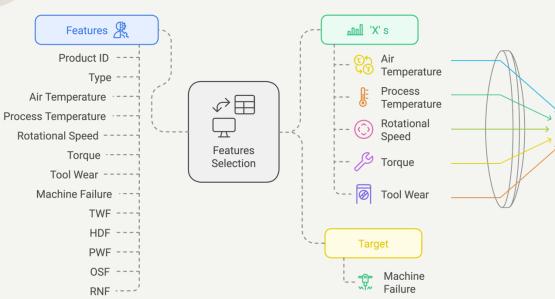
Check data for missing values, identify patterns and clean data according to patterns
(13 features, 10,000 rows)

Anomaly Detection using Autoencoder

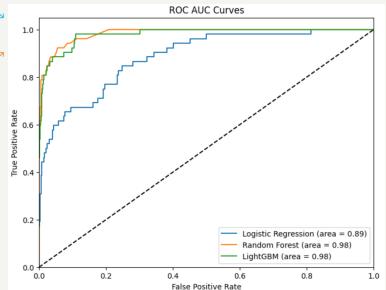
Defining the optimal parameters for machine learning to identify anomalies

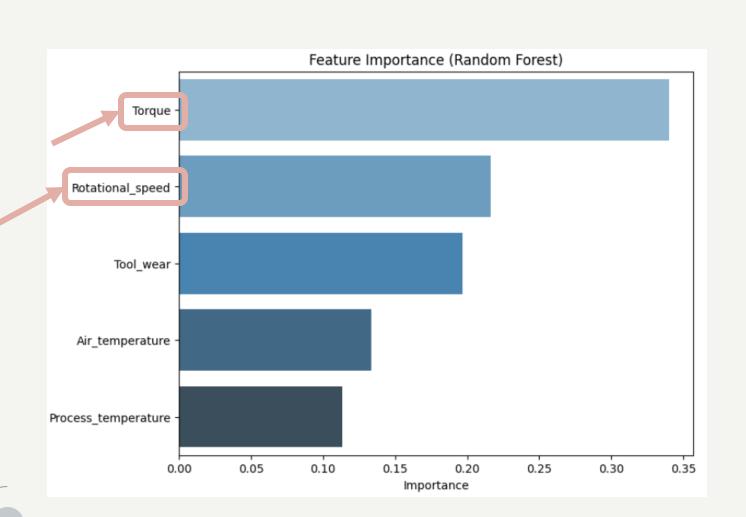


Failure Prediction

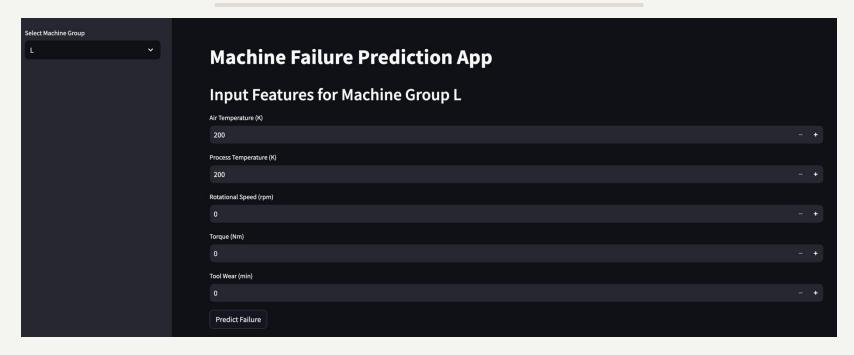


	Model	Precision	Recall	F1-Score	Accuracy
0	Random Forest	0.980159	0.980833	0.978770	0.980833
1	LightGBM	0.977803	0.979167	0.977208	0.979167
2	Logistic Regression	0.960905	0.965833	0.958135	0.965833





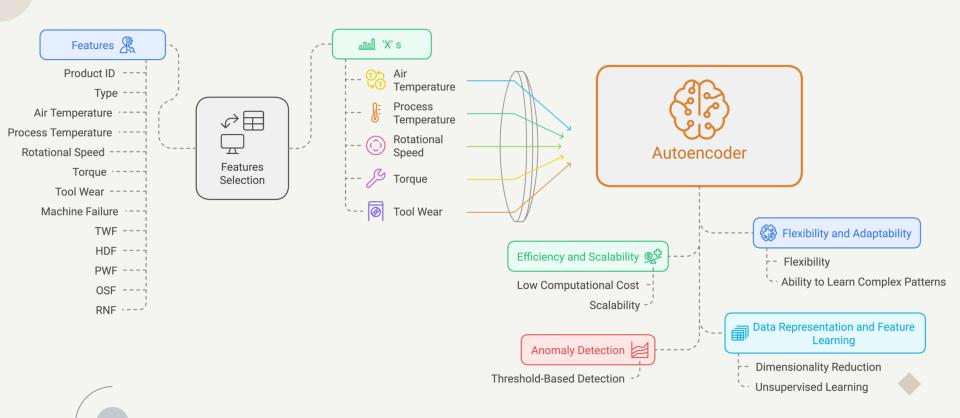
Demo – Predicting Failure



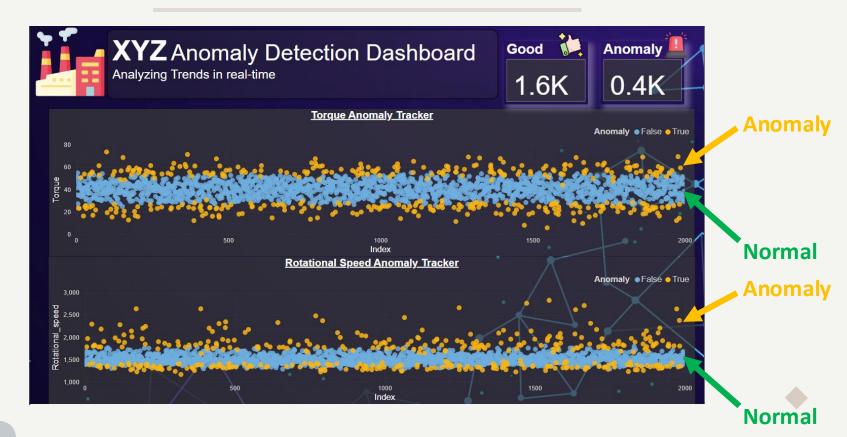
https://machine-prediction.streamlit.app/



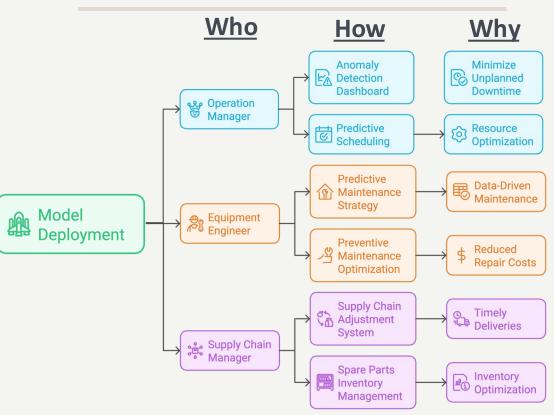
Anomaly Detection



Dashboard Deployment (PowerBI)



Enhancing Business Strategies

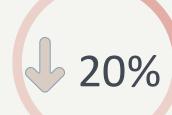


Next Steps

Machine Downtime



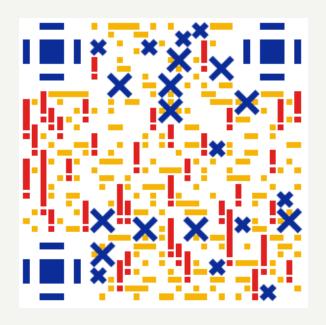
Maintenance Cost



- 1. Incorporate time-series features to make time-to-failure predictions.
- 2. Integrate **sensor data & IoT technologies** to collect real-time data. **Automated maintenance alerts** could be setup to notify the stakeholders when certain threshold are exceeded.
- 3. Expanding the dashboard features that facilitates **decision-making**. Features such as past **historical trends** and **visualization of the machine health** can be included.



Let's build smarter solutions for a brighter tomorrow!





Professional in Data Science & Artificial Intelligence

+65 9871 5633

lionel.lowyy@gmail.com

https://www.linkedin.com/lionel-low-yy

https://github.com/lioneltechjourney