### **MACHINE LEARNING PROJECT**

# PREDICTIVE MAINTENANCE OF INDUSTRIAL MACHINERY

#### **Presented By:**

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### **OUTLINE**

- Problem Statement (Should not include solution)
- Proposed System/Solution
- System Development Approach (Technology Used)
- Algorithm & Deployment
- Result (Output Image)
- Conclusion
- Future Scope
- References



# PROBLEM STATEMENT

Develop a predictive maintenance model for a fleet of industrial machines to anticipate failures before they occur. This project will involve analyzing sensor data from machinery to identify patterns that precede a failure. The goal is to create a classification model that can predict the type of failure (e.g., tool wear, heat dissipation, power failure) based on real-time operational data. This will enable proactive maintenance, reducing downtime and operational costs.



# **PROPOSED SOLUTION**

The proposed system aims to address the challenge of predicting the type of failure in industrial machines before it occurs. This enables timely maintenance and reduces operational disruptions. The solution will involve using machine learning algorithms to analyze sensor data and identify patterns associated with different types of failures. The solution will consist of the following components:

- Data Collection:
- Gather historical sensor data from the industrial machines including vibration, temperature, pressure, rotation speed, and torque.
- Collect machine status logs indicating operational conditions, failure events, maintenance records, and component usage.

## **PROPOSED SOLUTION**

- Data Preprocessing:
- Clean and preprocess the data to remove missing values, noise, duplicates, and inconsistent records.
- Perform feature engineering to extract meaningful features such as rolling averages, thresholds, and change detection patterns that correlate with failure types.
- Machine Learning Algorithm:
- Implement classification algorithms such as Random Forest, XGBoost, or Neural Networks to classify the type of failure.
- Experiment with time-series or sequential models (e.g., LSTM, GRU) to capture temporal dependencies in sensor data.

### PROPOSED SOLUTION

- Deployment:
- Develop a web-based dashboard or mobile application using IBM Cloud Lite services for real-time prediction and alerts.
- Automate model inference to run on real-time data streams and trigger early warning notifications.
- Ensure system scalability and reliability, considering factors such as processing time, API response time, and uptime.



# SYSTEM APPROACH

- The "System Approach" section outlines the overall strategy and methodology for developing and implementing the power system fault detection and classification.
  Here's a suggested structure for this section:
- System requirements:
- IBM Cloud (mandatory)
- IBM Watson Studio for model development and deployment
- IBM Cloud Object Storage for dataset handling



# **ALGORITHM & DEPLOYMENT**

- ALGORITHM & DEPLOYMENT
- Algorithm Selection:

Random Forest Classifier (or SVM based on performance)

Data Input:

Voltage, current, and phasor measurements from the dataset

Training Process:

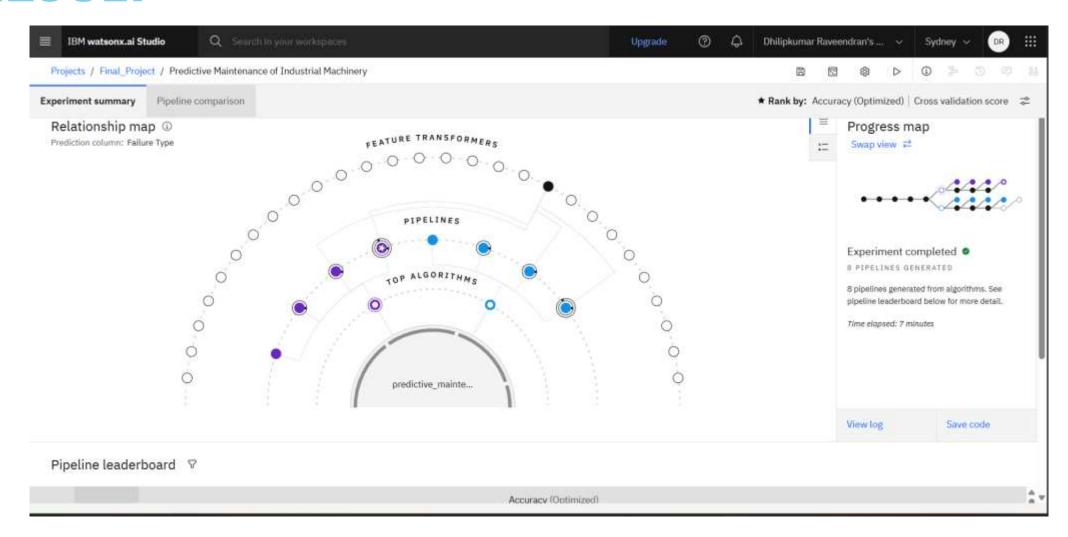
Supervised learning using labeled fault types

Prediction Process:

Model deployed on IBM Watson Studio with API endpoint for real-time predictions

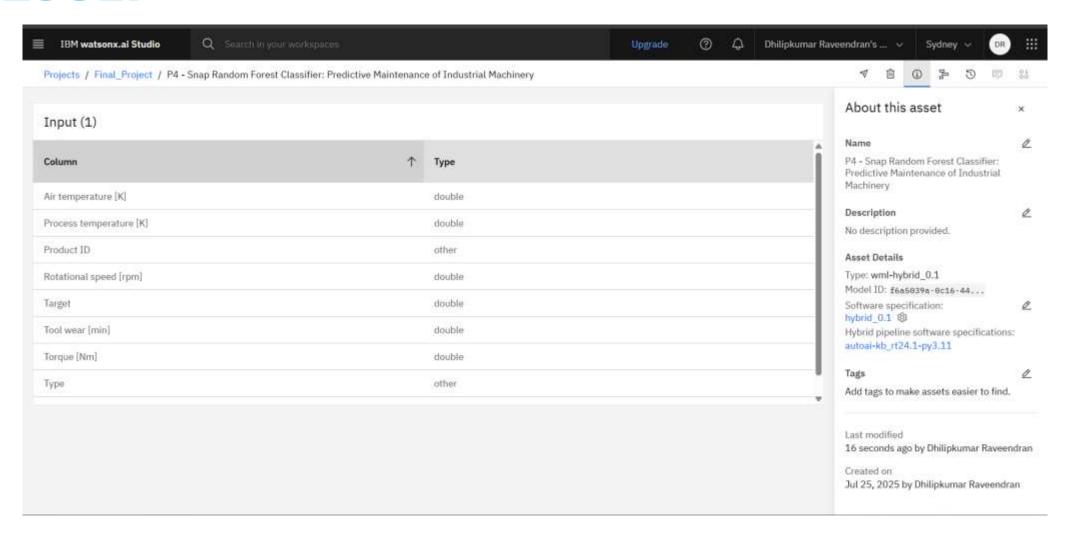


# **RESULT**



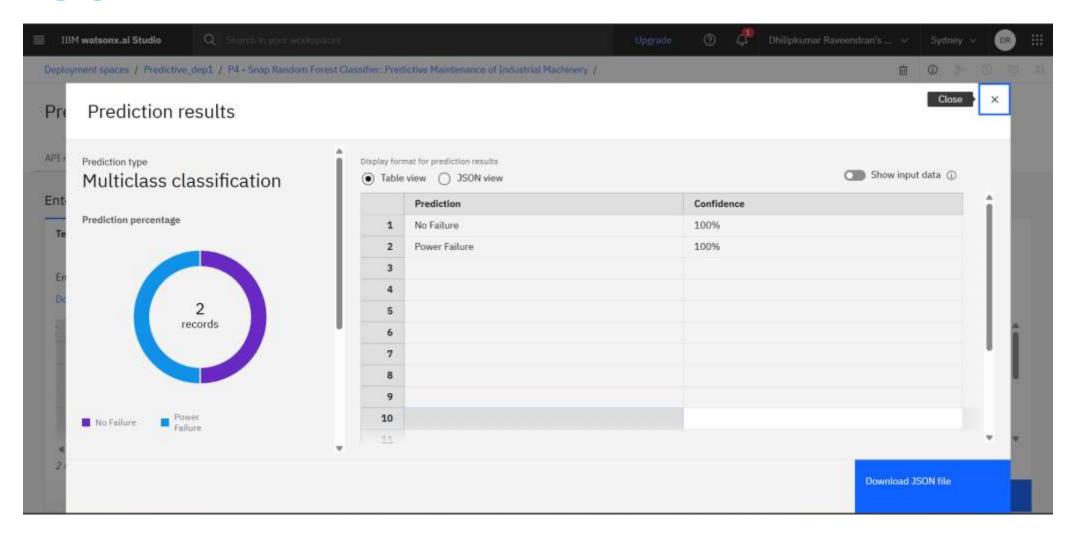


# **RESULT**





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# CONCLUSION

- •The predictive maintenance solution built on **IBM Cloud** provides a **scalable** and **reliable** approach to anticipate industrial machinery failures.
- •The model enables **proactive maintenance**, reducing **downtime**, **repair costs**, and **unexpected failures**.
- •Demonstrated the potential of **AI + Cloud** in enhancing traditional mechanical systems through **smart maintenance planning**.
- •Future enhancements may include **integration with IoT devices**, **real-time dashboards**, and expanding the model to cover **diverse machine types** across the industry.



### **FUTURE SCOPE**

The proposed system provides a strong foundation for real-time fault detection in power systems. However, there are several opportunities for future enhancement:

### Integration with IoT Devices

Real-time data acquisition can be enhanced by integrating IoT-enabled sensors directly with the model for live fault detection and faster response.

#### **Expansion to Multiple Fault Types**

The current model can be extended to classify more complex and rare fault types, improving grid reliability and robustness.

### **Use of Advanced Deep Learning Models**

Incorporating deep learning architectures such as CNNs or LSTMs could improve accuracy, especially in high-dimensional time-series data.



# REFERENCES

- •IBM Cloud Official Website
- https://www.ibm.com/cloud
- •IBM Watson Studio Model Development and Deployment
- https://www.ibm.com/cloud/watson-studio
- •IBM Cloud Object Storage Dataset Handling
- https://www.ibm.com/cloud/object-storage
- •IBM Machine Learning Documentation
- https://www.ibm.com/docs/en/cloud-paks/cp-data/4.6.x?topic=services-machine-learning
- •Kaggle Dataset for Predictive Maintenance (if used):
- https://www.kaggle.com/datasets/shivamb/machine-predictive-maintenance-classification

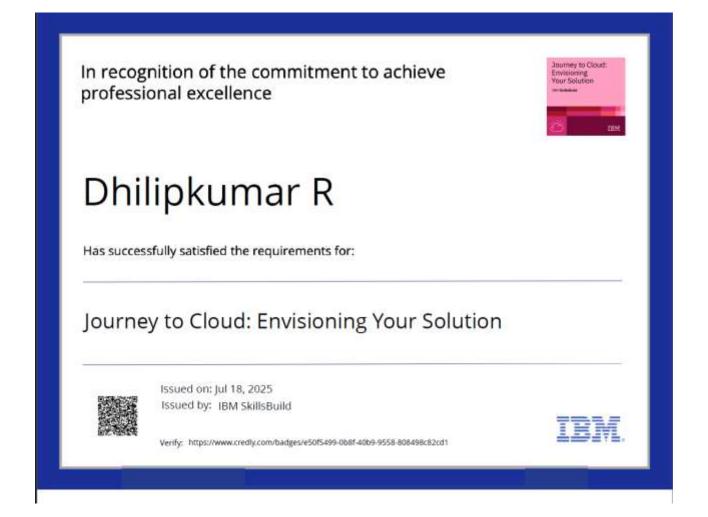


### **IBM CERTIFICATIONS**





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7/24/25, 2:54 PM Completion Certificate | SkillsBuild IBM SkillsBuild Completion Certificate This certificate is presented to Dhilipkumar R for the completion of Lab: Retrieval Augmented Generation with LangChain (ALM-COURSE\_3824998) According to the Adobe Learning Manager system of record Completion date: 24 Jul 2025 (GMT) Learning hours: 20 mins



### **THANK YOU**

