## **CAPSTONE PROJECT**

# PREDICTIVE MAINTENANCE OF INDUSTRIAL MACHINERY

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### **OUTLINE**

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# PROBLEM STATEMENT

### **Predictive Maintenance of Industrial Machinery**

- 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

#### Sensor Data Collection:

Collect real-time and historical sensor data from industrial machines, including parameters such as temperature, torque, speed, and tool wear.

### Data Preprocessing:

Clean and preprocess the data to handle missing values and outliers. Apply normalization and feature encoding as needed.

### • Feature Engineering:

Extract meaningful features from raw sensor inputs that contribute to machinery failures (e.g., tool wear thresholds, speed anomalies).



# PROPOSED SOLUTION

### Machine Learning Model:

Utilize IBM AutoAI to automatically test, tune, and select the best classification model (e.g., Random Forest) for predicting failure types.

### • Deployment:

Deploy the best-performing model using IBM Watsonx.ai Studio. Enable users to input real-time machine data and get instant failure predictions.

#### • Evaluation:

Assess model accuracy using cross-validation metrics (Accuracy: 99.5%). Continuously monitor and retrain if needed for improved reliability.



# SYSTEM APPROACH

The System Approach outlines the technical environment and tools used to build the predictive maintenance system.

### **System Requirements**

#### **Software:**

- IBM Cloud account
- IBM Watsonx.ai Studio
- IBM Cloud Object Storage
- Watson Online Deployment

#### **Input Data:**

Industrial sensor data (CSV format) which includes features such as Air Temp, Process Temp, Torque, Speed, Tool Wear, etc.

#### **Output:**

Predicted machine failure type (e.g., No Failure, Heat Dissipation, Power Failure)



# **ALGORITHM & DEPLOYMENT**

#### **Algorithm Used:**

- Snap Random Forest Classifier
- It has high accuracy (99.5%), Robust with structured sensor data, Good for multi-class classification, Handles complex relationships between features

#### **Input Features:**

• Product Name, Machine Type, Air & Process Temperature, Rotational Speed, Torque, Tool Wear etc.

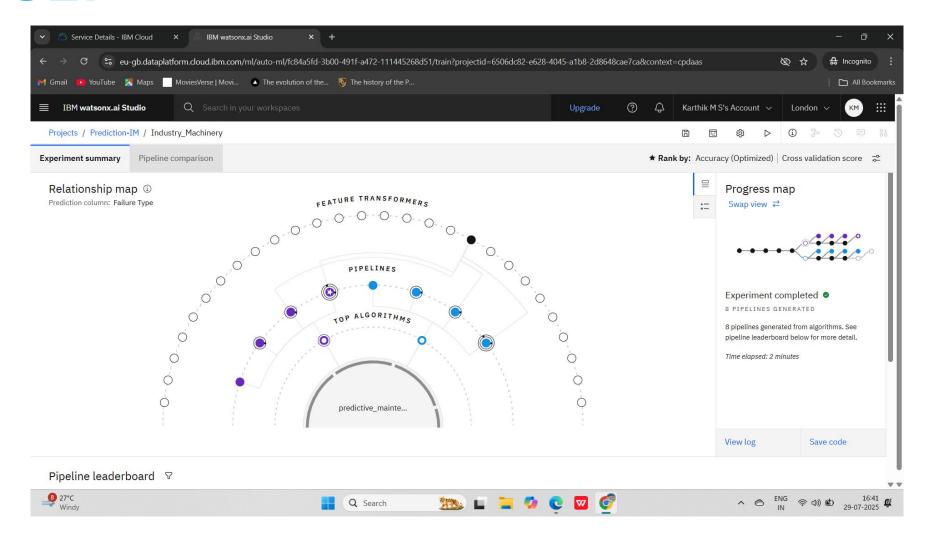
#### **Training Process:**

- AutoAI handled preprocessing, feature engineering, and model tuning
- Used cross-validation and hyperparameter optimization (HPO)

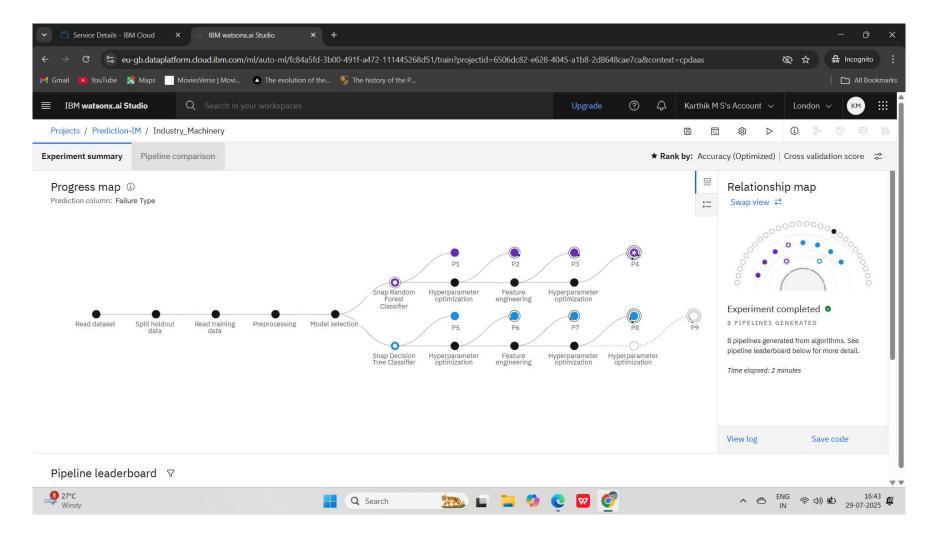
#### **Prediction:**

- Real-time failure type prediction
- Deployed on IBM Cloud for live input and output

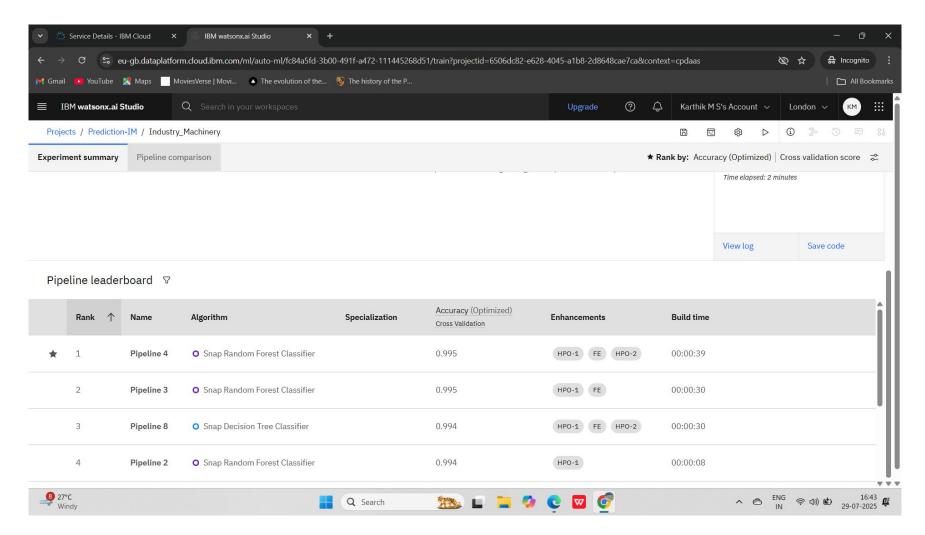




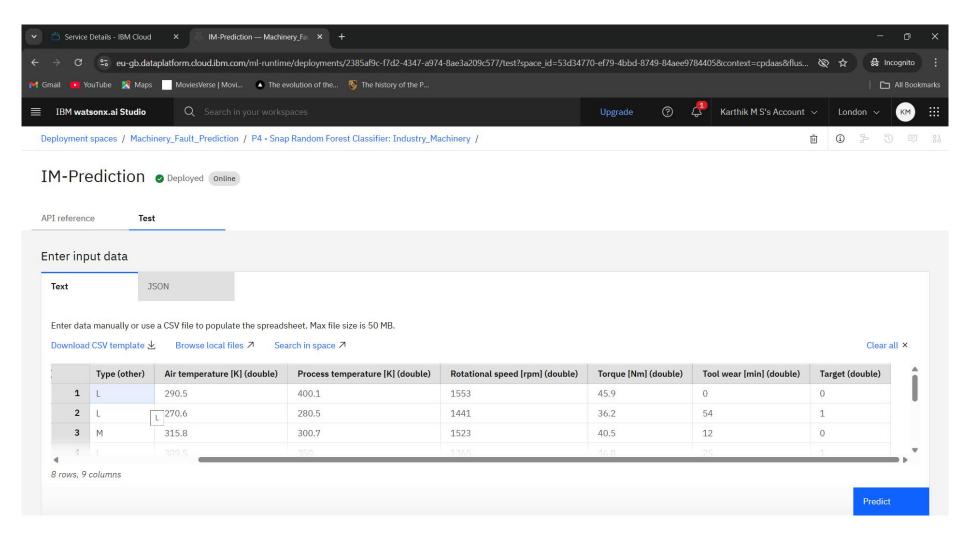










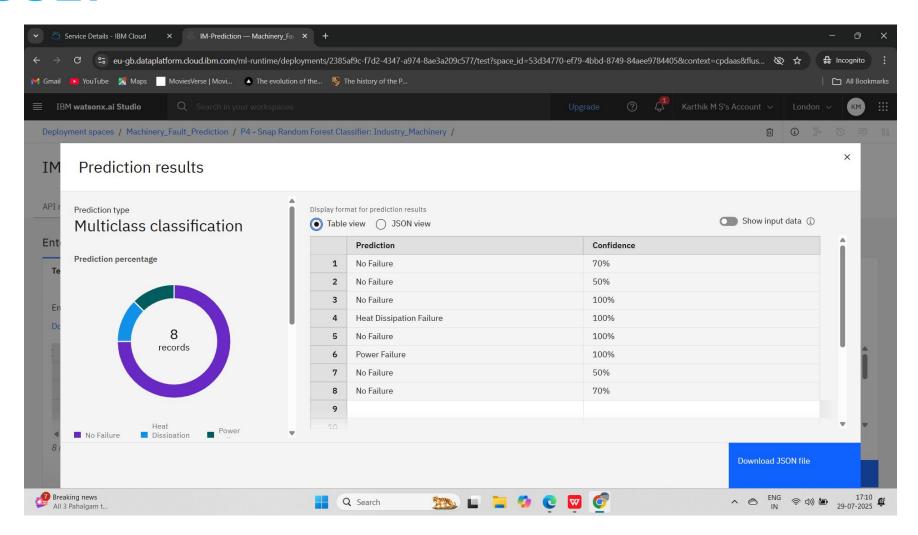




### Test data

UDI	Product ID	Type	Air tempera	Process tem	Rotationals	Torque [Nm	Tool wear [17	Target
	1 L58642	L	290.5	400.1	1553	45.9	0	0
	2 L89421	L	270.6	280.5	1441	36.2	54	1
	3 M98723	M	315.8	300.7	1523	40.5	12	0
	4 L15489	L	309.5	350	1365	46.8	25	1
	5 M36478	M	296.1	309	1523	36.9	24	0
	6 L47230	L	298	309	2861	4.6	143	1
	7 M25846	M	309.2	299	1400	43.7	19	1
	8 H73485	Н	320.7	290	1652	38.5	13	1







## CONCLUSION

- Achieved 99.5% accuracy in predicting machine failures using IBM AutoAI.
- Enabled proactive maintenance, reducing machine downtime and operational costs.
- Ensured accurate classification of different failure types (e.g., tool wear, power failure).
- Faced challenges in understanding AutoAI workflow and interpreting multi-class predictions.
- Demonstrated the effectiveness of AI in improving industrial machinery reliability.



### **FUTURE SCOPE**

- Add more data sources like vibration and maintenance logs.
- Use advanced ML models for better performance.
- Expand the system to multiple factories or regions.
- Explore deep learning and edge computing for real-time predictions.
- Build a dashboard with live alerts for maintenance teams.



# REFERENCES

Kaggle Dataset:

Machine Predictive Maintenance Classification Dataset

https://www.kaggle.com/datasets/shivamb/machine-predictive-maintenance-classification

IBM Documentation:

IBM Watsonx.ai – AutoAI User Guide

https://www.ibm.com/docs/en/watsonx

IBM AutoAI Overview

https://www.ibm.com/cloud/watson-studio/autoai



### **IBM CERTIFICATIONS**

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# **THANK YOU**

