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# **CAPSTONE PROJECT**

## **PREDICTIVE MAINTENANCE OF INDUSTRIAL MACHINERY USING MACHINE LEARNING**

**Presented By:**

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# OUTLINE

- Problem Statement
- Proposed System/Solution
- System Development Approach
- Algorithm & Deployment
- Result (Output Image)
- Conclusion
- Future Scope
- References

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# PROBLEM STATEMENT

Industrial machines are prone to unexpected failures, leading to costly downtime and inefficiencies. This project aims to develop a **predictive maintenance model** that uses data to identify patterns leading to machine failures such as **tool wear**, **power failure**, and **heat dissipation issues**. The goal is to build a **classification model** that predicts the **type of failure in advance**, enabling **proactive maintenance** and reducing operational costs.

# PROPOSED SOLUTION

- The proposed solution aims to predict potential machinery failures in advance by analyzing real-time sensor data, enabling proactive maintenance and reducing downtime. The system will be developed using **IBM Watsonx.ai Studio** and deployed on the **IBM Cloud** platform. The key components include::
  - **1. Data Collection**

Use a labeled dataset containing historical **sensor readings** from industrial machines.

Key features include: **air temperature**, **process temperature**, **torque**, **rotational speed**, and **tool wear**.

The dataset also includes labels for types of failures such as **tool wear**, **power failure**, **heat dissipation failure**, and **no failure**.
  - **2. Data Preprocessing**

Handle **missing values**, **outliers**, and normalize the sensor data.

Apply **label encoding** for categorical features (e.g., failure types).
  - **3. Machine Learning Model (AutoAI)**

Use **IBM Watsonx.ai Studio's AutoAI** to automate model training, testing, and pipeline creation.

Multiple pipelines are generated and evaluated automatically.
  - **4. Deployment**

Deploy the selected model to **IBM Watson Machine Learning** as an **online REST API** service.

Enable **real-time predictions** for use in production environments.
  - **5. Evaluation**

Evaluate model performance using classification metrics such as: **Accuracy**, **Recall**, **Precision**
  - **6 Result :** The solution enables industries to **predict specific types of failures** in advance, helping them schedule **proactive maintenance**, avoid **unexpected breakdowns**,

# SYSTEM APPROACH

The **System Approach** section outlines the overall strategy and tools used to develop, train, and deploy a predictive maintenance model that can classify potential machine failures based on sensor data.. **Here's a suggested structure for this section:**

- System requirements –
  - 1) IBM CLOUD (Mandatory)
  - 2) IBM Watson studio for model development and deployment
  - 3) IBM cloud object storage for data set handling

# ALGORITHM & DEPLOYMENT

- This section describes the machine learning algorithm used for classifying machinery failures and the process of deploying the trained model using IBM Watsonx.ai.

- **Algorithm Selection:**

- The project uses **IBM AutoAI** to automatically select and train the most effective machine learning classification algorithm based on the dataset, the chosen algorithm was **Decision Tree Classifier**, as it provided the highest accuracy for classifying failure types like tool wear, heat dissipation failure, and power failure.

- **Data Input:**

The following features from the sensor data were used as input for training the model: **Air Temperature, Process Temperature, Rotational Speed, Torque, Tool Wear**

- **Training Process**

Automatically split the data into **training and test sets**

Performed **feature engineering** and **normalization**

Applied **hyperparameter tuning**

Evaluated multiple classification pipelines using metrics like **Accuracy, Precision, Recall, and F1-Score**

- **Prediction Process**

Once trained, the model predicts the **type of failure** that may occur given a set of real-time sensor inputs. The model analyzes these inputs and classifies them into predefined failure categories. The deployment is designed for **real-time prediction**, allowing industries to act before an actual failure happens.

- **Deployment**

After training and selecting the best model pipeline:

The model was **saved as an asset** in IBM Watsonx.ai Studio.

It was then **promoted to a deployment space** using Watson Machine Learning (WML).

# RESULT

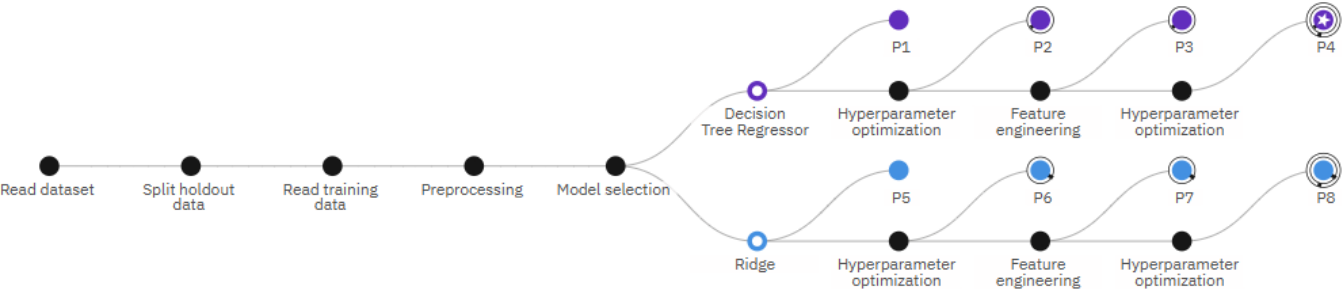
Projects / Predictive Maintenance ML model for Industrial Machinery / Predictive Maintenance ML Model

Experiment summary Pipeline comparison

★ Rank by: Root mean squared error (RMSE) (...) Cross validation score

## Progress map ⓘ

Prediction column: Tool wear [min]



## Relationship map

Swap view ↺



Experiment completed 🟢

8 PIPELINES GENERATED

8 pipelines generated from algorithms. See pipeline leaderboard below for more detail.

Time elapsed: 3 minutes

View log

Save code

## Pipeline leaderboard ▾

	Rank	↑	Name	Algorithm	RMSE (Optimized) Cross Validation	Enhancements	Build time
★	1		Pipeline 4	Decision Tree Regressor	62.145	HPO-1 FE HPO-2	00:00:34
	2		Pipeline 3	Decision Tree Regressor	62.171	HPO-1 FE	00:00:29
	3		Pipeline 2	Decision Tree Regressor	62.171	HPO-1	00:00:05

# RESULT

IBM watsonx.ai Studio

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Projects / ... / P4 - Decision Tree Regressor: Predictive Maintenance ML Model

Input (1)

Column	Type
Air temperature [K]	double
Failure Type	other
Process temperature [K]	double
Product ID	other
Rotational speed [rpm]	double
Target	double
Torque [Nm]	double
Type	other

About this asset

Name

P4 - Decision Tree Regressor: Predictive Maintenance ML Model

Description

This model predicts the type of failure in industrial machinery using sensor data. It is built using AutoAI with a Decision Tree Regressor pipeline, trained on the Predictive Maintenance dataset. The model helps anticipate failures like tool wear, power loss, or heat dissipation issues for proactive maintenance.

Asset Details

Type: wml-hybrid\_0.1

Model ID: ef088cd0-f222-4a...

Software specification: hybrid\_0.1

Hybrid pipeline software specifications: autoai-kb\_rt24.1-py3.11

Tags

Add tags to make assets easier to find.

Last modified

1 hour ago by Raj Singh



# RESULT

IBM watsonx.ai Studio

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Deployment spaces / Predictive Maintenance Deployment / P4 - Decision Tree Regressor: Predictive Maintenance ML Model /

PredictiveMaintenanceDeployment Deployed Online

API reference

Test

Enter input data

Text

JSON

Enter data manually or use a CSV file to populate the spreadsheet. Max file size is 50 MB.

Download CSV template

Browse local files

Search in space

Clear all

	UDI (double)	Product ID (other)	Type (other)	Air temperature [K] (double)	Process temperature [K] (double)	Rotational speed [rpm] (double)	Torque [Nm] (double)	Target (c
1	1	M17890	M	298.7	307.1	1508	49	29
2								
3								
4								
5								

1 row, 9 columns

Predict

# RESULT

Prediction results

Prediction type

Regression

Prediction distribution

Number of predictions

Prediction value

Display format for prediction results

☒ Table view

☐ JSON view

☒ Show input data

?

	Prediction	UDI	Product ID	Type	Air temperature
1	14.4	1	M17890	M	298.7
2					
3					
4					
5					
6					
7					
8					
9					
10					

# CONCLUSION

- In this project, a **predictive maintenance model** was successfully developed using IBM Watsonx.ai's **AutoAI** capabilities to classify machinery failures based on sensor data. By inputting various operational parameters (such as air temperature, torque, and rotational speed), the system was able to provide a prediction
- The model offers significant potential for **reducing machine downtime**, minimizing operational costs, and enabling **proactive maintenance scheduling**. The AutoAI pipeline efficiently automated the processes of model selection, training, and hyperparameter optimization, allowing for a robust and scalable solution.

# FUTURE SCOPE

- The system can eventually become a core component of **smart factories** and **Industry 4.0** solutions, where machines autonomously monitor their health and request maintenance, enabling near-zero downtime and optimized operational efficiency. the predictive maintenance system can be enhanced by integrating real-time IoT sensor data and historical maintenance records to improve accuracy.

# REFERENCES

Here is a sample **References** section tailored for your **Predictive Maintenance ML Project**, using AutoAI and IBM Cloud, with citations relevant to **machine learning**, **predictive maintenance**, and **data science best practices**:

1) IBM Watsonx.ai Documentation

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

2) 2) IBM cloud link - [IBM watsonx.ai Studio](#)

3) *Machine Predictive Maintenance Classification Dataset*, Kaggle-

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

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**THANK YOU**