
CAPSTONE PROJECT

PREDICTIVE MAINTENANCE OF INDUSTRIAL MACHINERY

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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

Example: 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

- Develop a machine learning-powered predictive maintenance system.
- Collect and analyze real-time sensor data (temperature, vibration, voltage, current).
- Detect early warning signs of three failure types:
 - Tool Wear
 - Heat Dissipation Issues
 - Power Failures
- Data Preprocessing:
 - Clean and preprocess the collected data to handle missing values, outliers, and inconsistencies.
 - Feature engineering to extract relevant features from the data that might impact bike demand.
- Machine Learning Algorithm:
 - Implement a machine learning algorithm, such as a time-series forecasting model (e.g., ARIMA, SARIMA, or LSTM), to predict bike counts based on historical patterns.
 - Consider incorporating other factors like weather conditions, day of the week, and special events to improve prediction accuracy.
- Deployment:
 - Develop a user-friendly interface or application that provides real-time predictions for bike counts at different hours.
 - Deploy the solution on a scalable and reliable platform, considering factors like server infrastructure, response time, and user accessibility.

SYSTEM APPROACH

The "System Approach" section outlines the overall strategy and methodology for developing and implementing the . “Predictive Maintenance of Industrial Machinery” 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 Selection:**
 - Random Forest Classifier(or SVM based on performance).
- **Data Input:**
 - Air temperature [K] , Process temperature [K], Torque [Nm] and many more..
- **Training Process:**
 - Supervised learning using labelled fault types..
- **Prediction Process:**
 - Model deployed on IBM Watson Studio with API endpoint for real-time predictions.

RESULT

IBM watsonx.ai Studio

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Projects / predictive_maintenance1 / Maintenance_ML

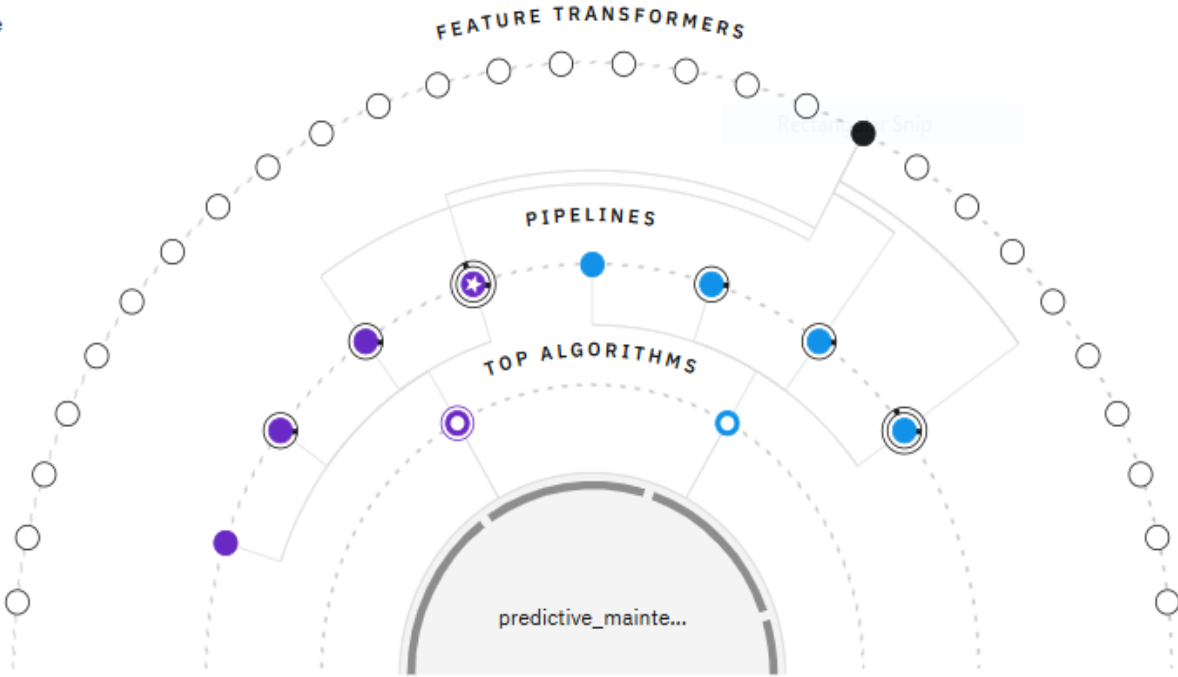
Experiment summary

Pipeline comparison

★ Rank by: Accuracy (Optimized) | Cross validation score


Relationship map ⓘ

Prediction column: Failure Type



Progress map

Swap view ↺



Experiment completed ✓

8 PIPELINES GENERATED

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

Time elapsed: 2 minutes

View log

Save code

Projects / predictive_maintenance1 / Maintenance_ML



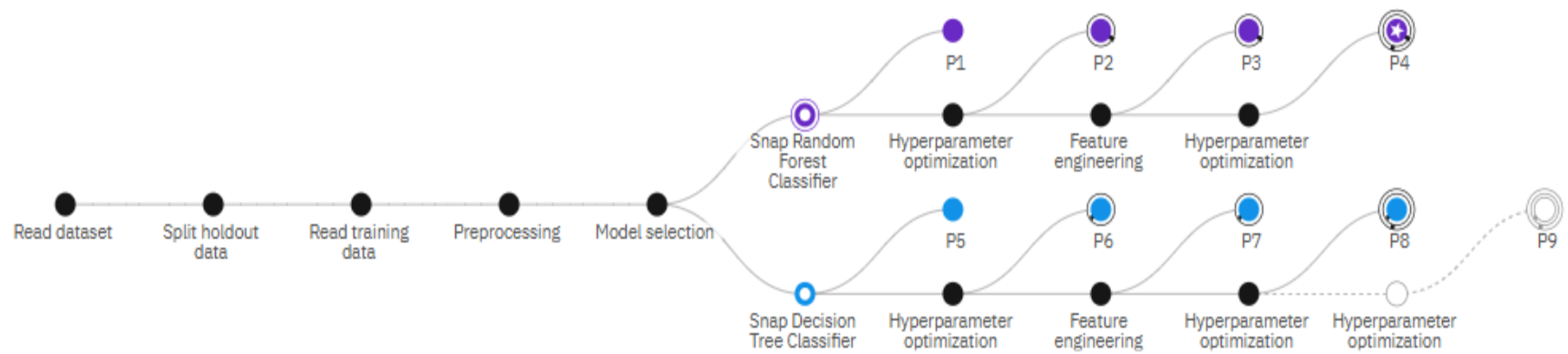
Experiment summary

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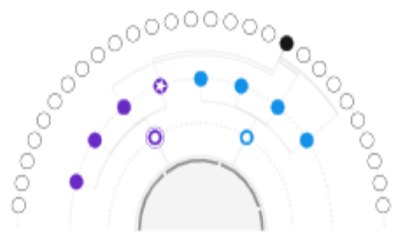
Progress map ⓘ

Prediction column: Failure Type



Relationship map

[Swap view](#)



Experiment completed ✓

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[View log](#)

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PIPELINE LEADERBOARD

Projects / predictive_maintenance1 / Maintenance_ML



Experiment summary

Pipeline comparison

★ Rank by: Accuracy (Optimized) | Cross validation score

Time elapsed: 2 minutes

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Pipeline leaderboard

	Rank ↑	Name	Algorithm	Specialization	Accuracy (Optimized) Cross Validation	Enhancements	Build time
★	1	Pipeline 4	● Snap Random Forest Classifier		0.995	HPO-1 FE HPO-2	00:00:41
	2	Pipeline 3	● Snap Random Forest Classifier		0.995	HPO-1 FE	00:00:33
	3	Pipeline 8	● Snap Decision Tree Classifier		0.994	HPO-1 FE HPO-2	00:00:27
	4	Pipeline 2	● Snap Random Forest Classifier		0.994	HPO-1	00:00:10

Prediction results

Prediction type

Multiclass classification

Prediction percentage



■ No Failure ■ Heat Dissipation Failure ■ Tool Wear Failure

Confidence level distribution

4

Display format for prediction results

☒ Table view ☐ JSON view

☐ Show input data ⓘ

	Prediction	Confidence
1	No Failure	100%
2	Heat Dissipation Failure	90%
3	Tool Wear Failure	90%
4	No Failure	100%
5		
6		
7		
8		
9		
10		
11		
12		
13		

Download JSON file

DATA INPUT

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Deployment spaces / Maintenance / P4 - Snap Random Forest Classifier: Maintenance_ML /

Maintenance 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)	Tool wear [min] (double)	Ta
1	1	M14860	M	298.1	308.6	1551	42.8	0	0
2	747	L47257	L	308.1	1289	1324	33.9	1	0
3	9577	L56756	L	299.6	310.7	1922	23.3	205	1
4	4410	M19269	M	302.4	310.1	1358	54.6	61	1
5	3944	L51123	L	302.4	311.4	1333	66.7	205	1
6									
7									

5 rows, 9 columns

Predict

OUTPUT

Prediction results

Display format for prediction results

☒ Table view ☐ JSON view

☒ Show input data ⓘ

	prediction	probability
1	No Failure	[0,1,0,0,0,0]
2	No Failure	[0.4,0.5,0,0,0,0.09999999999999998]
3	Tool Wear Failure	[0,0.1,0,0,0,0.9]
4	Heat Dissipation Failure	[0.7000000000000001,0.2,0.1,0,0,-2.220446049250313e-16]
5	Power Failure	[0.1,0,0,0.9,0,0]
6		
7		

CONCLUSION

The **predictive maintenance model** helps **anticipate machine failures** (tool wear, heat dissipation, power issues) before they occur.

FUTURE SCOPE

- **Integrate Deep Learning models** (LSTM, Transformers) for more accurate **time-series predictions** of failures.
- **Mobile Application Development** to provide real-time alerts and remote monitoring for operators.
- **Automated Maintenance Scheduling** – directly trigger work orders and manage spare part inventory.
- **Cloud Integration** for scalable data storage, model deployment, and analytics across multiple sites.
- **Cross-Industry Expansion** – adapt the solution for sectors like manufacturing, energy, transportation, and smart factories.

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