
CAPSTONE 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 is a **Predictive Maintenance Model** designed to forecast potential machinery failures before they occur. It leverages historical sensor data from industrial machines and utilizes machine learning algorithms to classify and predict failure types. The system ensures minimal downtime and proactive maintenance by alerting stakeholders about the type of machinery fault in advance.
- **Key Features:**
 - Reads and preprocesses training and test datasets.
 - Automatically selects and optimizes suitable models.
 - Applies advanced feature engineering and hyperparameter tuning.
 - Classifies failure types accurately based on learned patterns.

SYSTEM APPROACH

- **System Requirement : Window 10/11,8GB RAM**
- **Technology Used :IBM Cloud Lite Version**

ALGORITHM & DEPLOYMENT

Algorithm Used:

- Primary Models:

- Snap Random Forest Classifier
- Snap Decision Tree Classifier

- Optimization Techniques:

- Hyperparameter Optimization (multiple stages)
- Feature Engineering to enhance prediction accuracy

ALGORITHM & DEPLOYMENT

- The best-performing pipeline (P4) uses Random Forest Classifier with repeated tuning and feature engineering. It was **ranked highest based on optimized accuracy**.
- **Deployment:**
- Built using **IBM Watsonx.ai Studio**, which provides an automated machine learning (AutoAI) pipeline.
- 8 pipelines were generated and compared using **cross-validation** and accuracy metrics.
- The final optimized model can be **deployed as an API** or **integrated into manufacturing systems** for real-time failure prediction.

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RESULT

IBM watsonx.ai Studio

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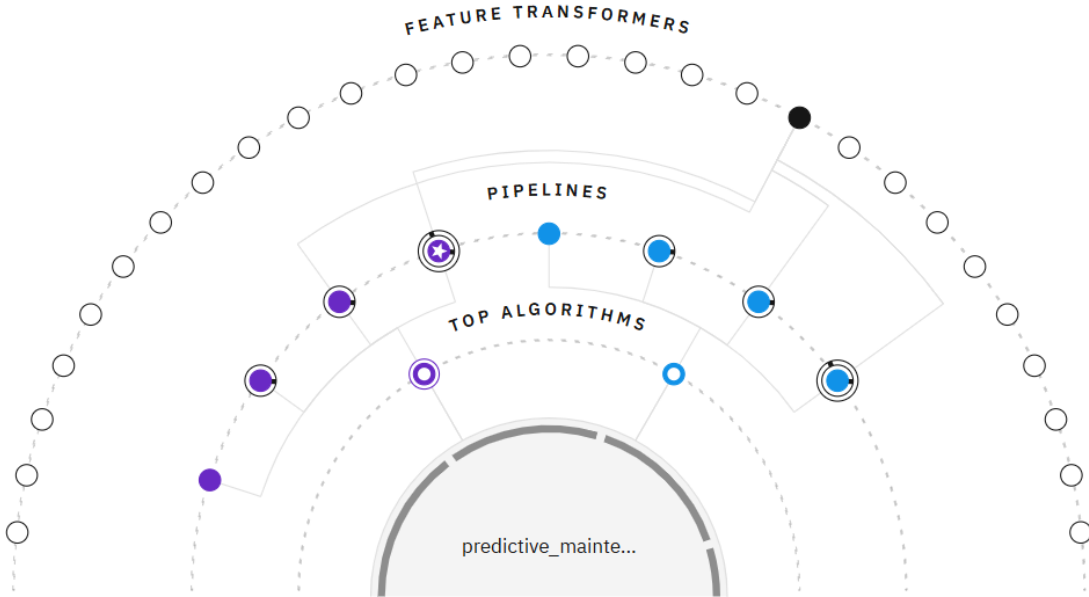
Projects / Industrial Machinery Predictor / Fault_predict

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: 3 minutes

[View log](#)

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Pipeline leaderboard ⌵

RESULT

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Progress map ⓘ

Prediction column: Failure Type

```
graph LR; A[Read dataset] --> B[Split holdout data]; B --> C[Read training data]; C --> D[Preprocessing]; D --> E[Model selection]; E --> F[Snap Random Forest Classifier]; E --> G[Snap Decision Tree Classifier]; F --> H[Hyperparameter optimization]; H --> I[Feature engineering]; I --> J[Hyperparameter optimization]; J --> K[P1]; K --> L[P2]; L --> M[P3]; M --> N[P4]; G --> O[Hyperparameter optimization]; O --> P[Feature engineering]; P --> Q[Hyperparameter optimization]; Q --> R[P5]; R --> S[P6]; S --> T[P7]; T --> U[P8]; U -.-> V[P9];
```

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

RESULT

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	Save as
	2	Pipeline 3	 Snap Random Forest Classifier		0.995	HPO-1 FE	00:00:32	
	3	Pipeline 8	 Snap Decision Tree Classifier		0.994	HPO-1 FE HPO-2	00:00:30	
	4	Pipeline 2	 Snap Random Forest Classifier		0.994	HPO-1	00:00:08	

RESULT

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

Deployment spaces / fault_dep / P4 - Snap Random Forest Classifier: Fault_predict

Deployments

Model details

Search

New deployment

Name	Type	Status	Tags	Last modified	
 dep2	Online	 Deployed		21 seconds ago Keerthana Muruganantham (You)	

Items per page: 20

1-1 of 1 items

1 of 1 pages

Name

P4 - Snap Random Forest Classifier: Fault_predict

Description

No description provided.

Asset Details

Type: wml-hybrid_0.1
Model ID: a6a98b3c-a71e-49...
Software specification: hybrid_0.1
Hybrid pipeline software specifications: autoai-kb_rt24.1-py3.11

Tags

Add tags to make assets easier to find.

Source asset details

Last modified
43 seconds ago by Keerthana Muruganantham
Created on
Aug 1, 2025 by Keerthana Muruganantham

RESULT

Prediction results

Prediction type

Multiclass classification

Prediction percentage



■ No Failure

Confidence level distribution

Display format for prediction results

☒ Table view ☐ JSON view

☒ Show input data ⓘ

	Prediction	Confidence
1	No Failure	100%
2		
3		
4		
5		
6		
7		
8		
9		
10		
11		

Download JSON file

CONCLUSION

- The predictive maintenance system effectively demonstrates how machine learning can revolutionize industrial operations by minimizing unplanned downtimes and optimizing asset life. Using IBM Watsonx's AutoAI, the model pipeline was automatically built, optimized, and evaluated within minutes, making it a scalable solution for the manufacturing industry. The system provides accurate fault classification and sets a solid foundation for more intelligent, connected, and proactive machinery maintenance in the future.

FUTURE SCOPE

- **Integration with IoT Devices:** Real-time sensor data from IoT-enabled machinery can be fed directly into the model.
- **Adaptive Learning:** Implement online learning techniques to allow the model to improve over time with new data.
- **Edge Deployment:** Optimize the model for edge devices to ensure predictions even in low-connectivity areas.
- **Explainability and Alerts:** Integrate tools like SHAP or LIME to explain predictions and trigger alerts for critical failures.
- **Dashboard Integration:** Build dashboards for monitoring machine health status visually using tools like IBM Cognos or Power BI.

REFERENCES

- **IBM watsonx.ai Studio Documentation**

IBM Cloud. *IBM watsonx.ai - Build, Train, and Deploy AI Models*.

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IBM Knowledge Center. *Understanding AutoAI pipelines and model evaluation*.

Available at:

<https://dataplatform.cloud.ibm.com/docs/content/wsj/autoai/>

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