

CAPSTONE PROJECT

PREDICTIVE MAINTENANCE OF INDUSTRIAL MACHINERY USING MACHINE LEARNING

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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

- Develop a machine learning-based predictive maintenance model that analyzes real-time sensor data (e.g., temperature, vibration, power) to classify potential failures (tool wear, heat dissipation, etc.). The system will predict failure types before they occur, enabling proactive maintenance to minimize downtime and costs. The model will be deployed for real-time monitoring, providing actionable alerts to prevent machine breakdowns
- Data Collection: Use the Kaggle dataset on power system faults
- Data Preprocessing: Clean and normalize sensor data.
- Model Training: Use algorithms like Random Forest, XGBoost, or LSTM (for time-series data) to classify failure types.
- Deployment: Integrate the model into a real-time monitoring system for proactive alerts.
- Evaluation: Validate the model using accuracy, precision, recall and F1-score

SYSTEM APPROACH

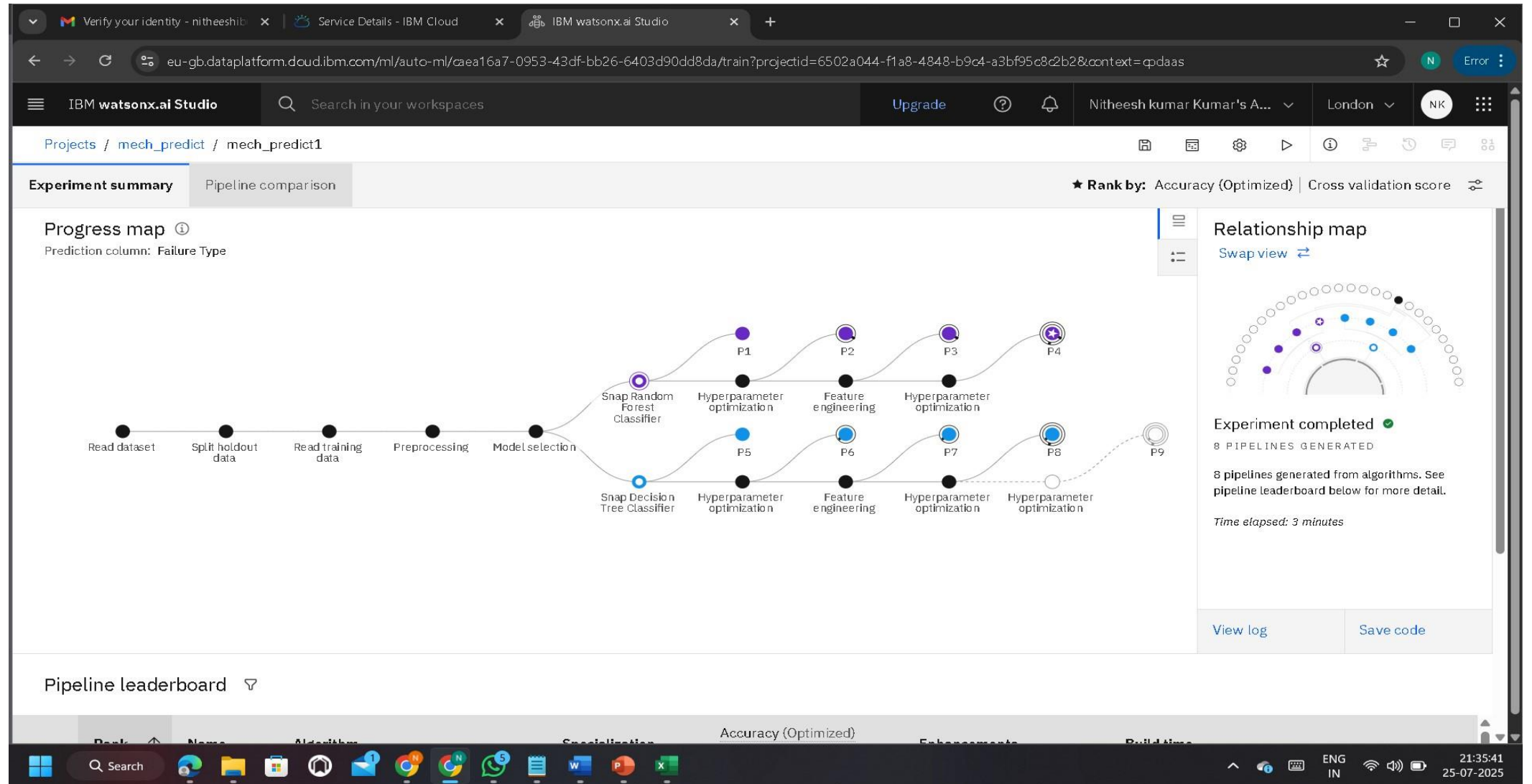
The "System Approach" section outlines the overall strategy and methodology for developing and implementing the Predictive Maintenance of Industrial Machinery. Here's a suggested structure for this section:

- System requirements :
 - IBM Cloud (mandatory).
 - IBM Watsonx.AI studio for model development and deployment.
 - IBM cloud object storage for dataset handling.
 - IBM Watsonx. Ai runtime for both software and hardware components.

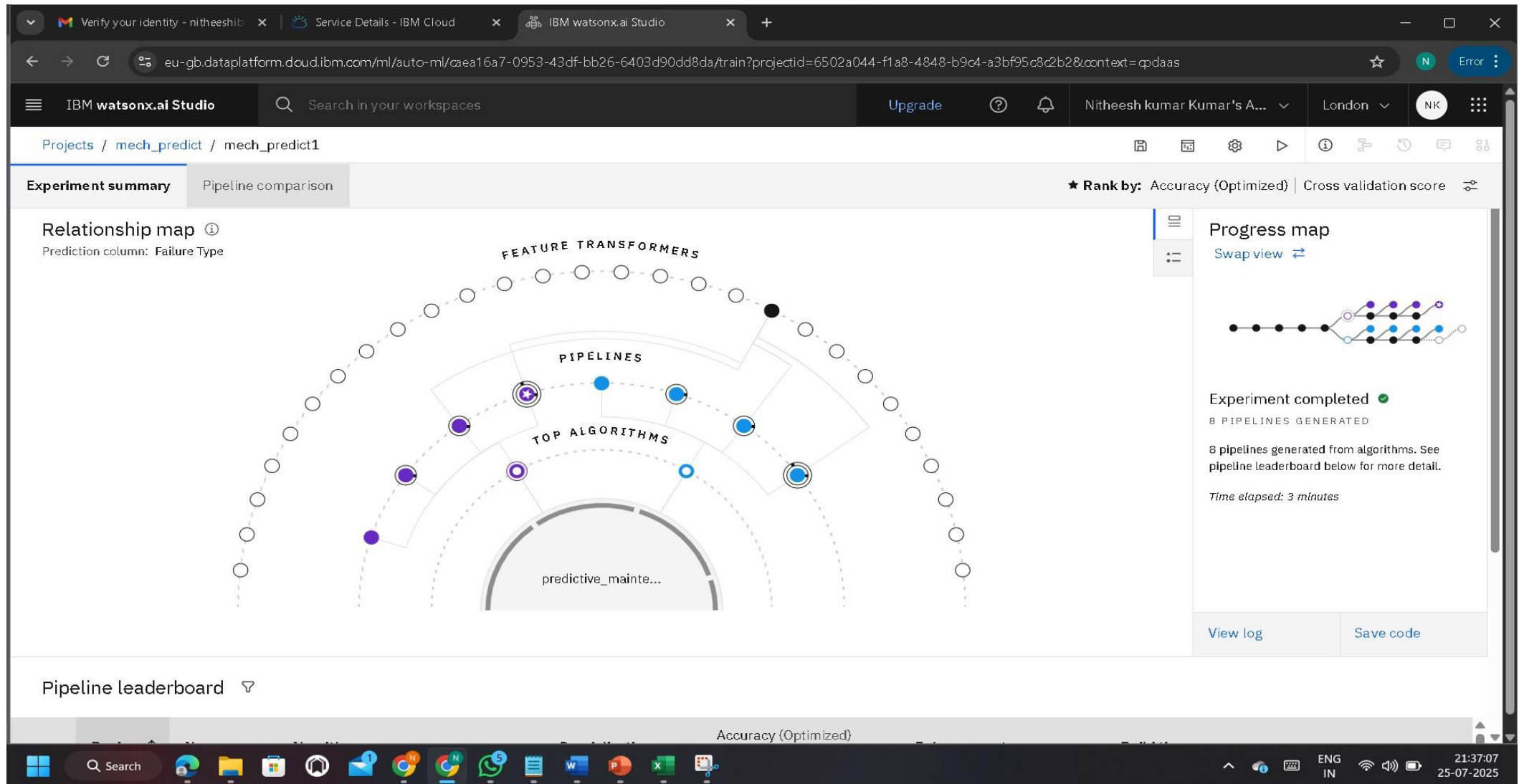
ALGORITHM & DEPLOYMENT

- **Algorithm Selection:**
 - Use Random Forest/XGBoost for tabular sensor data or LSTM for time-series patterns to classify failure types.
- **Data Input:**
 - Collect real-time sensor data (temperature, vibration, power) and historical maintenance logs with labeled failures.
- **Training Process:**
 - Preprocess data, extract features, train model on labeled failures, and validate using precision/recall metrics.
- **Prediction Process:**
 - Deploy model as REST API or edge service to predict failures in real-time and trigger maintenance alerts.

RESULT



RESULT



RESULT

IBM watsonx.ai Studio

Projects / mech_predict / mech_predict1

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

Time elapsed: 3 minutes

View log | Save code

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	00:00:42
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	00:00:08

RESULT

IBM watsonx.ai Studio

Deployment spaces / mech_predict_dep1 / P4 - Snap Random Forest Classifier: mech_predict1 /

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	[0,1,0,0,0,0]
3	No Failure	[0,1,0,0,0,0]
4	No Failure	[0,1,0,0,0,0]
5	No Failure	[0,1,0,0,0,0]
6	No Failure	[0,1,0,0,0,0]
7	No Failure	[0,1,0,0,0,0]
8	No Failure	[0,1,0,0,0,0]
9	No Failure	[0,1,0,0,0,0]
10	No Failure	[0,1,0,0,0,0]
11	No Failure	[0,1,0,0,0,0]
12	No Failure	[0,1,0,0,0,0]
13	No Failure	[0,1,0,0,0,0]

Download JSON file

22:04:35
25-07-2025

RESULT

IBM watsonx.ai Studio

Deployment spaces / mech_predict_dep1 / P4 - Snap Random Forest Classifier: mech_predict1 /

Prediction results

Display format for prediction results

☒ Table view ☐ JSON view

☐ Show input data ⓘ

	prediction	probability
69	No Failure	[0,1,0,0,0,0]
70	Power Failure	[0,0,0,1,0,0]
71	No Failure	[0,0.9998846590518952,0,0,0.00011534024961292744,6.984919309616089e-10]
72	No Failure	[0,1,0,0,0,0]
73	No Failure	[0,0.9988764047622681,0,0,0.001123595517128706,-2.793969500203275e-10]
74	No Failure	[0,0.9998846590518952,0,0,0.00011534024961292744,6.984919309616089e-10]
75	No Failure	[0,1,0,0,0,0]
76	No Failure	[0,1,0,0,0,0]
77	No Failure	[0,0.9997901380062104,0,0,0.00020986357703804971,-1.5832484212552345e-9]
78	Tool Wear Failure	[0,0,0,0,0,1]
79	No Failure	[0,1,0,0,0,0]
80	No Failure	[0,1,0,0,0,0]
81	No Failure	[0,1,0,0,0,0]
82	No Failure	[0,1,0,0,0,0]

Download JSON file

CONCLUSION

- This predictive maintenance system leverages machine learning to analyze real-time sensor data, accurately forecasting equipment failures before they occur. By deploying an optimized classification model (Random Forest/XGBoost/LSTM) through cloud or edge computing, it enables proactive maintenance—significantly reducing downtime, cutting costs, and enhancing operational efficiency in industrial settings. The solution's scalable architecture ensures continuous improvement through periodic model retraining, making it a robust, data-driven approach for modern industrial maintenance.

FUTURE SCOPE

- The system can be enhanced with IoT integration for real-time analysis, deep learning for better accuracy, and explainable AI (XAI) for transparency. Adding more failure modes, using digital twins for simulations, and implementing reinforcement learning for adaptive scheduling could further improve predictions and scalability.

REFERENCES

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

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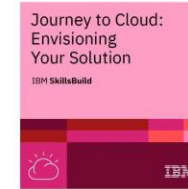
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THANK YOU