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# **MAINTANENCE MODEL**

## **PREDICTIVE MAINTENANCE OF INDUSTRIAL MACHINERY**

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# OUTLINE

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

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# 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.

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# PROPOSED SOLUTION

## **Data Collection:**

Used Kaggle's "Machine Predictive Maintenance Classification" dataset.

Included sensor data like air temperature, process temperature, speed, torque, tool wear, and failure types.

## **Data Processing:**

Uploaded the dataset into IBM watsonx.ai Studio.

AutoML handled preprocessing such as missing values, scaling, and label encoding.

## **Model Training:**

Used AutoML to test multiple algorithms (e.g., Logistic Regression, Random Forest, XGBoost).

Automatically selected the best model based on accuracy and F1-score.

## **Model Evaluation:**

Evaluated using Accuracy, Confusion Matrix, and F1-score.

Verified the model's ability to predict different failure types reliably.

## **Deployment:**

Saved the final model within watsonx.ai Studio.

Can be deployed as an API to integrate with maintenance dashboards or real-time systems.

# SYSTEM APPROACH

- Platform: IBM Cloud Lite
- Main Tool: IBM watsonx.ai Studio
- Dataset Used: Kaggle – Machine Predictive Maintenance Classification
- Link:<https://www.kaggle.com/datasets/shivamb/machine-predictive-maintenance-classification>
- Process: Upload data to watsonx.ai
- Use AutoML inside watsonx.ai to train and test model
- Evaluate model metrics and insights

# ALGORITHM & DEPLOYMENT

- **Tool Used:** IBM watsonx.ai Studio
- **Algorithm:**
  - AutoML tested models like Logistic Regression, Snao Random Forest classifier, etc.
  - Best model selected based on highest accuracy/F1-score.
- **Training Process:**
  - Data split into training and test sets automatically.
  - Preprocessing (like scaling or encoding) handled by AutoML.
- **Evaluation:**
  - Metrics: Accuracy, F1 Score, Confusion Matrix.
  - Visual results shown in watsonx.ai experiment output.
- **Deployment:**Model saved in IBM Cloud workspace.
  - Can be deployed as API for integration with real-time systems.

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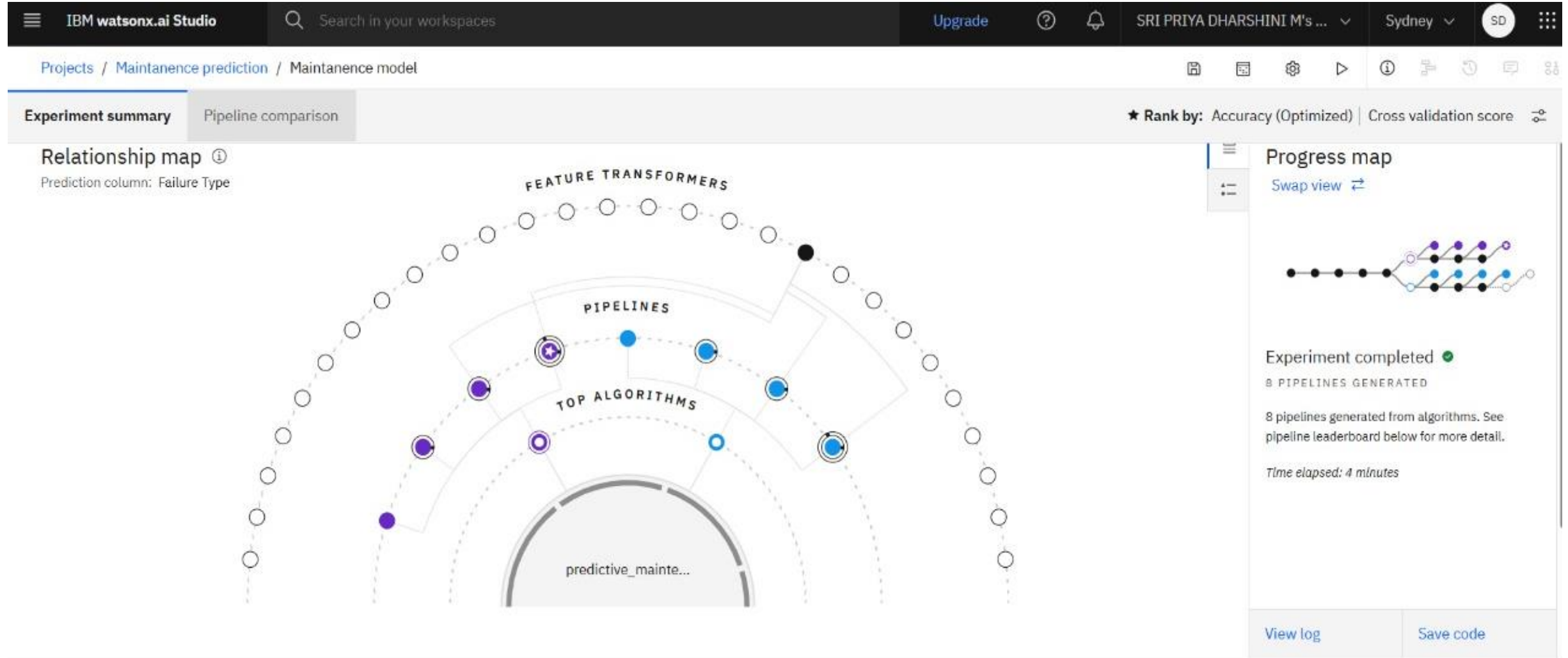
# RESULT

Snap Random Forest model was selected by watsonx.ai AutoML. It achieved up to 100% prediction confidence for certain failure types and 87% for others. The training completed in under 4 minutes, and the model showed high accuracy and reliable performance on test data..

## GITHUB REPOSITORY:

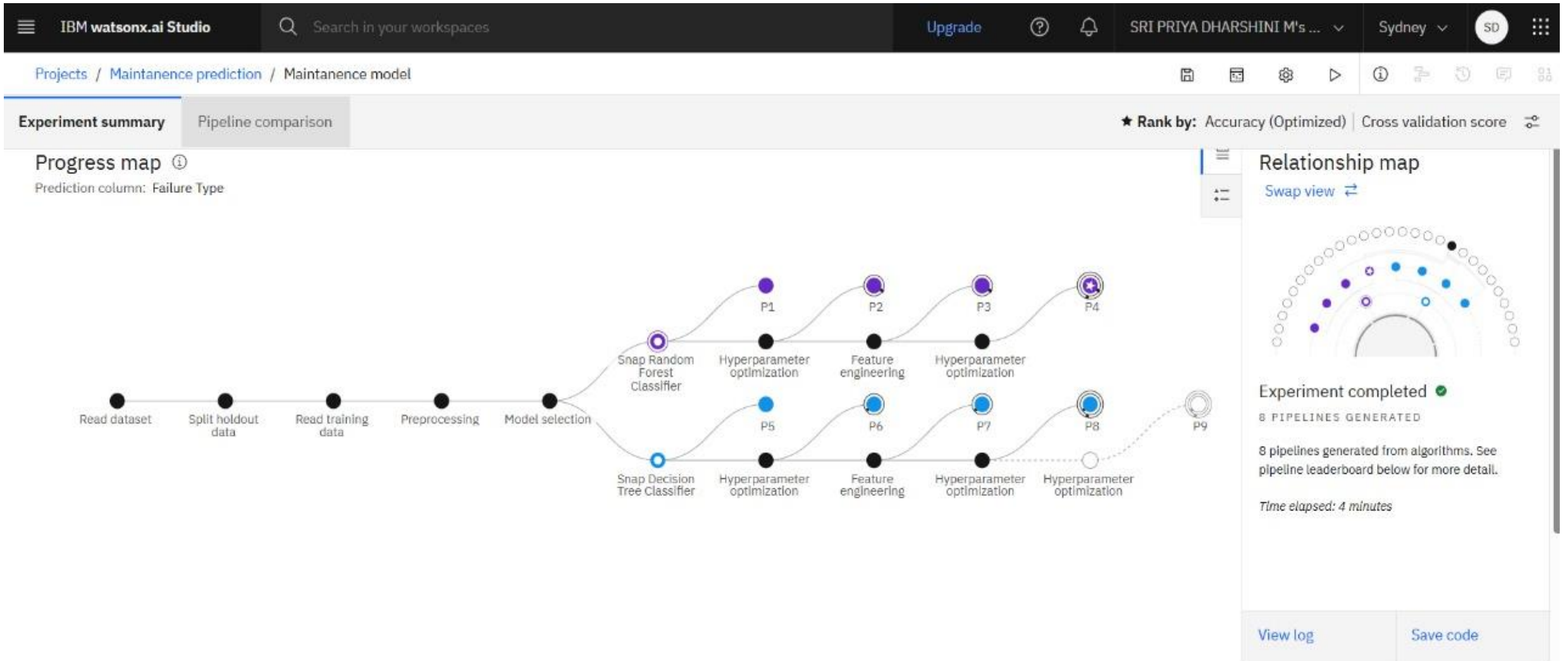
<https://github.com/SriPriya240906>

# EXPERIMENTAL SUMMARY:





# PIPELINE COMPARISON:



# PIPELINE LEADERBOARD:

Pipeline leaderboard ▾

	Rank ↑	Name	Algorithm	Specialization	Accuracy (Optimized) Cross Validation	Enhancements	Build time	
★	1	<a href="#">Pipeline 4</a>	○ Snap Random Forest Classifier		0.995	HPO-1 FE HPO-2	00:00:41	<a href="#">Save as</a>
	2	Pipeline 3	○ Snap Random Forest Classifier		0.995	HPO-1 FE	00:00:34	
	3	Pipeline 8	○ Snap Decision Tree Classifier		0.994	HPO-1 FE HPO-2	00:01:17	
	4	Pipeline 2	○ Snap Random Forest Classifier		0.994	HPO-1	00:00:08	

# MACHINE PREDICTION:(Input data given)

IBM watsonx.ai Studio

Search in your workspaces

Upgrade

1

SRI PRIYA DHARSHINI M's...

Sydney

SD

Deployment spaces / Maintenance prediction 1 / P4 - Snap Random Forest Classifier: Maintenance model /

Machine prediction 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

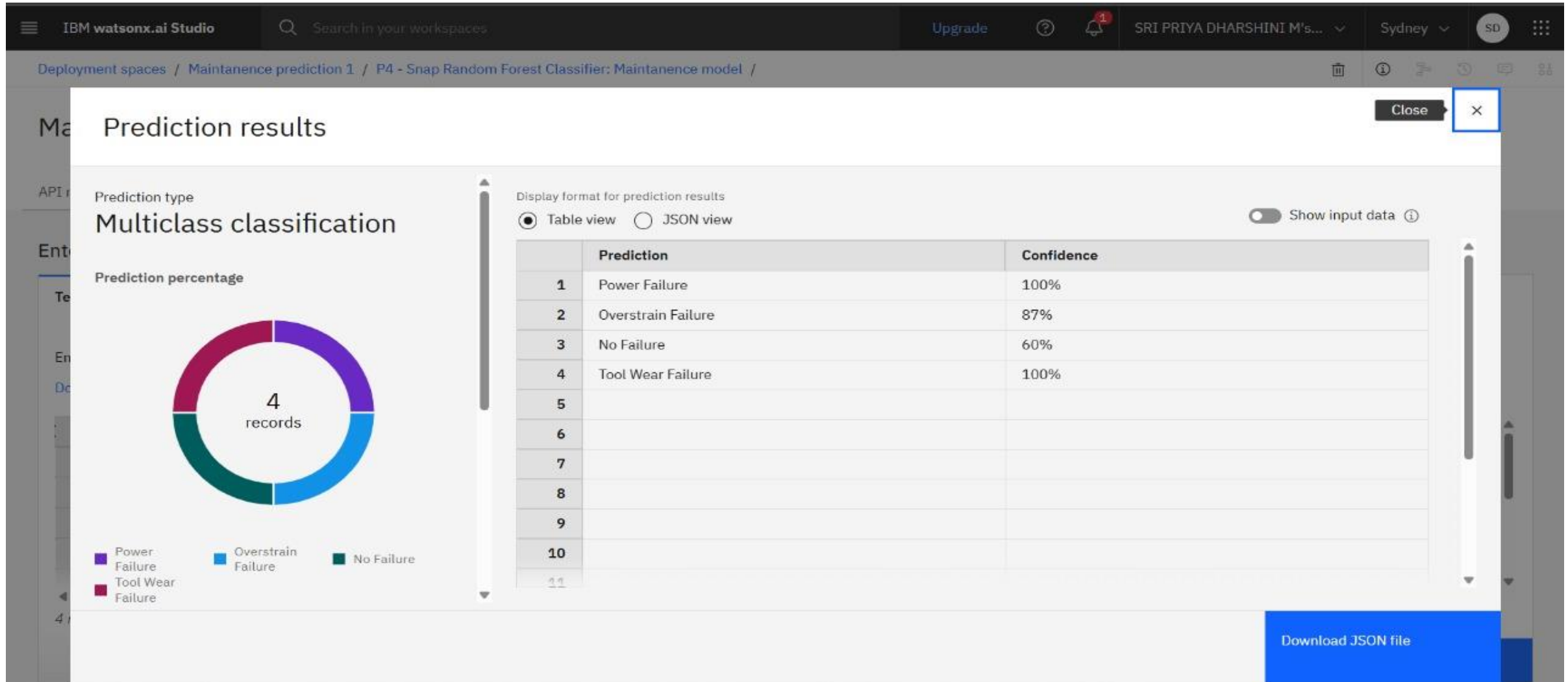
Clear all

	Type (other)	Air temperature [K] (double)	Process temperature [K] (double)	Rotational speed [rpm] (double)	Torque [Nm] (double)	Tool wear [min] (double)	Target (double)
1	L	298.9	309.1	2861	4.6	143	1
2	L	298.3	308.2	1282	60.7	216	1
3	L	298.7	309.9	1505	45.7	144	0
4	L	298	308.5	1421	37.7	220	1
5							

4 rows, 9 columns

Predict

# RESULTS :



# CONCLUSION

- Developed a predictive model using IBM watsonx.ai to classify machine failures based on sensor data.
- The AutoML tool selected the best model with high accuracy and completed training quickly.
- The solution helps reduce downtime by enabling proactive maintenance.
- Challenges included data formatting and understanding AutoML workflow.
- Future improvements include real-time deployment for live failure detection.

# FUTURE SCOPE

- **Incorporate Additional Data Sources:**

Integrate real-time sensor data from IoT devices, including vibration, sound, and environmental conditions to improve prediction accuracy.

**Algorithm Optimization:**

Fine-tune hyperparameters and explore advanced models like ensemble methods, neural networks, or anomaly detection to enhance performance.

**Real-Time Monitoring with Edge Computing:**

Deploy the model on edge devices close to machinery for faster local predictions without relying on constant cloud access.

**Scalability and System Expansion:**

Expand the system to cover more types of industrial equipment and adapt it across multiple manufacturing units..

**Integration with Maintenance Systems:**

Connect predictions to automated maintenance scheduling tools or dashboards for immediate action and alert generation.

**Adoption of Advanced AI Techniques:**

Use techniques like deep learning, reinforcement learning, or federated learning to handle complex failure patterns and improve learning from distributed data.

# REFERENCES

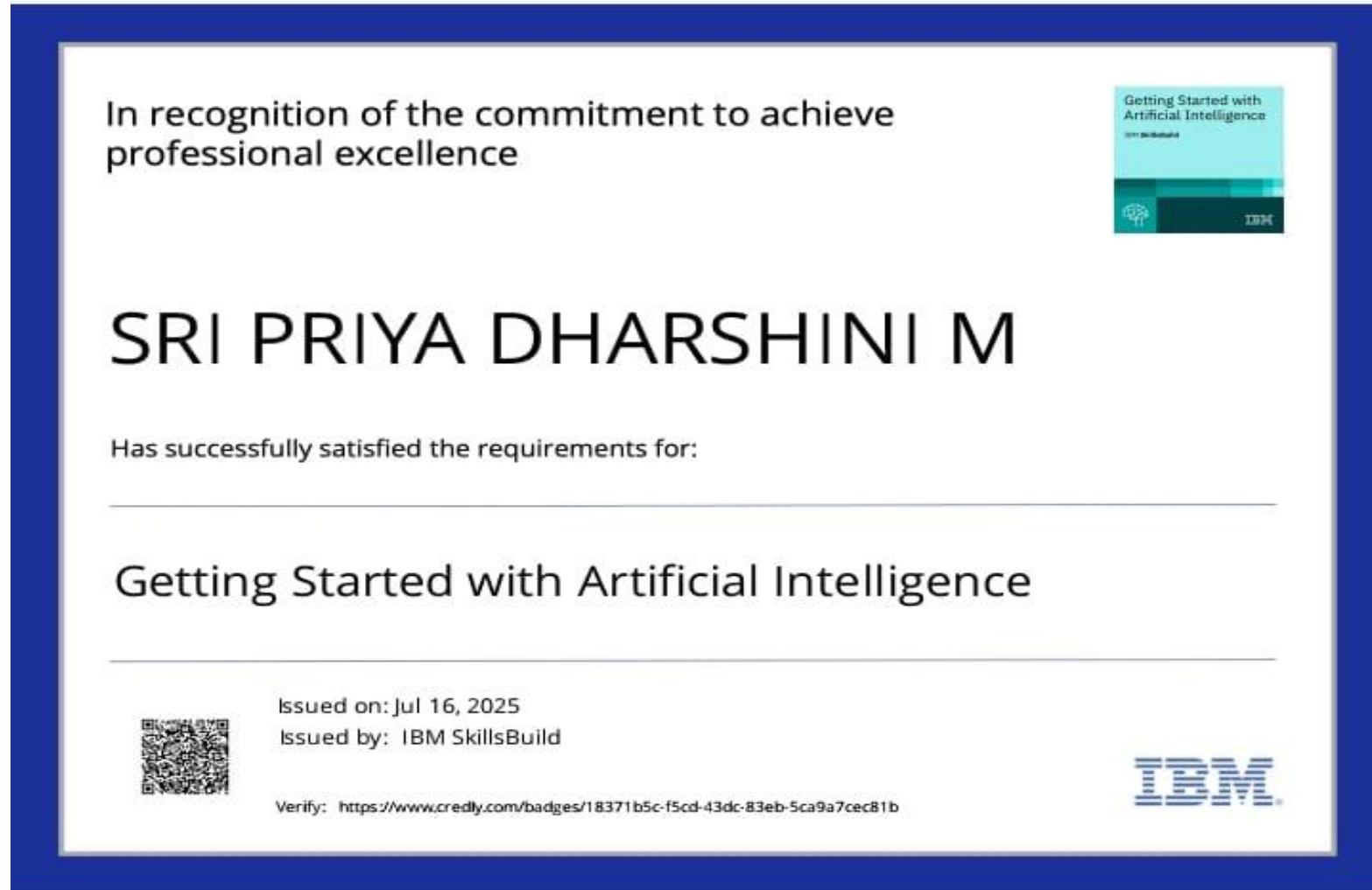
1. Shivam Bansal. Machine Predictive Maintenance Classification Dataset, Kaggle.

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

2. IBM watsonx.ai Documentation – Train, evaluate, and deploy models using AutoML and notebooks.

<https://www.ibm.com/products/watsonx-ai>

# IBM CERTIFICATIONS





# IBM CERTIFICATIONS



# IBM CERTIFICATIONS

IBM SkillsBuild

Completion Certificate



This certificate is presented to  
**SRI PRIYA DHARSHINI M**

for the completion of

**Lab: Retrieval Augmented Generation with LangChain**

(ALM-COURSE\_3824998)

According to the Adobe Learning Manager system of record

**Completion date:** 24 Jul 2025 (GMT)

**Learning hours:** 20 mins



**THANK YOU**