IBM PROJECT

► PROJECT TITLE -Problem statement No.39 – Predictive Maintenance of Industrial Machinery

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- **▶** Department- Computer Engineering.

OUTLINE

- Problem Statement
- Proposed System/Solution
- System Development Approach
- ► Algorithm & Deployment
- Result (Output Image)
- **▶** Conclusion
- **▶** Future Scope
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PROBLEM STATEMENT

Industrial machines are prone to unexpected failures such as tool wear, heat dissipation issues, and power failures. These failures cause unplanned downtime, leading to production delays and increased maintenance costs. Traditional maintenance schedules are reactive or time-based, lacking real-time failure prediction capabilities. There is a need to predict failures before they occur using sensor data and machine learning to enable proactive maintenance.

PROPOSED SOLUTION

- The solution to the problem of unexpected machine failures is a Machine Learning-based predictive maintenance model built and deployed using IBM Watson.ai Studio. Objective of the Solution:-
- Goal: Predict machine failure type in advance to enable preventive actions.
- Outcome: Reduce unplanned downtime, optimize maintenance schedules, and lower operational costs. Below is the step-by-step explanation
- ▶ 1) Data Collection: Utilize Kaggle's predictive maintenance dataset containing real sensor data from machines.
- ► Features include: Air temperature, process temperature, Rotational speed, torque, Tool wear time, Machine type and quality metrics. This data acts as input for the predictive model.
- **2) Data Preprocessing**: Data cleaning to remove missing and duplicate entries.
- Detect outliers that may affect predictions.
- Normalize features for consistent model training.
- ▶ Feature selection to identify variables influencing failures most strongly.

PROPOSED SOLUTION

- **3)Model Development in Watson.ai Studio:** Use Watson.ai Studio's environment for uploading and preparing data.
- Configure AutoAI pipelines to automatically select the best performing algorithm based on the dataset.
- Train the model to classify machine conditions: Tool wear failure, Heat dissipation failure, Power failure, Normal condition (no failure).
- ▶ **4)Deployment on Watson.ai Studio**: Deploy the trained model directly as a web service API from Watson.ai Studio.
- **5)Real-Time Prediction Process**: Live machine sensor data is sent to the deployed Watson.ai Studio API endpoint.
- The model returns a predicted failure category with associated probabilities. Alerts and recommendations are triggered automatically to maintenance teams.
- ▶ 6) Benefits of the Solution: Proactive Maintenance: Early detection of failures using AI predictions.
- Cost Efficiency: Minimized downtime and reduced emergency repair costs.
- Scalable Solution: Can be applied to multiple factories or machine types with minimal adjustments.

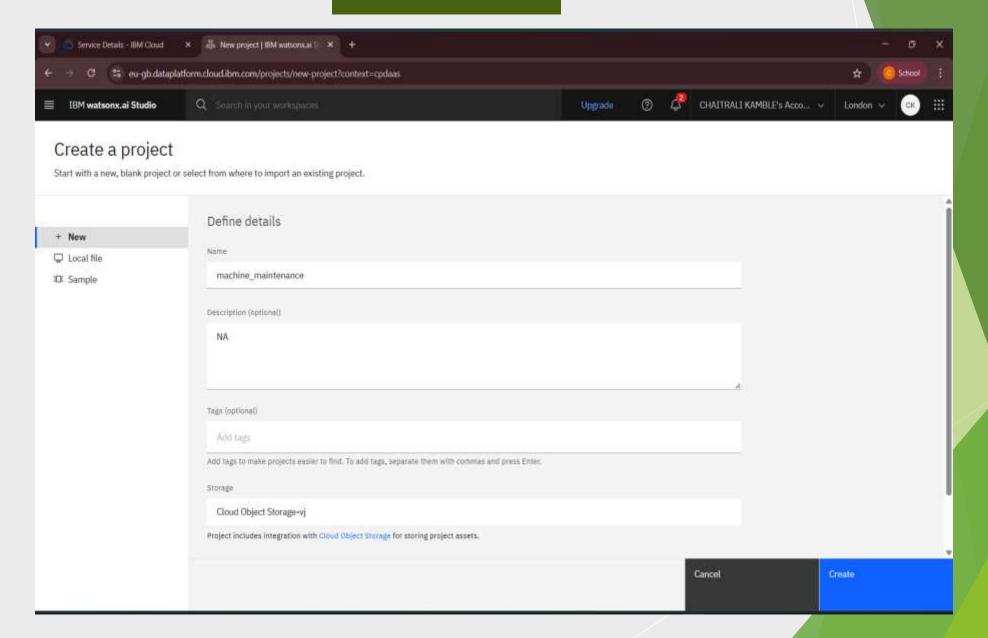
SYSTEM APPROACH

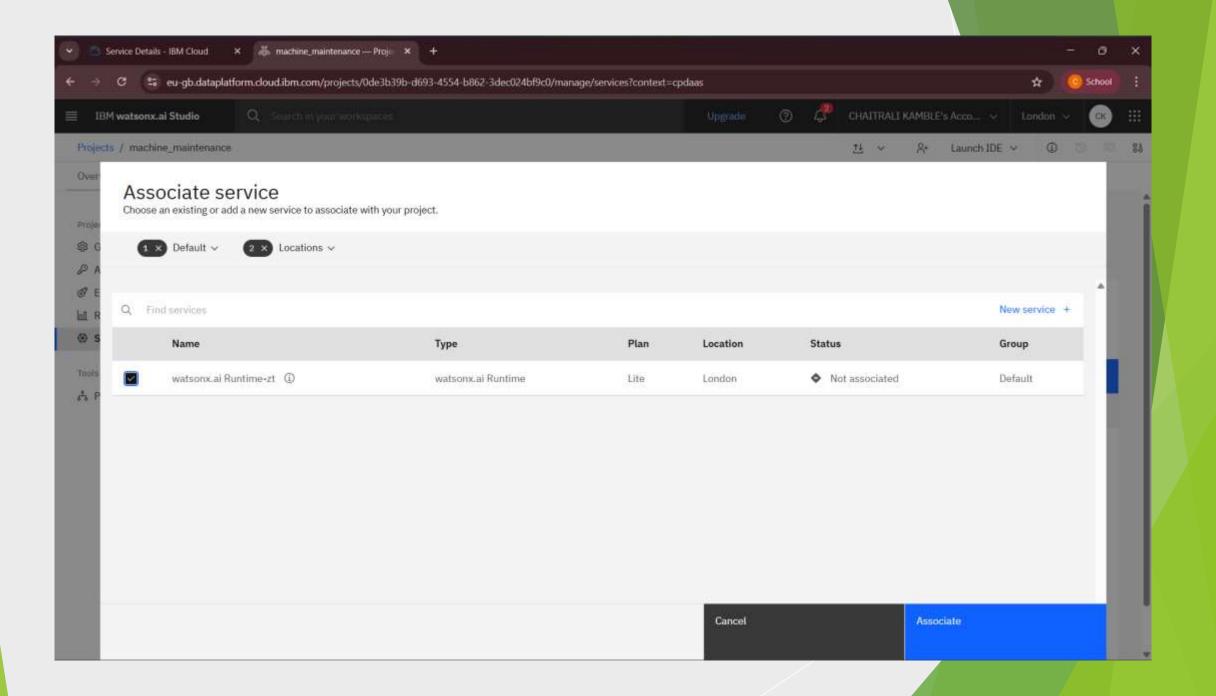
- The System Approach defines the methodology, resources, and processes followed to develop and deploy the predictive maintenance model.
- ▶ 1)System Requirements: Software: IBM Watson.ai Studio, Python libraries (Pandas, Scikit-learn, Matplotlib), IBM Watson Machine Learning service.
- ▶ Dataset: Kaggle predictive maintenance dataset with sensor readings.
- ► Cloud Platform: IBM Cloud Lite account for hosting and deployment.
- > 2) Data Flow Architecture: Data Acquisition: Sensor data collected from machines or historical datasets.
- Data Storage: Data uploaded to IBM Cloud Object Storage or Watson.ai Studio environment.
- Data Preprocessing: Cleaning, transformation, and feature engineering of raw data.
- Model Training: AI models developed in Watson.ai Studio using AutoAI.
- Model Deployment: Trained model deployed as a REST API service on IBM Watson Machine Learning.
- ▶ Real-Time Prediction: Live data fed to the deployed model, predicting failure types.
- ▶ Alert System: Automated notifications sent to maintenance teams for early action.

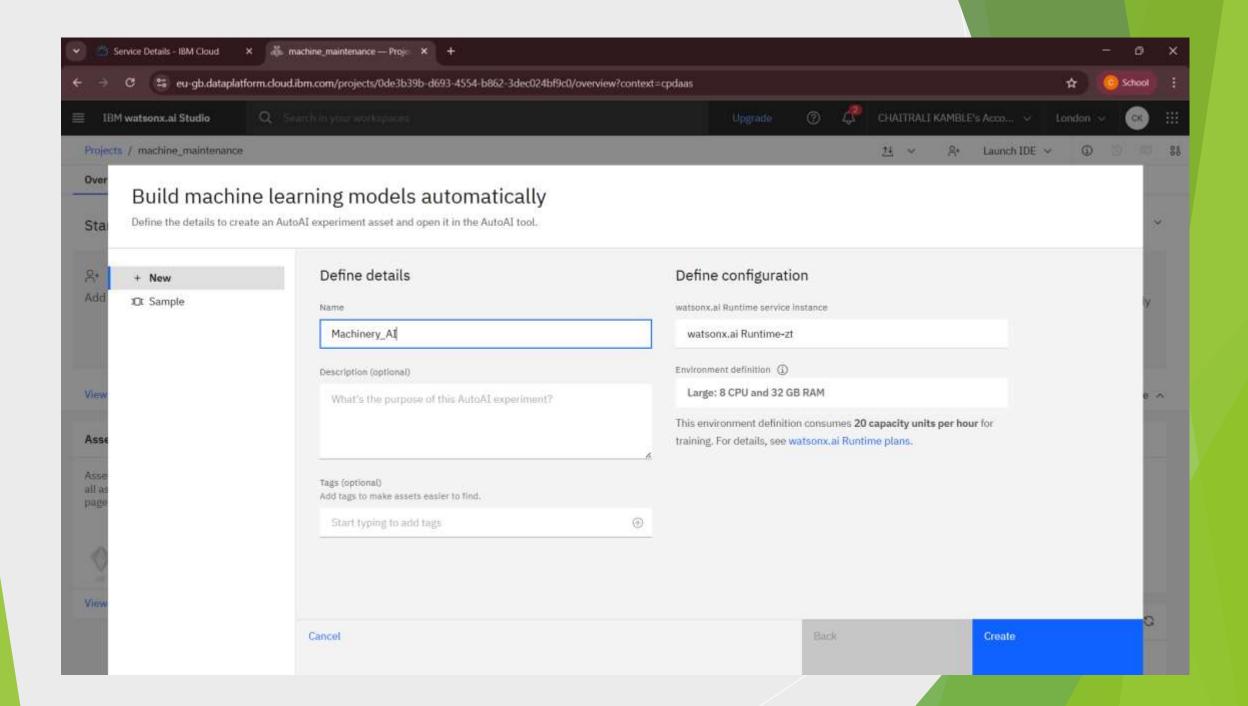
ALGORITHM & DEPLOYMENT

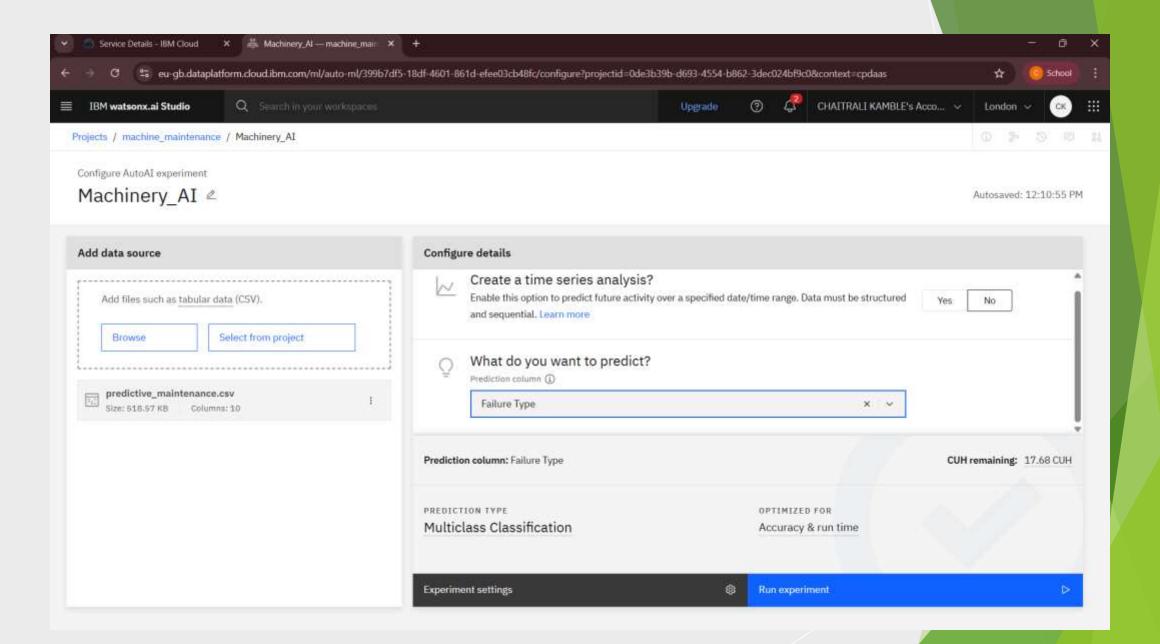
- Algorithm and deployment framework ensures a robust, scalable, and automated predictive maintenance system that can efficiently prevent unexpected machinery failures.
- ▶ **Algorithm Selection**: Random Forest Classifier for multi-class classification of failure types.
- Input Features: Air temperature, process temperature, rotational speed, torque, tool wear time, and quality measures.
- ► Training: Model trained on historical failure data, optimizing accuracy and precision.
- Prediction: Real-time sensor data fed into the model to predict failure type.
- **Deployment:** Model deployed on IBM Watson Machine Learning service, accessible via API for integration into factory systems.

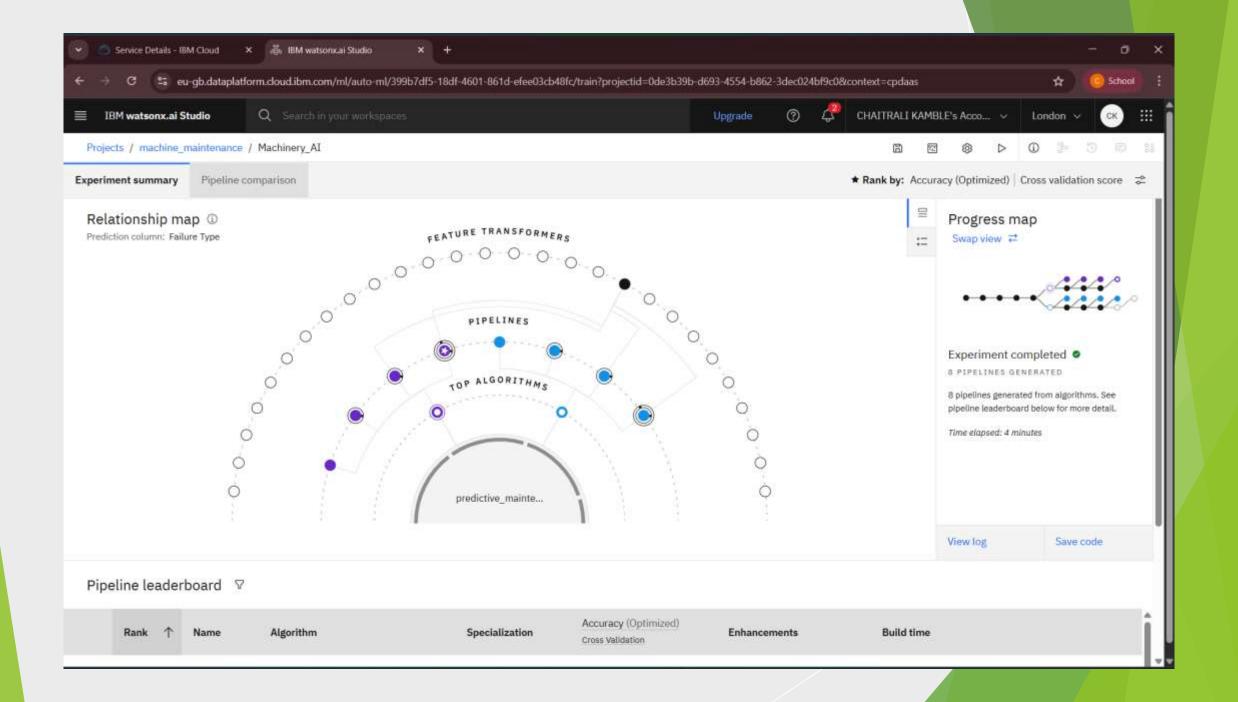


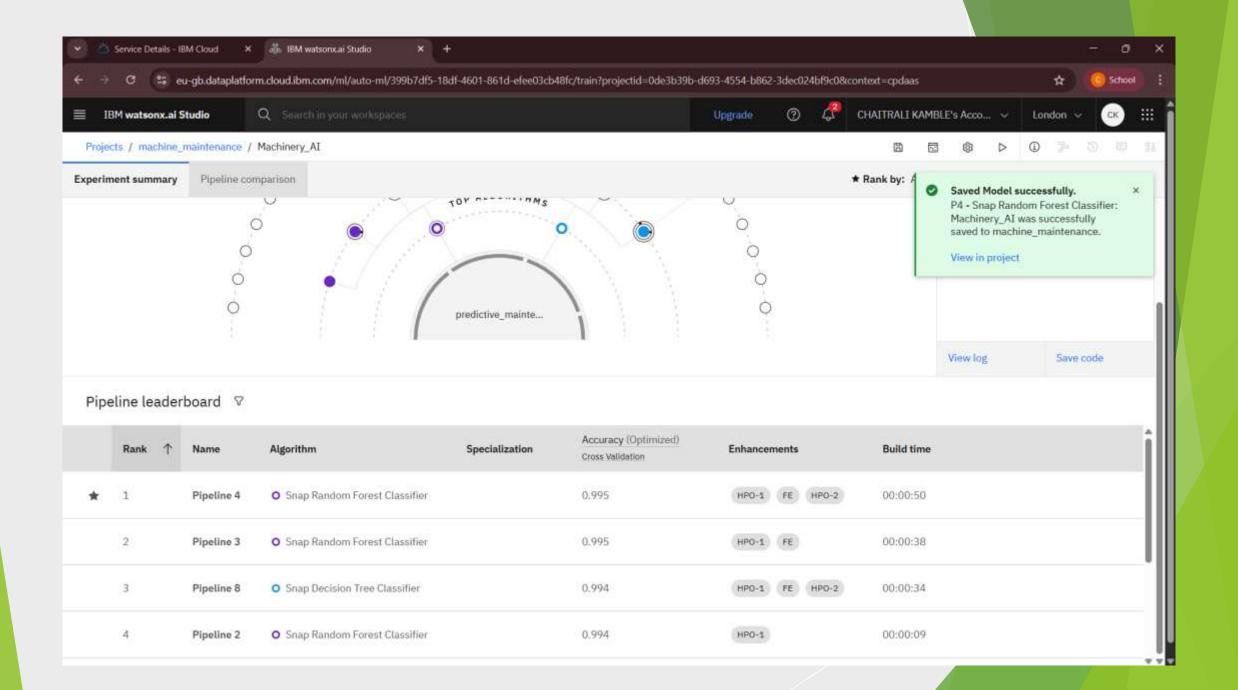


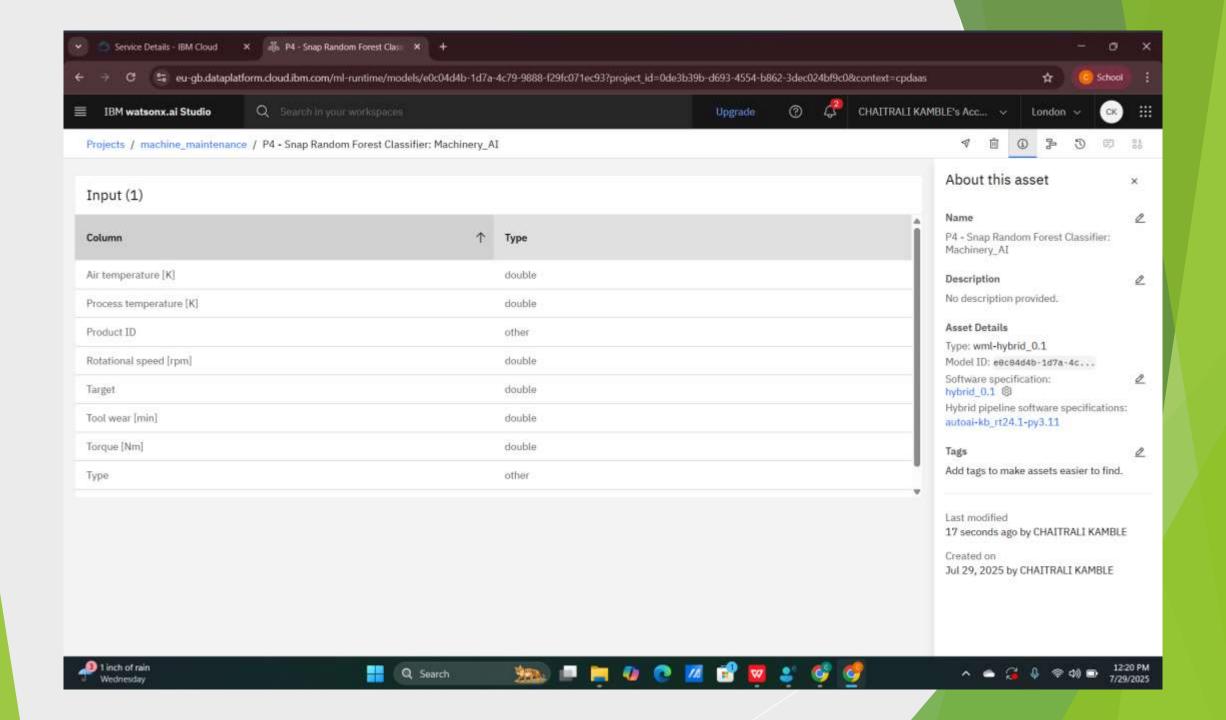


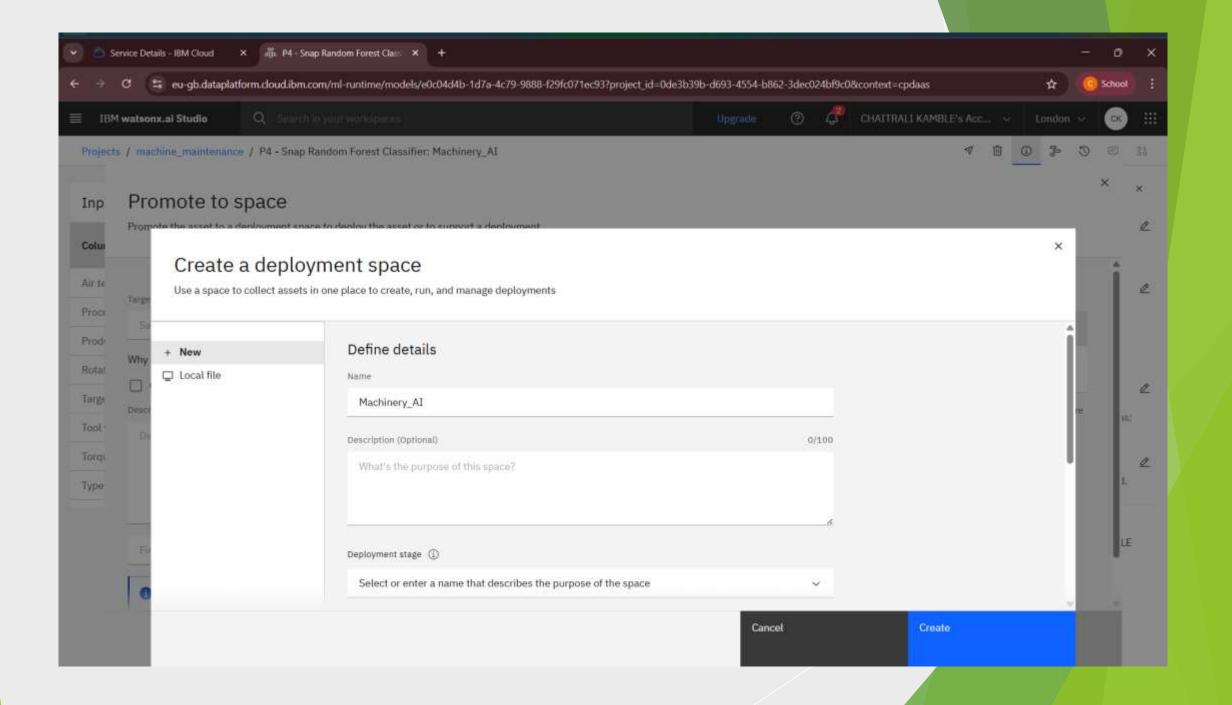


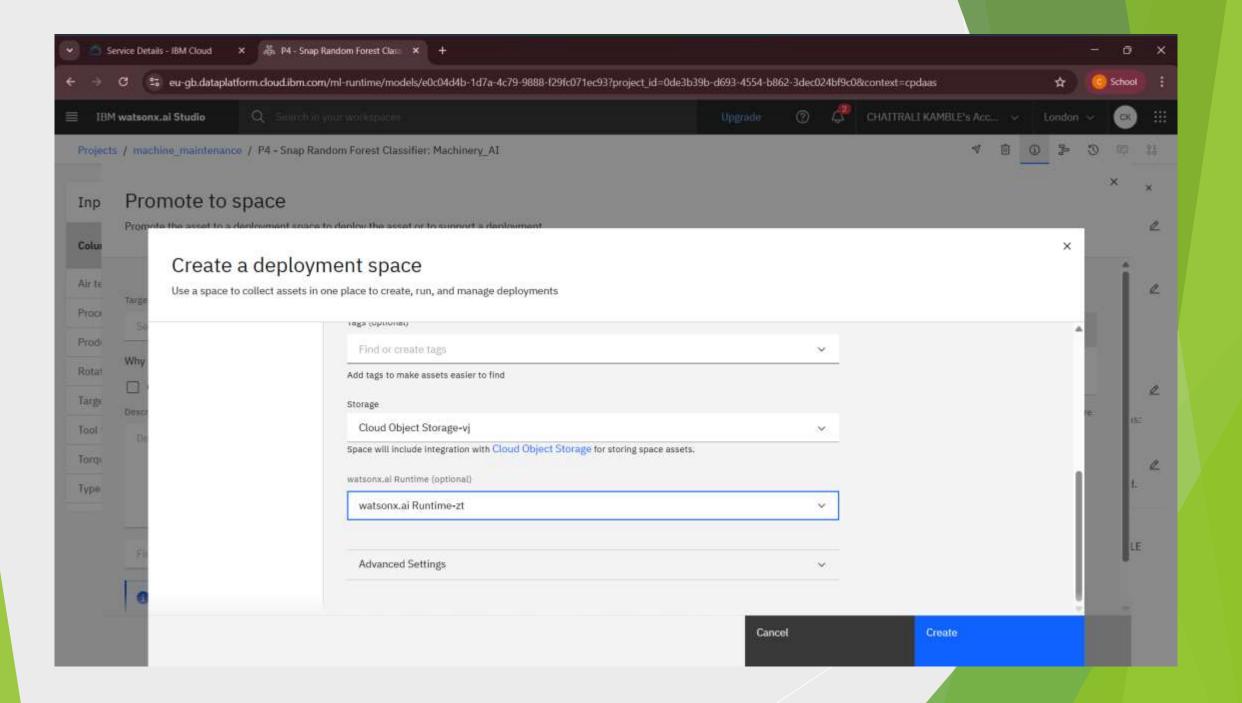


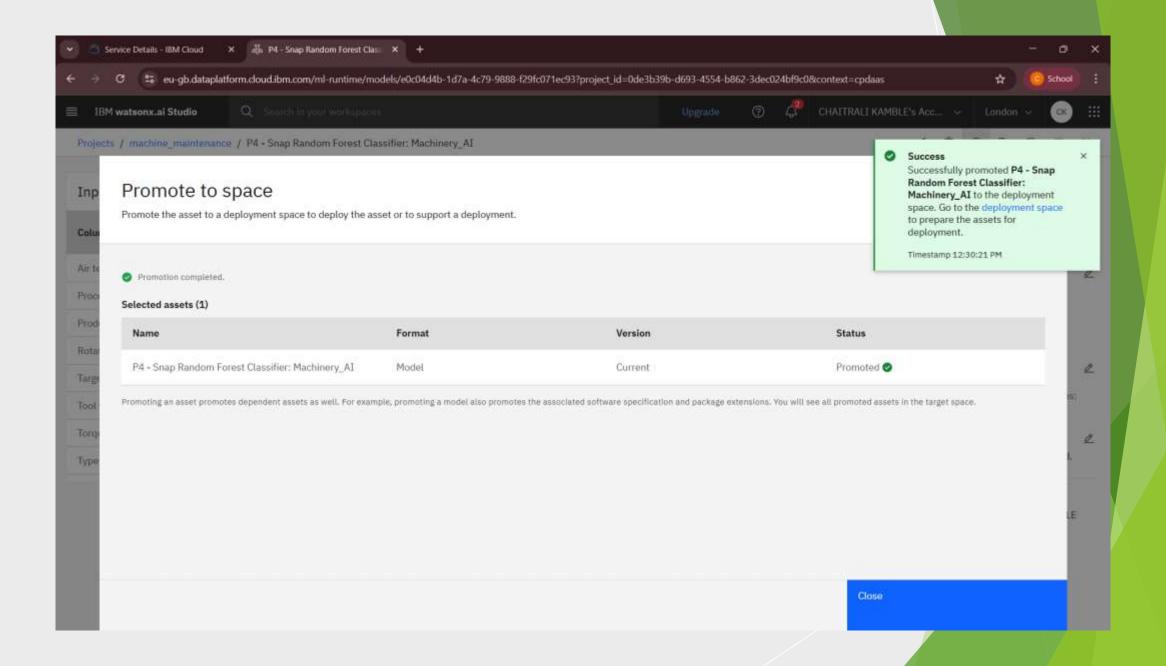


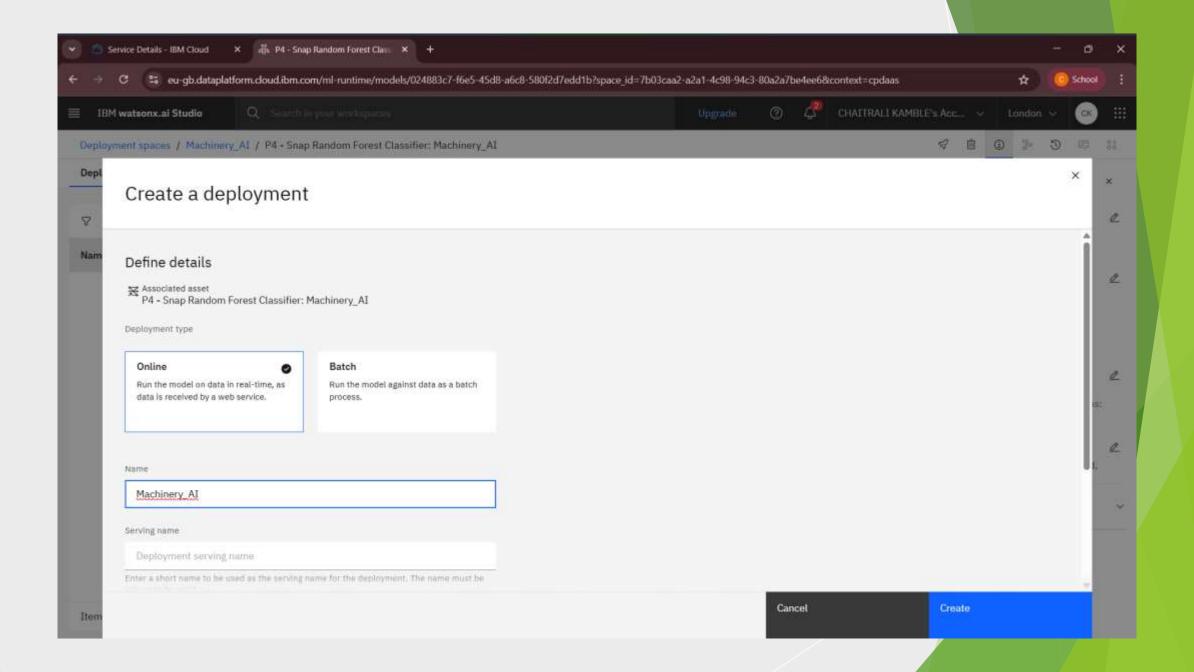


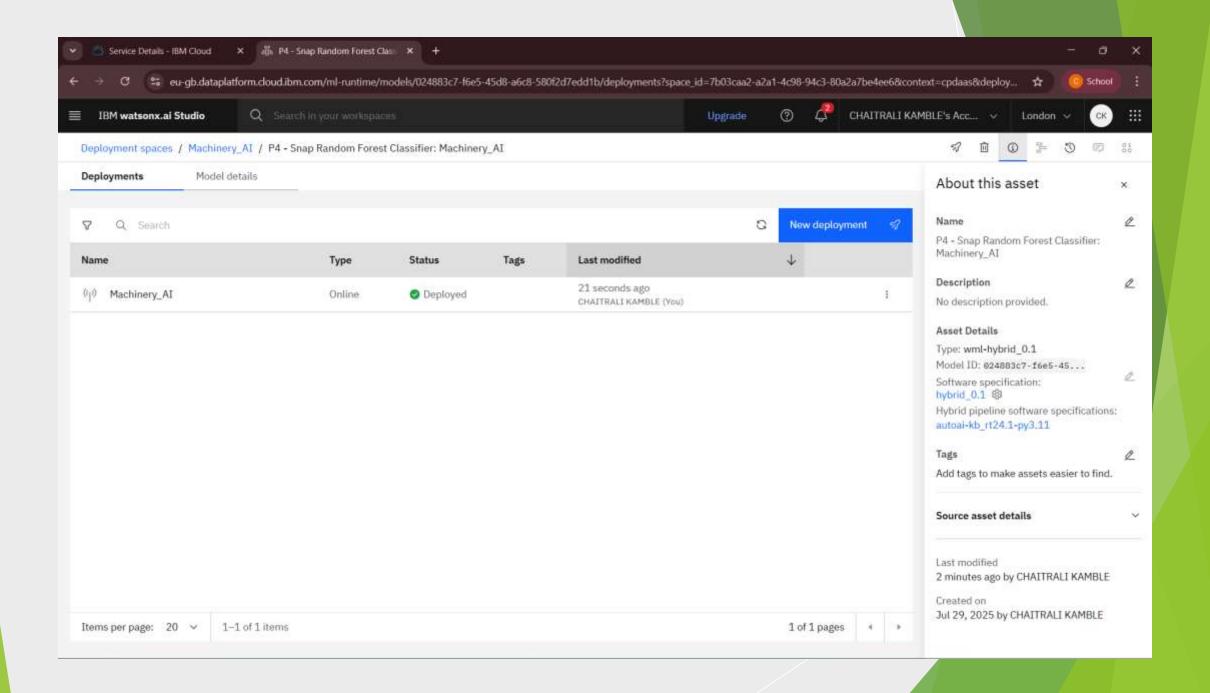


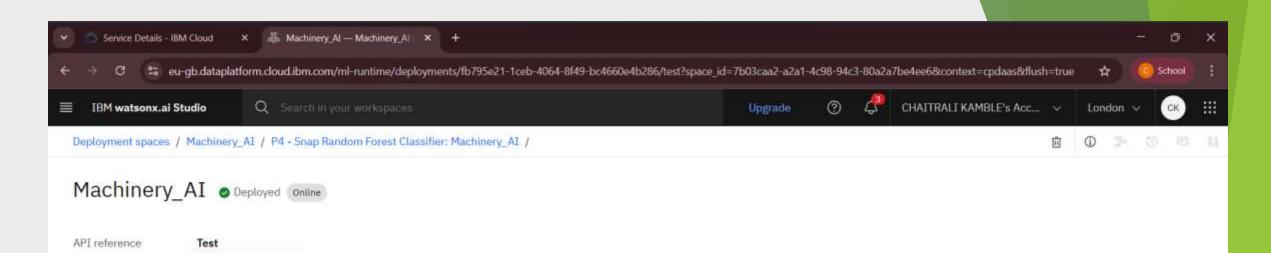












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Text JSON

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

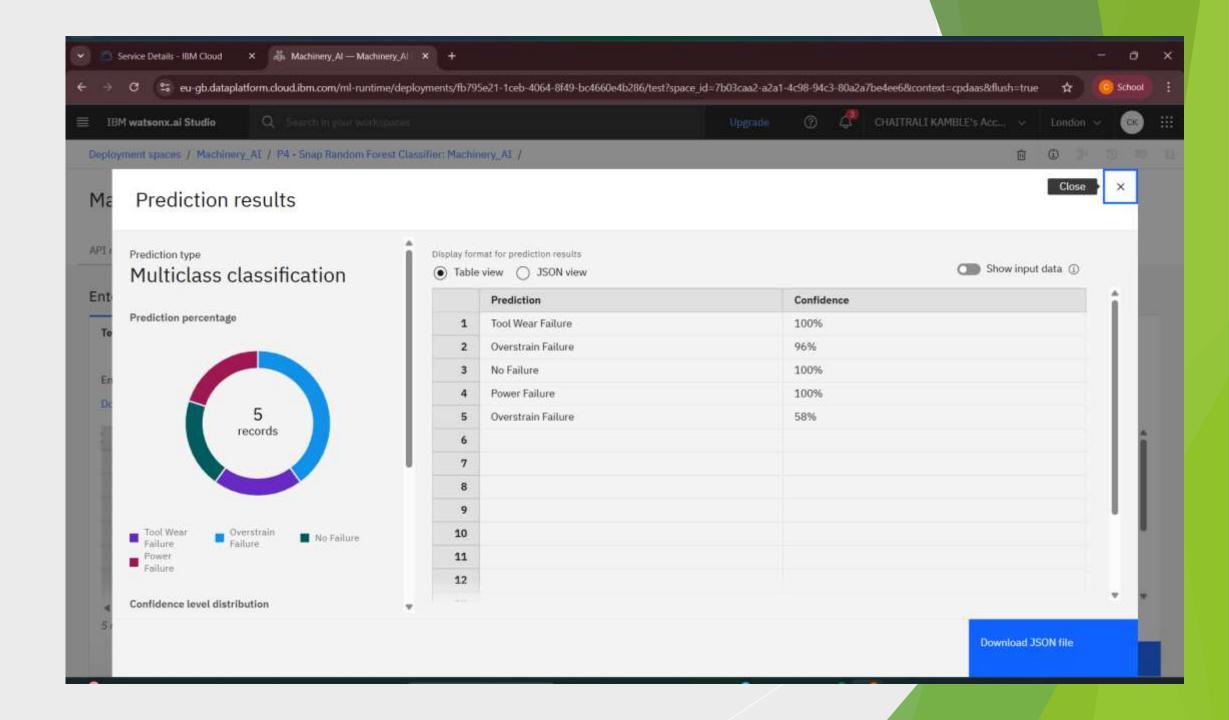
Download CSV template ₹ Browse local files 7 Search in space 7

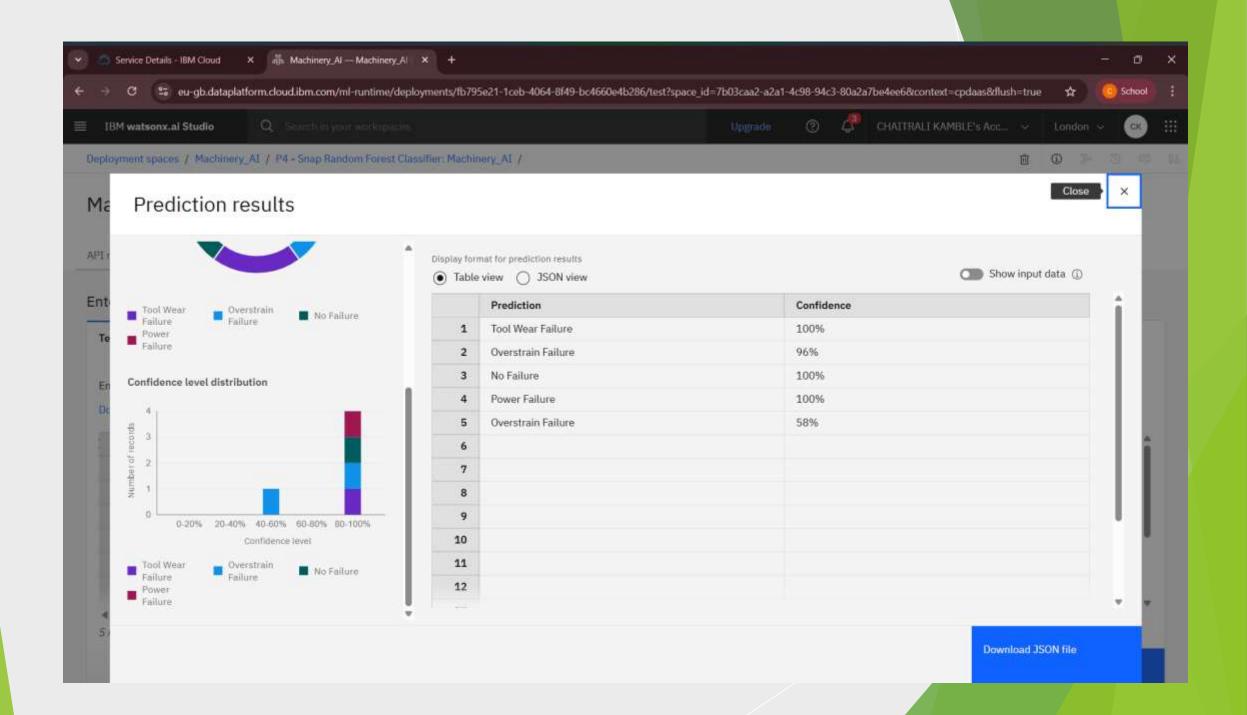
	UDI (double)	Product ID (other)	Type (other)	Air temperature [K] (double)	Process temperature [K] (double)	Rotational speed [rpm] (double)	Torque [Nm] (double)	Tool wea
1	78	L47257	L	300	310	1500	41	208
2	161	L47340	L	298.4	308.2	1282	60.7	216
3	182	H29595	Н	298.2	308.3	1824	24.2	49
4	169	L47348	L	298.4	308.3	1433	62.3	20
5	243	L47422	L	290	300	1300	55	200
6								

5 rows, 9 columns

Predict

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CONCLUSION

- Predictive maintenance models are essential for minimizing unexpected machine breakdowns.
- ▶ Use of machine learning and IBM Cloud services ensures scalability and real-time analysis.
- Reduces operational costs, increases machine lifespan, and improves production efficiency.

FUTURE SCOPE

- Integration with IoT sensors for real-time data streaming.
- Use of Deep Learning models for more accurate failure prediction.
- Expansion to multiple industrial sectors like manufacturing, automotive, aerospace.

REFERENCES

- Kaggle Dataset Machine Predictive Maintenance Classification: https://www.kaggle.com/datasets/shivamb/machinepredictive-maintenance-classification
- Watsonx.ai.Studio Service of IBM Cloud Lite.
- Python libraries Scikit-learn, Pandas, Matplotlib, Seaborn.

IBM CERTIFICATIONS

Screenshot/ credly certificate(getting started with Al)



Screenshot/ credly certificate (Journey to Cloud)

In recognition of the commitment to achieve professional excellence



CHAITRALI KAMBLE

Has successfully satisfied the requirements for:

Journey to Cloud: Envisioning Your Solution



Issued on: Jul 16, 2025 Issued by: IBM SkillsBuild





Screenshot/ credly certificate(RAG Lab)

IBM SkillsBuild

Completion Certificate



This certificate is presented to

CHAITRALI KAMBLE

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