
CAPSTONE PROJECT

PREDICTIVE MAINTENANCE OF INDUSTRIAL MACHINERY

Presented By:

Student Name- Sakshi Punjaji Patil

PRN :- 202301070173

College Name- MIT Academy Of Engineering, Alandi

Department – Electronics and Telecommunication

OUTLINE

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

PROBLEM STATEMENT

- The goal is to develop a predictive maintenance model for a fleet of industrial machines to **anticipate failures before they occur**.
- The system must analyze **sensor data from machinery** to identify patterns preceding failures.
- The model should predict the **type of failure** (e.g., tool wear, heat dissipation failure, power failure) based on **real-time operational data**.
- The solution aims to support **proactive maintenance**, reduce downtime, and lower operational costs.

PROPOSED SOLUTION

The proposed system aims to address the challenge of predicting machine failures in industrial machinery. This involves leveraging data analytics and machine learning techniques to identify failure patterns and enable proactive maintenance. The solution consists of the following components:

- ◆ **Data Collection:**

- Gather historical sensor data from machinery, including temperature, rotational speed, torque, and tool wear.

- Store and access the data securely using **IBM Cloud Object Storage**.

- ◆ **Data Preprocessing:**

- Clean and preprocess the collected data to handle missing values, outliers, and inconsistencies.

- Perform feature engineering to create new variables indicating failure conditions.

- ◆ **Machine Learning Algorithm:**

- Implement a **Random Forest Classifier** with class weighting to handle imbalanced failure classes.

- Use **SMOTE (Synthetic Minority Oversampling Technique)** to balance minority failure categories.

- Train the model on processed features and validate its performance.

- ◆ **Deployment:**

- Host the trained model on **IBM Watson Machine Learning** for scalable predictions.

- Integrate with real-time data pipelines to enable continuous monitoring and prediction.

- ◆ **Evaluation:**

- Evaluate the model using classification metrics such as precision, recall, and F1-score.

- Analyze feature importance to understand critical factors influencing failures.

SYSTEM APPROACH

Languages & Libraries:

- Python
- Pandas, NumPy
- scikit-learn
- Matplotlib, Seaborn

•Cloud Services:

- IBM Cloud Object Storage
- IBM watsonx.ai Studio
- IBM Watson Machine Learning

•Machine Learning Techniques:

- Random Forest Classifier
- Class Weight Balancing
- SMOTE Oversampling for minority classes

ALGORITHM & DEPLOYMENT

- **Algorithm:**

- Random Forest Classifier:**

- Ensemble of decision trees

- Balanced class weights to mitigate imbalance

- Feature importances computed

- SMOTE:**

- Synthetic Minority Oversampling Technique to balance class distribution before training

- **Deployment:**

- Model trained in **IBM watsonx.ai Studio**

- Data stored in **IBM Cloud Object Storage**

- Model hosted via **IBM Watson Machine Learning API**

RESULT


- **Classification Report Highlights:**


- High accuracy across most failure types

- Effective handling of class imbalance



- Key features contributing most to prediction: Torque, Tool Wear, Rotational Speed

RESULT IMAGES

 IBM watsonx.ai Studio


 Search in your workspaces

Upgrade







 

Sakshi Patil's Account ▾

Sydney ▾

SP 

Projects / PM AI / Predictive_Maintenance

✓ Dataset loaded successfully from Google Drive!

Columns in the dataset:
['UDI', 'Product ID', 'Type', 'Air temperature [K]', 'Process temperature [K]', 'Rotational speed [rpm]', 'Torque [Nm]', 'Tool wear [min]', 'Target', 'Failure Type']

✓ 'FailureType' column already exists.

✓ Data cleaning and target variable setup complete.
✓ Data split into training and testing sets.
✎ Applying SMOTE to balance classes...

🔍 Class distribution BEFORE SMOTE:

FailureType	
No Failure	7722
Heat Dissipation Failure	90
Power Failure	76
Overstrain Failure	62
Tool Wear Failure	36
Random Failures	14

Name: count, dtype: int64

✓ Class distribution AFTER SMOTE:

FailureType	
No Failure	7722
Tool Wear Failure	7722
Random Failures	7722
Power Failure	7722
Heat Dissipation Failure	7722
Overstrain Failure	7722

Name: count, dtype: int64

✎ Training RandomForestClassifier with balanced class weights...

✓ Model training complete.

RESULT IMAGES

Projects / PM AI / Predictive_Maintenance

```
--- Classification Report ---
              precision    recall  f1-score   support

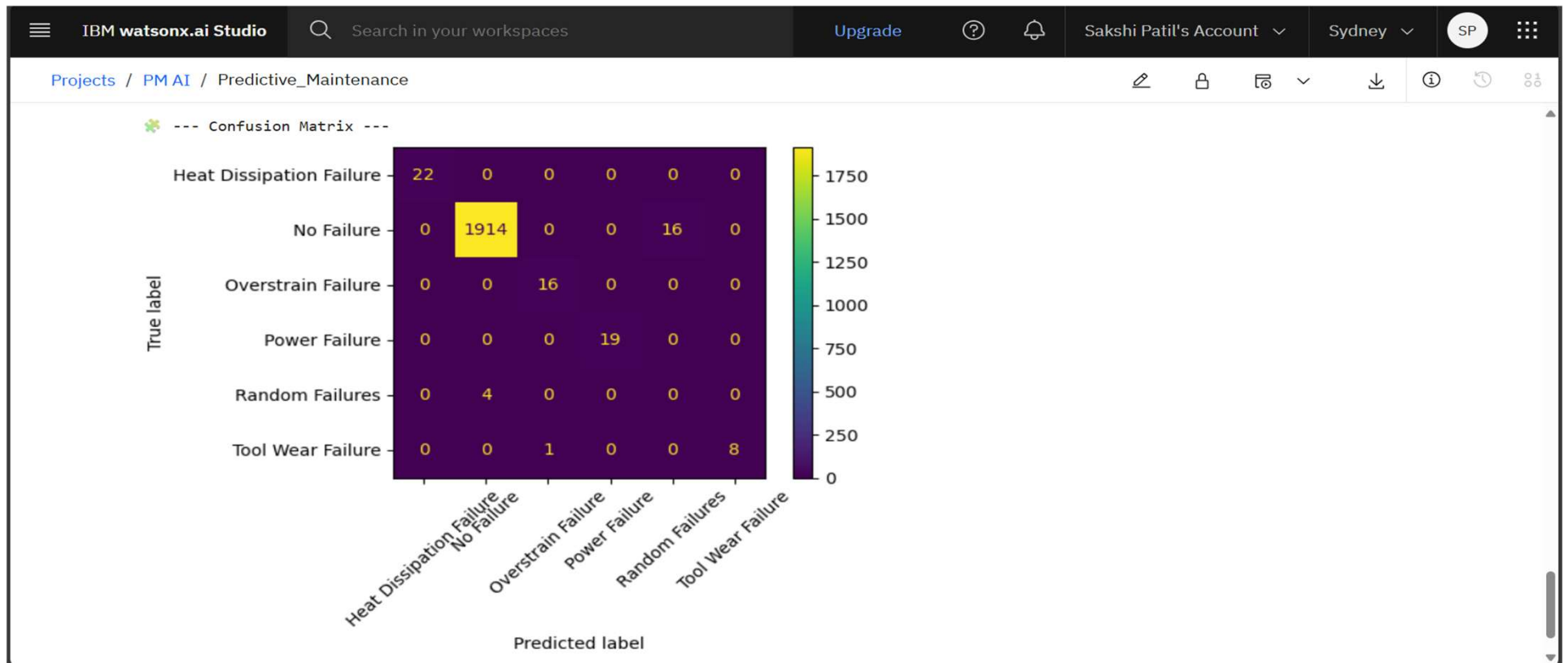
Heat Dissipation Failure      1.00      1.00      1.00        22
      No Failure              1.00      0.99      0.99       1930
Overstrain Failure            0.94      1.00      0.97        16
      Power Failure          1.00      1.00      1.00        19
      Random Failures         0.00      0.00      1.00         4
      Tool Wear Failure        1.00      0.89      0.94         9

              accuracy              0.99       2000
              macro avg          0.82      0.81      0.98       2000
              weighted avg        1.00      0.99      0.99       2000
```

```
--- Feature Importances ---
Torque      0.226019
ToolWear    0.201681
RotationalSpeed 0.176879
Target      0.175756
AirTemperature 0.111896
ProcessTemperature 0.054710
Type_L      0.032978
Type_M      0.020082
dtype: float64
```

```
--- Confusion Matrix ---
```

RESULT IMAGES



CONCLUSION

- Successfully developed a **predictive maintenance system** using machine learning.
- The model can reliably predict multiple types of machine failures.
- Integration with IBM Cloud ensures **scalable deployment and real-time inference**.
- Implementation of **SMOTE** significantly improved prediction performance on minority classes.

FUTURE SCOPE

- Incorporate **real-time streaming data pipelines** for live prediction.
- Experiment with other advanced algorithms (e.g., XGBoost, Neural Networks).
- Develop a **web dashboard** to visualize predictions for maintenance teams.
- Extend the system to **predict failure severity and maintenance recommendations**.

REFERENCES

- Kaggle Dataset :-

https://drive.google.com/file/d/1QQ67oHcgYZTpqrJ05EmSplmAYXN_nc2Y/view?usp=sharing

- IBM Cloud Lite Documentation

- scikit-learn Documentation

- imbalanced-learn Documentation

PROJECT LINK

Project Link :- https://github.com/Sakshi2004-29/Machine_Maintenance

IBM CERTIFICATIONS

In recognition of the commitment to achieve professional excellence



Sakshi Patil

Has successfully satisfied the requirements for:

Getting Started with Artificial Intelligence



Issued on: Jul 19, 2025
Issued by: IBM SkillsBuild

Verify: <https://www.credly.com/badges/83885c8b-c443-4777-9c6d-b7a8e44eff57>



IBM CERTIFICATIONS

In recognition of the commitment to achieve
professional excellence



Sakshi Patil

Has successfully satisfied the requirements for:

Journey to Cloud: Envisioning Your Solution



Issued on: Jul 19, 2025

Issued by: IBM SkillsBuild

Verify: <https://www.credly.com/badges/32b0c73b-4ade-4eed-9988-6e249e488798>



IBM CERTIFICATIONS

IBM **SkillsBuild**

Completion Certificate



This certificate is presented to

Sakshi Patil

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