
CAPSTONE PROJECT

POWER SYSTEM FAULT DETECTION AND CLASSIFICATION USING MACHINE LEARNING

Presented By:

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OUTLINE

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PROBLEM STATEMENT

Develop a machine learning-based system to detect and classify faults in a power distribution network using voltage and current phasor data. The model should accurately differentiate between normal conditions and faults such as line-to-ground, line-to-line, and three-phase faults. This enables fast, intelligent fault identification, ensuring improved grid reliability and operational efficiency.

PROPOSED SOLUTION

To build an intelligent fault detection system, we will leverage machine learning techniques trained on electrical measurement data. The solution pipeline involves:

- **Data Preprocessing:**
 - Clean and preprocess the collected data to handle missing values, outliers, and inconsistencies.
- **Preprocessing & Feature Engineering:**
 - Clean, normalize, and extract meaningful features to enhance model performance.
- **Model Development:**
 - Train and optimize fault classification models using algorithms like Random Forest, Support Vector Machine(SVM), or Neural Networks.
- **Validation & Performance Benchmarking:**
 - Evaluate model effectiveness using precision metrics like accuracy, recall, and F1-score to ensure reliable fault detection.

SYSTEM APPROACH

The "System Approach" section describes the overall methodology used to design, develop, and deploy the fault detection system using machine learning.

- **System requirements :**

- Laptop with minimum 8GB RAM & i5 Processor
- Reliable internet connection for cloud integration

- **Library required to build the model :**

- IBM Cloud (Essential platform)
- IBM Watsonx.ai Studio for model development & deployment
- IBM Cloud Object Storage for dataset access & management
- Kaggle dataset : fault.csv

ALGORITHM & DEPLOYMENT

- **Model Selection**

Random Forest used for fault classification; SVM considered for performance benchmarking.

- **Data Input**

Voltage, current, and phasor values extracted from dataset and pre-processed using IBM Cloud Storage.

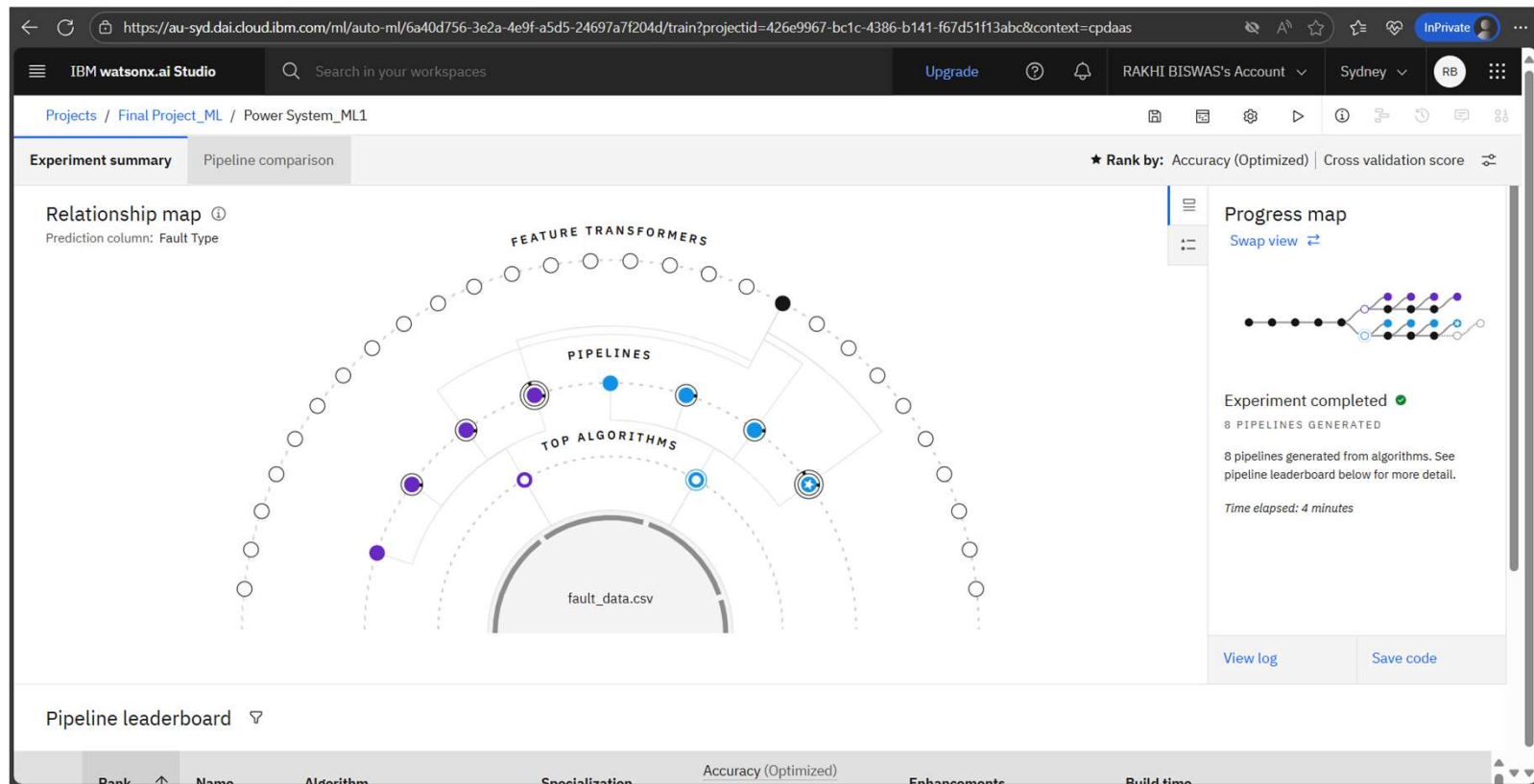
- **Training**

Supervised learning trains the model to identify labeled fault types efficiently.

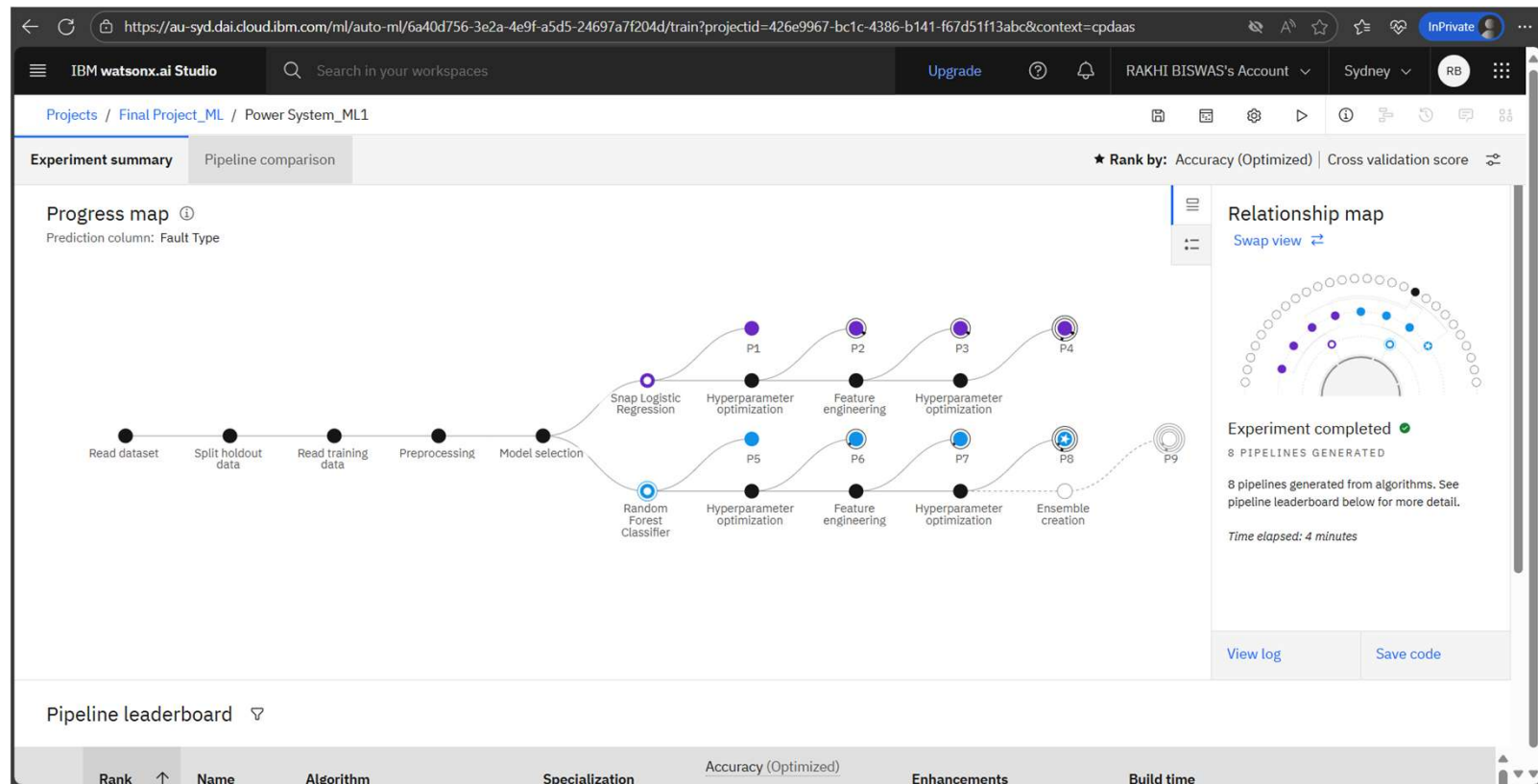
- **Deployment**

Model deployed on IBM Watson Studio with real-time API for predictions and insights.

RESULT



RESULT



RESULT

The screenshot displays the IBM Watson AI Studio interface. The top navigation bar includes the IBM Watson AI Studio logo, a search bar, and user account information (RAKHI BISWAS's Account, Sydney). The main content area shows the deployment 'Power system_Deploy2' in a 'Test' state. Below this, there is a section for 'Enter input data' with tabs for 'Text' and 'JSON'. A table with 5 rows and 12 columns is displayed, containing input data for a Random Forest Classifier. The table has columns for Fault ID, Fault Location (Latitude, Longitude), Voltage (V), Current (A), Power Load (MW), and Temperature (°C). The 'Predict' button is visible at the bottom right of the table.

Power system_Deploy2 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](#)

	Fault ID (other)	Fault Location (Latitude, Longitude) (other)	Voltage (V) (double)	Current (A) (double)	Power Load (MW) (double)	Temperature (°C) (double)
1	F001	(34.0522, -118.2437)	2200	250	50	25
2	F017	(34.9346, -118.9658)	2263	229	55	21
3	F024	(34.8432, -118.2489)	1877	249	54	33
4	F145	(34.7156, -118.2575)	1820	287	45	22
5	F160	(34.2934, -118.8373)	1948	221	51	20

5 rows, 12 columns

[Predict](#)

About this deployment

Name
Power system_Deploy2

Description
No description provided.

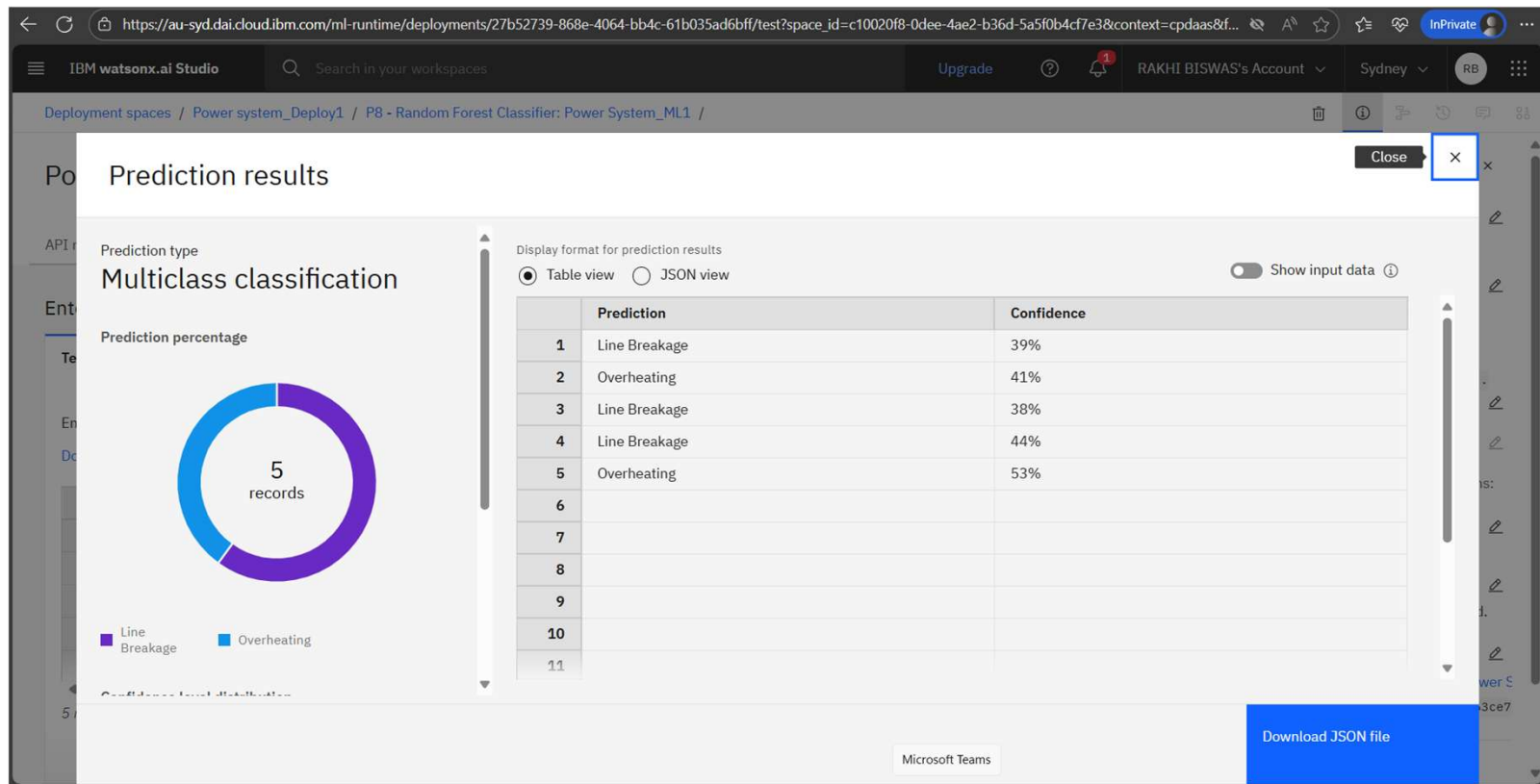
Deployment Details
Deployment ID: 27b52739-868e-40...
Serving name:
No serving name.
Software specification:
hybrid_0.1
Hybrid pipeline software specifications:
autoai-kb_rt24.1-py3.11
Copies:
1

Tags
Add tags to make assets easier to find.

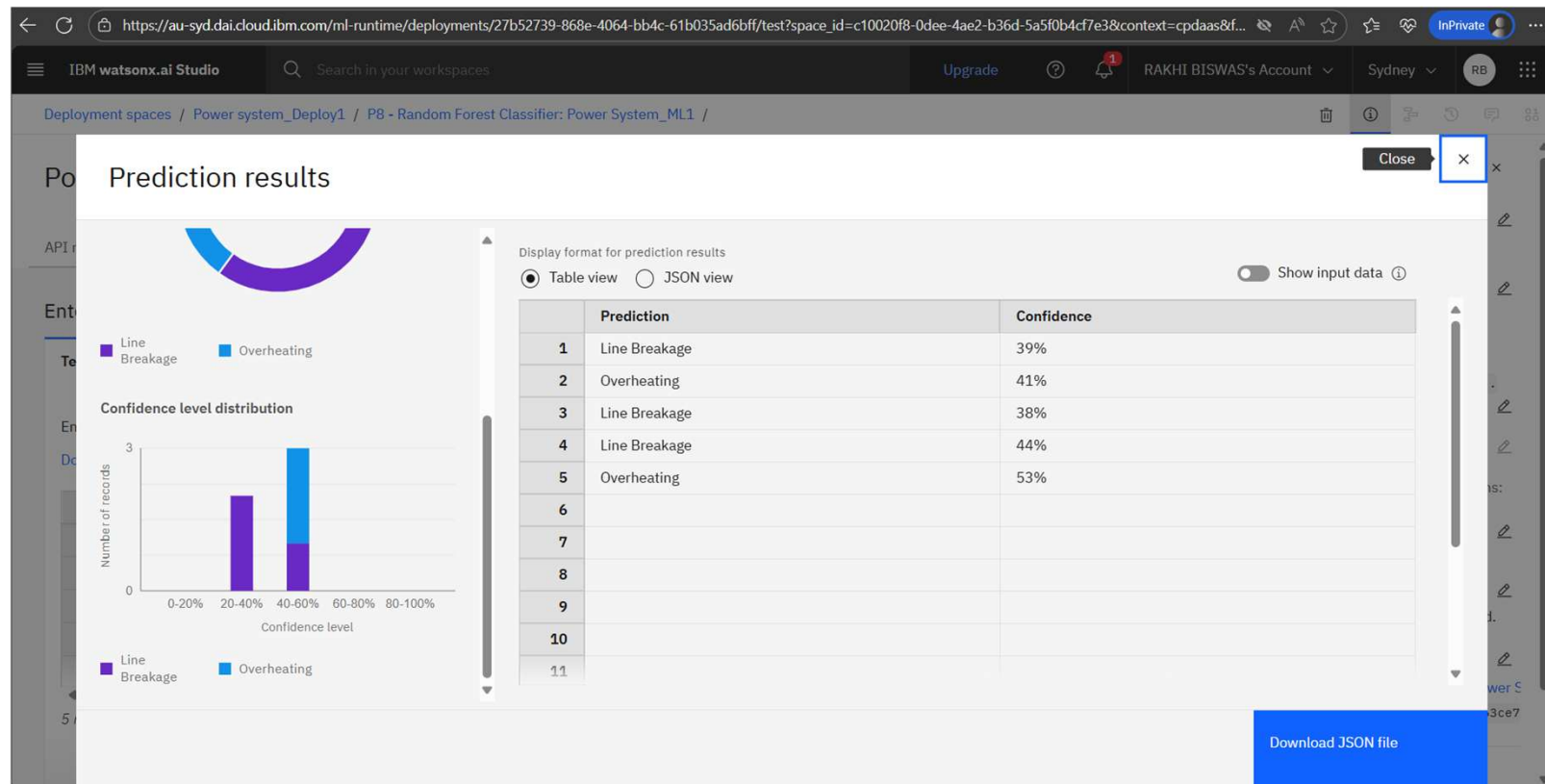
Associated asset
[P8 - Random Forest Classifier: Power S](#)
d4e5bf76-f1ce-4620-b1c8-753617a63ce7

Last modified

RESULT



RESULT



CONCLUSION

- This project successfully demonstrates the potential of machine learning for intelligent fault detection and classification in power distribution systems. By leveraging IBM Cloud Watsonx.ai Studio, we built accurate and efficient models capable of identifying various fault types using voltage and current phasor data. The use of AutoAI accelerated model development and optimization, enabling real-time fault detection critical for ensuring grid reliability and safety. Despite challenges like data preprocessing and cloud deployment, the system showed strong performance. Future enhancements may include integration with real-time SCADA data, deep learning for time-series analysis, and interactive dashboards. Overall, this solution highlights the importance of combining ML with cloud platforms to create smart, scalable, and proactive power grid monitoring systems.

FUTURE SCOPE

- To elevate the effectiveness and scalability of the power system fault detection model, future enhancements may include integrating real-time SCADA, IoT, and environmental data to improve fault prediction accuracy. Advanced ML techniques like LSTM, CNN, and anomaly detection can boost performance, especially for complex or rare fault patterns. Optimizing algorithms through hyperparameter tuning, ensemble models, and feature selection will further refine detection. The system can also be expanded geographically to monitor faults across multiple cities or regions, adapting to diverse grid configurations. Leveraging edge computing and digital twins will enable faster, localized responses, while interactive dashboards using tools like Streamlit can improve operator decision-making. Together, these enhancements pave the way for a smart, resilient, and future-ready power grid.

REFERENCES

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Learning hours: 20 mins

THANK YOU