

CAPSTONE PROJECT

POWER SYSTEM FAULT DETECTION AND CLASSIFICATION USING MACHINE LEARNING

Presented By:

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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**

PROBLEM STATEMENT

Design a machine learning model to detect and classify different types of faults in a power distribution system. Using electrical measurement data (e.g., voltage and current phasors), the model should be able to distinguish between normal operating conditions and various fault conditions (such as line-to-ground, line-to-line, or three-phase faults). The objective is to enable rapid and accurate fault identification, which is crucial for maintaining power grid stability and reliability.

PROPOSED SOLUTION

- Develop a machine learning model that classifies power system faults using the dataset provided . The model will process electrical measurements to identify the type of fault rapidly and accurately. This classification will help automate fault detection and assist in quicker recovery actions , ensuring system reliability.
- Key components:
 1. Data collection: Use the Dataset from Kaggle on power system faults.
 2. Preprocessing: Clean and Normalize the dataset.
 3. Model Training: Train a classification model(e.g, Decision Tree , Random Forest , or SVM).
 4. 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 rental bike prediction system. Here's a suggested structure for this section:

- System requirements:

- IBM Cloud(mandatory)

- IBM Watson studio for model development and deployment.

- IBM cloud object storage for dataset handling.

ALGORITHM & DEPLOYMENT

- Algorithm Selection:

Random Forest Classifier (or SVM based on performance)

- Data Input:

Voltage, current, and phasor measurements from the dataset.

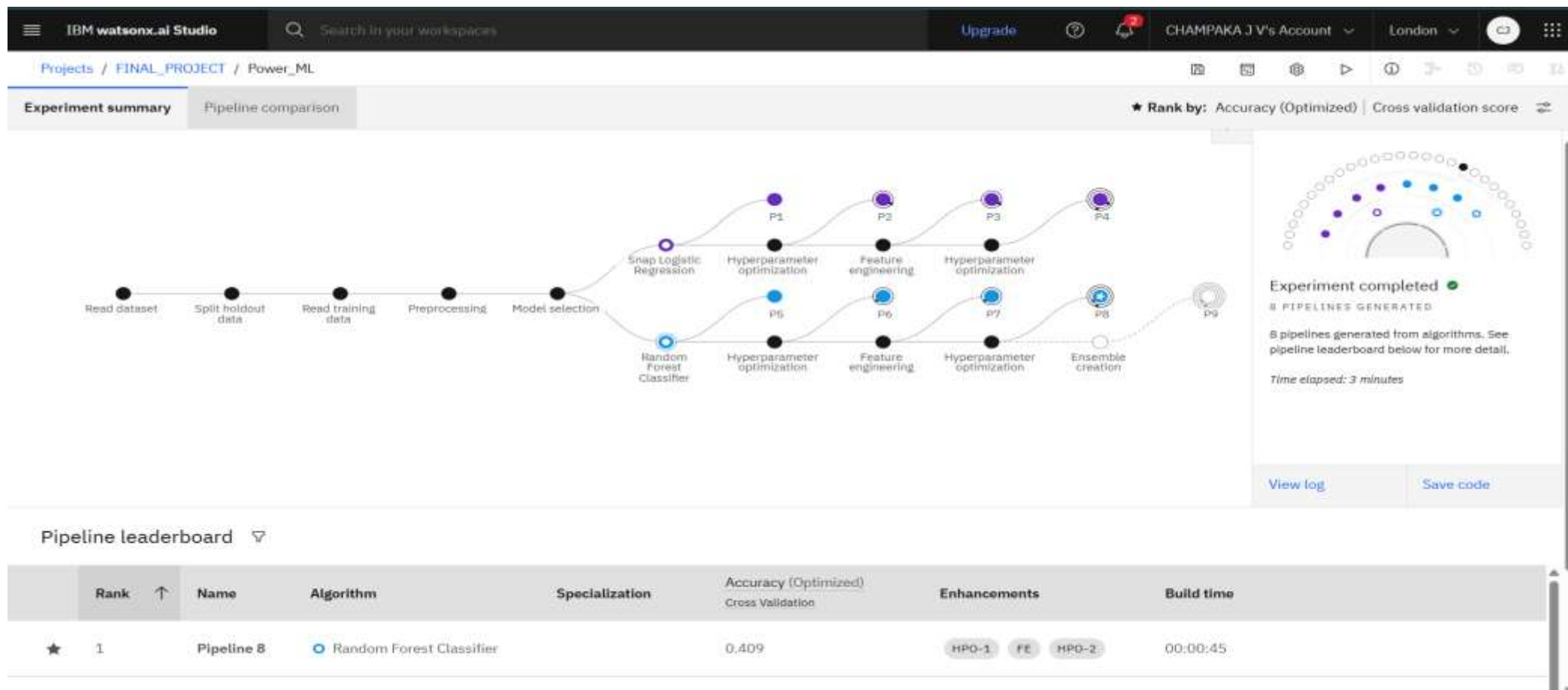
- Training Process:

Supervised learning using labelled fault types.

- Prediction Process:

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

RESULT



RESULT

IBM watsonx.ai Studio

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03

Deployment spaces / power_deployment1 / P8 - Random Forest Classifier: Power_ML /

power_dep2 Deployed Delete

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)	Wind Speed (km/h) (double)	Weather Condition (other)	Maintenance Status (other)	
1	F001	(34.0522, -118.2437)	2200	250	50	25	20	Clear	Scheduled	Fix
2	F005	(34.0545, -118.243)	1900	190	50	30	18	Snowy	Scheduled	Fix
3	F045	(34.4833, -118.5611)	2084	250	54	26	23	Rainy	Pending	On
4	F073	(34.9502, -118.4453)	2054	205	47	36	11	Windstorm	Completed	Fix
5	F079	(34.6843, -118.0941)	1801	187	51	38	13	Rainy	Pending	Fix
6	F102	(34.5828, -118.0344)	2296	238	55	30	13	Thunderstorm	Scheduled	Fix
7										
8										
9										
10										

6 rows, 12 columns

Predict

RESULT

Prediction results

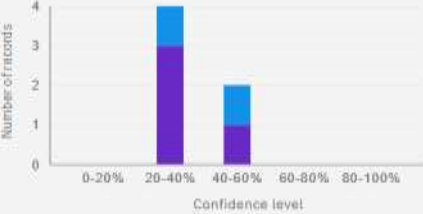
Prediction type
Multiclass classification

Prediction percentage



Line Breakage Transformer Failure

Confidence level distribution



Line Breakage Transformer Failure

Display format for prediction results

☒ Table view ☐ JSON view

☐ Show input data ⓘ

	Prediction	Confidence
1	Line Breakage	39%
2	Transformer Failure	38%
3	Line Breakage	37%
4	Line Breakage	39%
5	Line Breakage	43%
6	Transformer Failure	46%
7		
8		
9		
10		
11		
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16		

Download JSON file

CONCLUSION

- This project developed a machine learning-based system to detect and classify power system faults using voltage, current, and phasor data. After preprocessing the dataset, models like Random Forest and SVM were trained to accurately identify fault types. The best-performing model was deployed on IBM Watson Studio, enabling real-time fault detection through an API. This solution enhances power grid reliability by ensuring faster and more accurate fault response.

FUTURE SCOPE

- The developed fault detection system can be enhanced further by integrating real-time data from smart grid sensors for live monitoring. Future work may involve using advanced deep learning models like LSTM or CNN to capture more complex fault patterns. The system can also be extended to detect multiple or cascading faults and deployed on edge devices for faster, low-latency responses. Adding explainable AI features would help increase operator trust, while scaling the model for larger power grids and diverse regional topologies would make it suitable for widespread smart grid applications.

IBM CERTIFICATIONS




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IBM SkillsBuild

Completion Certificate



This certificate is presented to

Champaka J V

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