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# 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**
- **References**

# PROBLEM STATEMENT

Power distribution systems are prone to various types of faults such as Overheating, Transformer failure, line-to-ground, line-to-line and three-phase faults. These faults can disrupt power supply and reduce system reliability. The challenges lies in accurately detecting and classifying these faults using electrical measurement data(voltage, current, phasors) to differentiate them from normal operation condition, thereby ensuring the stability of the power grid.

# PROPOSED SOLUTION

Develop a ML 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. The classification will help automate fault detection and assist in quicker recovery actions, ensuring system reliability.

## Key Components

**Data Collection** - Use the Kaggle dataset on power system faults.

**Pre-processing** - Clean and normalize the dataset.

**Model training** - Train a classification model ( Decision tree, Random Forest, or SVM)

**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 endpoints for real time predictions.

Service Details - IBM Cloud

IBM watsonx.ai Studio

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Projects / Final\_Project / Power\_ML1

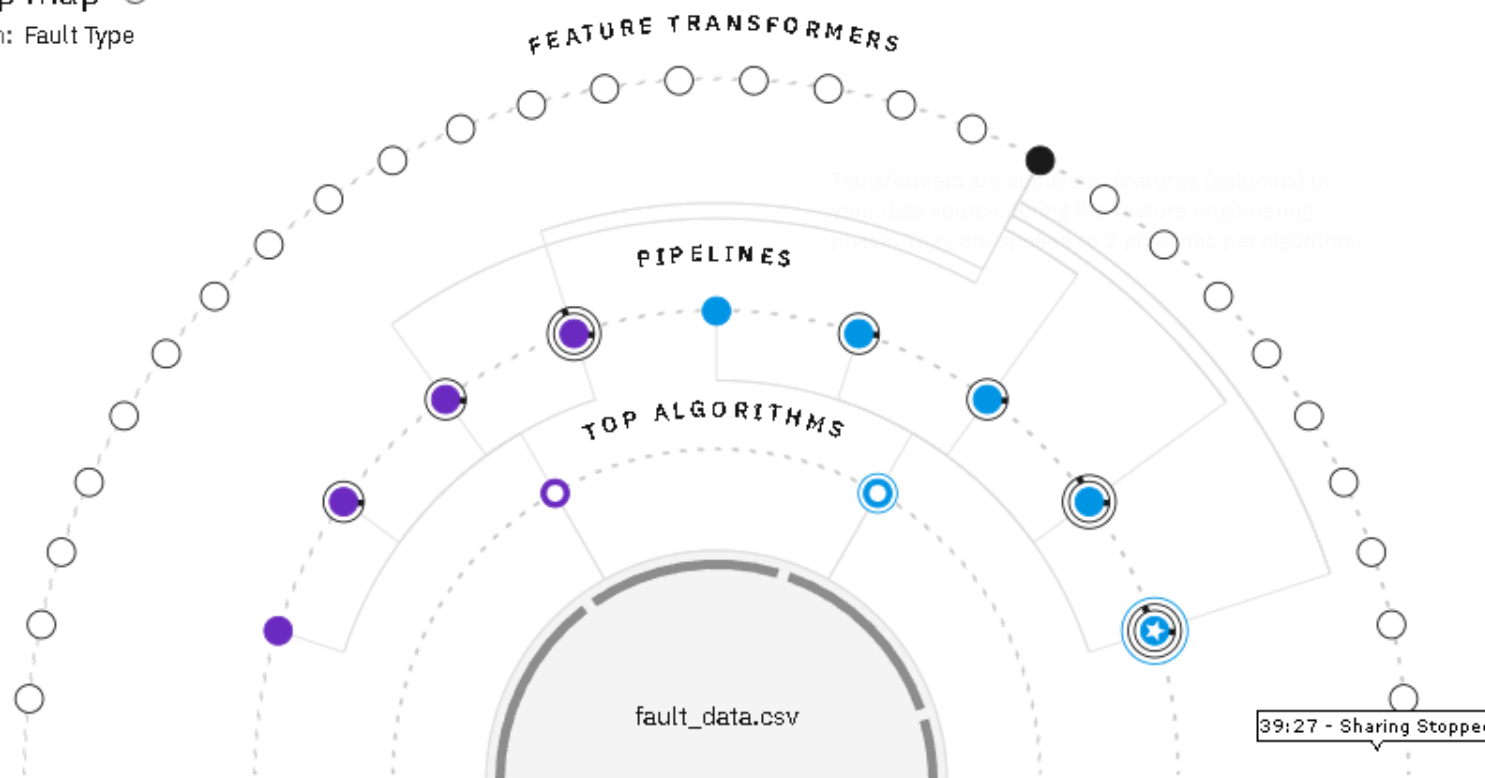
Experiment summary

Pipeline comparison

★ Rank by: Accuracy (Optimized) | Cross validation score


Relationship map ⓘ

Prediction column: Fault Type



Progress map

[Swap view](#)



Experiment completed ✓

9 PIPELINES GENERATED

9 pipelines generated from algorithms. See pipeline leaderboard below for more detail.

Time elapsed: 11 minutes

[View log](#)

[Save code](#)

Service Details - IBM CloudPower\_DEP2 — Power\_DEPSettings | IBM watsonx.ai

au-syd.dai.cloud.ibm.com/ml-runtime/deployments/af30b527-3db6-4df6-883e-4f5d03529251/test

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Prediction results

Close

Prediction type

Multiclass classification

Prediction percentage

5 records

Line Breakage

Overheating

Transformer Failure

Display format for prediction results

Table view

JSON view

Show input data

	Prediction	Confidence	Fault ID	Fault Location (Latitude, L	Voltage
1	Line Breakage	39%	F001	Line Breakage	2200
2	Overheating	37%	F003	Overheating	2100
3	Transformer Failure	47%	F008	Transformer Failure	2133
4	Line Breakage	42%	F015	Line Breakage	1848
5	Line Breakage	38%	F016	Transformer Failure	2102
6					
7					
8					
9					
10					

Download JSON file



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Prediction results

CloseX

Line BreakageOverheatingTransformer Failure

Confidence level distribution

Number of records

3

0

0-20%20-40%40-60%60-80%

Confidence level

Line BreakageOverheatingTransformer Failure

101

Display format for prediction results

Table viewJSON view

Show input data ⓘ

	Prediction	Confidence	Fault ID	Fault Location (Latitude, Longitude)	Voltage (V)
1	Line Breakage	39%	F001	Line Breakage	2200
2	Overheating	37%	F003	Overheating	2100
3	Transformer Failure	47%	F008	Transformer Failure	2133
4	Line Breakage	42%	F015	Line Breakage	1848
5	Line Breakage	38%	F016	Transformer Failure	2102
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Download JSON file

# CONCLUSION

The machine learning model developed for classifying power system faults effectively enhances fault detection and response in electrical distribution networks. Using voltage, current, and phasor data from a Kaggle dataset, the model distinguishes between various fault types such as overheating, transformer failure, and line faults. Built and deployed on IBM Cloud using Watson Studio and Cloud Object Storage, the system offers scalability and real-time prediction through API integration.

Random Forest outperformed other models in terms of accuracy, precision, recall, and F1-score, proving to be the most reliable classifier. Preprocessing and data normalization were critical for model performance, and supervised learning ensured accurate training using labeled data. Challenges included class imbalance and similar fault signatures, addressed through algorithm tuning and feature engineering. The approach not only strengthens grid stability but also reflects the versatility of ML systems, similar to predicting rental bike availability in urban settings for efficient resource planning.

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# FUTURE SCOPE

- Additional Data Sources:**

Incorporate weather data, smart meter readings, thermal sensors, and maintenance records to improve fault detection accuracy.

- Algorithm Optimization:**

Use advanced ML models like XGBoost, LSTM, or hybrid deep learning approaches for better performance and time-series analysis.

- Geographic Expansion:**

Extend the system to multiple cities or regions with localized training and calibration for different grid infrastructures.

- Edge Computing Integration:**

Deploy models on edge devices near transformers or substations for real-time, low-latency fault detection.

# REFERENCES

- Cloud Deployment & IBM Watson
- IBM Cloud Docs.

*Getting Started with IBM Watson Studio and Journey to the Cloud.*

<https://cloud.ibm.com/docs/watson-studio>

*(Official IBM documentation on developing and deploying ML models using Watson Studio.)*

# IBM CERTIFICATIONS



# IBM CERTIFICATIONS

In recognition of the commitment to achieve  
professional excellence



Dev yadav

Has successfully satisfied the requirements for:

Journey to Cloud: Envisioning Your Solution



Issued on: Jul 20, 2025

Issued by: IBM SkillsBuild

Verify: <https://www.credly.com/badges/0c3c9e5d-5b6f-4b3f-b38b-101b2dcb40a9>



# IBM CERTIFICATIONS

IBM SkillsBuild

Completion Certificate



This certificate is presented to

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**Lab: Retrieval Augmented Generation with LangChain**

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According to the Adobe Learning Manager system of record

Completion date: 24 Jul 2025 (GMT)

Learning hours: 20 mins



**THANK YOU**