
CAPSTONE PROJECT

POWER SYSTEM FAULT DETECTION AND CLASSIFICATION

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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

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

- **Data Acquisition & Understanding**

Utilize the Kaggle dataset containing electrical parameters (Voltage, Current, etc.) under normal and various fault conditions.

- **Data Preprocessing**

- Handle missing or inconsistent values
- Normalize/standardize voltage & current readings
- Encode fault types for classification
- Perform feature selection to focus on key electrical signatures

- **Model Design**

- Implement supervised machine learning algorithms (e.g., Random Forest, SVM, or Neural Networks)
- Train models to classify:
 - Line-to-Ground Fault
 - Line-to-Line Fault
 - Three-Phase Fault
 - No Fault (Normal)

- **Cloud Integration – IBM Cloud Lite**

- Use **IBM Watson Studio** for model development and training
- Deploy trained model as a **REST API** using **IBM Cloud Functions** or **IBM Cloud Foundry**
- Monitor performance and enable online predictions

- **Evaluation & Optimization**

- Assess using accuracy, precision, recall, and F1-score
- Use confusion matrix for detailed fault classification analysis
- Fine-tune hyperparameters for optimal results

- **Outcome**

A reliable ML system that can automatically detect and classify faults, enabling **faster fault response** and **power grid stability**.

SYSTEM APPROACH

The "System Approach" section outlines the overall strategy and methodology for developing and implementing the Power System Fault Detection and Classification model. Here's a suggested structure for this section:

- **System requirements – Windows 10 OS, 8GB RAM, Python 3.x**
- **Resources required on IBM Cloud platform to run AutoAI experiment – Watson Studio, Cloud Object Storage, Watson Machine Learning**

ALGORITHM & DEPLOYMENT

- **Step 1** – Allocate Cloud Object Storage and Watson Studio in IBM Cloud
- **Step 2** – Create a new project and enter project name, then click on create
- **Step 3** – Associate Watson Machine Learning service
- **Step 4** – Select AutoAI to automatically build models
- **Step 5** – Enter a name and create the AutoAI experiment
- **Step 6** – Upload the power system faults dataset from Kaggle to IBM Cloud and select it in AutoAI
- **Step 7** – Select the target column (Fault Type) to classify fault types based on voltage and current readings
- **Step 8** – Click on "Run Experiment"
- **Step 9 – Algorithm Selection:**
AutoAI will automate model selection, hyperparameter tuning, and recommend the best classification model for power system faults
- **Step 10** – Save the model with the highest accuracy
- **Step 11** – Click "Promote to Deployment Space" to deploy the model
- **Step 12** – Evaluation and testing using new voltage/current data for fault prediction

RESULT



IBM watsonx.ai Studio



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Projects / Power System Fault Detection and Classification / Power System



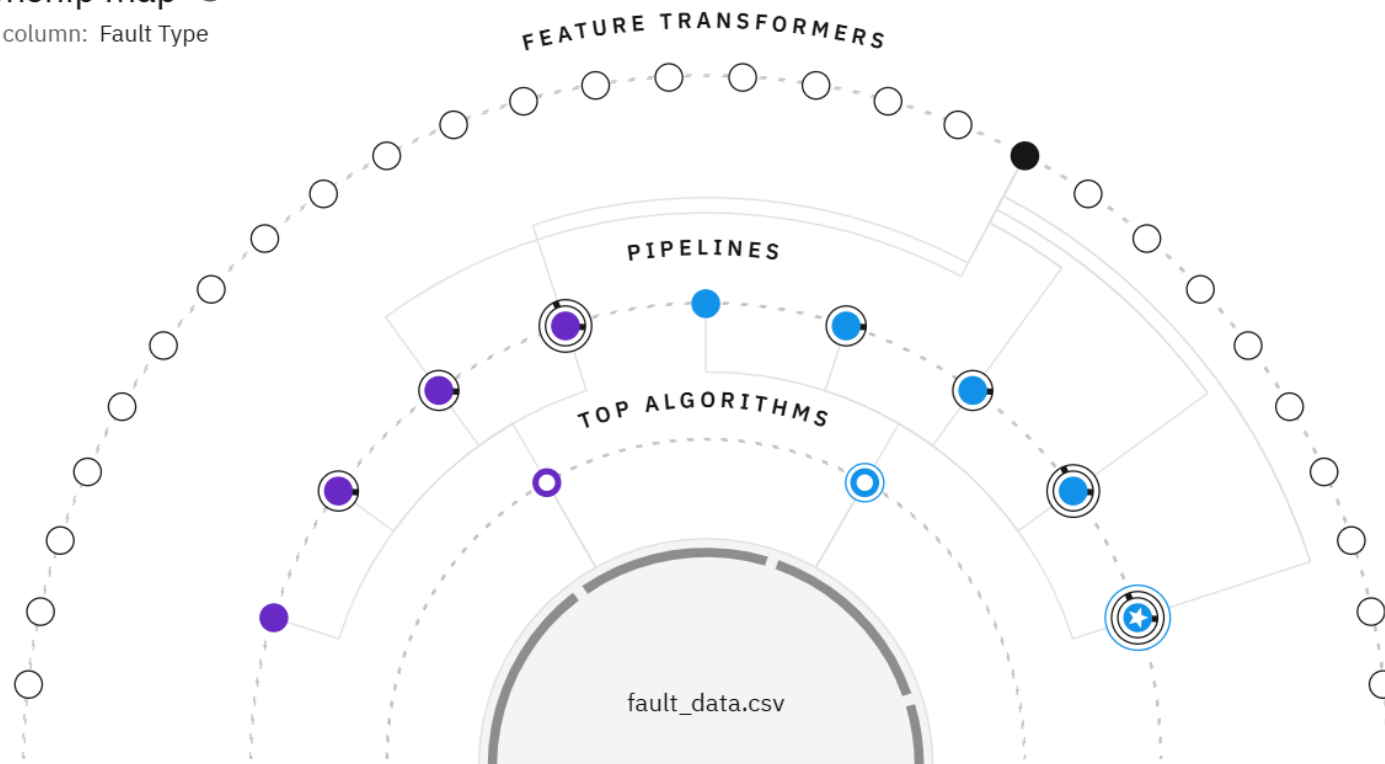
Experiment summary

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: 8 minutes

Experiment summary

Pipeline comparison

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

View log

Save code

Pipeline leaderboard

	Rank <div>↑</div>	Name	Algorithm	Accuracy (Optimized) Cross Validation	Enhancements	Build time
★	1	Pipeline 9	<div></div> Batched Tree Ensemble Classifier (Random Forest Classifier)	0.409	HPO-1 FE HPO-2 BATCH	00:01:41
	2	Pipeline 8	<div></div> Random Forest Classifier	0.409	HPO-1 FE HPO-2	00:01:38
	3	Pipeline 4	<div></div> Snap Logistic Regression	0.393	HPO-1 FE HPO-2	00:00:26
	4	Pipeline 3	<div></div> Snap Logistic Regression	0.393	HPO-1 FE	00:00:23

Power_DEP1 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](#)

	Power Load (MW) (double)	Temperature (°C) (double)	Wind Speed (km/h) (double)	Weather Condition (other)	Maintenance Status (other)	Component Health (other)	Duration of Fault (hrs) (double)	Down time (hrs) (double)
1	45	28	15	Rainy	Completed	Faulty	3	5
2	51	23	21	Snowy	Normal	Normal	3,7	6,3
3								
4								
5								
6								
7								
8								
9								

2 rows, 12 columns

Predict

Prediction results

Display format for prediction results

☒ Table view ☐ JSON view

☐ Show input data

	prediction	probability
1	Transformer Failure	[0.305630902371415,0.3409365463246582,0.3534325513039267]
2	Line Breakage	[0.40681209449076794,0.28061456440540333,0.3125733411038282]
3		
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14		
15		

Download JSON file

Power_DEP1

✔ Deployed

Online

API reference

Test

Endpoints for scoring

Private endpoint

`https://private.au-syd.ml.cloud.ibm.com/ml/v4/deployments/2461d35a-678c-4c46`

Bearer <token>

IAM

Public endpoint

`https://au-syd.ml.cloud.ibm.com/ml/v4/deployments/2461d35a-678c-4c46-83eb-24`

[Learn more](#) about the 2021-05-01 version query parameter

Code snippets

cURL

Java

JavaScript

Python

Scala

About this deployment

Name

Power_DEP1

Description

No description provided.

Deployment Details

Deployment ID: 2461d35a-678c-4c...

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.

CONCLUSION

- The system successfully detects and classifies different types of power system faults using machine learning techniques, ensuring rapid and accurate fault identification.
- This approach enables electrical utilities to respond faster to faults, reduce downtime, and maintain grid stability and reliability.
- The results highlight the significance of integrating machine learning into power systems for real-time monitoring and intelligent fault management. By leveraging such data-driven solutions, power distribution networks can become more efficient, resilient, and future-ready.

FUTURE SCOPE

- **Real-Time Integration:** Connect with SCADA/smart meters for live fault detection.
- **Fault Location Estimation:** Extend model to pinpoint fault locations on the grid.
- **Edge Deployment:** Deploy lightweight models on substations for on-site analysis.
- **Renewable Support:** Adapt system for dynamic behavior from solar and wind sources.
- **Scalability:** Scale to large power grids using cloud-native and distributed systems.
- **Automated Alerts:** Use IBM Cloud to trigger real-time alerts and maintenance actions.

REFERENCES

- ChatGPT
- IBM cloud platform
- **Research Paper** – *“Power System Fault Detection and Classification Using Machine Learning”* – [Link to paper](#)
- Project Link –

<https://dataplatform.cloud.ibm.com/projects/846904b4-ea60-4dee-b506-01ca07e1daa7?context=cpdaas>

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THANK YOU