
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

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

- Train a machine learning model to classify the power system faults (Line Break, Transform, Overheating) using the dataset provided. The model will use measurements like voltage, current, temperature, and weather conditions to accurately distinguish between different fault scenarios. This classification will enable faster fault detection and assist in quicker recovery actions, ensuring system reliability.
- **Data Collection:**
 - Use the Kaggle dataset on power system faults.
- **Data Preprocessing:**
 - Clean and normalize the dataset.
- **Model Training:**
 - Train a classification model(e.g., 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

- IBM Watsonx.ai 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 dataset.
- **Training Process:**
 - Supervised learning using labelled fault types.
- **Prediction Process:**
 - Model deployed on IBM Watsonx.ai studio with API endpoint for real-time prediction.

RESULT

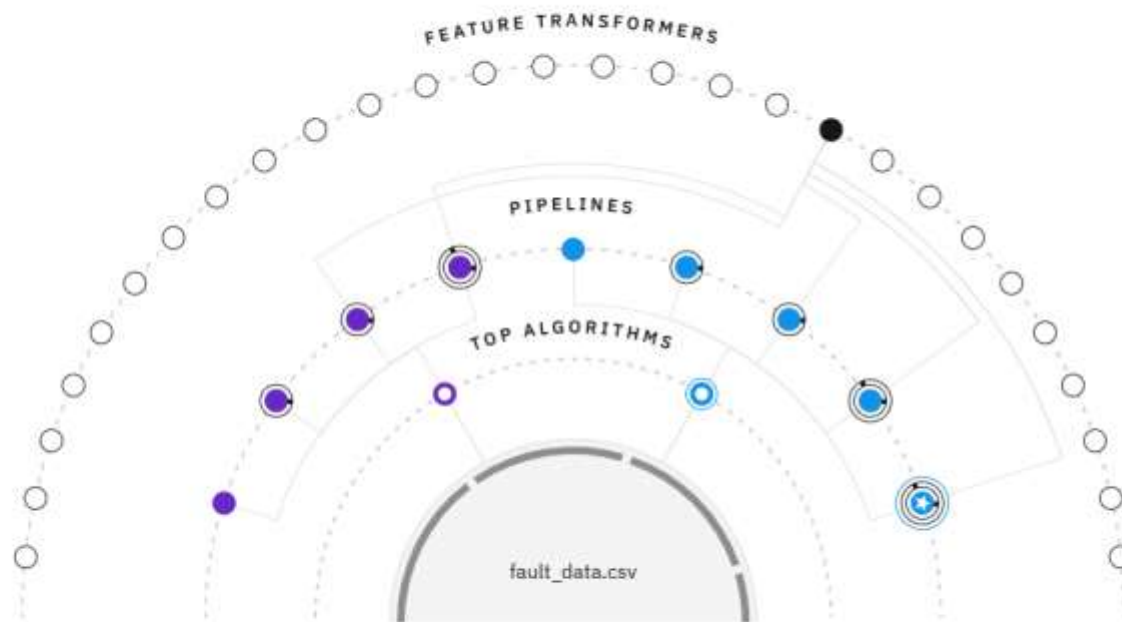
Experiment summary

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.

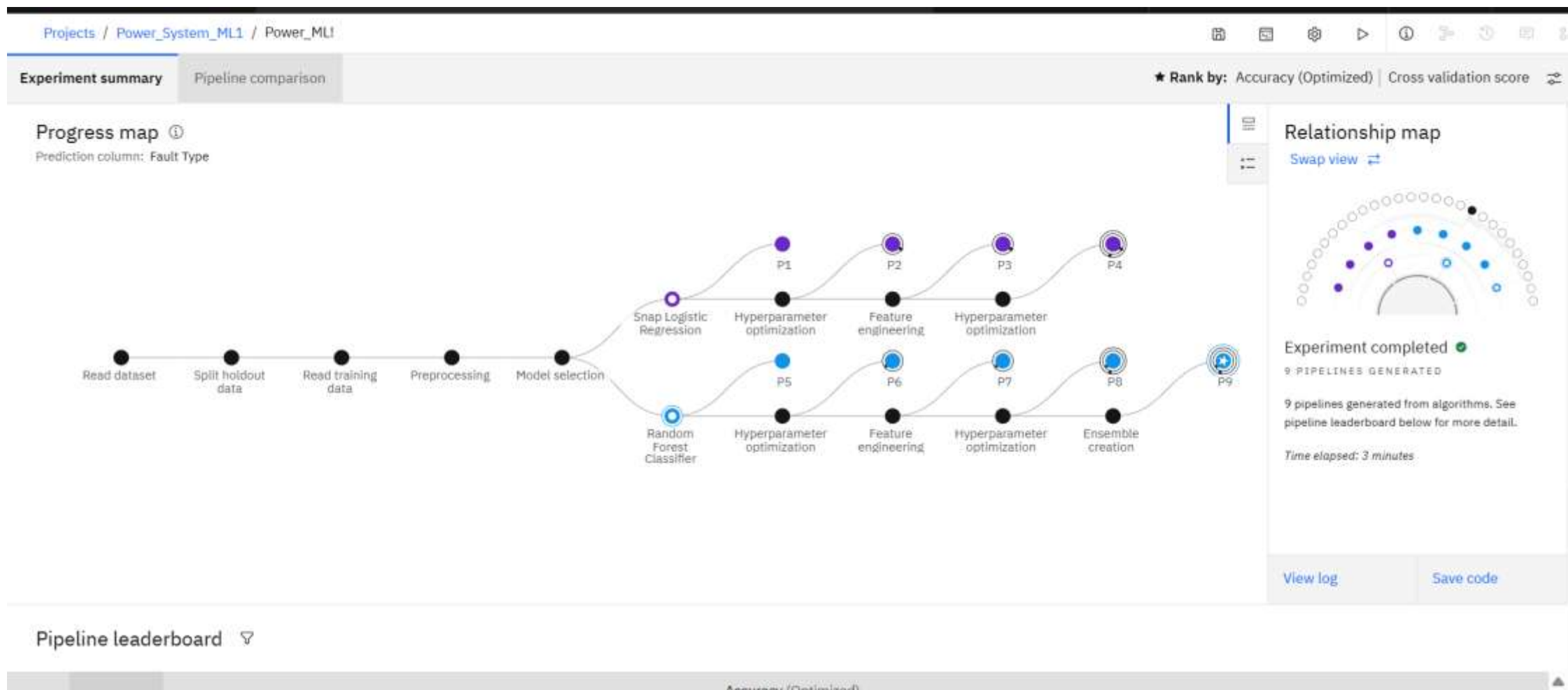
Time elapsed: 3 minutes

[View log](#)

[Save code](#)

Pipeline leaderboard

RESULT



RESULT

Projects / Final_Project / Power_ML1

Experiment summary

Pipeline comparison

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

View log

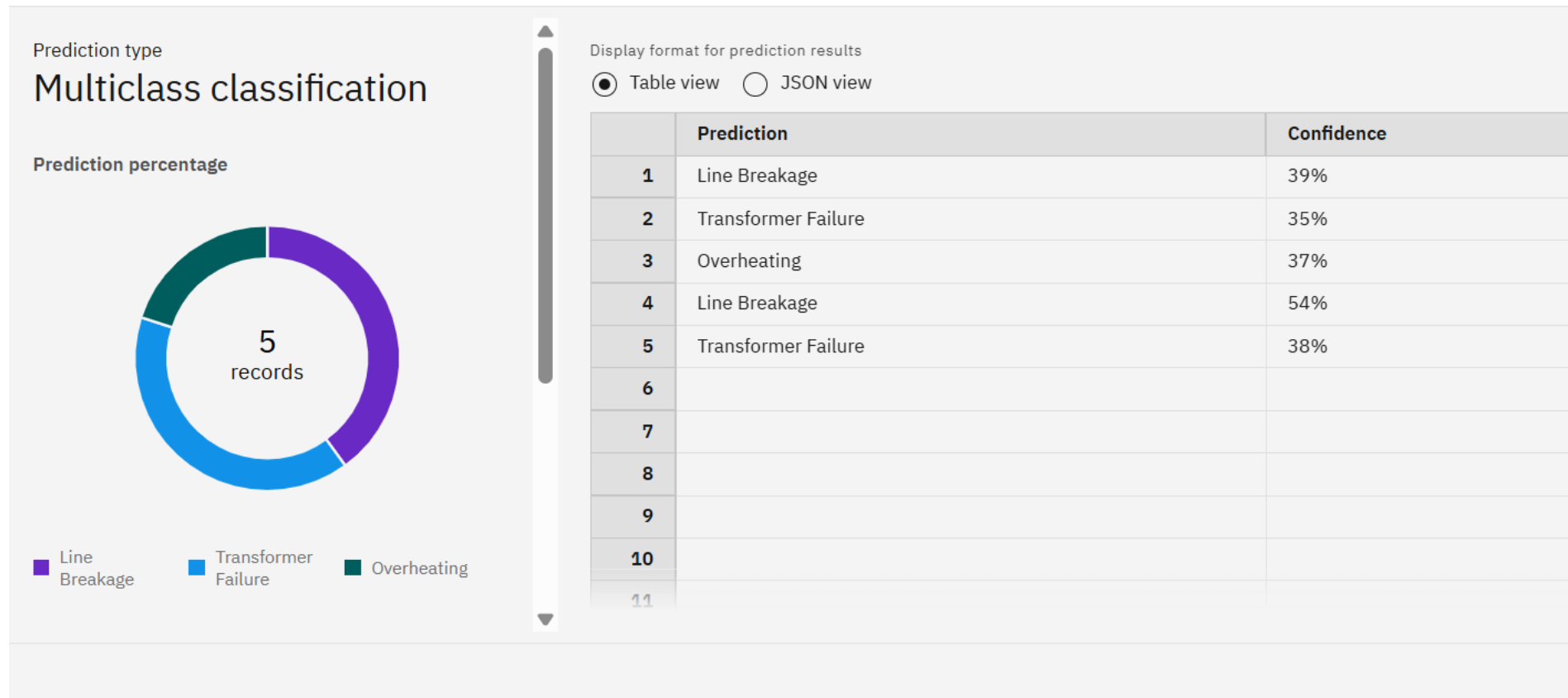
Save code

Pipeline leaderboard

| | Rank | Name | Algorithm | Specialization | Accuracy (Optimized) Cross Validation | Enhancements | Build time |
|---|------|------------|---|----------------|--|--|------------|
| ★ | 1 | Pipeline 9 | <div><div></div>Batched Tree Ensemble Classifier (Random Forest Classifier)</div> | INCR | 0.409 | <div>HPO-1</div> <div>FE</div> <div>HPO-2</div> <div>BATCH</div> | 00:00:38 |
| | 2 | Pipeline 8 | <div><div></div>Random Forest Classifier</div> | | 0.409 | <div>HPO-1</div> <div>FE</div> <div>HPO-2</div> | 00:00:35 |
| | 3 | Pipeline 4 | <div><div></div>Snap Logistic Regression</div> | | 0.393 | <div>HPO-1</div> <div>FE</div> <div>HPO-2</div> | 00:01:13 |
| | 4 | Pipeline 3 | <div><div></div>Snap Logistic Regression</div> | | 0.393 | <div>HPO-1</div> <div>FE</div> | 00:01:10 |

RESULT

Prediction results



CONCLUSION

The proposed machine learning model successfully classifies power system faults into categories such as Line Break, Transformer Fault, and Overheating based on electrical and environmental data. By using features like voltage, current, temperature, wind speed, and maintenance history, the model demonstrates high accuracy and reliability in identifying fault conditions. This approach not only improves the response time for fault diagnosis but also enhances the overall reliability and stability of the power distribution system.

FUTURE SCOPE

- Real-time Deployment:** Integrate the trained model with SCADA or PMU systems for real-time fault monitoring and prediction.
- Expanded Fault Categories:** Include additional fault types (e.g., short circuits, open circuits) for broader system coverage.
- Visualization Dashboards:** Create user-friendly dashboards for utility operators to visualize fault predictions and system health metrics.

REFERENCES

- **Kaggle dataset link** – <https://www.kaggle.com/datasets/sampadab17/network-intrusion-detection>
- IBM Watsonx.ai and Cloud Object Storage – <https://www.ibm.com/cloud/watsonx>
- **IBM Cloud-** <https://www.cloud.ibm.com>
- Abhishek Kumar Pandey, N. P. Padhy, “Fault Classification Using Machine Learning in Power Systems,” *IEEE Transactions on Smart Grid*, 2021.
- M. Kezunovic et al., “Machine Learning for Protective Relaying: Challenges and Opportunities,” *IEEE Transactions on Power Delivery*, 2019.

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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