CAPSTONE PROJECT

POWER SYSTEM FAULT DETECTION AND CLASSIFICATION USING MACHINE LEARNING

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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 line-to-ground, line-to-line, and three- phase faults. These faults can disrupt power supply and reduce system reliability. The challenge lies in accurately detecting and classifying these faults using electrical measurements data (voltage, current, phasors) to differentiate them form normal operating conditions, thereby ensuring the stability



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:

- Data Collection: Use the Kaggle dataset on power system faults.
- 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 power system fault detection and classification. 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

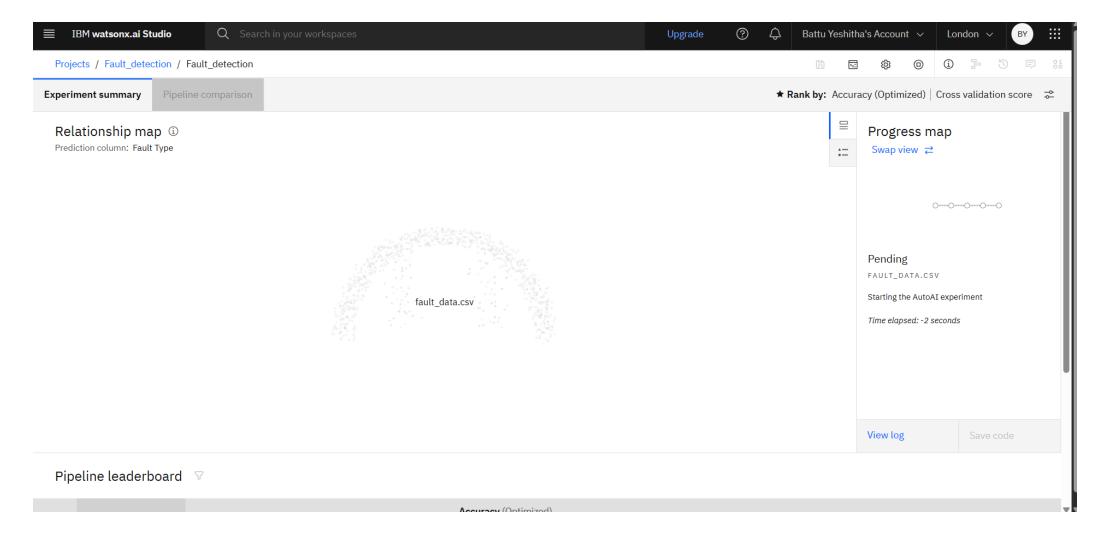
Training Process :

Supervised learning using labeled fault types

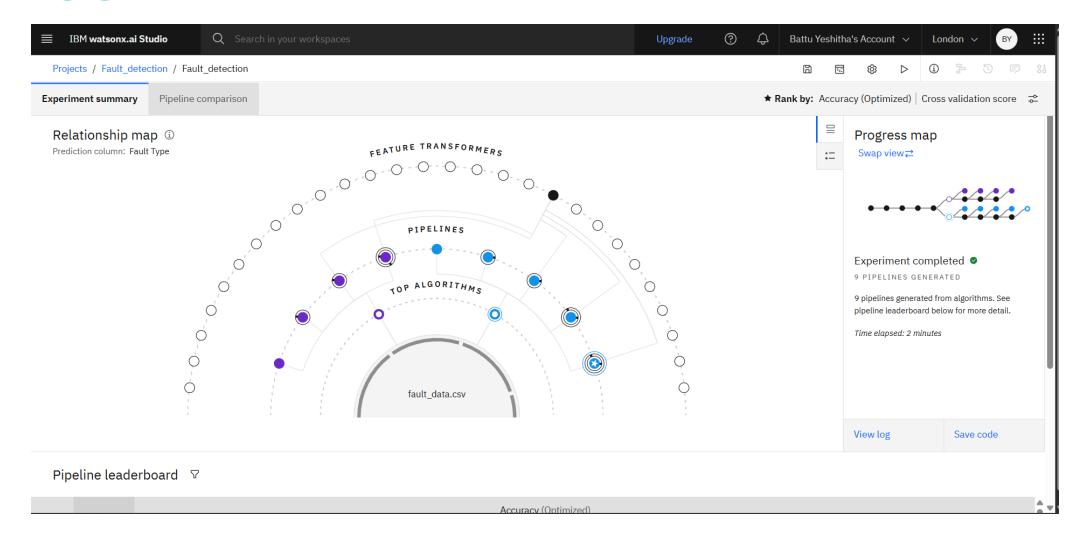
Prediction Process :

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

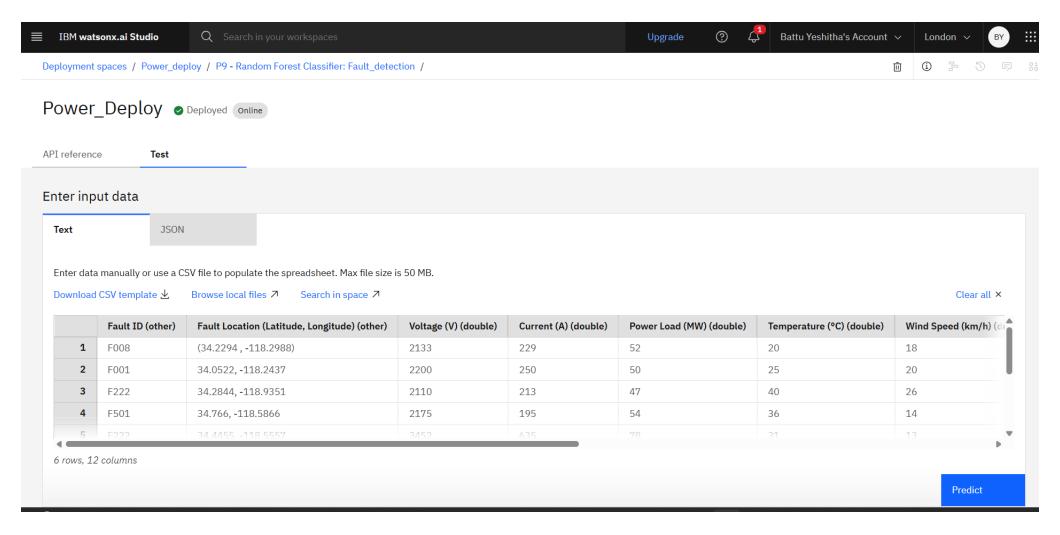




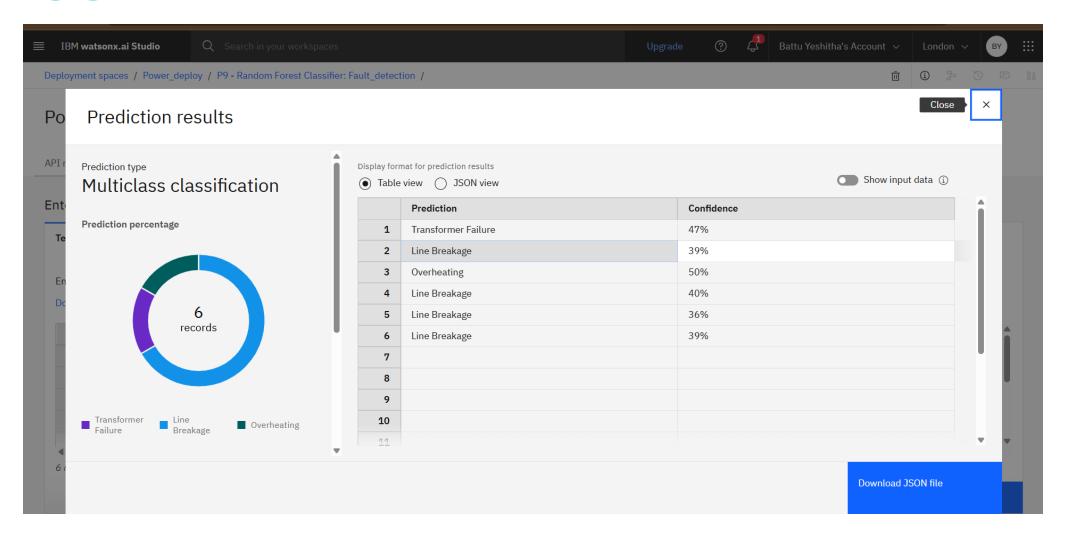














CONCLUSION

The proposed fault detection and classification system effectively identifies various types of faults—line-to-ground, line-to-line, and three-phase—by analyzing electrical measurements such as voltage, current, and phasor data. This approach ensures enhanced reliability and minimizes power disruption in distribution networks. The use of machine learning and signal processing techniques significantly improves accuracy and responsiveness, making the system more robust in distinguishing fault conditions from normal operations.



FUTURE SCOPE

- Integration with real-time monitoring and control systems for automated grid responses.
- Expansion to include renewable energy sources and smart grid components.
- Incorporation of advanced deep learning models to improve fault classification precision.
- Development of edge computing-based fault detection for faster local decision-making.
- Use of IoT-enabled sensors for continuous grid health monitoring and predictive maintenance.



REFERENCES

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THANK YOU

