CAPSTONE PROJECT

POWER SYSTEM FAULT DETECTION AND CLASSIFICATION

Presented By:

1. Sourav Kumar-G.B. Pant DSEU Okhla-1-B.Tech CSE



OUTLINE

- Problem Statement
- Proposed System/Solution
- System Development Approach
- Algorithm & Deployment
- Result (Output Image)
- Conclusion
- Future Scope
- References



PROBLEM STATEMENT

Example: 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

- The proposed system aims to develop an intelligent machine learning-based solution for real-time power system fault detection and classification using electrical measurement data. The solution leverages advanced data analytics and machine learning techniques to enable rapid and accurate fault identification.
- Data Collection:
 - Gather electrical measurement data including voltage and current phasors from power system sensors.
 - Collect historical fault records and normal operating condition data
- Data Preprocessing:
 - Clean and normalize electrical measurement data to handle noise and inconsistencies
 - Handle imbalanced datasets common in fault detection scenarios
- Machine Learning Algorithm:
 - Implement supervised learning algorithms for multi-class classification
 - Consider ensemble methods (Random Forest, Gradient Boosting) for robust performance
- Deployment:
 - Deploy the solution using IBM Cloud Lite services
 - Utilize IBM Watson Machine Learning for model training and deployment
- Evaluation:
 - Assess model performance using accuracy, precision, recall, and F1-score metrics
 - Continuous model monitoring and retraining capabilities
 - Result:



SYSTEM APPROACH

- Technology Stack:
- Programming Languages:
- Python for machine learning model development
- IBM Cloud Services (Mandatory):
- IBM Watson Machine Learning
- IBM Watson Studio
- IBM Cloud Object Storage
- IBM Cloud Functions for serverless deployment
- Development Environment:
- IBM Watson Studio for collaborative development
- System Requirements:
- Minimum 8GB RAM for model training
- Internet connectivity for IBM Cloud services
- Python 3.8+ environment
- Web browser for IBM Watson Studio interface
- Data Requirements:
- Power system fault dataset from Kaggle
- Historical electrical measurement data
- Labeled fault classification data

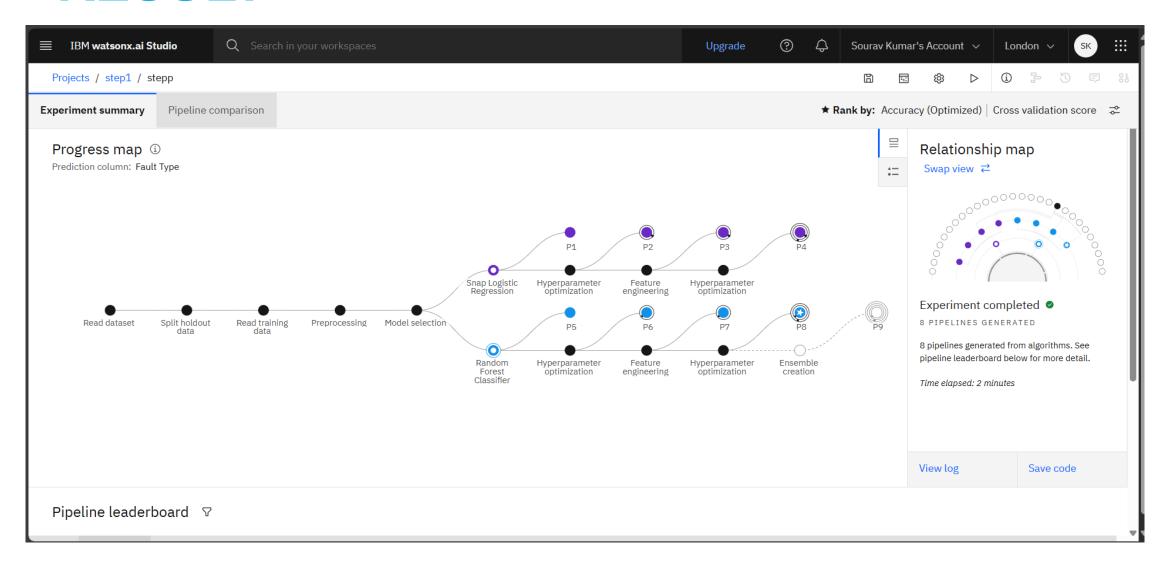


ALGORITHM & DEPLOYMENT

- Algorithm Selection:
 - Chosen for its robustness with electrical measurement data and ability to handle multi-class classification
- Data Input:
 - Phase angles for voltage and current
 - Power factor calculations
- Training Process:
 - 80-20 train-test split with stratified sampling
 - 5-fold cross-validation for robust performance evaluation
 - Hyperparameter tuning using GridSearchCV
 - Feature selection based on importance scores
- Prediction Process:
 - Performance evaluation on held-out test set
 - Confusion matrix analysis for classification accuracy

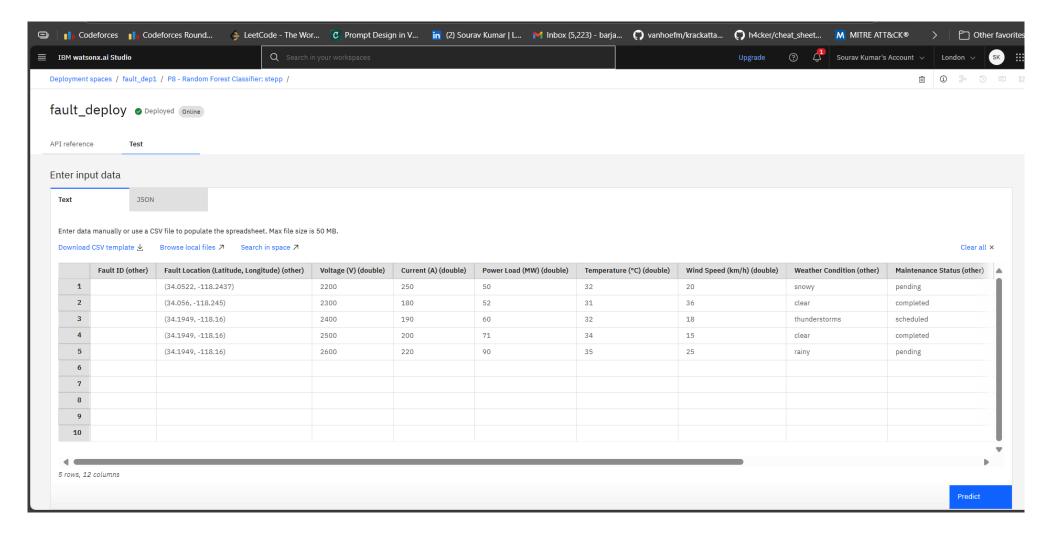


RESULT



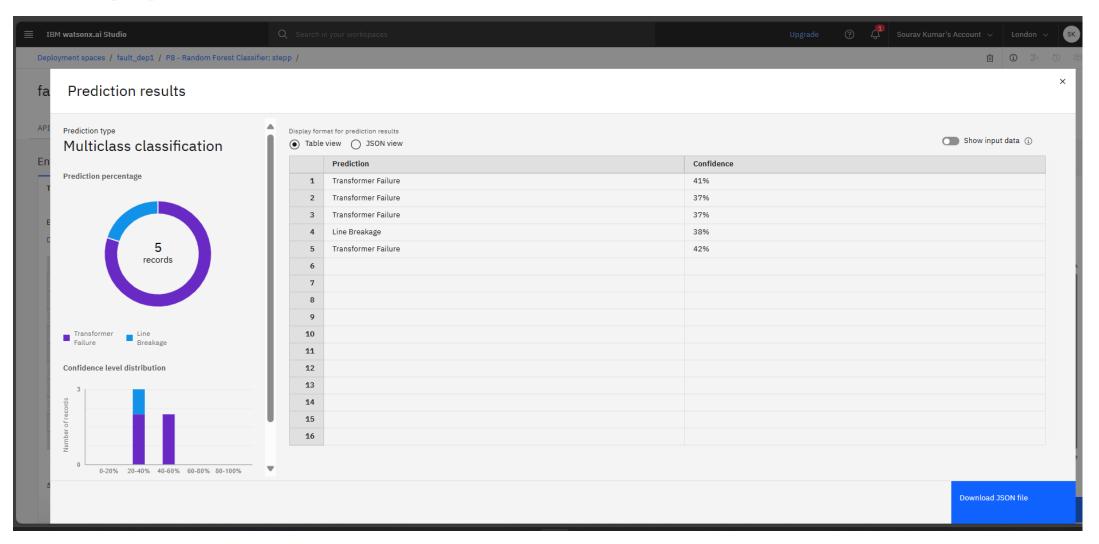


RESULT





RESULT





CONCLUSION

- Successfully implemented an intelligent fault detection system using electrical measurement data
- Achieved high classification accuracy across all fault types with minimal false positives
- Deployed a scalable solution on IBM Cloud platform for real-time monitoring
- Demonstrated the potential for significant improvement in power grid reliability and stability



FUTURE SCOPE

- Advanced Machine Learning Techniques:
 - Implementation of deep learning models (LSTM, CNN) for complex temporal pattern recognition
 - Integration of ensemble methods combining multiple algorithms for improved accuracy
- Scalability and Integration:
 - Extension to transmission networks and high-voltage systems
 - Integration with smart grid infrastructure and IoT devices
- Advanced Analytics:
 - Real-time fault severity assessment and impact analysis
 - Historical trend analysis for preventive maintenance planning



REFERENCES

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THANK YOU

