

---

# **CAPSTONE PROJECT**

## **POWER SYSTEM FAULT DETECTION AND CLASSIFICATION**

**Presented By:**

**1. Jagan Palaparthi-Raghu Engineering college-Datascience**

# 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

- **Example:** In modern power distribution systems, ensuring grid stability and reliability is critically dependent on the rapid and accurate detection of faults. Power system faults, such as **line-to-ground**, **line-to-line**, and **three-phase faults**, can cause severe disruptions, equipment damage, and widespread outages if not identified and addressed promptly.
- This project aims to develop a machine learning-based approach to **automatically detect and classify various types of electrical faults** using real-time measurement data. By leveraging electrical parameters such as **voltage and current phasors**, the proposed system should be capable of distinguishing between **normal operating conditions** and **abnormal fault conditions** in the distribution network.
- The ultimate objective is to enhance fault diagnosis capabilities to enable **faster protective actions**, reduce downtime, and improve the **operational efficiency and resilience** of the power grid.

# PROPOSED SOLUTION

- The system uses machine learning to detect and classify faults in a power distribution network based on voltage and current phasor data.
- **Data Collection:**  
Gather phasor data under normal and fault conditions from simulations or sensors.
- **Data Preprocessing:**  
Clean and normalize the data; extract key features relevant to fault types.
- **Modeling:**  
Train a classification model (e.g., Random Forest or Neural Network) to distinguish between normal, line-to-ground, line-to-line, and three-phase faults.
- **Deployment:**  
Implement the model in a real-time monitoring system for fault identification.
- **Evaluation:**  
Use metrics like accuracy, precision, and recall to evaluate model performance.
- **Result:**  
The system effectively classifies different fault types, enabling faster response and improving grid stability.

# SYSTEM APPROACH

The "System Approach" section outlines the overall strategy and methodology for developing and implementing the fault diagnosis system.

- System requirements**

- A computer system with a minimum of 8GB RAM, i5 or Ryzen 5 processor, and 64-bit OS
- Internet connection for accessing data and libraries
- Simulation tools (e.g., MATLAB, PSCAD) or sensors for data generation

- Libraries required to build the model**

- NumPy and Pandas – for data manipulation
- Matplotlib or Seaborn – for data visualization
- Scikit-learn – for machine learning models (e.g., Random Forest, SVM)
- TensorFlow or PyTorch – if deep learning models are used
- SciPy – for signal processing tasks (if needed)

# ALGORITHM & DEPLOYMENT

- In the Algorithm section, describe the machine learning algorithm chosen for classifying faults in the power distribution system. Here's an example structure for this section:
- **Algorithm Selection:**  
Provide a brief overview of the chosen algorithm (e.g., classification models like Random Forest, SVM, or a Neural Network) and justify its selection based on the problem statement and nature of phasor data.
- **Data Input:**  
Specify the input features used by the algorithm, such as voltage and current phasors, phase angles, and frequency deviations during both normal and faulty conditions.
- **Training Process:**  
Explain how the algorithm is trained using labeled data from fault simulations or historical system records. Highlight any specific techniques or considerations, such as data normalization, cross-validation, or hyperparameter tuning.
- **Prediction Process:**  
Detail how the trained algorithm identifies the fault type (e.g., line-to-ground, line-to-line, three-phase) using real-time input phasor data. Mention any real-time preprocessing done before classification.

# RESULT

- The developed machine learning model successfully detects and classifies different types of faults—such as line-to-ground, line-to-line, and three-phase faults—with high accuracy.
- The system was tested using labeled phasor data under both normal and fault conditions. It demonstrated strong performance in identifying fault types in real-time, with minimal false classifications.
- The results confirm that the proposed approach can significantly reduce fault detection time and improve the reliability and stability of the power distribution network.

# CONCLUSION

- This project successfully demonstrated the application of machine learning techniques for detecting and classifying different types of faults in a power distribution system using voltage and current phasor data.
- By leveraging classification algorithms and real-time data inputs, the system can accurately identify fault types such as line-to-ground, line-to-line, and three-phase faults. The approach enhances the speed and reliability of fault diagnosis, which is essential for maintaining grid stability.
- Overall, the model provides a promising foundation for intelligent power grid monitoring systems and opens avenues for further enhancement using deep learning and real-time deployment in smart grid environments.



# FUTURE SCOPE

- The model can be further enhanced in the following ways:
- **IoT Integration:** Use real-time data from smart sensors for faster fault detection.
- **Deep Learning:** Apply CNNs or LSTMs for better accuracy in complex fault patterns.
- **Smart Grid Deployment:** Scale the system for automated fault response in smart grids.
- **Multi-Fault Detection:** Extend the model to detect and classify multiple faults simultaneously.
- **Live Monitoring:** Add dashboards for real-time visualization and quick decision-making.

# REFERENCES

- Kundur, P. *Power System Stability and Control*. McGraw-Hill, 1994.
- Abur, A., & Exposito, A. G. *Power System State Estimation*. CRC Press, 2004.
- Scikit-learn Documentation – <https://scikit-learn.org>
- TensorFlow Documentation – <https://www.tensorflow.org>
- Open Power System Data – <https://data.open-power-system-data.org>
- IEEE papers on power system fault detection using machine learning

# IBM CERTIFICATIONS

In recognition of the commitment to achieve  
professional excellence



## JAGAN Palaparthi

Has successfully satisfied the requirements for:

---

### Getting Started with Artificial Intelligence

---



Issued on: Jul 20, 2025  
Issued by: IBM SkillsBuild

Verify: <https://www.credly.com/badges/023e3e89-c201-439c-bdc7-6b9493342d68>



# IBM CERTIFICATIONS

In recognition of the commitment to achieve  
professional excellence



## JAGAN Palaparthi

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/db33f2b6-6edc-486d-9e76-61f39a36aa42>



# IBM CERTIFICATIONS

7/24/25, 6:41 PM

Completion Certificate | SkillsBuild

**IBM SkillsBuild**

Completion Certificate



This certificate is presented to

**JAGAN Palaparthi**

for the completion of

**Lab: Retrieval Augmented Generation with LangChain**

(ALM-COURSE\_3824998)

According to the Adobe Learning Manager system of record

**Completion date:** 24 Jul 2025 (GMT)

**Learning hours:** 20 mins



**THANK YOU**