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# **PROJECT TITLE: POWER SYSTEM FAULT DETECTION AND CLASSIFICATION**

**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**

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# 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

## Proposed System

The proposed system aims to address the challenge of **detecting and classifying power system faults** to maintain grid stability and reduce downtime. It leverages data analytics and machine learning to automate fault identification. The solution consists of the following components:

### ◆ Data Collection:

Gather historical data on power distribution faults, including voltage, current, load, temperature, weather conditions, maintenance status, and component health.

### ◆ Data Preprocessing:

Clean the data to handle missing values and outliers.

Apply feature engineering to extract and encode key features that influence fault types.

### ◆ Machine Learning Algorithm:

Use **Snap Logistic Regression** to classify faults (e.g., Line Breakage, Overheating, Transformer Failure) based on input features.

Train and optimize the model using AutoML (Watsonx.ai) with feature selection and cross-validation.

### ◆ Deployment:

Deploy the trained model on IBM Watsonx.ai for real-time or batch inference.

Enable easy integration with grid monitoring systems for proactive maintenance alerts.

# SYSTEM APPROACH

- **Platform:** IBM Watsonx.ai Studio (AutoML)
- **Dataset:** Electrical measurement & fault classification data
- **Preprocessing:**
  - Feature Engineering (FE)
  - Hyperparameter Optimization (HPO)
- **Features Used:**
  - Voltage, Current, Power Load
  - Temperature, Wind Speed
  - Weather Condition, Maintenance Status, Component Health

# ALGORITHM & DEPLOYMENT

## ◆ Algorithm Selection:

Snap Logistic Regression was chosen for its speed, simplicity, and effectiveness in multiclass classification. It is ideal for structured electrical fault data and supports fast training with good accuracy.

## ◆ Data Input:

Input features include:

- Voltage (V), Current (A), Power Load (MW)
- Temperature (°C), Wind Speed (km/h)
- Weather Condition, Maintenance Status, Component Health

## ◆ Training Process:

The model was trained using IBM Watsonx.ai AutoML with:

- Cross-validation
- Feature Engineering (FE)
- Hyperparameter Optimization (HPO-1, HPO-2)

## ◆ Prediction Process:

The trained model predicts the fault type (e.g., Line Breakage, Overheating) based on real-time input data, enabling quick fault classification for power system reliability



# RESULT



[Deployment spaces](#) / [Power System\\_Deploy](#) / P9 - Random Forest Classifier: Power System








## Deployments

Model details

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 New deployment 

Name	Type	Status	Tags	Last modified	
 Power_Final	Online	 Initializing		6 seconds ago P J RAKSHITHA - (You)	

Items per page: 20 ▾ 1-1 of 1 items 1 of 1 pages  

## About this asset

### Name

P9 - Random Forest Classifier: Power System

### Description

No description provided.

### Asset Details

Type: wml-hybrid\_0.1

Model ID: d34a6a86-a1da-4b...

Software specification:

[hybrid\\_0.1](#) 

Hybrid pipeline software specifications:

[autoai-kb\\_rt24.1-py3.11](#)

### Tags

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### Source asset details

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Experiment summary

Pipeline comparison

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

Progress map ⓘ

Prediction column: Fault Type

```
graph LR; A[Read dataset] --> B[Split holdout data]; B --> C[Read training data]; C --> D[Preprocessing]; D --> E[Model selection]; E --> F[Hyperparameter optimization]; E --> G[Hyperparameter optimization]; F --> H[Feature engineering]; G --> I[Feature engineering]; H --> J[Hyperparameter optimization]; I --> K[Hyperparameter optimization]; J --> L[Ensemble creation]; K --> M[Ensemble creation]; L --> N[Random Forest Classifier]; M --> O[Random Forest Classifier]; N --> P[Snap Logistic Regression]; O --> P;
```

Relationship map

Swap view ↔

Experiment completed ✓

9 PIPELINES GENERATED

9 pipelines generated from algorithms. See pipeline leaderboard below for more detail.

Time elapsed: 2 minutes

View log

Save code

Pipeline leaderboard ▾



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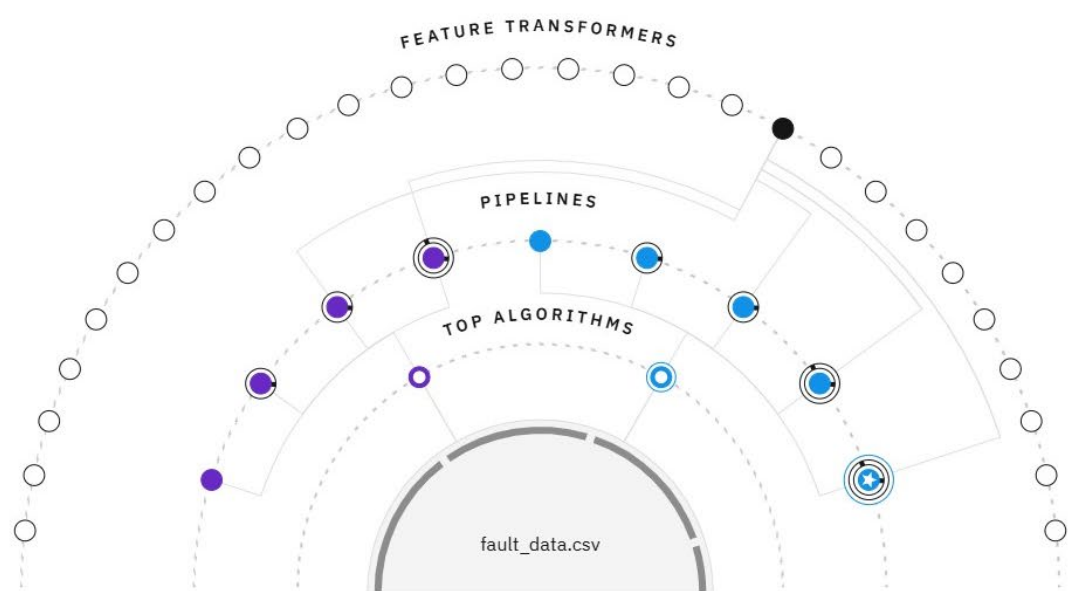
Experiment summary

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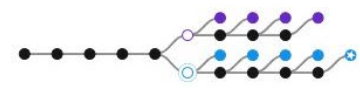
Prediction column: Fault Type



The diagram illustrates the machine learning pipeline structure. At the base is a semi-circle labeled 'fault\_data.csv'. Above it is a layer of dashed lines labeled 'TOP ALGORITHMS'. Above that is a layer of solid lines labeled 'PIPELINES'. At the top is a layer of dashed lines labeled 'FEATURE TRANSFORMERS'. The flow is indicated by arrows pointing upwards from the data source through the algorithms and pipelines to the feature transformers.

Progress map

Swap view



The progress map shows a sequence of steps represented by colored circles (black, purple, blue, green) connected by lines, indicating the flow of the experiment from data input to final output.

Experiment completed

9 PIPELINES GENERATED

9 pipelines generated from algorithms. See pipeline leaderboard below for more detail.

Time elapsed: 2 minutes

View log

Save code

Pipeline leaderboard

# Prediction results

Display format for prediction results

☒ Table view ☐ JSON view

☐ Show input data ⓘ

	prediction	probability
1	Overheating	[0.28970419343567033,0.3839578339127503,0.32633797265157927]
2		
3		
4		
5		
6		
7		
8		
9		
10		
11		

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Deployments

Model details

Input (1)

Column	Type
Component Health	other
Current (A)	double
Down time (hrs)	double
Duration of Fault (hrs)	double
Fault ID	other
Fault Location (Latitude, Longitude)	other
Maintenance Status	other
Power Load (MW)	double

About this asset

Name

P9 - Random Forest Classifier: Power System

Description

No description provided.

Asset Details

Type: wml-hybrid\_0.1

Model ID: d34a6a86-a1da-4b...

Software specification: hybrid\_0.1

Hybrid pipeline software specifications: autoai-kb\_rt24.1-py3.11

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Source asset details

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25 minutes ago by P J RAKSHITHA -

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Aug 4, 2025 by P J RAKSHITHA -

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# CONCLUSION

The proposed machine learning-based system effectively automates the **detection and classification of power distribution faults**, improving the speed and accuracy of fault diagnosis. By using Snap Logistic Regression and IBM Watsonx.ai's AutoML capabilities, the model successfully classifies different fault types based on electrical and environmental inputs. Although the initial accuracy is modest, the system demonstrates strong potential for real-time deployment, contributing to enhanced grid reliability, reduced downtime, and smarter energy management. Continuous data updates and model refinements can further improve system performance over time.

# FUTURE SCOPE

- **Model Enhancement:**  
Improve prediction accuracy by using advanced algorithms like Random Forest, XGBoost, or deep learning models (e.g., LSTM, CNN).
- **Fault Localization:**  
Extend the system to not just classify but also **pinpoint the exact location** of faults in the grid.
- **Integration with SCADA/IoT Systems:**  
Connect the model with real-time monitoring systems for **automated fault alerts** and proactive maintenance.
- **Time-Series & Streaming Data:**  
Incorporate live data streams to enable **real-time fault detection and continuous learning**.
- **Mobile & Web Interface:**  
Develop a dashboard or mobile app for engineers to view predictions, alerts, and analytics on the go.
- **Scalability & Cloud Deployment:**  
Deploy the solution on scalable cloud platforms for **wider geographic coverage** and reliability.

# REFERENCES

- IBM Watsonx.ai Documentation – <https://www.ibm.com/products/watsonx-ai>
- Scikit-learn: Machine Learning in Python – <https://scikit-learn.org>
- IEEE Papers on Power System Fault Classification
- “Introduction to Machine Learning” by Andreas C. Müller & Sarah Guido
- Project Dataset – *Custom fault data collected for power distribution analysis*

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**Learning hours:** 20 mins



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