

---

# **CAPSTONE PROJECT**

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

**Presented By:**

**1. Anand Jha –Trinity Institute Of Innovation In Professional  
Studies – CSE**

# OUTLINE

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

---

# PROBLEM STATEMENT

Design a machine learning model that can detect and classify different types of faults in a power distribution system. By using electrical measurement data like voltage and current phasors, the model will learn to tell the difference between normal conditions and various fault types, such as line-to-ground, line-to-line, or three-phase faults. This will help in quickly and accurately identifying faults, which is important for keeping the power grid stable and reliable.

# PROPOSED SOLUTION

The goal of the proposed system is to detect and classify faults in a power distribution system using machine learning. By analyzing electrical measurement data such as voltage and current phasors, the system can distinguish between normal operation and various fault types. The solution will involve the following components:

- **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 ) .
- **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 [Power System Fault Detection and Classification](#) prediction system. 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

- In the Algorithm section, describe the machine learning algorithm chosen for predicting Power System Fault Detection and Classification. Here's an example structure for this section:
- **Algorithm Selection:**

Random forest classification (or SVM based on performance) .
- **Data Input:**

Voltage, current and phasor measurements from the dataset.
- **Training Process:**

Supervised learning using labelled fault types.
- **Prediction Process:**

Model deployed on IBM watsonx studio with API endpoint for real-time prediction.

# RESULT

IBM watsonx.ai Studio

Search in your workspaces

Upgrade

Anand Jha's Account

Sydney

AJ

Projects / faulty\_detector10 / fault1

Configure AutoAI experiment

fault1

Autosaved: 9:22:47 pm

Add data source

Add files such as tabular data (CSV).

Browse

Select from project

fault\_data.csv

Size: 47.62 KB | Columns: 13

Configure details

Enable this option to predict future activity over a specified date/time range. Data must be structured and sequential. [Learn more](#)

YesNo

What do you want to predict?

Prediction column ⓘ

Fault Type

Prediction column: Fault Type

CUH remaining: 15.21 CUH

PREDICTION TYPE

Multiclass Classification

OPTIMIZED FOR

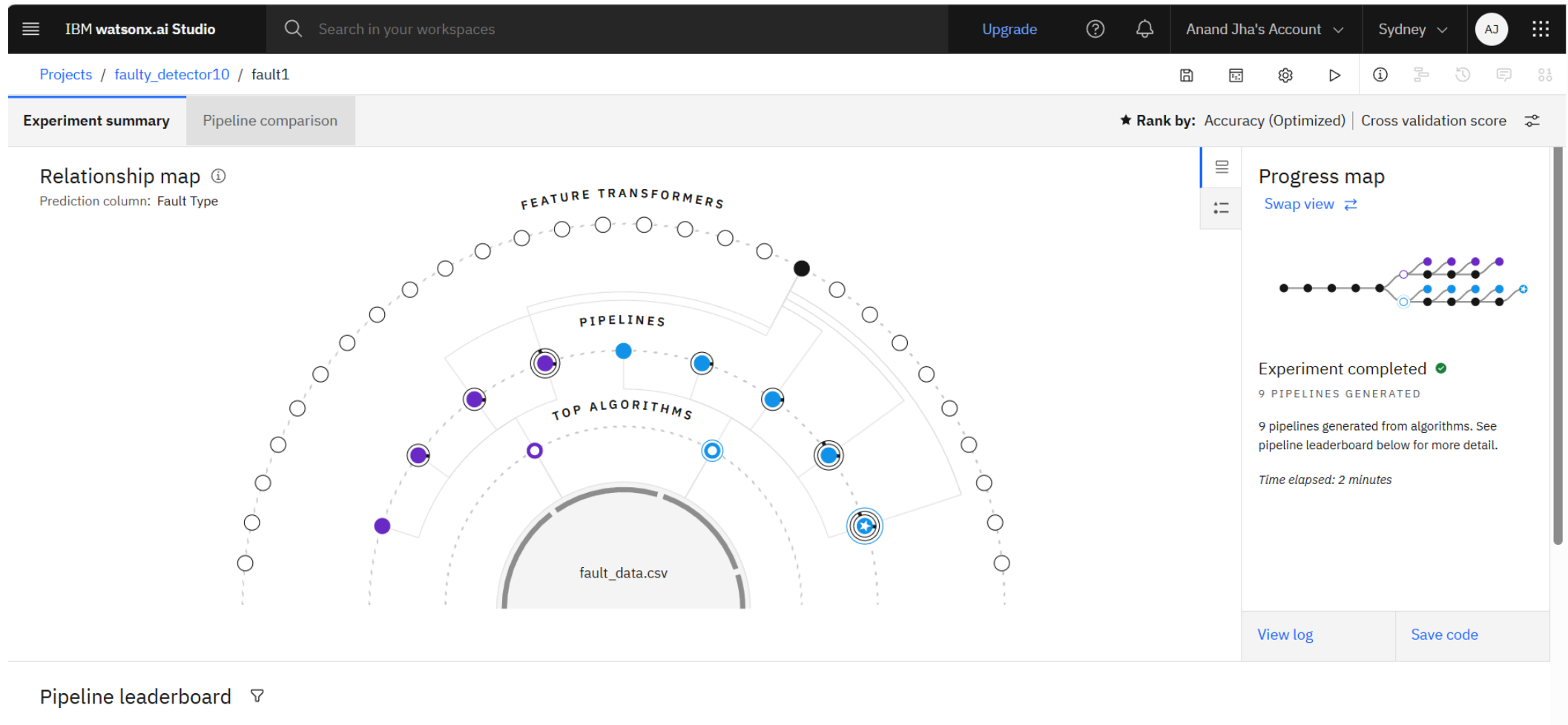
Accuracy & run time

Experiment settings

Run experiment

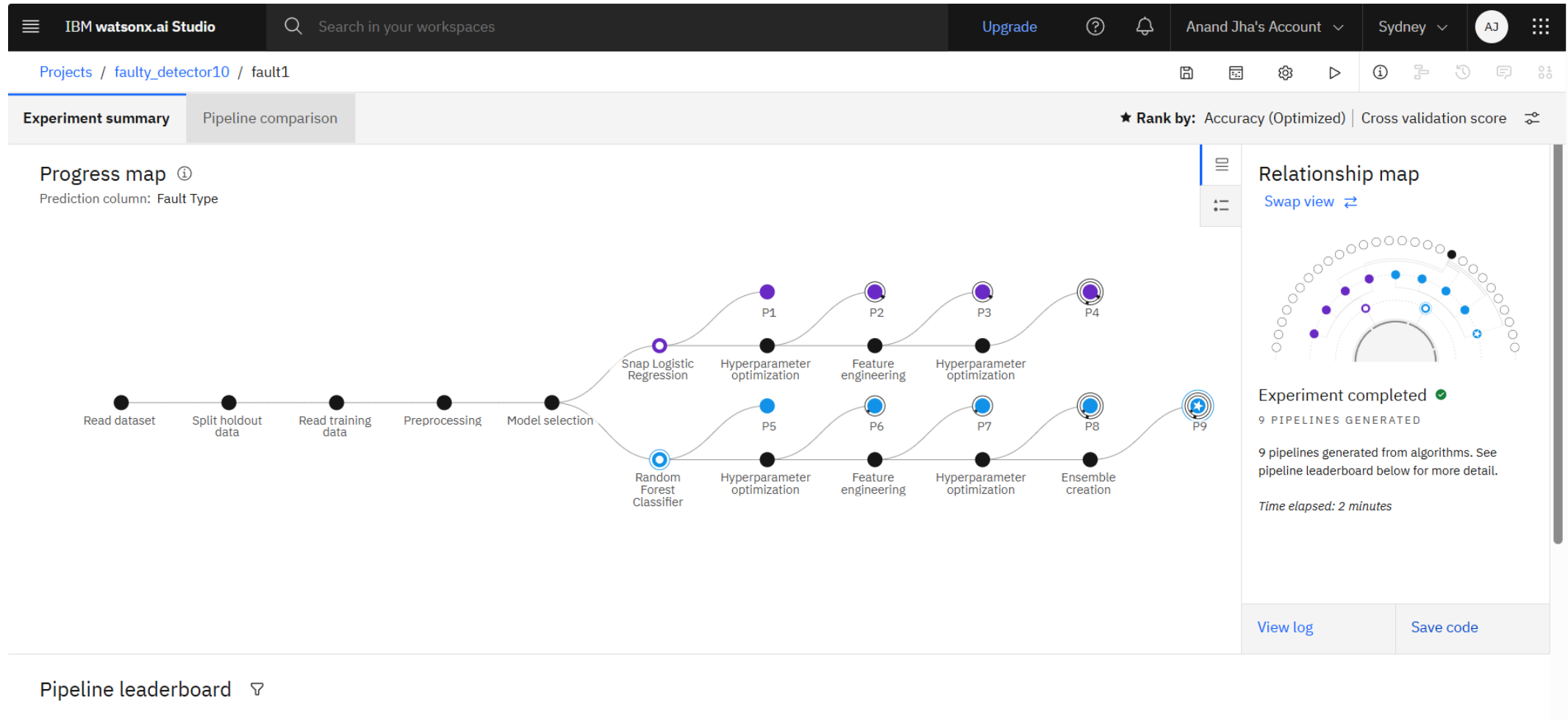
edunet  
foundation

# RESULT





# RESULT



# RESULT

IBM watsonx.ai Studio

Search in your workspaces

Upgrade

Anand Jha's Account

Sydney

AJ

Projects / faulty\_detector10 / fault1

Experiment summary

Pipeline comparison

★ Rank by:

Random Forest Classifier

Hyperparameter optimization

Feature engineering

Hyperparameter optimization

Ensemble creation

✓ Saved Model successfully.

P9 - Random Forest Classifier: fault1 was successfully saved to faulty\_detector10.

View in project

View log

Save code

Pipeline leaderboard

	Rank ↑	Name	Algorithm	Specialization	Accuracy (Optimized) Cross Validation	Enhancements	Build time
★	1	Pipeline 9	Batched Tree Ensemble Classifier (Random Forest Classifier)	INCR	0.409	HPO-1 FE HPO-2 BATCH	00:00:40
	2	Pipeline 8	Random Forest Classifier		0.409	HPO-1 FE HPO-2	00:00:37
	3	Pipeline 4	Snap Logistic Regression		0.393	HPO-1 FE HPO-2	00:00:26
	4	Pipeline 3	Snap Logistic Regression		0.393	HPO-1 FE	00:00:23

# RESULT

IBM watsonx.ai Studio

Search in your workspaces

Upgrade

Anand Jha's Account

Sydney

AJ



Deployment spaces / fault\_deploy / P9 - Random Forest Classifier: fault1

Deployments

Model details

Search

New deployment

Name	Type	Status	Tags	Last modified	
 fault_deploy2	Online	 Deployed		14 seconds ago Anand Jha (You)	<div></div>

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

About this asset

Name

P9 - Random Forest Classifier: fault1

Description

No description provided.

Asset Details

Type: wml-hybrid\_0.1

Model ID: 816c514c-ebb1-4e...

Software specification:  
hybrid\_0.1

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

Tags

Add tags to make assets easier to find.

Source asset details

Last modified  
1 minute ago by Anand Jha

Created on  
Aug 4, 2025 by Anand Jha

# RESULT

IBM watsonx.ai Studio

Search in your workspaces

Upgrade

1

Anand Jha's Account

Sydney

AJ

Deployment spaces / fault\_deploy / P9 - Random Forest Classifier: fault1

fault\_deploy2

Deployed

Online

API reference

Test

Enter input data

Text

JSON

Enter data manually or use a CSV file to populate the spreadsheet. Max file size is 50 MB.

Download CSV template

Browse local files

Search in space

Clear all

	Fault ID (other)	Fault Location (Latitude, Longitude) (other)	Voltage (V) (double)	Current (A) (double)	Power Load (MW) (double)	Temperature (°C) (double)	Wind Speed (km/h) (double)
1	F003	(34.0525, -118.244)	2100	230	55	35	25
2	F004	(34.055, -118.242)	2050	240	48	23	10
3	F005	(34.0545, -118.243)	1900	190	50	30	18
4	F006	(34.05, -118.24)	2150	220	52	32	22
5							

4 rows, 12 columns

Predict

# RESULT

## Prediction results

Close



Display format for prediction results

☒ Table view ☐ JSON view

☐ Show input data

	prediction	probability
1	Overheating	[0.29108219796175333,0.37123284986298116,0.33768495217526545]
2	Line Breakage	[0.5449969393042704,0.27473275868914016,0.18027030200658925]
3	Transformer Failure	[0.35086020700535114,0.27274207234586495,0.37639772064878374]
4	Line Breakage	[0.3460042759167417,0.3280694101262557,0.32592631395700233]
5		
6		
7		
8		
9		
10		
11		

Download JSON file

---

# CONCLUSION

- The machine learning model effectively identifies and classifies power system faults , improving the fault search time.
- This contribute as a more stable and reliable power grid.

# FUTURE SCOPE

- Use real-time sensor data instead of static dataset
- Integrate with smart grid system
- Future systems can automatically isolate faults and restore power quickly.
- It can be upgraded to predict faults before they happen, reducing power outages.
- The model can keep learning from new data to handle new fault types and changes in the grid.

# REFERENCES

- Kaggle document : -<https://www.kaggle.com/datasets/ziya07/power-system-faults-dataset>
- IBM CLOUD and watsonx documentation
- **Razaque & Rizwan (2021)** – Applied ML algorithms in smart grids for fault detection.
- **Abdullah et al. (2020)** – Reviewed various ML methods for fault detection in power systems.



# IBM CERTIFICATIONS

In recognition of the commitment to achieve  
professional excellence



Anand Jha

Has successfully satisfied the requirements for:

Getting Started with Artificial Intelligence



Issued on: Jul 21, 2025  
Issued by: IBM SkillsBuild

Verify: <https://www.credly.com/badges/6ffe1fc7-bd56-4148-bda2-c6a592385086>



# IBM CERTIFICATIONS

In recognition of the commitment to achieve  
professional excellence



Anand Jha

Has successfully satisfied the requirements for:

Journey to Cloud: Envisioning Your Solution



Issued on: Jul 21, 2025  
Issued by: IBM SkillsBuild

Verify: <https://www.credly.com/badges/4fbfabb6-fa63-4f90-8074-d946083c9961>



# IBM CERTIFICATIONS

IBM SkillsBuild	Completion Certificate
-----------------	------------------------



This certificate is presented to

Anand Jha

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**