

# CatBoost Algorithm for Transmission line Fault Detection and Classification

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# Thesis Outline

Title	01
Introduction	03
Literature Review	03
Methodology	04
Results	05
Discussion	06
Conclusions	07
Questions	08

# Introduction

Typical fault detection procedure-

01

Needs long time

02

Hampers reliability

03

Create instability,  
discontinuity

Consequences,

01

Transmission & distribution  
loss 8.31%

02

Frequent power  
outages

03

Hampers businesses &  
discourages foreign investment

# Introduction

Machine learning procedure-

01

Needs short time

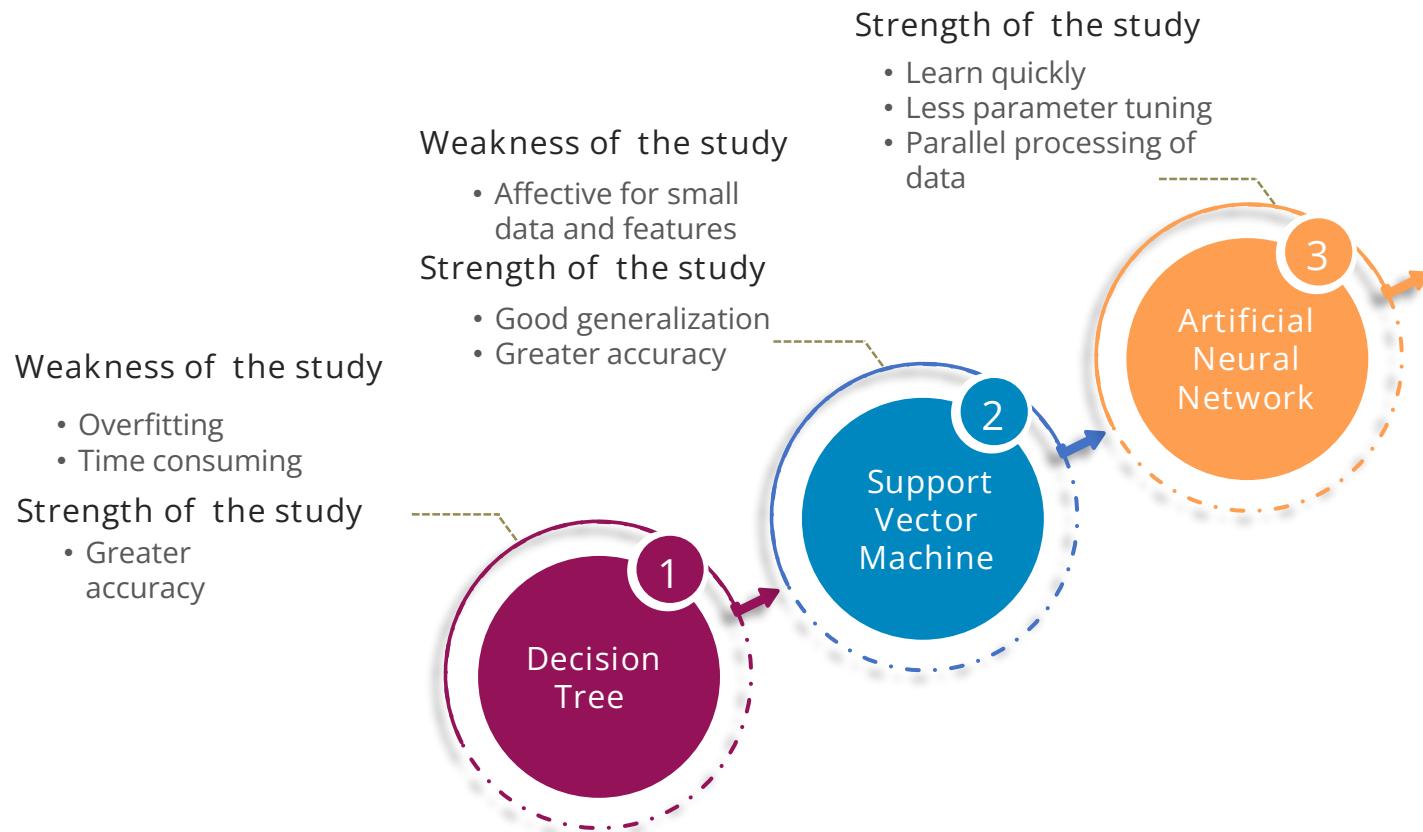
02

Increases reliability

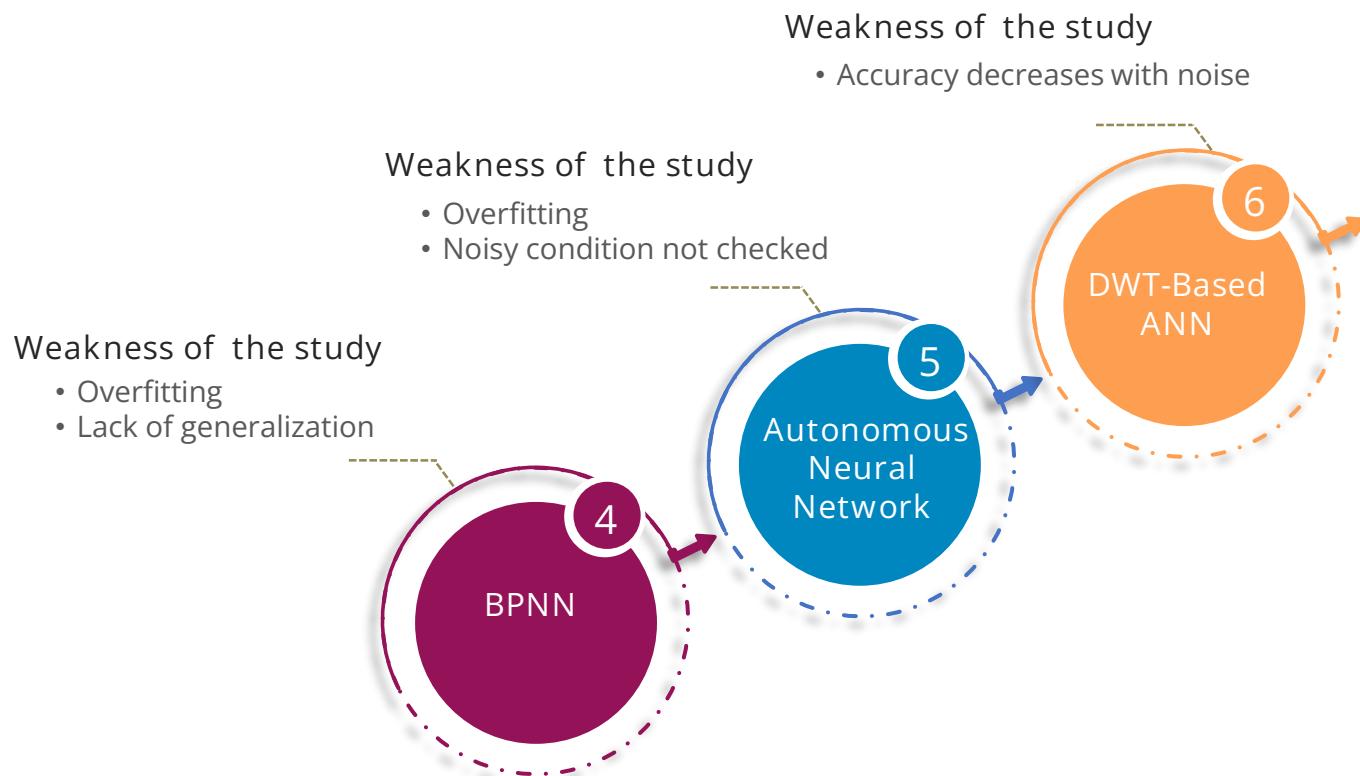
03

Provide stability,  
continuity

# Literature Review



# Literature Review



# Literature Review

## Weakness of the study

- High sampling rate for good accuracy
- Increase size and time

## Strength of the study

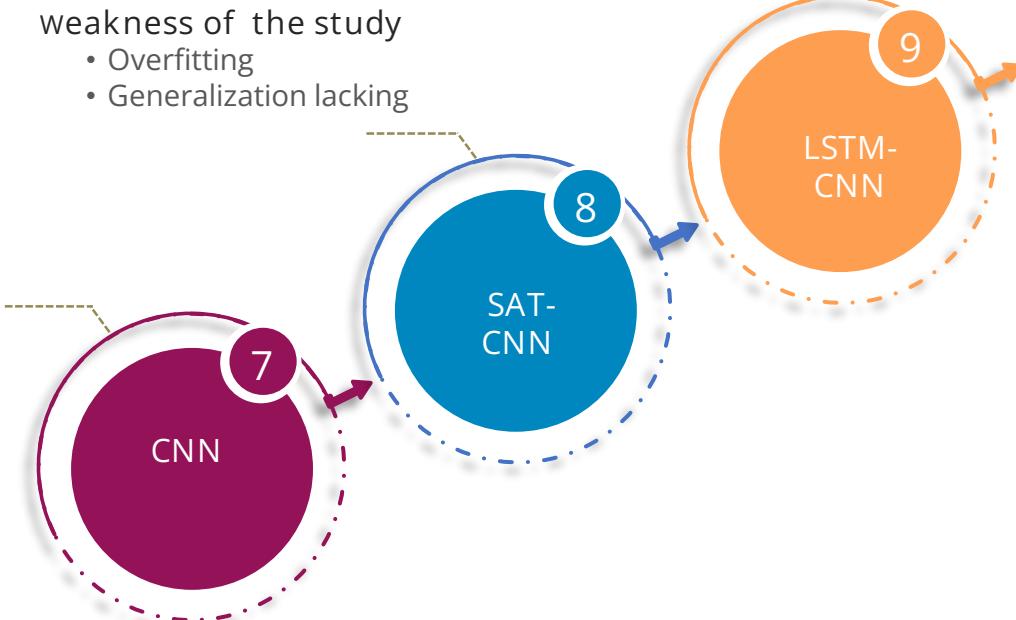
- Ultrafast
- Precise

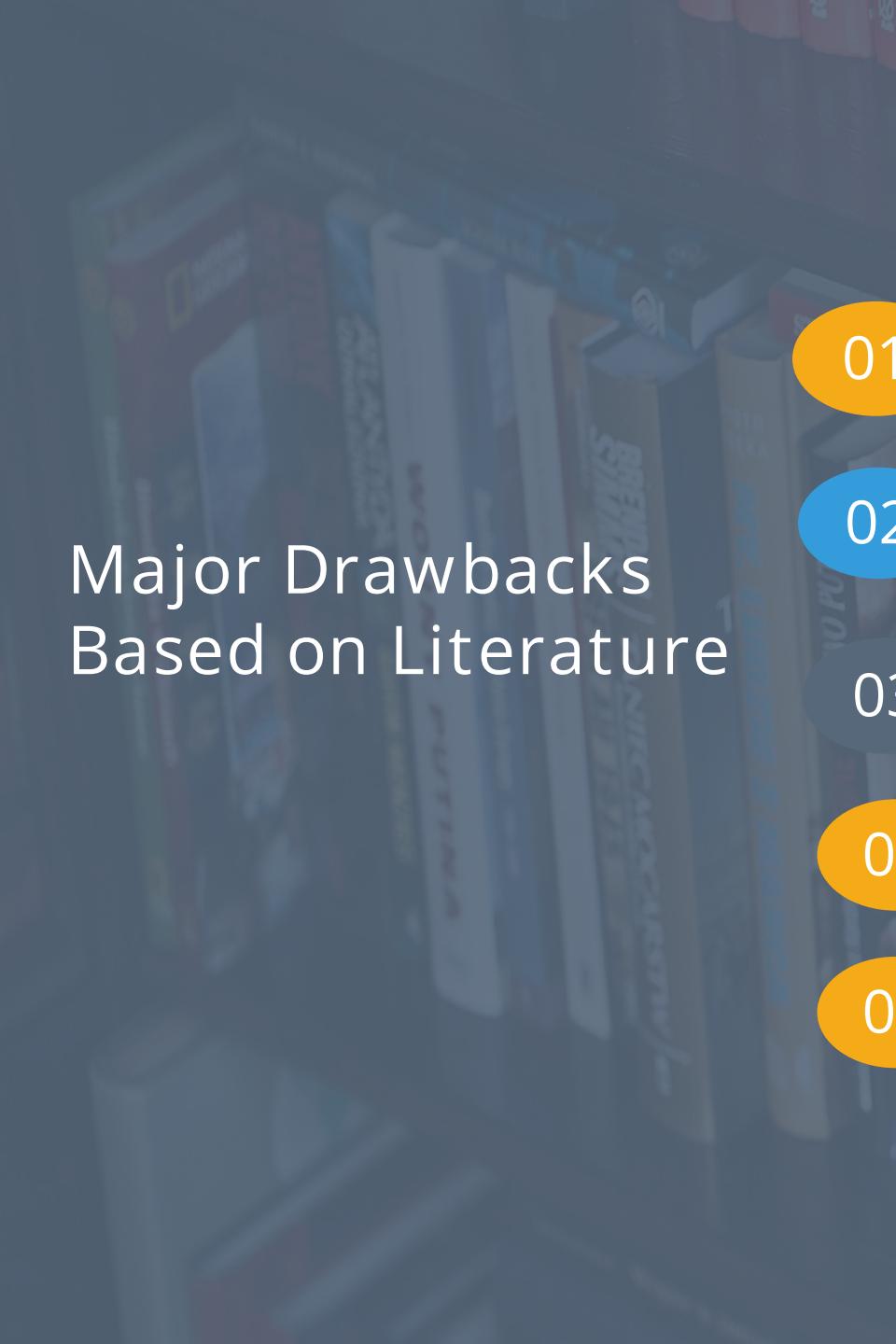
## weakness of the study

- Overfitting
- Generalization lacking

## Weakness of the study

- Learn less from small data
- Poor performance for imbalanced data





## Major Drawbacks Based on Literature

01

Overfitting

02

Accuracy decreases with noise

03

Less performance in data varieties

04

Less learning for small data

05

Implicit pre-processing of data

## Hypothesis (Proposed model)

01

Overfitting detector

02

Maintains accuracy with noise

03

Good performance in data varieties

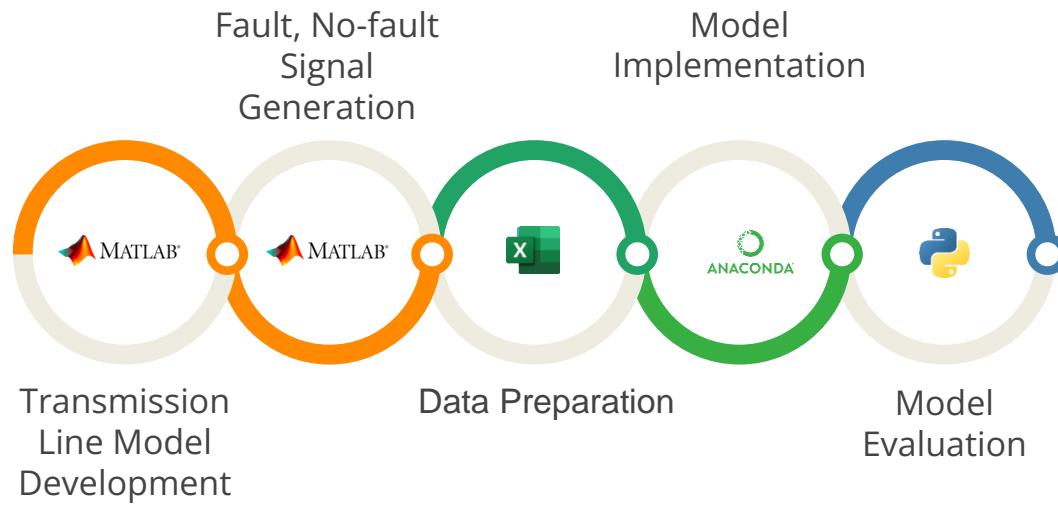
04

Good learning for small and big data

05

Explicit pre-processing of data

# Methodology



# Transmission Line Model Development

01

01

## Operating at

- I. 132kV,
- II. 50Hz
- III. length of 150km.

02

## Positive and zero sequence parameters

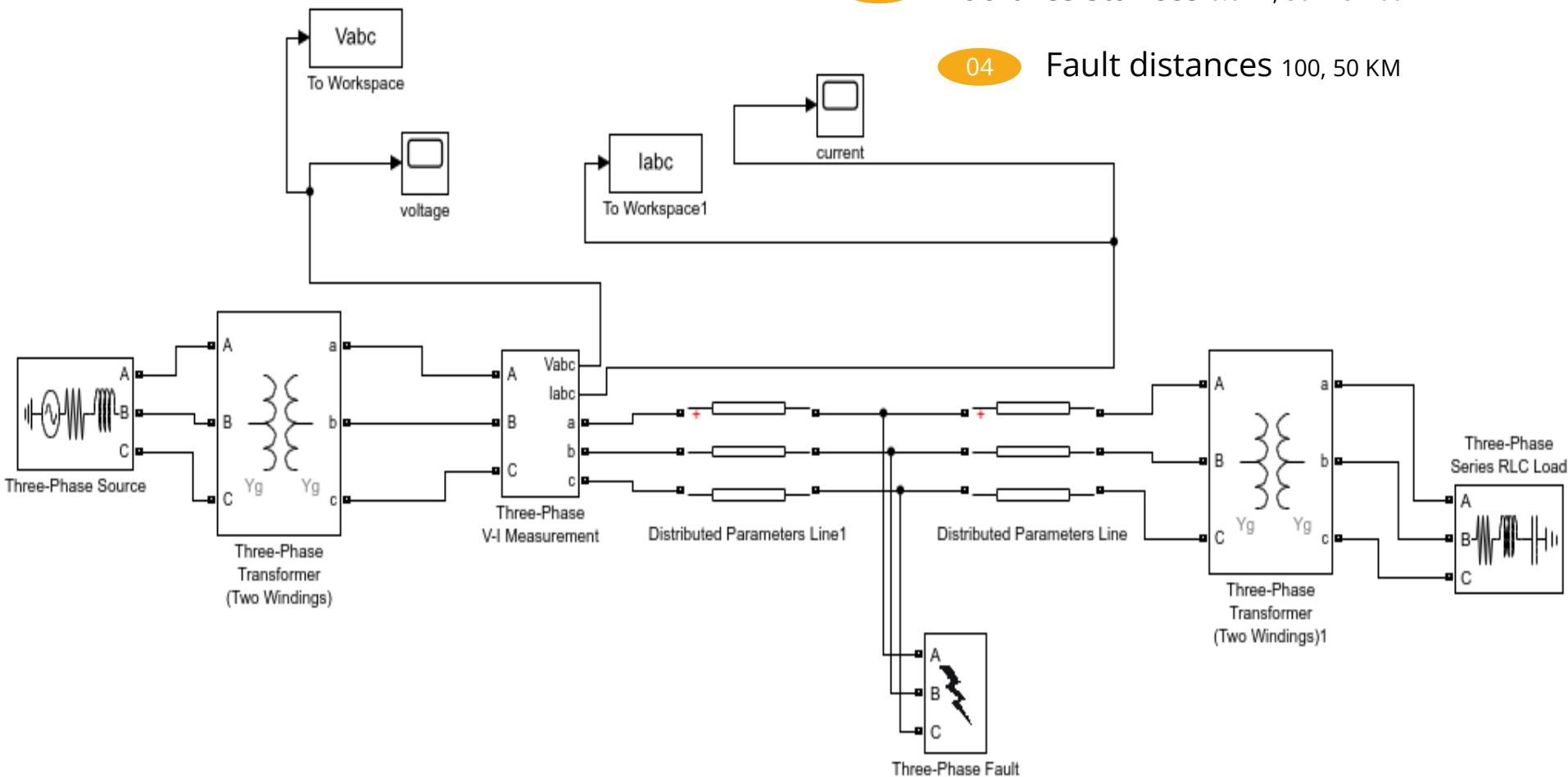
- I. 0.044965  $\Omega/\text{km}$  and 0.11240  $\Omega/\text{km}$  resistances
- II. 0.00101 H/km and 0.00202 H/km inductances
- III. 7.471nF/km and 4.394nF/km capacitances

03

Fault resistances 0.01 $\Omega$ , 50  $\Omega$  & 100

04

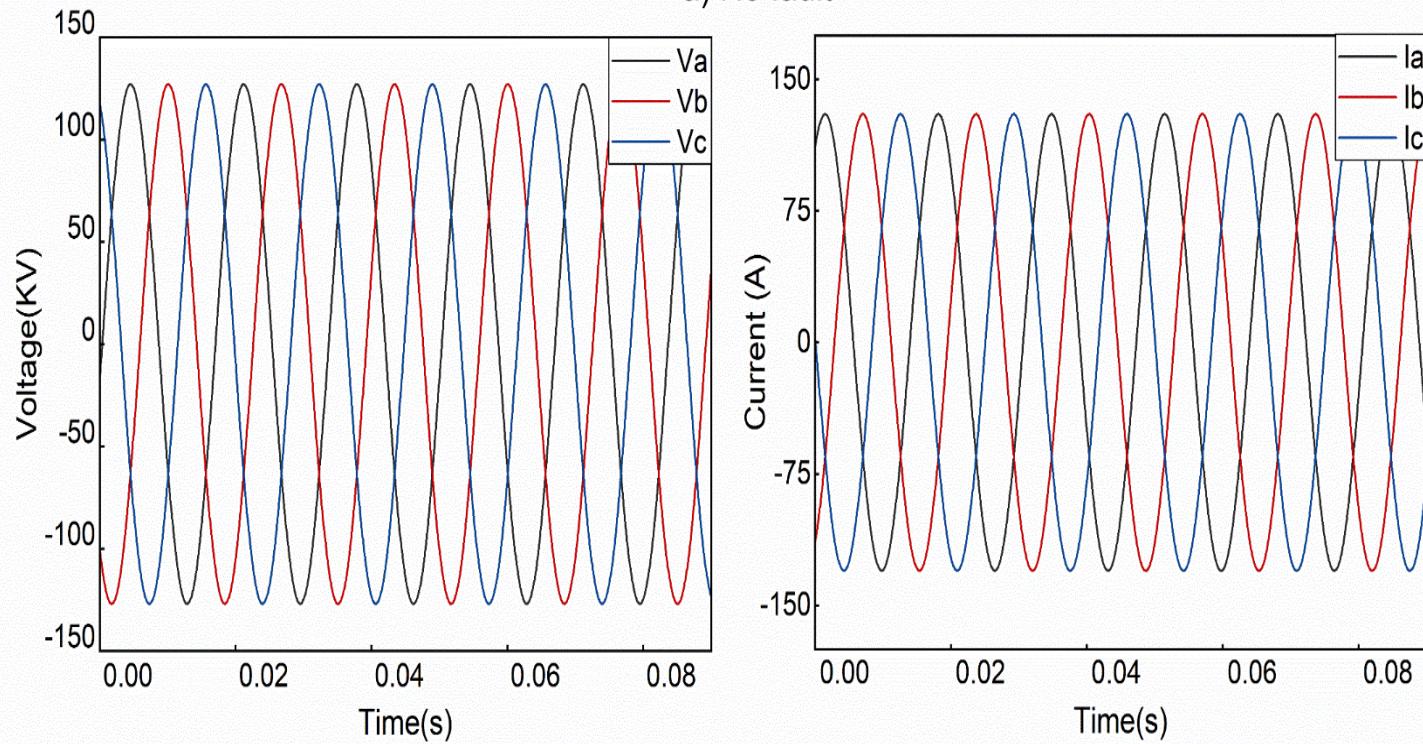
## Fault distances 100, 50 KM



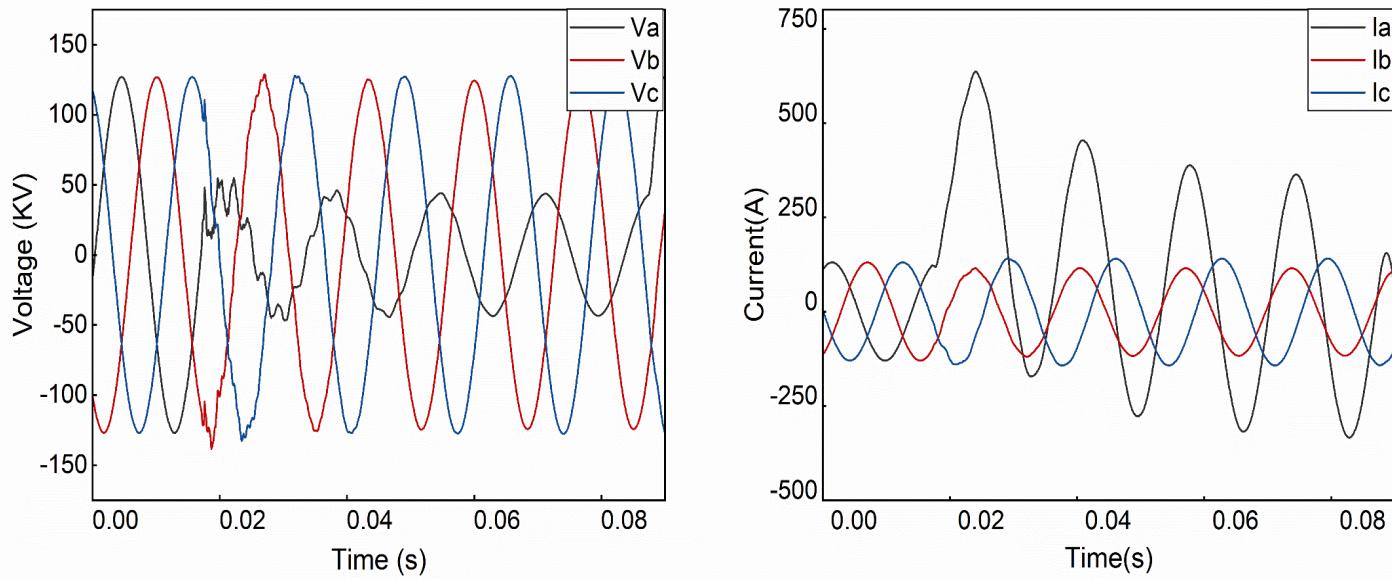
# Fault, No-fault Signal Generation

02

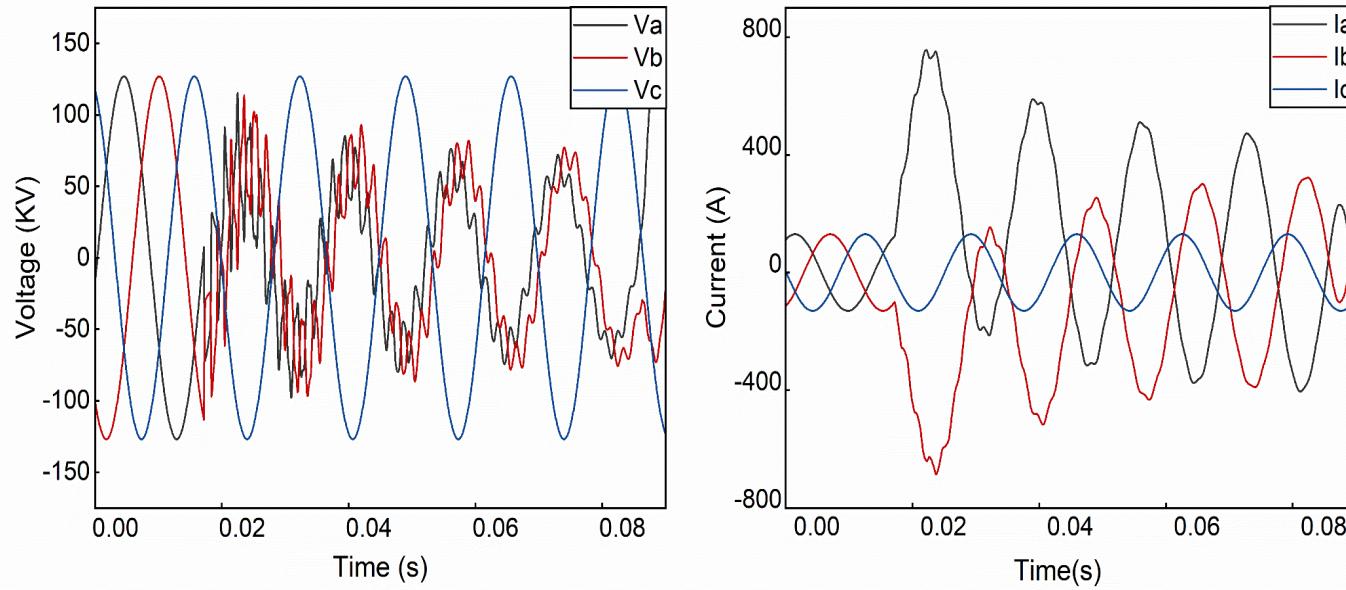
a) No fault



b) LG fault



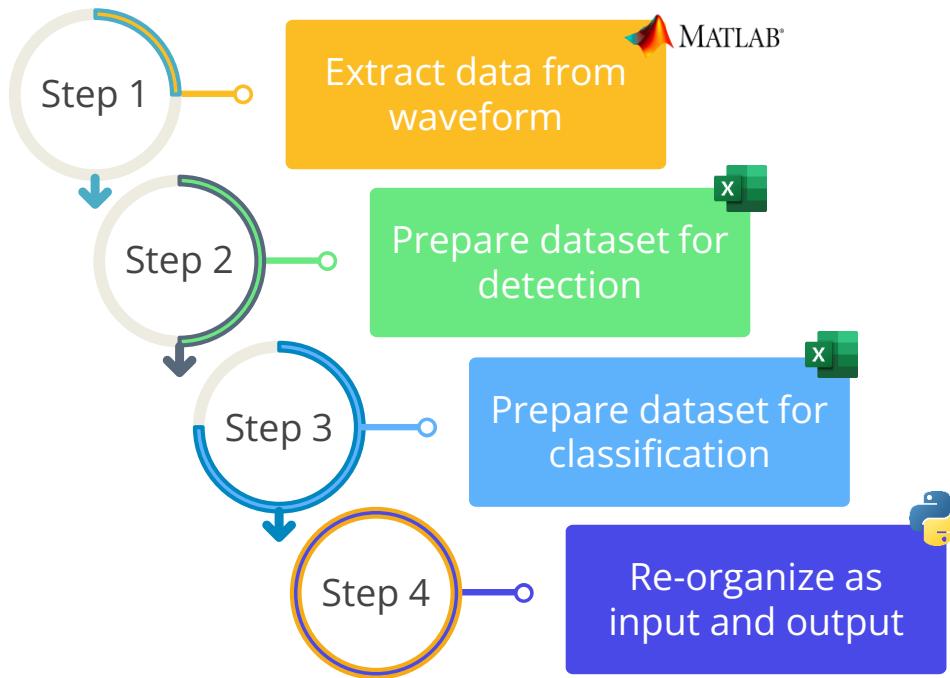
c) LL fault



# Data Preparation

03

✓ Sampling frequency 12khz



## Detection dataset

Input



Va	Vb	Vc	Ia	Ib	Ic	OUTPUT	
49.17	-	125.96	76.79	130.1718	-67.1407	-63.0311	0
53.55	-	126.47	72.92	130.1688	-62.8889	-67.2798	0
55.71	-	126.66	70.95	130.0978	-60.7291	-69.3687	0
14.24	-	125.99	76.76	135.0093	-71.3579	-67.2375	1
17.88	-	122.97	76.44	142.1678	-67.5876	-71.9696	1
15.78	-	123.58	74.05	146.0309	-65.7525	-74.3848	1

→ No fault

→ Fault

## Classification dataset

	Input										Output	
	Va	Vb	Vc	Ia	Ib	Ic	A	B	C	G	faultType	
0	49170.90541	-125961.168300	76790.262910	130.171817	-67.140715	-63.031102	0	0	0	0	0000	
1	53547.88759	-126472.806200	72924.918580	130.168754	-62.888916	-67.279838	0	0	0	0	0000	
2	55708.23313	-126661.265900	70953.032720	130.097849	-60.729111	-69.368738	0	0	0	0	0000	
3	59968.79332	-126903.134700	66934.341370	129.817420	-56.345527	-73.471893	0	0	0	0	0000	
4	64144.14075	-126964.684200	62820.543410	129.352520	-51.881845	-77.470675	0	0	0	0	0000	
...	...	...	...	...	...	...	...	...	...	...	...	
2495	-14967.52182	27368.963230	-12401.441410	484.939112	-88.531589	-396.407523	1	1	1	0	1110	
2496	-17222.85093	18231.551560	-1008.700629	477.660245	-70.840372	-406.819873	1	1	1	0	1110	
2497	-18100.79030	14054.227380	4046.562918	473.898117	-61.327590	-412.570527	1	1	1	0	1110	
2498	-19444.64246	6202.593350	13242.049110	466.062186	-41.094758	-424.967429	1	1	1	0	1110	
2499	-20291.56226	-1070.148109	21361.710360	457.737625	-19.381800	-438.355825	1	1	1	0	1110	

Categorical representation	Fault type(output)
0000	No fault
1001	LG fault
1100	LL fault
1101	LLG fault
1110	LLL fault

# Model Implementation

04

Tools  
1. Python  
2. Anaconda  
3. Jupiter notebook

**IMPORT LIBRARY** Matplotlib,  
Pandas, Numpy, CatBoost  
classifier, Sea-born, etc.

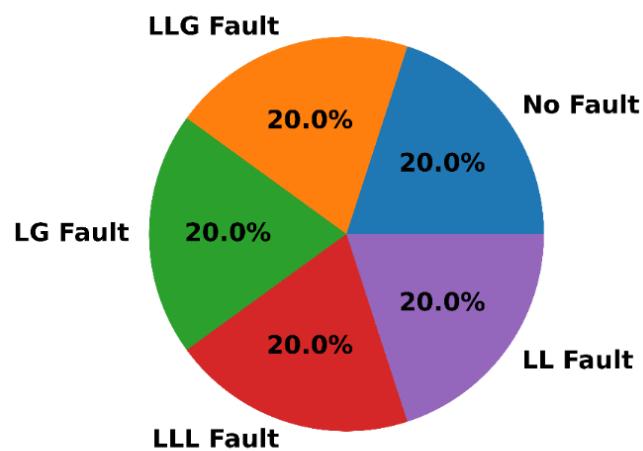
**SPLIT DATA** 80% for train &  
20% for test

**DATA VARIETIES** 1. Balanced  
2. Imbalanced 3. Noisy

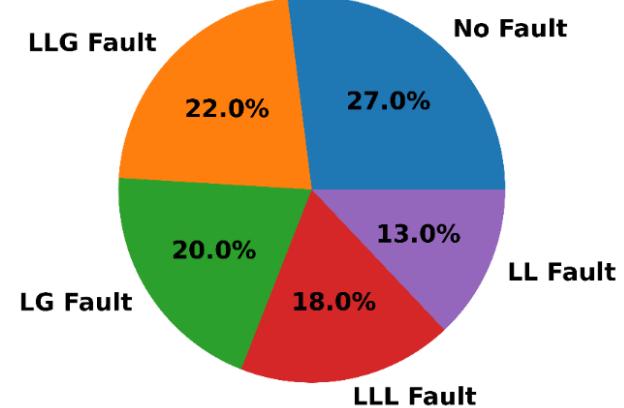
**MODEL OPTIMIZATION**  
Hyper-parameter tuning

**EVALUATE THE MODEL**

## Data state variation

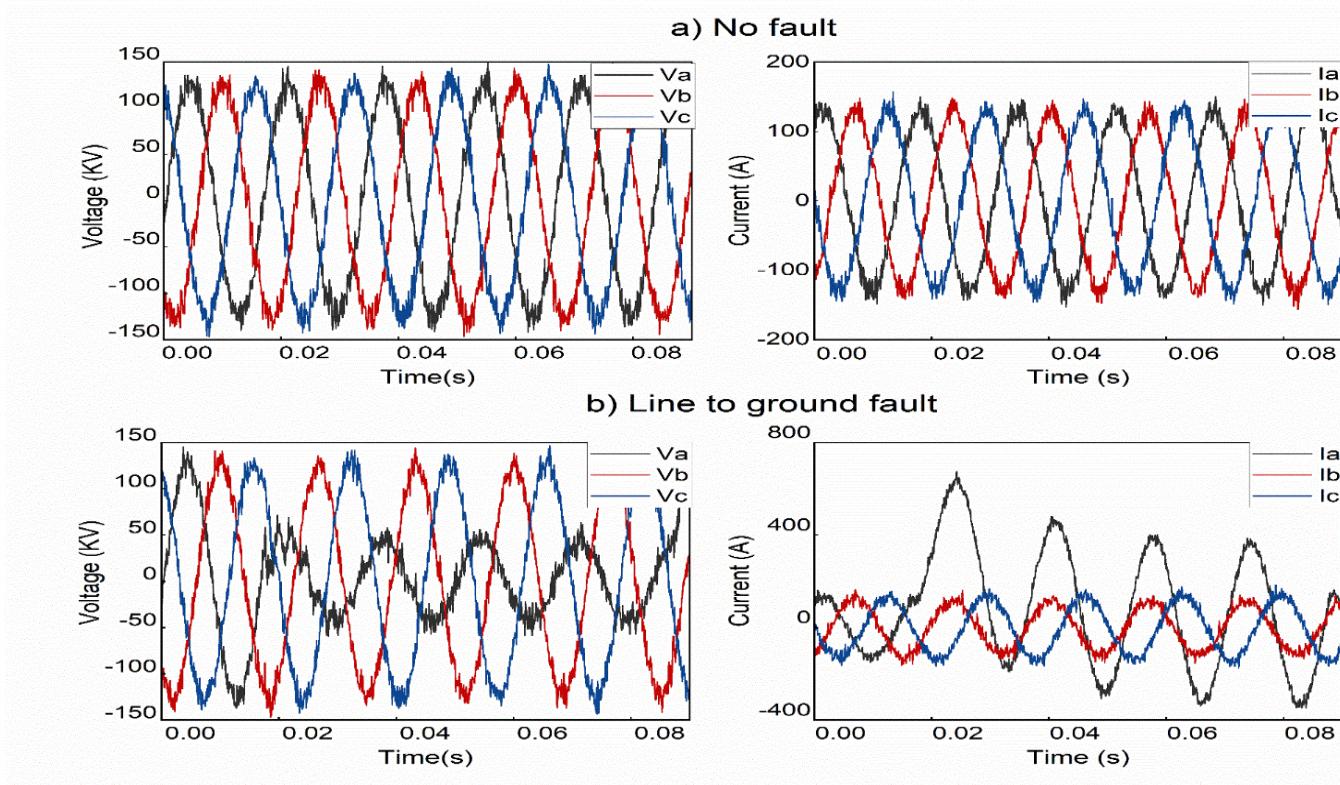


a) Balanced condition



b) Imbalanced condition

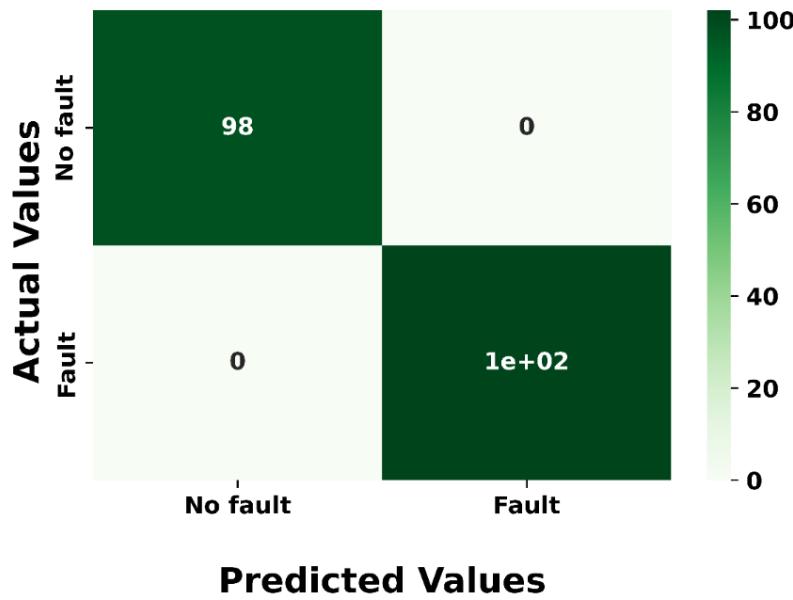
- ✓ Additive White Gaussian Noise
- ✓ 37,20 dB SNR



Sample waveform for noisy condition data collection

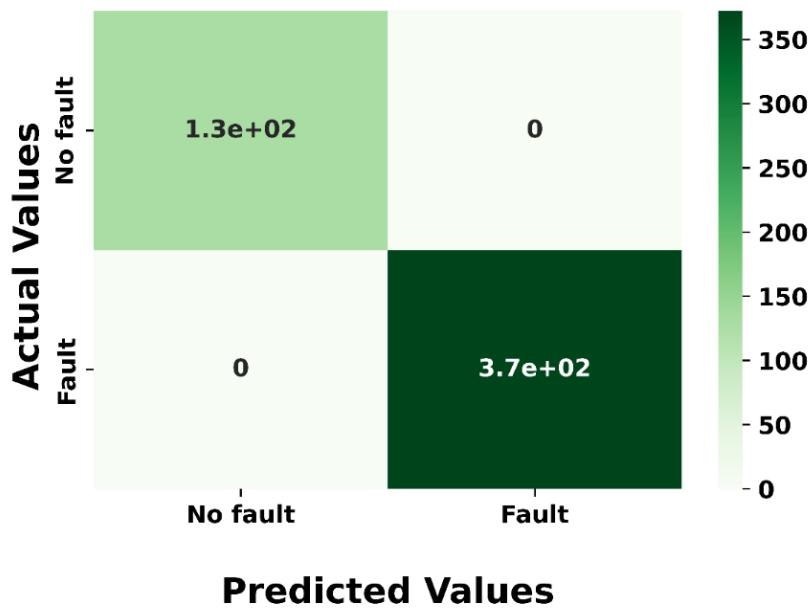
# Results (Detection)

**For Balanced**



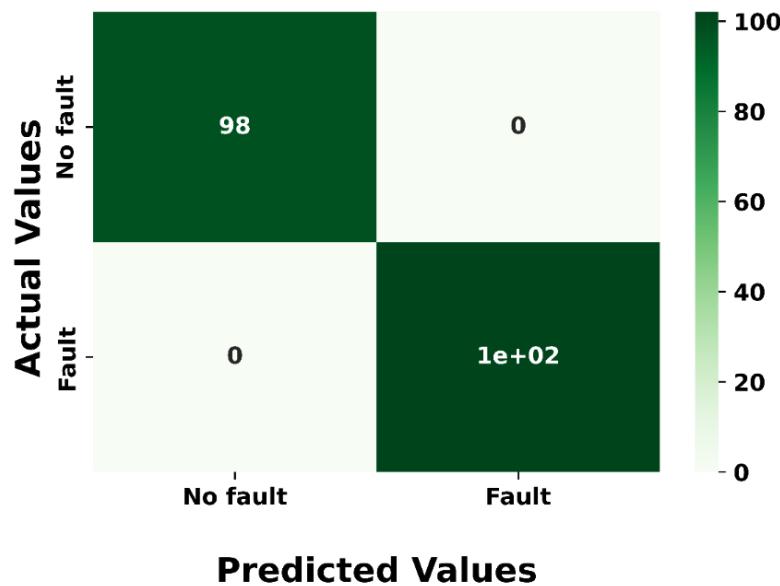
# Results (Detection)

**For Imbalanced**



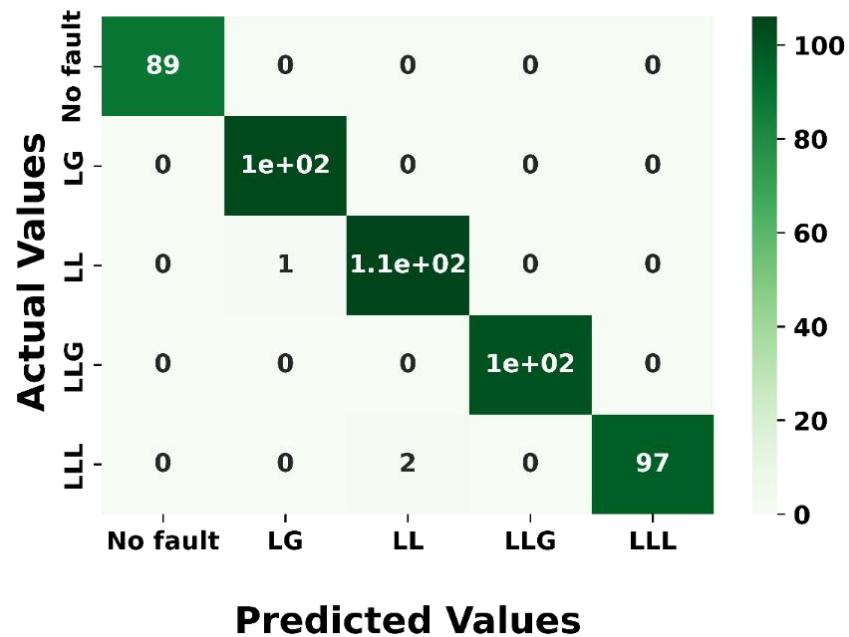
# Results (Detection)

**For Noise**



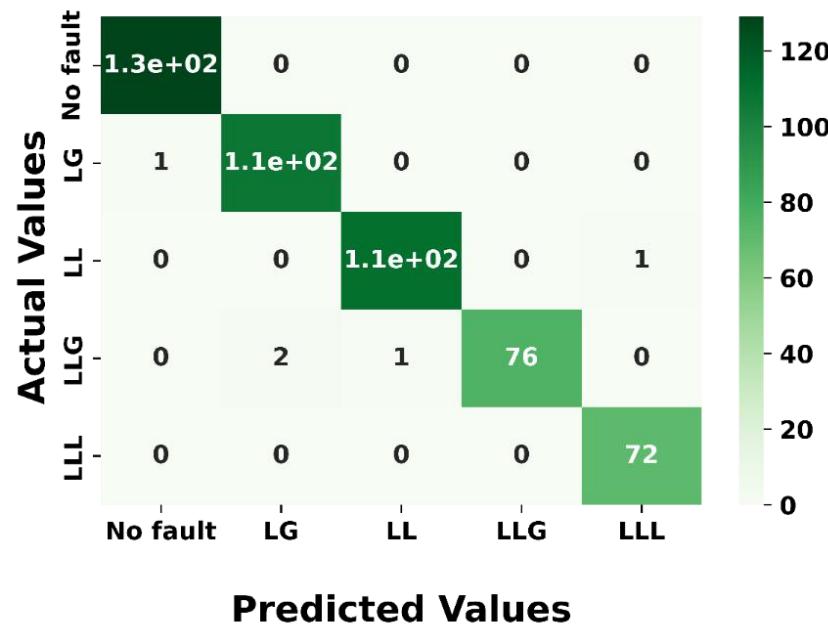
# Results (Classification)

For Balanced



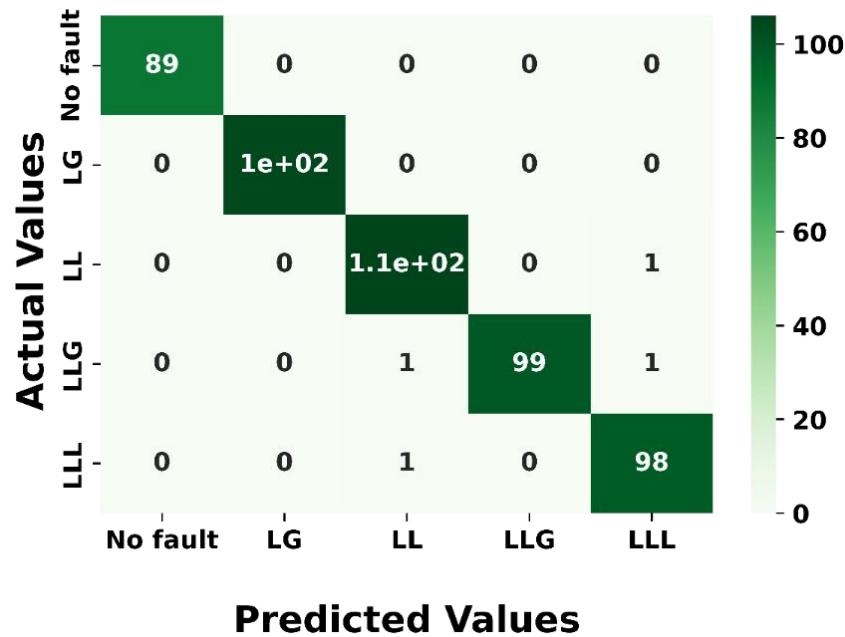
# Results Classification

**For Imbalanced**

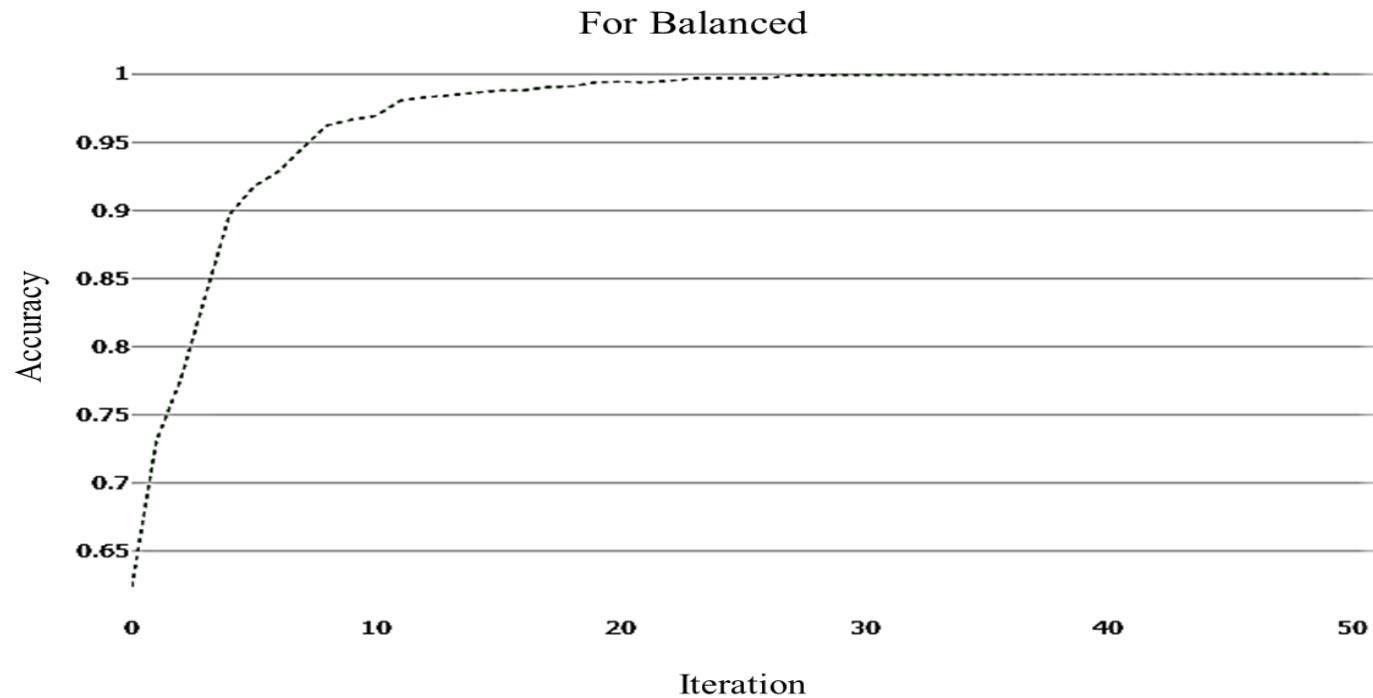


# Results Classification

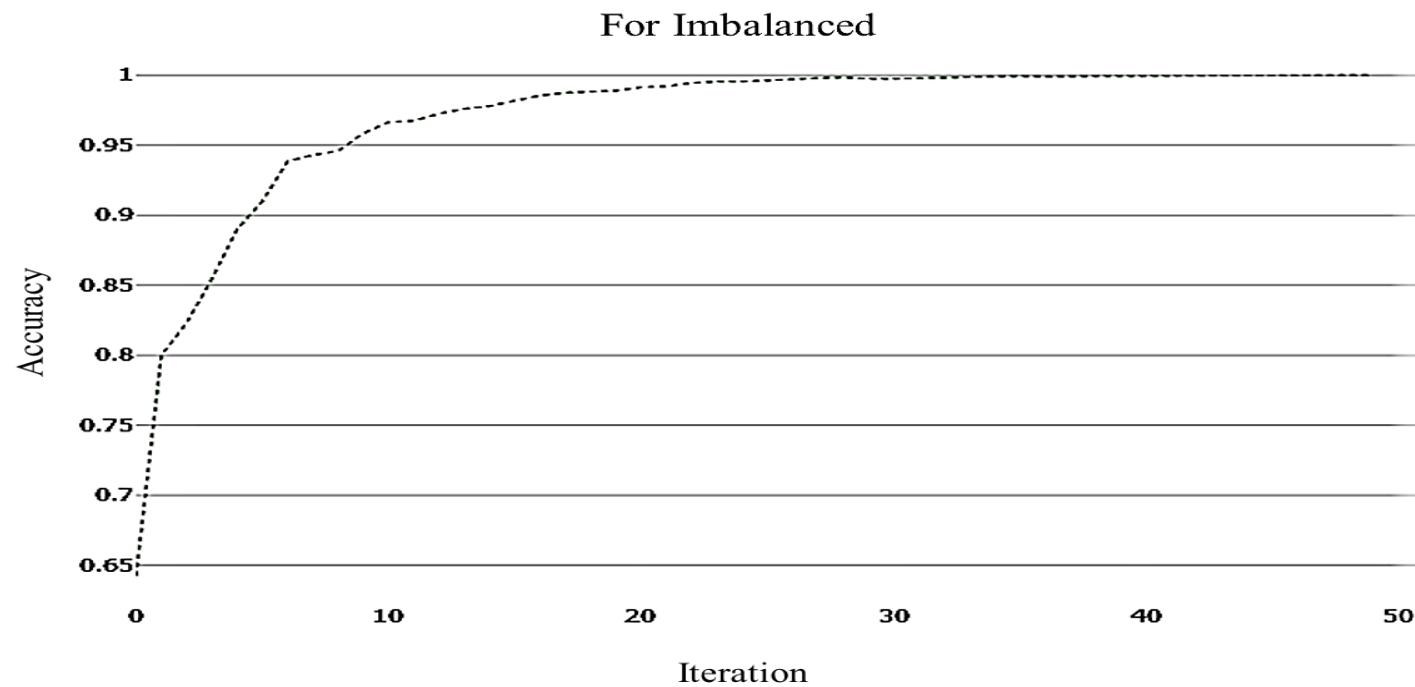
For Noise



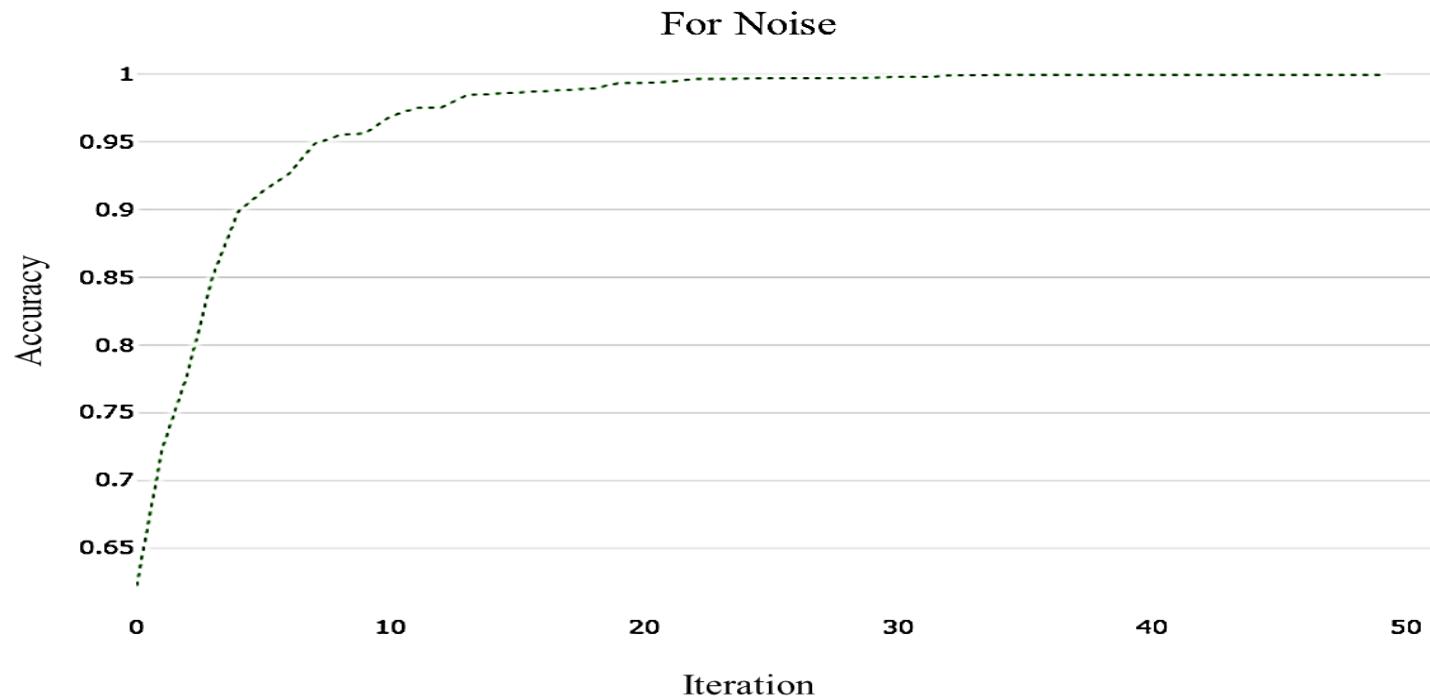
# Results (Classification)



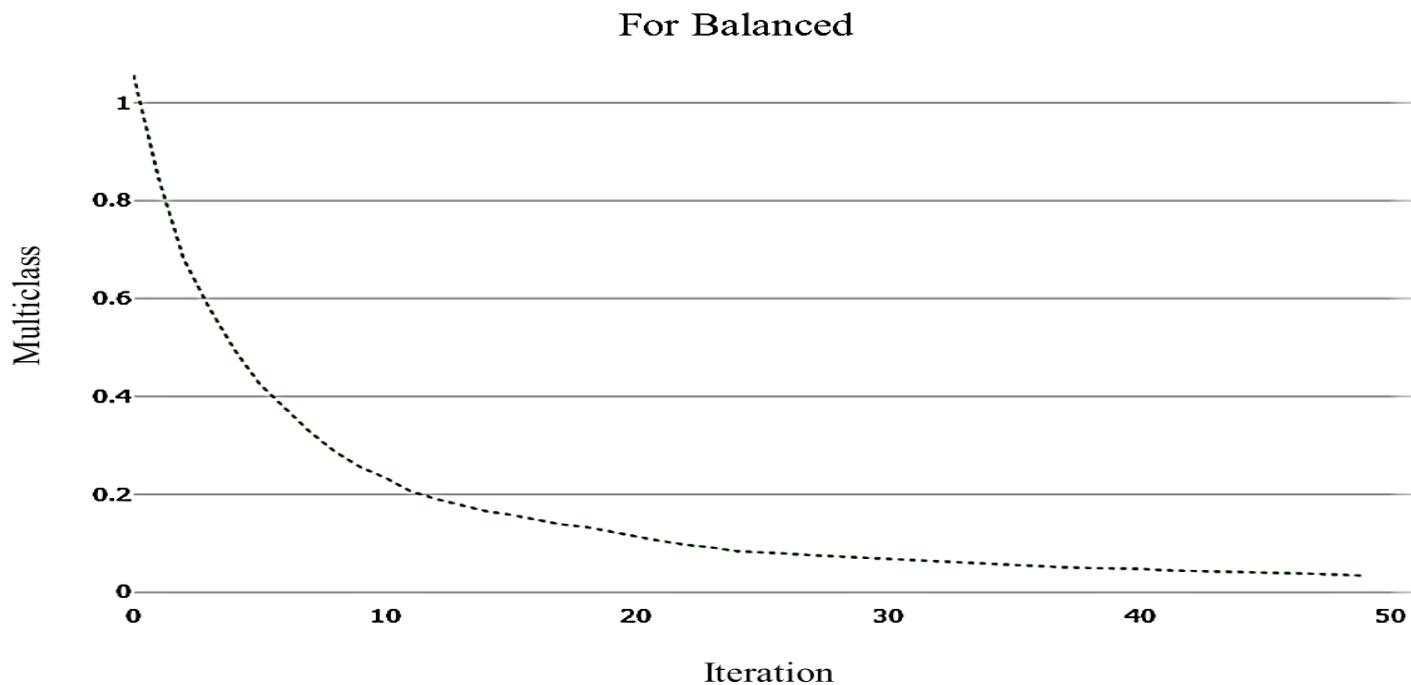
# Results Classification



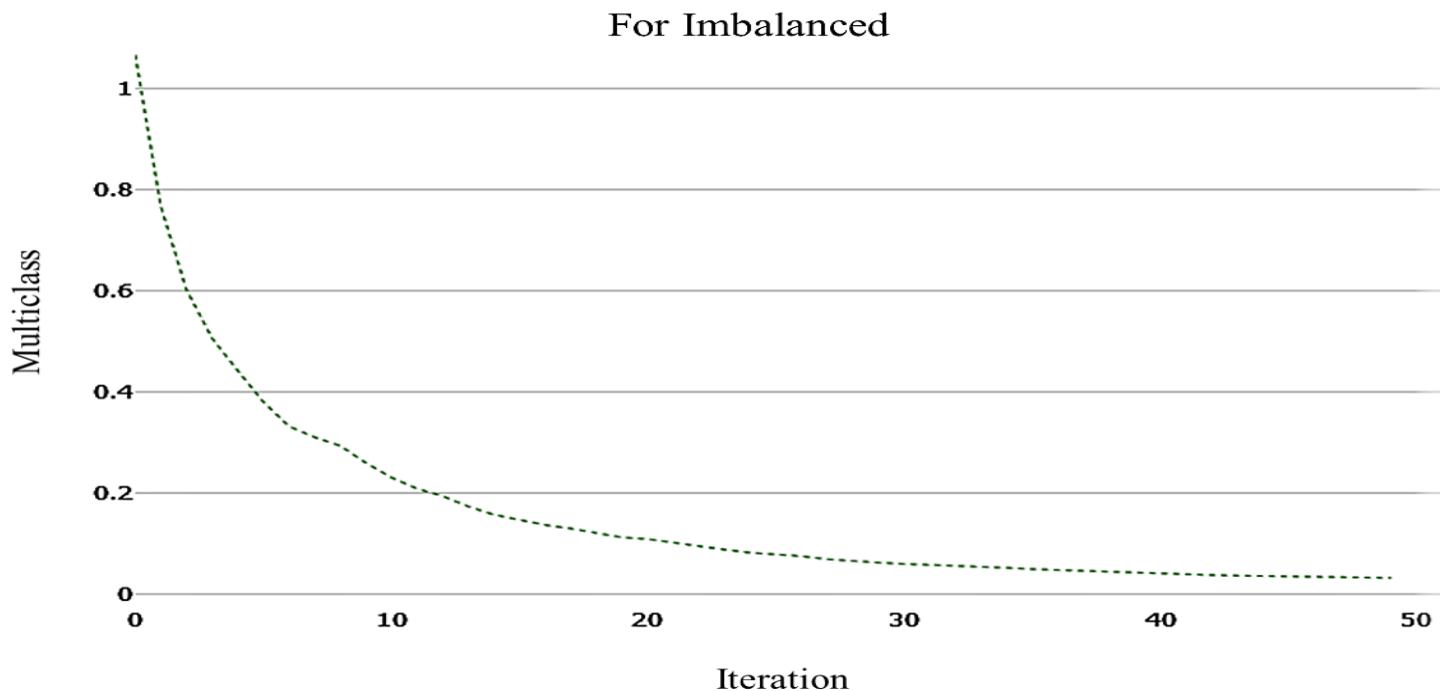
# Results Classification



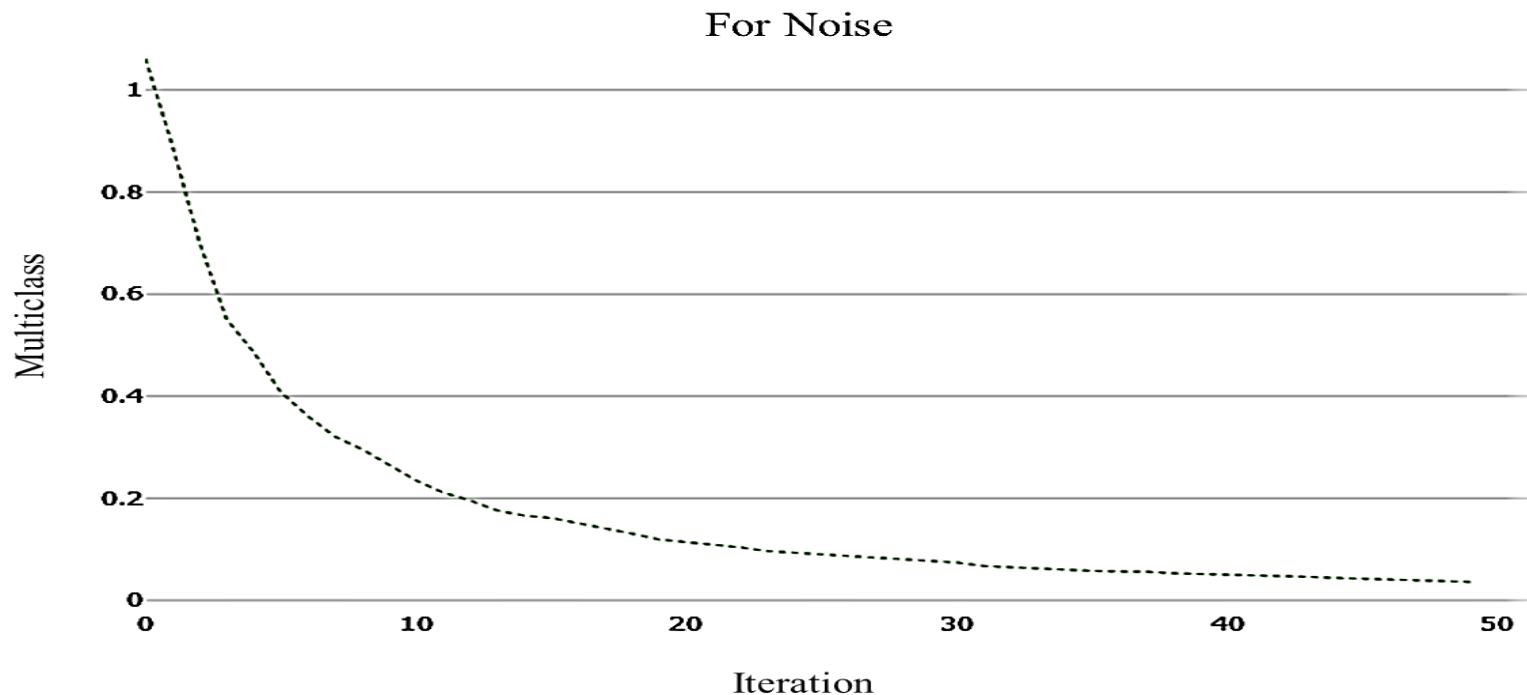
# Results Classification



# Results Classification



# Results Classification



# Discussion

01

Well-fitted

02

Good  
performance  
metrics

03

Maintains  
accuracy

Table 1: Performance measurement metrics

Task		Data type	Precision	Recall	F1	Accuracy
Detection	Balanced	1	1	1	1	1
	Noise	1	1	1	1	1
	Imbalanced	1	1	1	1	1
Classification	Balanced	0.99	0.99	0.99	0.99	0.99
	Noise	0.99	0.99	0.99	0.99	0.99
	Imbalanced	0.99	0.99	0.99	0.99	0.99

Table 2: Comparison with some reviewed literature

<b>Method</b>	<b>Input</b>	<b>Data Size</b>	<b>Noise Immunity</b>	<b>Accuracy</b>	<b>Operating condition Variation</b>	<b>Data Variation Check</b>
DWT-ANN	Current	48,396	Yes, but impact on accuracy	99.3%	Fault resistance are 0, 20 Ω, 100 Ω and 1000 Ω	Not mentioned
SAT-CNN	Voltage and Current	228,690	97% accuracy at 10 dB SNR	99.58%	Not mentioned	Not mentioned
BPNN	Voltage and Current	1188	Not mentioned	97.3%	Not mentioned	Not mentioned
<b>Proposed CatBoost Classifier</b>	<b>Voltage and Current</b>	<b>2500</b>	<b>99% accuracy at 20, 37 dB SNR</b>	<b>99%</b>	<b>Fault resistance are 0.01, 0.01, 50, 100 ohm, Fault distance are 50,100 Km</b>	<b>99% accuracy</b>

- Abdullah, A. Ultrafast Transmission Line Fault Detection Using a DWT-Based ANN. *IEEE Trans. on Ind. Applicat.* 2018, 54, pp. 1182–1193.
- Fahim, S.R.; Sarker, Md.R.I.; Arifuzzaman, Md.; Hosen, Md.S.; Sarker, S.K.; Das, S.K. A Novel Approach to Fault Diagnosis of High Voltage Transmission Line - A Self Attentive Convolutional Neural Network Model. In Proceedings of the 2020 IEEE Region 10 Symposium (TENSYMP); IEEE: Dhaka, Bangladesh, 2020; pp. 1329–1332.
- Jamil, M.; Sharma, S.K.; Singh, R. Fault Detection and Classification in Electrical Power Transmission System Using Artificial Neural Network. *SpringerPlus* 2015, 4, 334.

# Conclusions

01

No explicit data processing

02

High time performance

03

100% accuracy in detection

04

99% in classification

05

Maintains accuracy in data varieties

06

Resilient towards operating conditions

# Future Research

01

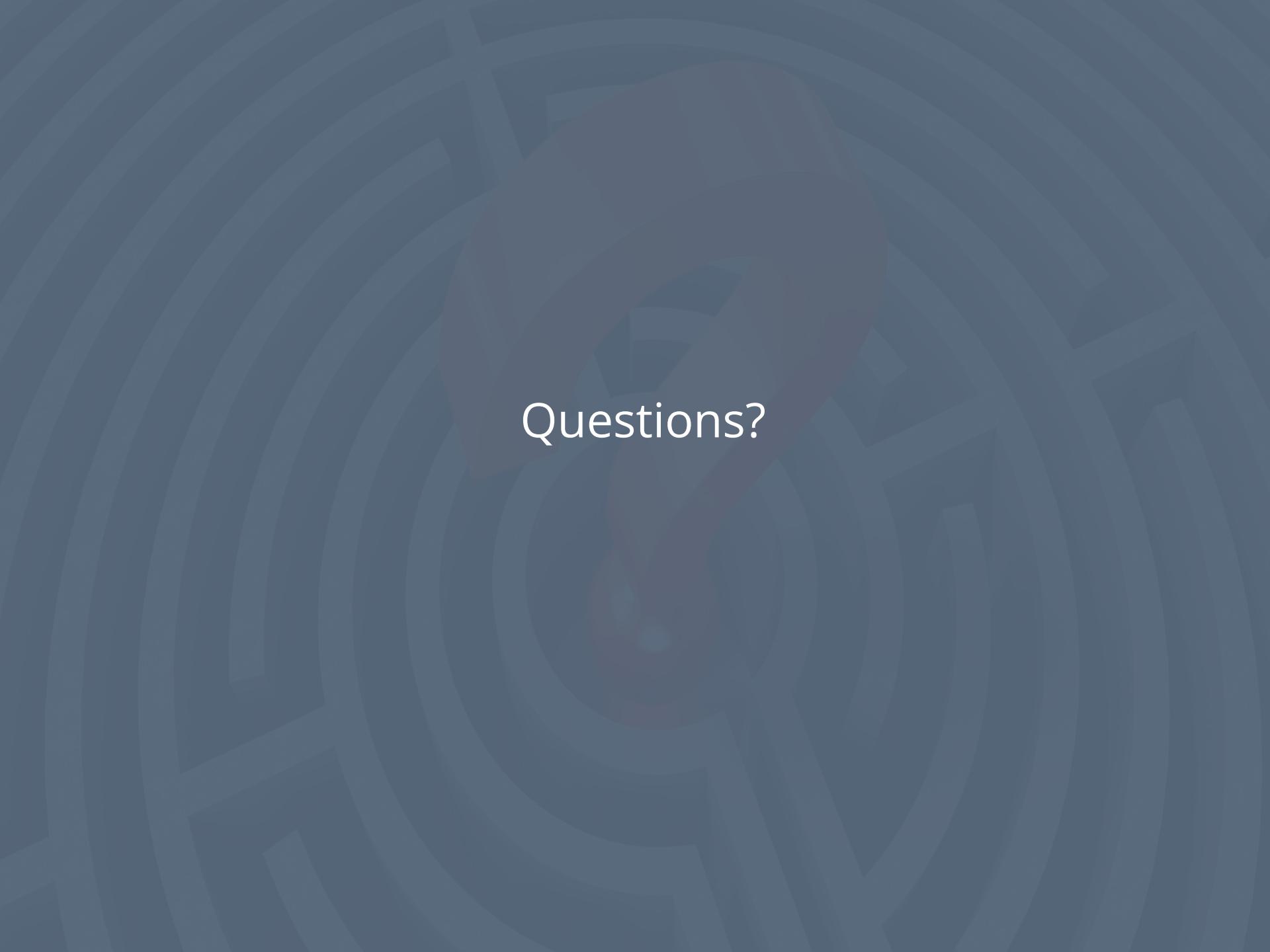
More optimized by real-world applications

02

Design protective system based on ML

03

Advanced computation technology



Questions?

Thank You!