

FAULT DETECTION AND CLASSIFICATION USING ARTIFFICIAL NEURAL
NETWORK (ANN)

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NETWORK (ANN)

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DEDICATION

This thesis is dedicated to my beloved parents, En. Mohd Lutpee bin Salleh and Pn. Faizah binti Ali, my supervisor, Assoc. Prof Ir Dr Saifulnizam bin Abd. Khalid, my family and all my supportive friends.

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ABSTRACT

Various reasons contribute to the occurrence of transmission line faults, which lead to system outages. To prevent power supply failure, a transmission line failure must be detected and isolated as soon as possible. This study aims to establish a system for identifying and classifying the various kinds of faults that might occur in power transmission lines, by using the Artificial Neural Network method. ANN is one of the most effective methods for identifying and diagnosing faults that have occurred. To develop ANN, backpropagation was used as a learning algorithm that could handle large volumes of data. These were the detection and classification training methods used by Levenberg-Marquardt (`trainlm`) and Scaled-Conjugate Gradient (`trainscg`). In order to implement this project, the ANN, feed-forward-networks-with-backpropagation algorithms for each phase (voltage and current) are selected. Based on the simulation results, fault voltages and currents are measured and allocated as input data for ANN. The training data set consists of with faults and without faults on three-phase lines and ground. MATLAB Simulink was used in this study to run tests on 14-bus, 30-bus, and 57-bus systems to generate fault parameters. This project examines neural network output for Line-to-Line-faults, Single-Line-to-Ground faults, and Double Line-to-Ground faults. Then, import the data into MATLAB as input for ANN to-detect-and-classify-the fault. Three critical phases are used in ANNs: training, validation, and testing. The mean squared error (MSE), correlation coefficient, and confusion matrix are all used to evaluate the detection and identification network efficiency. The detection in the 14-bus system achieves a satisfactory MSE of $1.8857e-15$, correlation of 1, and accuracy of 100%, indicating that the performance of the system is excellent. Meanwhile, the classification achieved a reasonable MSE of 0.48107, a correlation of 0.75643, and an accuracy of 80%, indicating that the system is acceptable.

ABSTRAK

Sistem perlindungan yang berkesan dan boleh dipercayai diperlukan untuk bekalan elektrik berkualiti tinggi kerana ia mesti mampu menangani kerosakan dalam talian penghantaran yang disebabkan oleh sumber rawak yang tidak dapat diramalkan. Untuk mengelakkan kegagalan bekalan kuasa, kegagalan talian penghantaran mesti dikesan dan diasingkan secepat mungkin. Kajian ini bertujuan untuk mewujudkan satu sistem untuk mengenal pasti dan mengklasifikasikan pelbagai jenis kerosakan yang mungkin berlaku dalam talian penghantaran kuasa, dengan menggunakan kaedah Rangkaian Neural Buatan. ANN adalah salah satu kaedah yang paling berkesan untuk mengenal pasti dan mendiagnosis kesalahan yang telah berlaku. Untuk membangunkan ANN, perambatan belakang digunakan sebagai algoritma pembelajaran yang boleh mengendalikan jumlah data yang besar. Ini adalah kaedah latihan pengesanan dan pengelasan yang digunakan oleh Levenberg-Marquardt (trainlm) dan Gradient Scaled-Conjugate (trainscg). Untuk melaksanakan projek ini, algoritma ANN, suapan ke hadapan rangkaian dengan perambatan belakang untuk setiap fasa (voltan dan arus) dipilih. Berdasarkan keputusan simulasi, voltan dan arus kerosakan diukur dan diperuntukkan sebagai data input untuk ANN. Set data latihan terdiri daripada dengan kerosakan dan tanpa kerosakan pada garisan dan tanah tiga fasa. MATLAB Simulink telah digunakan dalam kajian ini untuk menjalankan ujian pada sistem 14-bas, 30-bas, dan 57-bas untuk menjana parameter kerosakan. Projek ini mengkaji keluaran rangkaian saraf untuk ralat Talian ke Baris, ralat Talian Tunggal-ke-Ground, dan ralat Talian-ke-Ground Berganda. Kemudian, import data ke dalam MATLAB sebagai input untuk ANN untuk mengesan dan mengklasifikasikan kesalahan. Tiga fasa kritikal digunakan dalam ANN: latihan, pengesanan dan ujian. Purata ralat kuasa dua (MSE), pekali korelasi, dan matriks kekeliruan semuanya digunakan untuk menilai kecekapan rangkaian pengesanan dan pengenapastian. Pengesanan dalam sistem 14-bas mencapai MSE yang memuaskan iaitu $1.8857e-15$, korelasi 1, dan ketepatan 100%, menunjukkan bahawa prestasi sistem adalah cemerlang. Sementara itu, klasifikasi mencapai MSE munasabah 0.48107, korelasi 0.75643, dan ketepatan 80%, menunjukkan bahawa sistem itu boleh diterima..

TABLE OF CONTENTS

	TITLE	PAGE
	DECLARATION	iii
	DEDICATION	iv
	ACKNOWLEDGEMENT	v
	ABSTRACT	vi
	ABSTRAK	vii
	TABLE OF CONTENTS	viii
	LIST OF TABLES	xi
	LIST OF FIGURES	xii
	LIST OF ABBREVIATIONS	xiv
CHAPTER 1	INTRODUCTION	1
1.1	Research Background	1
1.2	Problem Statement	2
1.3	Research Objective	2
1.4	Scope of Work	3
1.5	Structure of Thesis	3
CHAPTER 2	LITERATURE REVIEW	5
2.1	Introduction	5
2.2	Types of Faults	5
2.2.1	Single Line-to-Ground Fault (SLG)	5
2.2.2	Double Line-to-Ground Fault (DLG)	6
2.2.3	Line-to-Line Fault (LL)	6
2.2.4	Three Phase Fault	7
2.3	Artificial Neural Network (ANN)	7
2.3.1	ANN Training Algorithm	9
2.4	Previous ANN-related research	9
2.5	Summary	13

CHAPTER 3	RESEARCH METHODOLOGY	14
3.1	Introduction	14
3.2	Flowchart of fault detection and classification using ANN	14
3.2.1	Create Fault program	15
3.2.1.1	IEEE 14 Bus system	16
3.2.1.2	IEEE 30 Bus system	16
3.2.1.3	IEEE 57 Bus system	17
3.2.2	Create ANN for Fault Detection	18
3.2.3	Create ANN for Fault Classification	19
3.3	Summary	20
CHAPTER 4	RESULTS AND DISCUSSION	21
4.1	Introduction	21
4.2	Fault Signal	21
4.2.1	No Fault	21
4.2.2	Single Line to Ground Fault (SLG)	22
4.2.3	Double Line to Ground Fault (DLG)	23
4.2.4	Line to Line Fault	23
4.3	Detection Results	24
4.3.1	IEEE 14-bus Detection	24
4.3.2	IEEE 30-bus Detection	28
4.3.3	IEEE 57-bus Detection	31
4.4	Classification Results	34
4.4.1	IEEE 14-bus Classification	34
4.4.2	IEEE 30-bus Classification	37
4.4.3	IEEE 57-bus Classification	40
4.5	Discussion	43
4.5.1	Overview of Fault Detection	44
4.5.2	Overview of Fault Classification	45
4.6	Summary	45

CHAPTER 5	CONCLUSION AND RECOMMENDATIONS	47
5.1	Conclusion	47
5.2	Future Works and Recommendation	47
REFERENCES		49

LIST OF TABLES

TABLE NO.	TITLE	PAGE
Table 2.1:	Literature Review based on ANN	10
Table 3.1:	ANN Fault Detection Target Output	18
Table 3.2:	ANN Fault Classifier Target Output	19
Table 4.1:	Summary three test case of Fault Detection	44
Table 4.2:	Summary three test case of Fault Classification	45

LIST OF FIGURES

FIGURE NO.	TITLE	PAGE
Figure 2.1:	Single Line-to-Ground Fault	6
Figure 2.2:	Double Line-to-Ground Fault	6
Figure 2.3:	Line-to-Line Fault	7
Figure 2.4:	Three Phase Fault	7
Figure 2.5:	Architecture Artificial Neural Network	8
Figure 3.1:	Flowchart of the system	15
Figure 3.2:	IEEE 14-bus system in MATLAB Simulink	16
Figure 3.3:	IEEE 30-bus system in MATLAB Simulink	17
Figure 3.4:	IEEE 57 bus system in MATLAB Simulink	17
Figure 4.1:	Normal Three Phase Voltage Waveform	22
Figure 4.2:	Single Line-to-Ground Fault Current and Voltage Waveform	22
Figure 4.3:	Double Line-to-Ground Fault Current and Voltage Waveform	23
Figure 4.4:	Line-to-Line Fault Current and Voltage Waveform	24
Figure 4.5:	14 bus, 6-5-1 configuration ANN fault detection	25
Figure 4.6:	Mean-square error performance 14-bus fault detection	26
Figure 4.7:	Linear regression curve 14-bus fault detection	27
Figure 4.8:	Confusion matrix 14-bus fault detection	28
Figure 4.9:	30 bus, 6-7-1 configuration ANN fault detection	29
Figure 4.10:	Mean-square error performance 30-bus fault detection	30
Figure 4.11:	Linear regression curve 30-bus fault detection	30
Figure 4.12:	Confusion matrix 30-bus fault detection	31
Figure 4.13:	57 bus, 6-10-1 configuration ANN fault detection	32
Figure 4.14:	Mean-square error performance 57-bus fault detection	33
Figure 4.15:	Linear regression curve 57-bus fault detection	33
Figure 4.16:	Confusion matrix 57-bus fault detection	34

Figure 4.17: 14 bus, 6-2-4 configuration ANN fault classification	35
Figure 4.18: Mean-square error performance 14-bus fault classification	36
Figure 4.19: Linear regression curve 14-bus fault classification	36
Figure 4.20: Confusion matrix 14-bus fault classification	37
Figure 4.21: 30 bus, 6-5-4 configuration ANN fault classification	38
Figure 4.22: Mean-square error performance 30-bus fault classification	39
Figure 4.23: Linear regression curve 30-bus fault classification	39
Figure 4.24: Confusion matrix 30-bus fault classification	40
Figure 4.25: 57 bus, 6-8-4 configuration ANN fault classification	41
Figure 4.26: Mean-square error performance 57-bus fault classification	42
Figure 4.27: Linear regression curve 57-bus fault classification	42
Figure 4.28: Confusion matrix 57-bus fault classification	43

LIST OF ABBREVIATIONS

ANN	-	Artificial Neural Network
AI	-	Artificial Intelligence
BPNN	-	Backpropagation Neural Network
MSE	-	Mean Squared Error
SLG	-	Single Line-to-Ground
DLG	-	Double Line-to-Ground
LL	-	Line-to-Line
V	-	Voltage
I	-	Current

CHAPTER 1

INTRODUCTION

1.1 Research Background

The electrical power system is split up into three subsystems: generation, transmission, and distribution. Transmission lines are the main components of electrical power systems that provide a pathway for electricity to be transferred from generation to distribution. Since transmission lines are constantly exposed to the environment, they are prone to various forms of damage. Fault to the power transmission line can cause power outages[1]. Faults are abnormal conditions that occur between two phases or between one phase and the ground, especially in transmission lines[2]. In order to avoid disruption, a failure on the transmission line must be identified and immediately isolated. To avoid unstable problems which can impact the system, it is necessary to detect, identify and clear defects with high speed, selectivity and accuracy.

An excellent fault detection system allows for an effective, dependable, fast, and safe relaying operation[3]. There are two categories of faults. Transmission lines can experience either symmetrical faults or unsymmetrical faults. Unsymmetrical faults can be broken down into three categories: Line-to-Line faults (LL), Single Line-to-Ground faults (SLG), and Double Line-to-Ground faults (DLG) three types of electrical interference[4]. The SLG fault is the most common, followed in frequency of occurrence by the LL fault, DLG fault, and three-phase fault[5].

There are several approaches for detecting and classifying faults, including artificial neural networks (ANN), Fast Fourier Transform (FFT) and Discrete Wavelet Transform (DWT), and others. When it comes to fault detection, each technique has benefits and drawbacks. ANN has excellent characteristics such as generalization,

noise immunity, robustness, and fault resistance[6]. ANN also are extremely effective at recognizing and classifying patterns using pattern recognition[7].

1.2 Problem Statement

Damage to the power system transmission line accounts for most of the voltage and current signal interference[8]. Interruption of the power supply might result in power transmission line failures and customer loss of power supply. However, simply detecting faults is insufficient. It is preferable to have a system that can identify different types of faults in order to take fast and effective corrective action [7].

Hence, the purpose of this research proposal is to identify the fault in the transmission line using the ANN. As a result of the accurate information about the fault provided by the automated system, the amount of time required to detect the fault can be reduced. [9][10]. It is also necessary to distinguish types of faults in the transmission line so that action speed up the process to fix these faults and hence manage to avoid damage the equipment [11].

1.3 Research Objective

The objectives of this project are to solve of all the problems that happened prior to the study. The research objective of the study is:

- (a) To diagnose the unsymmetrical fault in the transmission line by using an Artificial Neural Network (ANN).
- (b) To identify and classify the fault in the transmission line by using electrical parameter such as Voltage (V) and Current (I).
- (c) To compare the output performance of three test cases (14 bus, 30 bus and 57 bus) by using MATLAB Simulink 2021b.

1.4 Scope of Work

The scope of this study is to use an Artificial Neural Network (ANN) to classify and diagnose transmission line faults. In this project, Backpropagation Neural Network (BPNN) will be used since it has more advantages such as fast, simple and easy to program[12].

The limiting works are based on Single Line-to-Ground, Line-to-Line, and Double Line-to-Ground power transmission faults since the transmission line usually has these kinds of faults, the analysis will focus on the performance plot, regression fit plot, and confusion matrices are used to calculate the results[13]. To simulate a power transmission line failure, MATLAB R2021b software will be used.

1.5 Structure of Thesis

This work comprises five main chapters. The following contents were discussed in each chapter:

Chapter 1 discusses the background of the study, the statement of the problem to be solved, the objectives of the project, the scope of the study, and the organization of the thesis. Chapter 2 discusses the literature review on faults and Artificial Neural Network (ANN) which are used in this research.

Chapter 3 describes the overall steps and process, starting with the input and ending with the output. This chapter provides an overview of the research approach conducted to detect the fault using ANN. Chapter 4 presents the results and discussion obtained from doing simulation in MATLAB 2021b.

Lastly, chapter 5 summarises the project's development results and recommendations for future work.

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