



**UNIVERSITI PUTRA MALAYSIA**

**WAVELET DECOMPOSITION-NNARX MODEL FOR FLOOD  
PREDICTION OF KELANTAN RIVER, MALAYSIA**

**MOHD AZROL SYAFIEE BIN ANUAR**

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OF KELANTAN RIVER, MALAYSIA**

By

**MOHD AZROL SYAFIEE BIN ANUAR**

**Thesis Submitted to the School of Graduate Studies, Universiti Putra Malaysia, in  
Fulfillment of the Requirements for the Degree of Master of Science**

**December 2018**

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Abstract of thesis presented to the Senate of Universiti Putra Malaysia in fulfilment of the requirement for the degree of Master of Science

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**December 2018**

**Chair: Ribhan Zafira binti Abdul Rahman, PhD**

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Flood is a major disaster that happens around the world. It has caused the loss of many precious lives and massive destruction of property. The possibility of flood can be determined depends on many factors that consist of rainfall, structure of the river, flow rate of the river etc. One of the research challenges is to develop accurate prediction models and what improvement can be made to the forecasting model. The objective of this thesis is to improve the performance of the neural network model to predict the flood on the Kelantan River, Malaysia. A technique for modelling of nonlinear data of flood forecasting using wavelet decomposition-neural network autoregressive exogenous input (NNARX) approach is proposed.

This thesis discusses the identification of parameters that involved in the forecasting field as rainfall value, flow rate of the river and the river water level. With the original data acquired, the data had been processing through to wavelet decomposition and filtered to generate a new set of input data for NNARX prediction model. This proposed technique has been compared with the non-wavelet NNARX.

The experimental result show that the proposed approach provides better testing performance compared to its counterpart, which the mean square error obtained is  $2.0491e^{-4}$  while the normal NNARX is  $6.1642e^{-4}$ .

Abstrak tesis yang dikemukakan kepada Senat Universiti Putra Malaysia sebagai memenuhi keperluan untuk ijazah Master Sains

## **MODEL GELOMBANG DEKOMPOSISI-NNARX BAGI RAMALAN BANJIR DI SUNGAI KELANTAN, MALAYSIA**

Oleh

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Banjir adalah bencana besar yang berlaku di seluruh dunia. Ia telah menyebabkan kehilangan nyawa dan kemusnahan harta benda. Kemungkinan banjir dapat ditentukan dengan bergantung kepada banyak faktor yang terdiri daripada hujan, struktur sungai, aliran sungai dan lain-lain. Salah satu cabaran penyelidikan ini adalah untuk membangunkan model ramalan yang tepat dan penambahbaikan yang boleh dilaksanakan pada model ramalan. Objektif tesis ini adalah untuk meningkatkan prestasi model rangkaian neural dalam meramal banjir di Sungai Kelantan, Malaysia. Teknik pemodelan data ramalan banjir ini menggunakan penguraian gelombang kecil – rangkaian neural pendekatan input eksogen autoregressif (NNARX) adalah dicadangkan.

Tesis ini membincangkan pengenalpastian parameter yang terlibat dalam bidang ramalan seperti nilai hujan, kadar aliran sungai dan paras air sungai. Dengan data asal yang diperoleh, data telah melalui penguraian dan penapis gelombang kecil dalam menghasilkan set data input baru untuk model ramalan NNARX. Teknik yang dicadangkan ini telah dibandingkan dengan NNARX bukan gelombang kecil.

Keputusan eksperimen menunjukkan bahawa pendekatan yang dicadangkan memberikan prestasi ujian yang lebih baik berbanding dengan teknik NNARX yang biasa, yang mana ralat punca kuasa setara diperoleh adalah  $2.0491e^{-4}$  manakala NNARX biasa ialah  $6.1642e^{-4}$ .

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## TABLE OF CONTENTS

	<b>Page</b>
<b>ABSTRACT</b>	i
<b>ABSTRAK</b>	iii
<b>ACKNOWLEDGEMENTS</b>	vi
<b>APPROVAL</b>	vii
<b>DECLARATION</b>	viii
<b>LIST OF TABLES</b>	ix
<b>LIST OF FIGURES</b>	xi
<b>LIST OF ABBREVIATIONS</b>	xii
<b>CHAPTER</b>	
<b>1 INTRODUCTION</b>	<b>1</b>
1.1 Overview	1
1.2 Problem Statement	2
1.3 Objectives	2
1.4 Scope of study	3
1.5 List of contributions	3
1.6 Thesis organization	4
<b>2 LITERATURE REVIEW</b>	<b>5</b>
2.1 Overview	5
2.2 Parameters factors	5
2.3 Data processing	7
2.3.1 Training function	7
2.3.2 Hidden neurons	8
2.4 Existing model	9
2.4.1 Backpropagation neural network (BPNN)	9
2.4.2 Neural network autoregressive exogenous input	11
2.4.3 Neuro-wavelet	12
2.5 General applications	14
2.6 AI in hydrology	15
<b>3 METHODOLOGY</b>	<b>18</b>
3.1 Overview	18
3.2 Input-output of the model	20
3.2.1 Identifying input and its relation	20
3.2.2 Data selection and pre-processing	20

3.3	Neural network model	21
3.3.1	Activation function and performance index	23
3.3.2	Number of hidden neurons	24
3.3.3	Backpropagation neural network	24
3.3.4	Nonlinear autoregressive exogenous input neural network (NNARX)	24
3.4	Wavelet decomposition-NNARX model	26
<b>4</b>	<b>RESULTS AND DISCUSSION</b>	<b>30</b>
4.1	Overview	30
4.2	Training function and number of hidden neurons	30
4.3	Performance between BPNN and NNARX	31
4.4	Performance of wavelet decomposition-NNARX model	34
4.5	Performance in long-term forecasting	37
<b>5</b>	<b>SUMMARY, CONCLUSION AND RECOMMENDATIONS FOR FUTURE RESEARCH</b>	<b>38</b>
	<b>REFERENCES</b>	<b>39</b>
	<b>APPENDICES</b>	<b>43</b>
	<b>BIODATA OF STUDENT</b>	<b>47</b>
	<b>PUBLICATIONS</b>	<b>48</b>

## LIST OF TABLES

Table		Page
2.1	Justification of the NNARX approach	17
2.2	Wavelet-neuro approach	17
3.1	NNARX structure parameters	26
3.2	Original data that used in normal neural network	27
3.3	Generated data from the 1 <sup>st</sup> level wavelet decomposition.	28
3.4	Wavelet decomposition level based on Galas data	29
4.1	Comparison between training function	31
4.2	Comparing the performance between number of hidden neurons	31
4.3	Performance between BPNN and NNARX	31
4.4	Wavelet level decomposition	34
4.5	Comparison of the model with the wavelet addition	36
4.6	Comparison for long-term forecasting.	37

## LIST OF FIGURES

Figure		Page
1.1	Area of study	3
2.1	Classification of threat of water level at Kelantan River	7
2.2	The structure of BPNN	10
2.3	Wavelet of decomposition	13
3.1	Block diagram of the model	18
3.2	Flow chart of the research	19
3.3	The structure of the BPNN in this research	22
3.4	NNARX model structure for this study	25
3.5	Wavelet decomposition of Galas data	28
3.6	Process in developing NNARX model	29
4.1	BPNN model regression analysis	32
4.2	NNARX model regression analysis	33
4.3	NNARX model graph	33
4.4	Wavelet decomposition NNARX regression analysis	35
4.5	Wavelet decomposition BPNN regression analysis	36
4.6	Graph between the original data and predicted data by wavelet decomposition for long-term forecasting	37

## LIST OF ABBREVIATIONS

ANN	Artificial neural network
BPNN	Backpropagation neural network
NNARX	Nonlinear autoregressive exogenous input neural network
ARX	Autoregressive exogenous input
RNN	Recurrent Neural Network
MA	Moving Average
AI	Artificial intelligence
ARIMA	Autoregressive integrated moving average
ARMA	Autoregressive moving average
SVM	Support vector machine
RMSE	Root mean square error
MSE	Mean square error
ANFIS	Neuro fuzzy inference system
MARE	Mean absolute relative error
DNN	Deep Neural Network
R	Regression

# CHAPTER 1

## INTRODUCTION

### 1.1 Overview

Flood is a common natural disaster that affecting lots of countries around the world. Flood have posed a great threat up to the extent of causing massive destruction on public facilities, residential homes and even worse, death. Malaysia is one of the countries that have endure and encounter this natural disaster. Flash flood is the types of flood that usually occur in the urban area meanwhile where the monsoon flood is in the east coast of Malaysia which consist of three states that is Kelantan, Pahang and Terengganu. The east coast flood dependent on the monsoon where it is due to the result of wind carrying large amount of rain. The monsoon season can be divided into two, which is northeast monsoon and southwest monsoon. Normally, east coast monsoon floods occur from October to February and bring heavy rainfall to Malaysia, especially East Peninsular Malaysia, Sabah and Sarawak, while west coast monsoon floods happen from May to August and bring little amount of rainfall to the west coast of the peninsula, including Kuala Lumpur, Selangor, Melaka, Johor, Perak, and Penang [1]. The monsoons contribute about 86% of the annual rainfall in the east coast region [2].

Water flow rate, rainfall and water levels among the parameters that contribute to flood events. Thus, in this modern era monitoring or prediction system rely on fuzzy logic or neural network have become common [3]. Though every intelligent system such as fuzzy logic and neural network are best suit to be implemented for forecasting compare to its linear model counterparts, each of this intelligent do have its own advantages and disadvantages need to be considered. Through several literature study shown that the neural network has certain advantages when dealing environmental data. Thus, neural network was used in this research to analyse the flood pattern that occurs in Kelantan. The concept of the artificial neural network can be described as a computer model assumption of the biological brain. It consists of a set of interconnected simple processing units which combine to output a signal to solve a certain problem based on the input signals it received. In neural network, there many different types of model that can be applied to any certain applications and one of it is Neural Network Autoregressive with Exogenous Input (NNARX) which is based on linear Autoregressive Model with Exogenous Input (ARX) which is have become widely applied method in many prediction researches. ARX, and other model can be trained as neural network thus neural network also can be developed for system identification [4]. NNARX is a class of Recurrent Neural Network (RNN) that takes past input and output values to compute the current output.

Despite of the capabilities of the neural network in prediction, the model can still be improved from time to time by optimizing the system and combining with other technique or method that has been discussed in the literature review chapter. Wavelet decomposition is one of mathematical technique that transforms signal to spectral data in time and frequency [5]. The signal decomposes into approximation and detail, with passing previous coefficient through high and low pass filters. As such, in this research is to discover how this wavelet combination with neural network or in other term is called neuro-wavelet could be approached and be implemented in the forecasting field and especially for this river water level prediction.

## **1.2 Problem statement**

In the river forecasting application, many parameters must be involved in order to deliver an accurate reading of the flood situation. Therefore, many methods have been introduced as to overcome such inaccuracy in predicting, where certain method tends to have advantages over others. The flood forecasting models involved with non-linear data such as rainfall, flow rate etc. Artificial intelligence (AI) application has shown rising interest throughout the year in solving forecasting problem. Backpropagation neural network (BPNN) model are among used technique in prediction due to its simplicity and easy to understand but neural network has other technique that can improve the accuracy when handling time-series forecasting as such NNARX. There are researches of that used NNARX as the predictor whether in the flash flood forecasting and monsoon flood. Different from flash flood, NNARX that applied in this type of field able to achieve a higher performance compared to monsoon flood. Monsoon flood can also be dividing into short-term and long-term forecasting with each have different set of data. With lower performance obtained through a long time period, does NNARX will also affect a shorter time period on monsoon flood. This leads us to uncover the capabilities and its limitation of the NNARX model in time series forecasting. Wavelet decomposition was discovered that could help improve the performance of AI such as support vector machine, adaptive neuro fuzzy inference system (ANFIS) etc. With this discovery of the wavelet advantages, this will help the NNARX to improve its performance and achieve a better accuracy.

## **1.3 Objectives**

The aim of this project is to develop a prediction system through NNARX with Wavelet decomposition system. To achieve this aim, the following objectives are set:

- To identify the data from the related rivers and rainfall that causes the flood in the downstream of Kelantan River at Guillemard Bridge.
- To develop NNARX that have better accuracy compared to Backpropagation Neural Network.
- To combine NNARX with Wavelet decomposition method to improve the accuracy in flood prediction.



## 1.4 Scope of Study

The main concern of this project is classifying the river water level. For classifying water level, the required parameters are the water level of Kelantan River, Rainfall of surrounding area and Lebir river flow rate. The data of rainfall, flow rate, water level was obtained from the Department of Irrigation and Drainage Malaysia [6]. As status river water level is normal, the data for other two classes which is caution (moderate) and danger is based on the millimetre of rain dropping, and water flow cubic metre per seconds result in the water level changes. The data used in this research are hourly data of November until December for 2012, 2013 and 2014. Figure 1.1 is the representing the area of study for this project that is in Kelantan.

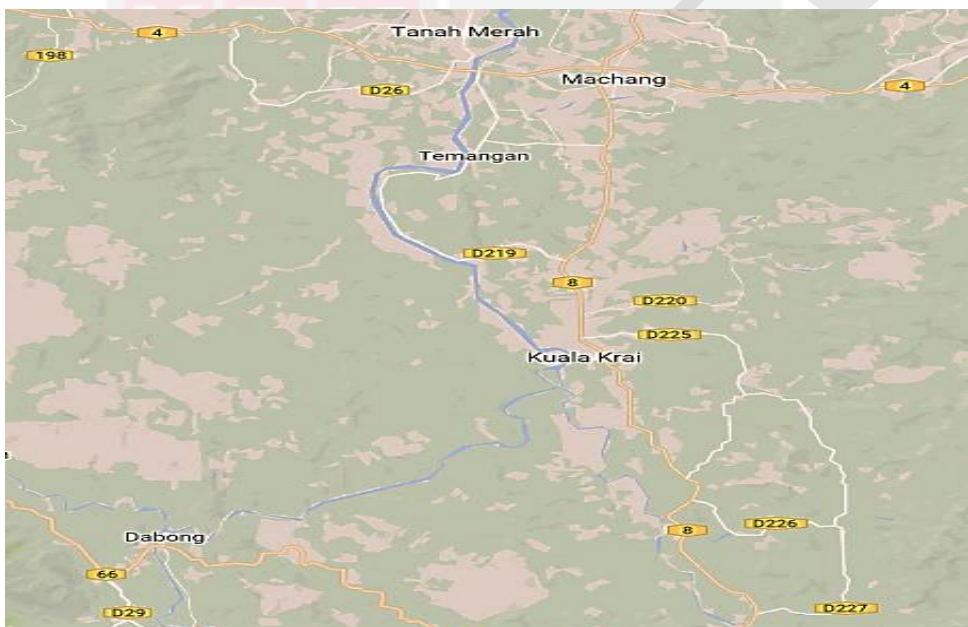


Figure 1.1: Area of Study (Kelantan)

## 1.5 List of contributions

This is list of contribution that obtained and discovered throughout this research:

- The wavelet decomposition helps NNARX model to achieve better testing accuracy compared to its normal NNARX counterparts.
- Wavelet-BPNN model achieved higher accuracy compared to normal BPNN with increase both in training and testing phase.
- Discover the NNARX model to be able to predict in both short- and long-term period, although long term forecasting does not obtain a good performance as compared to shorter study time period.

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## PUBLICATIONS

### Conference

M. A. S. Anuar, R. Z. A. Rahman, S. B. Mohd, A. C. Soh and Z. D. Zulkafli, 2017, Early prediction system using neural network in Kelantan River, Malaysia. *2017 IEEE 15th Student Conference on Research and Development (SCORED)*, Putrajaya, pp.104-109. doi: 10.1109/SCORED.2017.8305412

### Journals

M. A. S. Anuar, R. Z. A. Rahman, S. B. Mohd, A. C. Soh and Z. D. Zulkafli, "Kelantan River Flood Prediction based on Wavelet Decomposition-NNARX model". *International Journal of Engineering & Technology (IJET)*. (Accepted)



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