

PERPUSTAKAAN UMP



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**A STUDY ON FLOOD FORECASTING AT SUNGAI LEMBING USING
ARTIFICIAL NEURAL NETWORK**

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ABSTRACT

Industrial countries which are rapidly developing had to face environmental disaster. Flood occurs due to the excessive rainfall in the river catchment. The effects from flood are damage of properties and loss of life. Flood forecasting is a necessity which will help in reduce the effects of flood and help better management planning of flood events. Statistical method such as Auto Regressive Moving Average (ARMA) is commonly used, but it is only a rough estimation of the flow. There is an alternative computing model that has been successfully tested in flood forecasting studies called Artificial Neural Network (ANN). It helps to produce an accurate forecasting result. The study is conducted to make accurate prediction of the flood event using Artificial Neural Network(ANN). The objective of the study is also to gain more understanding about Artificial Neural Network in data forecasting. Besides that, the objective of the study is to issue the flood warning. In this study, three iterations were conducted which is 1000, 2000 and 5000 iterations with six datasets of network model. The performance of training and validation data were evaluated using Nash Sutcliffe(NSC), correlation coefficient(R^2), and Root Mean Square Error(RMSE). Error distribution graph are presented to show the accuracy and reliability of the forecasting models. The results showed that ANN able to provide accurate forecasting using sample historical datasets.

ABSTRAK

Negara-negara industri yang pesat membangun terpaksa menghadapi bencana alam sekitar. Banjir berlaku disebabkan oleh hujan yang berlebihan di kawasan tadahan sungai. Kesan-kesan dari banjir adalah kerosakan harta benda dan kehilangan nyawa. Ramalan banjir adalah satu keperluan yang akan membantu mengurangkan kesan-kesan banjir dan perancangan banjir. Kaedah statistik seperti Auto Regressive Moving Average (ARMA) biasanya digunakan, tetapi ia adalah hanya anggaran kasar aliran. Terdapat satu alternatif model komputer yang telah berjaya diuji dalam kajian ramalan banjir adalah Artificial Neural Network (ANN). Ia membantu untuk menghasilkan keputusan ramalan yang tepat. Kajian ini dijalankan untuk membuat ramalan yang tepat tentang kejadian banjir menggunakan ANN. Objektif kajian ini juga untuk mendapat lebih kefahaman tentang Artificial Neural Network dan data ramalan. Di samping itu, objektif kajian ini adalah untuk memberi amaran banjir. Dalam kajian ini, tiga lelaran telah dijalankan iaitu 1000, 2000, dan 5000 dengan enam set data model rangkaian. Prestasi latihan dan pengesahan data telah diuji menggunakan Nash Sutcliffe (NSC), correlation coefficient (R^2) dan Root Mean Square Error (RMSE). Graf pembahagian ralat dibentangkan untuk menunjukkan ketepatan dan kebolehpercayaan model ramalan tersebut. Hasil kajian menunjukkan ANN dapat menyediakan ramalan yang tepat menggunakan set data sampel terdahulu.

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LIST OF SYMBOLS

O_i	Observed value
F_i	Forecasted value
\bar{O}	Observed value mean
\bar{F}	Forecasted value mean
N	Number of records evaluated
R^2	Correlation coefficient
Q_o	Value of observed
Q_m	Value of forecasted

LIST OF ABBREVIATIONS

ANN	Artificial Neural Network
ARMA	Auto Regressive Moving Average
BASIC	Beginner's All-purposed Symbolic Instruction Code
DID	Department of Irrigation and Drainage
MLP	Multilayer Perceptron
NSC	Nash-Sutcliffe efficiency coefficient
RMSE	Root Mean Square Error

CHAPTER 1

INTRODUCTION

1.1 INTRODUCTION

An overflow of water that submerges land which is usually dry is called flood. From a strict hydrological sense, flood also can be defined as a rise, usually brief, in the water level in a stream to a peak from which the water level recedes at a slower rate (UNESCO-WMO 1974). It is a dangerous phenomenon which bring to economic losses and life losses. Floods can occur in any region, in the countryside or in cities. Floods can also occur because of the flow rate exceeds the capacity of the river.

Flood forecasting can be defined as the use of real-time precipitation and stream flow data in rainfall-runoff and stream flow models to forecast flow rates and water levels for periods ranging from a few hours to days ahead, depending on the size of the watershed or river basin. In order to make the most accurate flood forecasting, it is better to have a long time-series of historical data that relates stream flows to measured past rainfall events. Flood forecasting is significant to give flood warning to people. Flood warning purpose is to make decisions whether warnings of floods should be issued to the public or previous warnings should be invalidated.

The study are mainly about the hydrological flood forecasting using Artificial Neural Network(ANN). ANN is a model approach from the structure of the brain which well suited to certain tasks related to the hydrological. Numerous studies related to the application of ANNs in order to overcome water resources variables problems. Data used are depends on the study such as rainfall-runoff relationship(Hsu et al,1995),regional flood frequency analysis(Hall and Minns,1998) and river runoff

forecasting (Tokar and Johnson,1999,Cigizoglu,2003).In this study daily water level data will be used. Data training and data validation with forecasted value will be produced once the ANN is running. ANN application does not require hard learning process. In the following section will describe more on neural network model.

1.2 PROBLEM STATEMENT

Industrial countries which are rapidly developing had to faced environmental disaster. Flood occur due to the excessive rainfall in the river catchment. The effects from flood are damage of properties and loss of life. Flood forecasting is a necessity which will help in reduce the effects of flood and planning for flood events. Statistical method such as Auto Regressive Moving Average (ARMA) is commonly used, but it is only a rough estimation of the flow. There is an alternative computing model that has been successfully tested in flood forecasting studies called Artificial Neural Network (ANN). It helps to produce an accurate forecasting results.

1.3 OBJECTIVES OF STUDY

1. Make a more accurate prediction of the flood event using Artificial Neural Network (ANN)
2. To issue the flood warning more earlier to the peoples.
3. More understanding about Artificial Neural Network.

1.4 SCOPE OF STUDY

Commonly, flood forecasting be done in the river catchment area. So, scope of the study is focused at Sungai Lembing. Sungai Lembing located in northwest of Kuantan in Pahang, Malaysia. There is a station at Bukit Kenau, The station number is 3930401.Catchment area of Bukit Kenau is 582km².All the information are get from Department of Irrigation and Drainage.(DID) Sungai Lembing has a low level land area and risk of flood during heavy rain. Frequent flooding is usually happen because of exceed water level of the river catchment. Many other factors are related with the flood phenomena in Sungai Lembing such as uncontrolled logging activities and the depth of

the river could not support the capacity of excessive rainfall. Figure 1.1 shows Sungai Lembing location. The normal water levels at Bukit Kenau is 17.00m while flood levels at the station is 21.50m. Figure 1.2 shows the normal, alert and danger water levels at Sungai Lembing. All the flooding events at Bukit Kenau which happen from year 2003 to 2009 are shown in Figure 1.3.

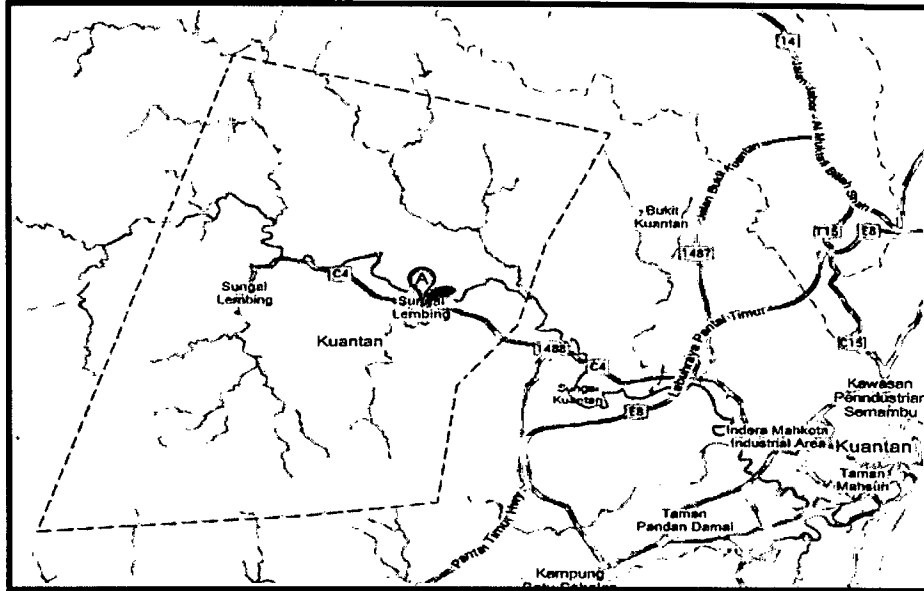


Figure 1.1 : Sungai Lembing, Kuantan Pahang map

(www.googlemap/Sg Lembing)

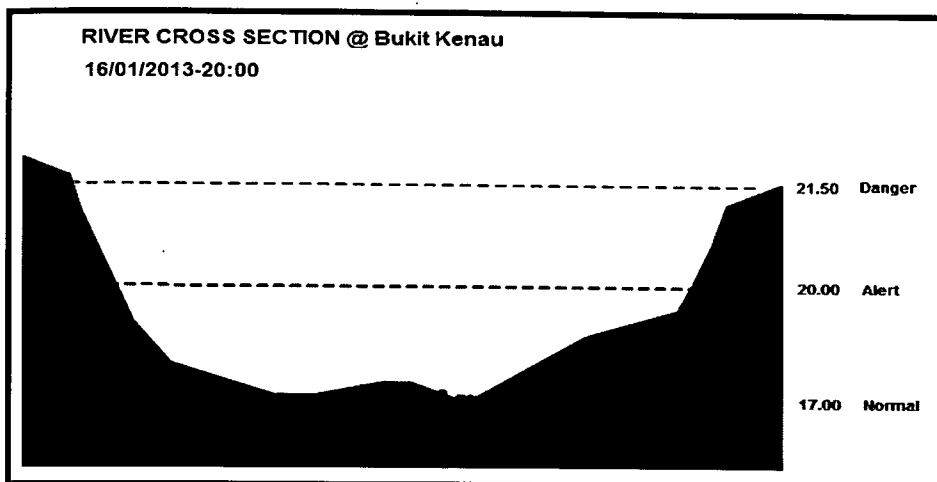


Figure 1.2 : River cross section at Bukit Kenau

(infobanjir.water.gov.my)

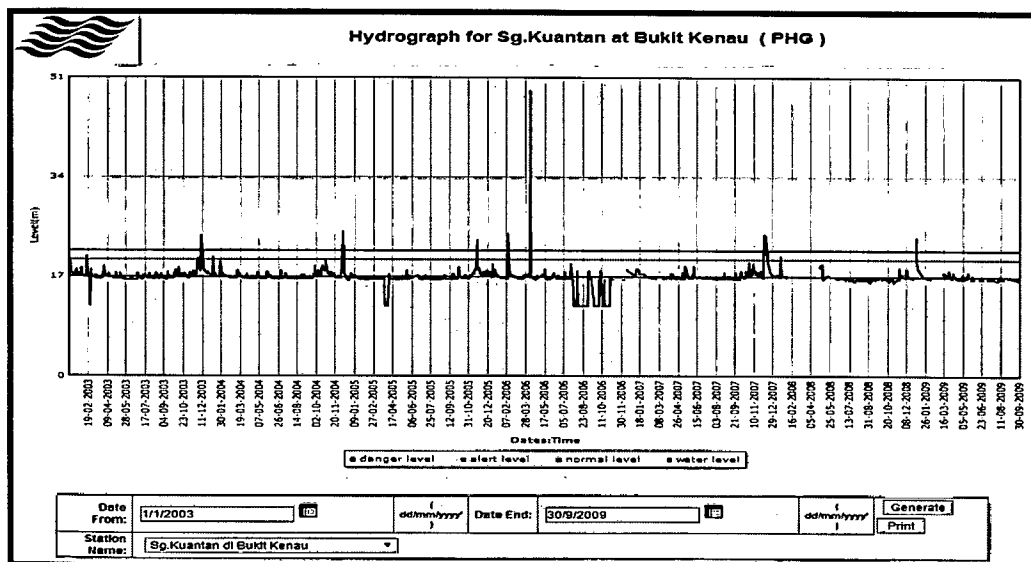


Figure 1.3 : Water level recorded at Bukit Kenau Station for January 2003 to December 2009

(infobanjir.water.gov.my)

CHAPTER 2

LITERATURE REVIEW

2.1 INTRODUCTION

This chapter briefly shows the previous studies about the topic. Flood forecasting using Artificial Neural Network have been conducted with some journal from other researchers. All the journals are made as a reference for my study in order to complete this thesis. However the different are only the scope of study and different modelling approaches use in each studies.

2.2 FLOOD FORECASTING

According to Ahmed El Shafie (2011), river flow forecasting is essential in water resources management because it can facilitate the management of water resources, thereby optimizing the use of water. The ability to forecast river flow also helps predict the occurrence of future flooding, enabling better preparation to avoid the loss of lives and minimize property damage.

Sungwon Kim et. Al (2000) also state that flood stage forecasting is an important subject in water resources planning and management. Flood stage forecasting is often used for warning cities of impending flood flows.

Forecasting river flow after heavy rain is important for several purpose.M. Campolo et Al(1999) in their studies explain the significant purpose of flood forecasting which are for public safety, environmental issues, and water management.

2.3 WHAT IS ARTIFICIAL NEURAL NETWORK

Current studies have shown that a parallel computing model has an ability to predict river flow which called Artificial Neural Network. Ahmed El Shafie et Al (2009) explain the definition of artificial neural network. He states that an artificial neural network or known as ANN is a non-linear mathematical approach that has been proven the ability to map inputs and output pattern and mimic different dynamic experienced in the data.

A simple definition is state by H. Kerem Cigizoglu(2003) that ANN is a model inspired from the structure of the brain, is well suited to such tasks as pattern recognition, combinatorial optimization and discrimination.

Many study about Artificial Neural Network. ANNs have been widely applied in many areas including the financial, mathematical, computer, medicinal, weather forecasting and engineering fields. In water resources studies, ANNs are employed to forecast daily river flow (Atiya et al.1999; Coulibaly et al. 2000; Ahmed and Sarma 2007; El-Shafie et al. 2008; Wu et al.2009)

Moreover, other studies by Hsu et al(1995) make a comparison between ANN models and traditional approach and conclude that ANN is more effective and more efficient.

Many reasonable reason why Artificial Neural Network is chosen for the study. The study by M Campolo et al(1999) conclude that ANN has an advantages of low cost and simplicity with respect to complex physically based model.

2.3.1 Multilayer Perceptron

There are many types of Artificial Neural Network, but commonly used is Multilayer Perceptron(MLP).It is because MLP is easy to be understand and suitable for our study.

As stated by Ahmed El Shafie (2009) in his journal, MLP is known as a supervised feed forward back propagation learning ANN model. It consists of an input layer, one or more hidden layers, and an output layer. The term supervised means that the model requires output data to learn the pattern of data input. There are no cycles or loops in the network. It is the most commonly used of artificial neural network in river flow study.

2.4 TYPES OF STATISTICAL MATHEMATICAL EQUATION

The flood forecasting method is change from day to day. There are also traditional approach research had be found in a few journals which called Statistical Mathematical Equation. However, the method is less used and has come out with poor results. Examples of Statistical Mathematical Equation used is Auto Regressive Moving Average(ARMA) and Linear Regression.

2.4.1 Auto Regressive Moving Average(ARMA)

There is a journal called 'Parameter Estimation of an ARMA Model For River Flow Forecasting Using Goal Programming' cited that there are still applications where the accurate estimation of linear processes such as Auto Regressive Moving Average models are sufficient especially when they are used for linear time series analysis(Hwarng,2001).

Others study which is by Abrahart and See(2000) make a comparison between ARMA models and Artificial Neural network(ANN) for river flow data forecasting data for two contrasting catchments. The result is quite similar relative performance between both approach.

The objective of the study between ARMA and ANN are the same which is to minimize the estimation error in forecasted time series as stated by H.R. Eslami et al(2006).

The weakness of ARMA using Goal Programming method to calculate the model coefficients is the high computational cost especially for a large number of model parameters.(K.Mohammadi et al,2006)

2.4.2 Linear Regression

Mahabir (2006) states that through multiple linear regression analysis, equations were developed to model the maximum water level during spring break up. The optimal model contained a combination of hydrological and meteorological data collected.

2.5 IMPORTANT OF USING WATER LEVEL

Mostly, study conducted is using rainfall data, sedimentation, and rainfall pattern. However, in our study we are using hourly water level data. Such developments aim to the real-time prediction of floods from available rainfall data collected in the basin and specific data collected in the reservoir and dam, during the event (R. Garcia Bartual).

From a journal 'Forecasting model of Chao Praya river flood levels at Bangkok', it explained about the important of using water level.Hourly flood forecasting is necessary in coastal rivers and estuaries for the purposes of flood control and mitigation. The water level is greatly affected by the movement of the tidal waves that continually fluctuate and by the upland flood discharge (Tawatchai Tingsanchali).

Besides that, El Shafie et al(2007,2008,2009) cited that data inputs can be generated from historical river flow data, rainfall, precipitation and sedimentation in river flow forecasting. Zealand et al(1999) also included precipitation, rainfall, and flow data in their forecasting study. So, it is clearly explained that not only water level data be using but other types of data can also be used.

CHAPTER 3

RESEARCH METHODOLOGY

3.1 INTRODUCTION

Type of method used in this study is modelling. This chapter will clearly explained about the method use in this study. The modelling details is describe in next section. Next , this chapter will represent on how the data are collected and a flowchart and Gantt Chart is shown as a tentative of the method used in this study.

3.2 MODELLING DETAILS

Artificial Neural Network will be used to forecast water level at Sungai Lembing. The performance of ANN data training and validation will be determine using statistical index called Nash-Sutcliffe efficiency coefficient. Correlation Coefficient(R^2) and Root Mean Square Error (RMSE) will also be used to measure the accuracy of the data performance. Multilayer Perceptron of ANN will be used to adjust the connection weight. BASIC256 software is used to run ANN.

3.2.1 Artificial Neural Network

There have been many recent papers and contributions proving the ability and potentials of artificial neural networks (ANN) modelling approaches in the field of rainfall-runoff modelling and time series forecasting, both rainfall and runoff series (Lachtermacher, G. and Fuller, J. D.,1994; Boogard et al, 1998; Zealand et al, 1999; Luck et al,2000; Coulibaly et al, 2000; Toth et al, 2000; DeoDeo et al, 2000).

The ANN is a non-linear mathematical computing model that can solve arbitrarily complex non-linear problems such as time series data forecasting. Type of Artificial Neural Network used is Multilayer Perceptron (MLP). Multilayer Perceptron (MLP) is used to adjust the connection weight. MLP consists of three layers which are input, hidden and output layer. There are only one input and output layer but more than one in hidden layer. Each layer contains neurons. The neurons are connected to each other. In the input layer, it can contain numbers of neurons as it is by the data input study from user selection. However, in the output layer, there must be only one neuron in order to gain output. Hidden layer is up to user selection. In the input layer, an additional dummy neuron known as bias neuron is to be found. Bias neurons act as one of the competitors in the network.

The activation function is used to compute the incoming value and produce the value in the output layer. So, the actual computational process may occur. In the input layer, a linear transfer function is applied. But in the hidden and output layers, a sigmoid function is commonly used.

After the activation transfer function, the connection weight can be determined by data training. The process of the weight adjustment starts from the output layer and proceeds backward toward the input layer. The process of feed-forward and back-propagation continues until the performance target is achieved. Once the data training process is completed, the weights of the connections are determined. Then a single feed-forward computation is employed. The results are then assessed by applying performance measures to the observed data versus the model output values. Figure 3.1 shows the Multilayer Perceptron with three layers as stated before.

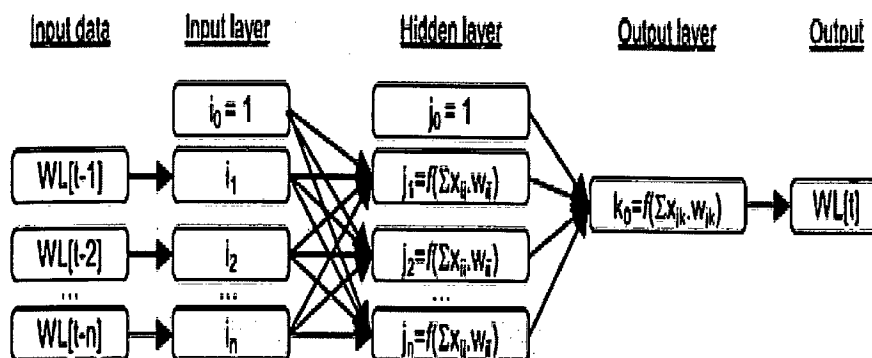


Figure 3.1 : Multilayer Perceptron

3.3 BASIC 256

BASIC 256 is a simple version of BASIC. In BASIC 256, edit window, text output, and graphic output are shown in the same screen. In this study, BASIC is an acronym for Beginner's All-purposed Symbolic Instruction Code which design to ease use for general purpose and high level programming. The acronym *BASIC* comes from the name of an unpublished paper by Thomas Kurtz. BASIC 256 is used to run ANN in arrangement of data. BASIC 256 has their own typical keyword and command that need to be study first in order to obtain our results in text output. BASIC 256 will run the arithmetic expressions such as multiplication, addition, division and subtraction. Besides that, BASIC 256 allows matrix manipulations, plotting of functions and data, implementation of algorithms, creation of user interfaces, and interfacing with programs written in other languages. BASIC 256 is widely used for beginner who interested in learning programming.

In this study, BASIC 256 software is used to run ANN. Include the suitable coding in order to gain the output. In this study, all data are arranged using BASIC 256. The advantages are to save time and easy calculation can be made using BASIC 256. Besides that, to find a suitable coding to be include in the text output is not easy. Example of coding used in this study are print, open, if, close, write line, string, read, and others. A few symbols also be used to show what we want to make as output.

3.4 DATA COLLECTION

Data that will be used in our study is daily water level data. Daily water level data will be obtained from Department of Irrigation and Drainage(DID).A formal letter is needed in order to make sure they give a cooperation. Data from year 2003 to 2009 are used for forecasting result. The data are then will be divided due to missing daily data in the historical records. The training and validation data were divided into smaller groups of continuous datasets. In this study, one third is for training data and two thirds for validation data.

3.5 FLOWCHART

Figure 3.2 shows the flow chart are used to clearly show the step in this study. The flow shows the steps from data collection until finding out performance measures.

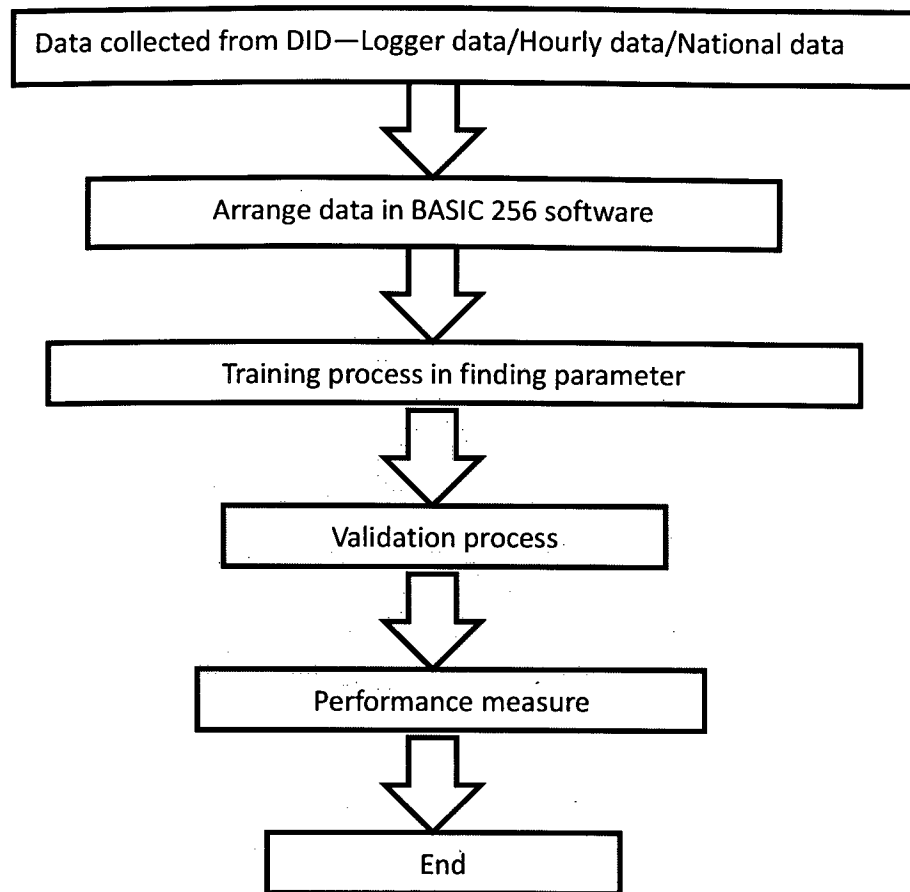


Figure 3.2 : Work flow in this study

3.6 GANTT CHART

Figure 3.3 shows the Gantt chart which is the progress of the work in order to finish this study.

No	Activities	FYP Part 1					FYP Part 2				
		SEPT	OCT	NOV	DEC	JAN	FEB	MAR	APR	MAY	JUNE
1	Find related articles and journals										
2	Background and problem statement										
3	Introduction and research objectives										
4	Literature Review										
5	Research Methodology										
6	Prepare research proposal and slides										
7	FYP 1 Presentation and approval										
8	Data Collection										
9	Review data collected										
10	Arrange data using BASIC 256										
11	Run ANN Software										
12	Gain Performance Measures										
13	Evaluate the result										
14	Conclusion										
15	Recheck thesis formats and references										
16	Prepare for Presentation PSM 2										
17	FYP 2 Presentation										
18	Amendment of corrections										
19	Submit Degree Final Year Project										

Figure 3.3 : Gantt Chart of the work progress

CHAPTER 4

DATA ANALYSIS

4.1 INTRODUCTION

The procedure of forecasting water level using Multilayer Perceptron(MLP) was described in the previous chapter. ANN model tested in this study is 2,3,4,5,6,and 7 input data values were investigated. The number of neurons in the hidden layer was equal to the number of input neurons. Data analysis had be done in two stages. First stages is to determine the best number of data input that produce the best performance. Second stage is to use best data input with increased number of iteration. The data training and data validation results for the six ANN architectures, with 2,3,4,5,6,7 input, seven data sets, and with 1000,2000 and 5000 epochs had been done. The architecture of the six network models is shown in Table 4.1

4.2 PERFORMANCE MEASURES

There are a few performance measures that are used to evaluate the results in this study. Three of these come from common statistical indices that evaluate the goodness of fit between two data series. These are the Nash-Sutcliffe efficiency coefficient (NSC), the correlation coefficient (R^2),and root mean square error (RMSE).The equations for these are shown below in Eq. (4.1), (4.2), and (4.3)

$$NSC = 1 - \frac{\sum_{i=1}^N (O_i - F_i)^2}{\sum_{i=1}^N (O_i - \bar{O})^2} \quad (4.1)$$