

**A HYBRID FORECASTING MODELS USING LEAST SQUARE
SUPPORT VECTOR MACHINE (LSSVM) AND GENETIC
ALGORITHM (GA)**

THIAN SWEE LIANG



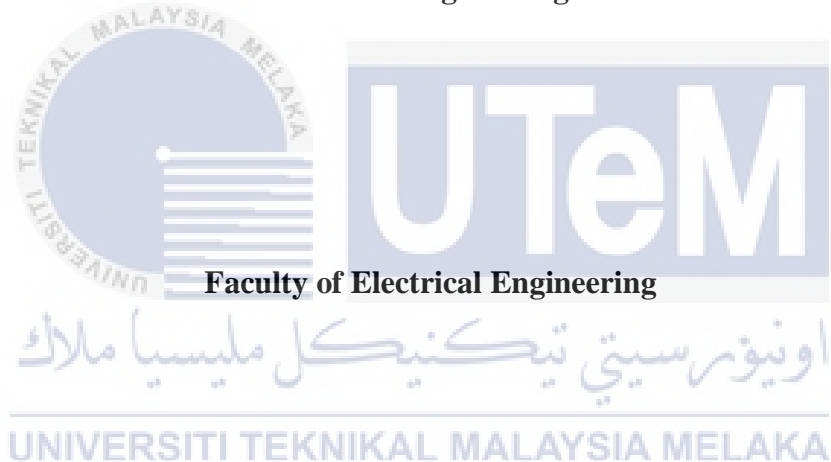
**BACHELOR OF ELECTRICAL ENGINEERING WITH HONOURS
UNIVERSITI TEKNIKAL MALAYSIA MELAKA**

2021

**A HYBRID FORECASTING MODELS USING LEAST SQUARE SUPPORT
VECTOR MACHINE (LSSVM) AND GENETIC ALGORITHM (GA)**

THIAN SWEE LIANG

**A report submitted
in partial fulfilment of the requirements for the degree of
Bachelor of Electrical Engineering with Honours**



UNIVERSITI TEKNIKAL MALAYSIA MELAKA

2021

DECLARATION

I declare that this thesis entitled “A HYBRID FORECASTING MODELS USING LEAST SQUARE SUPPORT VECTOR MACHINE (LSSVM) AND GENETIC ALGORITHM (GA) is the result of my own research except as cited in the references. The thesis has not been accepted for any degree and is not concurrently submitted in the candidature of any other degree.

Signature

:

Name

:

THIAN SWEE LIANG

Date

:

28/6/2021



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APPROVAL

I hereby declare that I have checked this report entitled “A Hybrid Forecasting Models Using Least Square Support Vector Machine (Lssvm) And Genetic Algorithm (GA)” and in my opinion, this thesis it complies with the partial fulfilment for awarding the award of the degree of Bachelor of Electrical Engineering with Honours

Signature

:



Supervisor Name

:

Dr. Intan Azmira Binti Wan Abdul Razak

Date

:

5/7/2021

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DEDICATIONS

To my beloved mother and father



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In preparing the report, I received a lot of help from my friends and supervisor. They have contributed to help me for more understanding about my project. I would like to express my special thanks to Dr Intan Azmira Binti Wan Abdul Razak act as my project supervisor who give me a lot of technical and knowledge support. I would like to convey my heartfelt thanks to my academic advisor Dr. Azrita Binti Alias for the guidance and advice. Without their support, the project would not to presented here.

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Lastly, I would like to thank my classmates and my friend for their opinions and their views during the process of completing the project.



ABSTRACT

The project is to develop electricity price forecasting models. Forecast price of electricity become an important issue in the electricity power generation field. The prediction price of electricity price can help the electricity provider decide their electricity price strategy. It can also help the electricity consumer know their electricity price more accurately and estimate their daily electricity usage a day ahead of 24-hour cost. The challenge face on forecasting prices is more on the accuracy and the efficiency of the forecasting model. Therefore, some researchers have developed a forecasting model by using the algorithm to optimize the forecast output. Hence, a combination of the Least Square Support Vector Machine (LSSVM) and Genetic Algorithm (GA) forecast model was developed in this project. The parameter used in the project was the hourly Ontario electricity price (HOEP) and Ontario demand in years 2010 provided by The Independent Electricity System Operator (IESO) website. This type of hybrid forecasting model will be tested on the Ontario power market, which is report as the most volatile market worldwide, to compare with the existing model and provide better accuracy of forecast price and performance. The hybrid forecasting model has completed developed come out with the forecasting results and performance on the hybrid forecasting model itself.

ABSTRAK

Projek ini adalah membina sesuatu model ramalan untuk meramalkan harga penggunaan elektrik. Ramalan harga elektrik merupakan satu isu yang amat penting dalam bidang penjanaan kuasa. Oleh itu, mempunyai ramalan harga elektrik akan bermanfaat kepada industri penjanaan kuasa dapatkan info penggunaan kuasa elektrik masa depan, ia juga akan membantu pengguna elektrik untuk mengetahui harga penggunaan elektrik yang lebih tepat dan lanjut bagi masa yang depan seperti mengetahui ramalan harga elektrik untuk hari keesokan. Walau bagaimanapun, cabaran atas model ramalan yang telah dicipta adalah ketepatan dan kecekapan model tersebut. Oleh itu, sesetengah penyelidik menggunakan pelbagai algoritma untuk optimumkan hasil yang dapat dalam model ramalan yang diciptakan. Dengan ini, model ramalan yang mempunyai gabungan antara algoritma 'Least Square Support Vector Machine (LSSVM)' dan algoritma genetik akan dihasilkan dalam projek ini. Data yang digunakan dalam projek ini adalah harga elektrik setiap jam Ontario pada tahun 2010 yang dikongsikan oleh The Independent Electricity System Operator (IESO) di halaman web rasmi mereka. Tujuan kombinasi dua algoritma ini akan mengoptimumkan ketepatan dan kecekapan hasilnya, dan ianya akan digunakan di tempat Ontario dengan menggunakan data kuasa elektrik di Ontario. Sebagai pengetahuan Ontario mempunyai gelaran antara harga elektrik yang paling tidak stabil di dunia, dengan mengambil kesempatan untuk mengkaji ketepatan keluaran dari projek yang sedia ada. Model ramalan gabungan antara algoritma 'Least Square Support Vector Machine (LSSVM)' dan 'algoritma genetik' berjaya dibina dengan mengeluarkan keluaran anggaran yang ditetapkan.

TABLE OF CONTENTS

	PAGE
DECLARATION	
APPROVAL	
DEDICATIONS	
ACKNOWLEDGEMENTS	2
ABSTRACT	3
ABSTRAK	4
TABLE OF CONTENTS	5
LIST OF TABLES	7
LIST OF FIGURES	8
LIST OF SYMBOLS AND ABBREVIATIONS	11
LIST OF APPENDICES	12
CHAPTER 1 INTRODUCTION	13
1.1 Background	13
1.2 Problem Statement	14
1.3 Objective	14
1.4 Project Scope	15
1.5 Thesis Outline	15
CHAPTER 2 LITERATURE REVIEW	16
2.1 Introduction	16
2.2 Support Vector Machine	16
2.3 Least Square Support Vector Machine (LSSVM)	18
2.4 Genetic Algorithm (GA)	19
2.5 The Ontario electricity system	Error! Bookmark not defined.
CHAPTER 3 METHODOLOGY	20
3.1 Introduction	20
3.2 Project Flowchart	21
3.3 Project Grantt Chart	22
3.4 Support Vector Machine and Least Square Support Vector Machine (LSSVM)	24
3.5 Training Data Selection of Forecasting	27
3.6 Testing Data Selection of Forecasting	40
3.7 Mean Absolute Error (MAE)	46
3.8 Mean Absolute Percentage Error (MAPE)	46
3.9 Data Normalization	47
CHAPTER 4 RESULTS AND DISCUSSIONS	49

4.1	Introduction	49
4.2	LSSVM Forecasting Model Forecast Result	49
4.3	LSSVM-GA Hybrid Forecasting Model Forecast Result	54
4.4	Performance Between LSSVM forecasting model and LSSVM GA Hybrid Forecasting Model	60
4.5	Comparison LSSVM Forecasting Model and LSSVM GA Hybrid Forecasting Model With Other Types of Forecasting Model	62
CHAPTER 5	CONCLUSION AND RECOMMENDATIONS	64
5.1	Conclusion	64
5.2	Recommendation and Future Works	64
	REFERENCES	65
	APPENDICES	66



LIST OF TABLES

Table 4-1: Daily MAPE Values of LSSVM Forecasting Model	53
Table 4-2: LSSVM GA Daily MAPE	59
Table 4-3: LSSVM Forecasting Model VS LSSVM GA Forecasting Model MAPE	60
Table 4-4: Comparison of MAPE with Different Forecasting Models	63
Table 5-1: LSSVM HOEP Actual and Forecast on 15/7/2010 and 16/7/2010	66
Table 5-2: LSSVM HOEP Actual and Forecast on 17/7/2010 and 18/7/2010	67
Table 5-3: LSSVM HOEP Actual and Forecast on 19/7/2010 and 20/7/2010	68
Table 5-4: LSSVM HOEP Actual and Forecast on 21/7/2010	69
Table 5-5: LSSVM-GA HOEP Actual Value and Forecast Values on 15/7/2010 and 16/7/2010	70
Table 5-6: LSSVM-GA HOEP Actual Value and Forecast Values on 17/7/2010 and 18/7/2010	71
Table 5-7: LSSVM-GA HOEP Actual Value and Forecast Values on 19/7/2010 and 20/7/2010	72
Table 5-8: LSSVM-GA HOEP Actual Value and Forecast Values on 21/7/2010	73

LIST OF FIGURES

Figure 2-1: Data Before Classification	16
Figure 2-2: Data After Classification	17
Figure 2-3: Different Of The High Sigma(Left) And Low Sigma(Right) Of The Data Separating	17
Figure 3-1: The Flowchart of Project	21
Figure 3-2: 7/5/2010 Training Data of Hourly Ontario Electricity Price (HOEP)	29
Figure 3-3: 7/6/2010 Training Data of Hourly Ontario Electricity Price (HOEP)	29
Figure 3-4: 7/7/2010 Training Data of Hourly Ontario Electricity Price (HOEP)	30
Figure 3-5: 7/8/2010 Training Data of Hourly Ontario Electricity Price (HOEP)	30
Figure 3-6: 7/9/2010 Training Data of Hourly Ontario Electricity Price (HOEP)	31
Figure 3-7: 7/10/2010 Training Data of Hourly Ontario Electricity Price (HOEP)	31
Figure 3-8: 7/11/2010 Training Data of Hourly Ontario Electricity Price (HOEP)	32
Figure 3-9: 7/12/2010 Training Data of Hourly Ontario Electricity Price (HOEP)	32
Figure 3-10: 7/13/2010 Training Data of Hourly Ontario Electricity Price (HOEP)	33
Figure 3-11: 7/14/2010 Training Data of Hourly Ontario Electricity Price (HOEP)	33
Figure 3-12: 7/5/2010 Training Data of Ontario Demand	35
Figure 3-13: 7/6/2010 Training Data of Ontario Demand	35
Figure 3-14: 7/7/2010 Training Data of Ontario Demand	36
Figure 3-15: 7/8/2010 Training Data of Ontario Demand	36
Figure 3-16: 7/9/2010 Training Data of Ontario Demand	37
Figure 3-17: 7/10/2010 Training Data of Ontario Demand	37
Figure 3-18: 7/11/2010 Training Data of Ontario Demand	38
Figure 3-19: 7/12/2010 Training Data of Ontario Demand	38

Figure 3-20: 7/13/2010 Training Data of Ontario Demand	39
Figure 3-21: 7/14/2010 Training Data of Ontario Demand	39
Figure 3-22: 7/15/2010 Testing Data of Hourly Ontario Electricity Price (HOEP)	42
Figure 3-23: 7/16/2010 Testing Data of Hourly Ontario Electricity Price (HOEP)	42
Figure 3-24: 7/17/2010 Testing Data of Hourly Ontario Electricity Price (HOEP)	43
Figure 3-25: 7/18/2010 Testing Data of Hourly Ontario Electricity Price (HOEP)	43
Figure 3-26: 7/19/2010 Testing Data of Hourly Ontario Electricity Price (HOEP)	44
Figure 3-27: 7/20/2010 Testing Data of Hourly Ontario Electricity Price (HOEP)	44
Figure 3-28: 7/21/2010 Testing Data of Hourly Ontario Electricity Price (HOEP)	45
Figure 3-30: Flowchart of Normalization	48
Figure 4-1: Actual VS Forecast graph for the testing period (15 July 2010 ~ 21 July 2010)	49
Figure 4-2: HOEP Actual Values VS Forecast Values on 15/7/2010	50
Figure 4-3: HOEP Actual Values VS Forecast Values on 16/7/201	50
Figure 4-4: HOEP Actual Values VS Forecast Values on 17/7/2010	51
Figure 4-5: HOEP Actual Values VS Forecast Values on 18/7/2010	51
Figure 4-6:: HOEP Actual Values VS Forecast Values on 19/7/2010	52
Figure 4-7: HOEP Actual Values VS Forecast Values on 20/7/2010	52
Figure 4-8:: HOEP Actual Values VS Forecast Values on 21/7/2010	53
Figure 4-9: Genetic Algorithm Plot Graph	54
Figure 4-10: Hybrid Forecasting Model (LSSVM-GA) Prediction Values VS Actual Values of HOEP Results	55
Figure 4-11: LSSVM-GA HOEP Actual Value VS Forecast Values on 15/7/2010	56
Figure 4-12: LSSVM-GA HOEP Actual Value VS Forecast Values on 16/7/2010	56
Figure 4-13: LSSVM-GA HOEP Actual Value VS Forecast Values on 17/7/2010	57

Figure 4-14: LSSVM-GA HOEP Actual Value VS Forecast Values on 18/7/2010	57
Figure 4-15: LSSVM-GA HOEP Actual Value VS Forecast Values on 19/7/2010	58
Figure 4-16: LSSVM-GA HOEP Actual Value VS Forecast Values on 20/7/2010	58
Figure 4-17: LSSVM-GA HOEP Actual Value VS Forecast Values on 21/7/2010	59
Figure 4-18: LSSVM Forecasting Model VS LSSVM GA Forecasting Model MAPE	61



LIST OF SYMBOLS AND ABBREVIATIONS

IESO	-	The Independent Electricity System Operator
LSSVM	-	Least Square Support Vector Machine
SVM	-	Support Vector Machine
MAE	-	Means Absolute Error
MAPE	-	Means Absolute Percentage Error
GA	-	Genetic Algorithm
FA	-	Fuzzy ARTMAP
FF	-	Firefly Algorithm
WT		Wavelet Transform



LIST OF APPENDICES

APPENDIX A	LIST OF LSSVM HOEP ACTUAL AND FORECAST	
RESULTS		66
APPENDIX B	LIST OF LSSVM GA HOEP ACTUAL AND FORECAST	
RESULTS		70



CHAPTER 1

INTRODUCTION

1.1 Background

Electricity is a flow of charge, and it was one of the essential daily needs for humans nowadays. Based on the data and statistics done by International Energy Agency, world electricity consumption for the year 2018 was archived over 24000 TW/h, the consumption of electricity still increasing every year.

Since the early of years 1990, Europe, Canada, North America, and Australia started introducing a competitive electricity market due to deregulation in the countries. By transforming competitive electricity market launching, the electricity was traded based on two common categories, which are sold under market rules spot and derivative contracts. This type of trade was made a most significant change on the traditional monopolistic and government-controlled power sectors.

Price forecasting becomes an important strategy, especially for the power generation company. Future electricity price forecasting can help the power generation company optimize the output and gain a good profit. The same as the electrical power consumption can also benefit from an electricity price forecast to control their energy consumption. At the same time, it will become more manageable for the consumer to choose which electricity supplier they need to choose.

Some researchers have successfully developed a forecasting model with a Support Vector Machine or hybrid forecasting model to predict the electricity price. However, forecasting electricity price was more challenging than load or demand because some factors can directly affect the cost of electricity, such as coal, fuel price, etc. In this project, the Least Square Support Vector Machine with optimization of the Genetic algorithm will be used. The project will be developed to improve the accuracy and the performance compared with the previous model.

1.2 Problem Statement

Nowadays, several forecasting models with different algorithms or methods exist in the market, such as straight-line method, moving line method, simple linear regression method, multiple regression method, etc.

The way to developing a good forecasting model will easier the user during the electricity price forecasting. The forecasting model should need a stable or friendly user platform for development to ensure that the forecasting model can be complete when running forecasting work with good forecasting results.

The multiple regression analysis methods with the support of LSSVM was one of the forecasting models available in the market. The result comes out with the forecasting model become a challenge to get an accurate forecasting result. The forecasting model also can improve its accuracy by supporting other optimization methods such as the Genetic algorithm.

1.3 Objective

The aim of this project to:

- i. Develop hybrid Forecasting model (LSSVM – GA) with MATLAB.
- ii. Optimize LSSVM parameters (gamma and sigma) by Genetic Algorithm.

1.4 Project Scope

- i. The implementation of LSSVM technique as main forecast engine.
- ii. The input used for the forecast is the historical price and demand of electricity per hours.
- iii. The output forecast for day-ahead 24-hour prices.
- iv. The objective function of the model is Mean Absolute Percentage Error (MAPE) and MAE; where minimum error should be achieved to obtain good predictive accuracy.
- v. The optimization process is held by the Genetic Algorithm (GA) to optimize the parameters of LSSVM and reduce the number of input for prediction.
- vi. The model developed is tested in the Ontario electricity market, which is one of the most volatile energy markets in the world

1.5 Thesis Outline

The thesis is organized into five chapters, which is:

Chapter 1 explained the background of the project and followed with the problem statement. In the background of the project, the overview of the project was discussed. After that, the problem statement was to state the problem more clearly and followed with the objective and scope of the project.

Chapter 2 will describe the literature review LSSVM and Genetic algorithm, which contained thesis, book or journal information related to the project needed to develop.

Chapter 3 discussed methodology for the LSSVM, Training data selection, testing data selection, Mean Absolute Error (MAE) and Means Absolute Percentage Error (MAPE).

Chapter 4 discuss the data after the forecasting model made the prediction and comparisons between the LSSVM, LSSVM-GA and other existing forecasting models.

Chapter 5 will discuss the conclusion and the recommendation based on the project.

CHAPTER 2

LITERATURE REVIEW

2.1 Introduction

This chapter will describe about the literature review on Support Vector Machine, Least Square Support Vector Machine (LSSVM), Genetic Algorithm. The information on this chapter come from various source such as thesis, books, journal, article and other academic websites. All the information were cited clearly and benefit on the road of complete the project.

2.2 Support Vector Machine

Support Vector Machine is one of the machines learning techniques and it was introduced by Vapnik [1]. The Support Vector Machine (SVM) commonly used for classifying or separating of two data. As an example using in the forecasting model to separating the demand and the price of electricity.

A simple example for easier understanding, when it had two of the data as shown as the figure 2.2.1 below, the data in the graph was mixing on the graph. In this case, separation step will be done by the SVM. SVM will separate the different data with the line or can called hyperplane as show in the **Figure 2-1**.

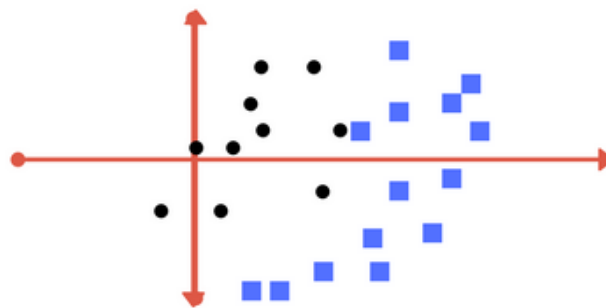


Figure 2-1: Data Before Classification

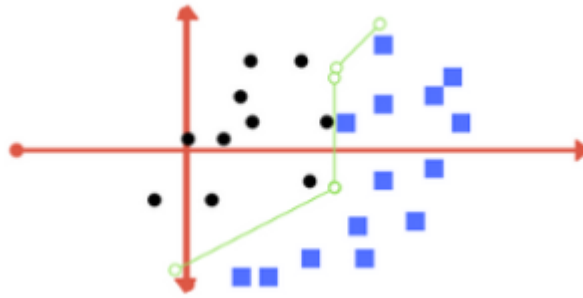


Figure 2-2: Data After Classification

When the operation of separating data done by SVM, there was few parameters need to be considered in our forecasting project especially the Gamma and Sigma. Gamma can be understanding as the how far or close of the separation line needed to consider on the separating line calculation. It can be known as the how long of our forecasting data range needed to be considered in the forecasting model to produce our output needed. Sigma also an important parameter guided forecasting model to archived an accurate result. Sigma can be explained as the **Figure 2-3** shown the different of the high sigma(left) and low sigma(right) of the data separating.



Figure 2-3: Different Of The High Sigma(Left) And Low Sigma(Right) Of The Data Separating

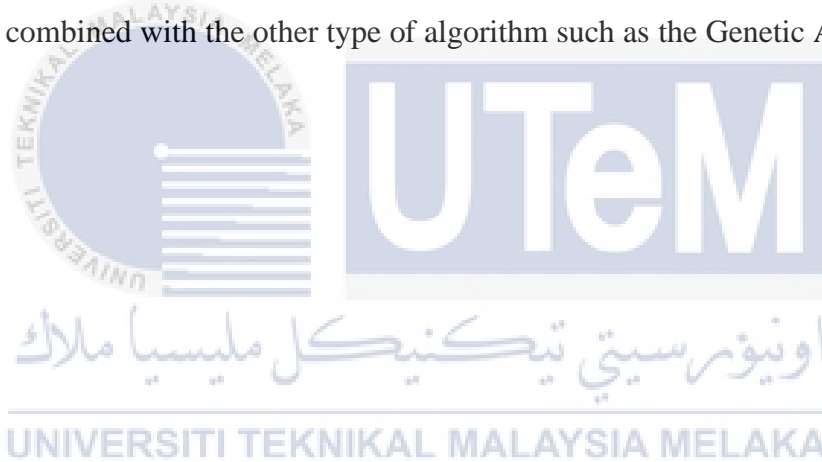
Lastly, the more important role of the SVM in the project was to produce a good margin during the forecasting process. A good margin of SVM needed to considered the equability of two data, closer gamma parameter and low sigma parameter as well.

2.3 Least Square Support Vector Machine (LSSVM)

LSSVM is the least-squares of SVM. It was first proposed by J.A.K. Suykens and J. Vandewalle as a classifier in 1999. [2] [3] Unlike the inequality constraints introduced in the standard SVM, LSSVM proposed equality constraints in the formulation. LSSVM solves a system of linear equations instead of quadratic programming (QP) problem that improves the computational speed. The solution is transformed from solving a quadratic program to a set of linear equations known as the linear Karush–Kuhn–Tucker (KKT) systems. [2] [4]

LSSVM become an efficient method to become a choice to apply on the forecasting model. Refer to the previous researcher, the LSSVM forecasting model is better than the SVM forecasting model during the training process. [5]

For improving the performance, a suitable LSSVM method in forecasting data can be enhanced as combined with the other type of algorithm such as the Genetic Algorithm.



2.4 Genetic Algorithm (GA)

Genetic algorithm has reflected the process as a natural selection where the fittest individual are selected for evolution process via reproduction. The genetic algorithm has 5 main phase on in process which stand for initial population, fitness function, selection, crossover and mutation. [6] [7]

The every single phase has their own characteristic, let start with the initial population phase. Initial population phase was a beginning step of transformation in this phase, it will have a set of individual (problem) then follow with a set of genes will joined into a string to form a chromosome. In other word means that a problem(individual) will be solved by a solution (Chromosome).

The next phase was represent as fitness function. Fitness function was a pahase to determine how to fit an individual. It means that these phases has the ability which able the an individual to compete with the other individual. For an example, genes A had more taller genetic compare genes B.

Next, Selection phase was represent the select step on the fitness individual to fulfil the enquiry. Same as the meaning on a situation had a lot of the individual, researcher will select the individual which had more closer the fitness score for reproduction.

In the transformation of the genetic algorithm, crossover also had happened on it. Crossover means the random chosen at genes. Lasty was the mutation. Mutation had low random probability happened because it will exchange the genes.

As the phases of genetic algorithm explain before, the concept on GA will act as 2nd step of optimization of data to increase the accuracy of the forecast mode.

CHAPTER 3

METHODOLOGY

3.1 Introduction

The design of the hybrid forecasting model by MATLAB will be explained in detail. The project flowchart also will present in this chapter. The Support Vector Machine (SVM), Least Square Support Vector Machine (LSSVM), and Genetic Algorithm (GA) forecast process will be explain more detail in this chapter. The mean absolute error (MAE) and Mean Absolute Percentage Error (MAPE) calculation method will show in this chapter to observes performance of designing forecast model.

