

Car Parking Demand Forecasting using Fuzzy Logic

by

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CERTIFICATION OF APPROVAL

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A project dissertation submitted to the
Electrical and Electronics Engineering Programme
Universiti Teknologi PETRONAS
in partial fulfilment of the requirement for the
BACHELOR OF ENGINEERING (Hons)
(ELECTRICAL AND ELECTRONICS ENGINEERING)

Approved by,



(Dr Zuhairi Baharudin)

UNIVERSITI TEKNOLOGI PETRONAS

TRONOH, PERAK

Sept 2011

CERTIFICATION OF ORIGINALITY

This is to certify that I am responsible for the work submitted in this project, that the original work is my own except as specified in the references and acknowledgements, and that the original work contained herein have not been undertaken or done by unspecified sources or persons.



(SOFIA DELAILA BINTI MOHD SAMIN)

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ABSTRACT

Forecasting in the retail sector can presents some unusual challenge due to special demand pattern. As central business districts, airports, transit stations, and shopping mall continue to become more crowded during peak times, demand for parking spaces has increasing. In this project, Car Parking Demand Forecasting Using Fuzzy Logic will employ the Short-range Demand Forecasting method to predict the future parking spaces demand for Wangsa Walk Mall. The historical data of Wangsa Walk Mall from January 2011 to June 2011 are gathered, analyzed and utilized for the forecasting model's development. A short-range demand forecasting model is developed, simulated and fine-tuning using a mathematical software, *fuzzyTECH*. The proposed model will be tested to forecast car parking demand by weekly with the actual parking data. The most accurate model is obtained by reducing the percentage error between the forecasted and actual value of the parking space demand, in the form of mean absolute percentage error (MAPE) results. This project is expected to develop short-range parking space demand forecasting with mean absolute percentage error (MAPE) lower than 10% for an optimal result and accurate model.

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CHAPTER 1

INTRODUCTION

“When King Solomon said there were a time and a place for everything he had not encountered the problem of parking his automobile.”

– Bob Edwards (1947-2005)

1.1 BACKGROUND

Parking is a key issue in all major cities like New York, Chicago, Seoul, London, Beijing and New Delhi. According to the Ministry of Transport Malaysia (MOT) statistics for the year of 2009 and 2010, there is an increase of approximately 1.2 million passenger vehicles from 19,016,782 to 20,188,565 in Malaysia itself [1]. These 19 million vehicles include cars, motorcycles, buses, taxis, rental cars, and pick-up trucks. The Bloomberg Bussinesweek magazine states that the increases of passenger vehicles that are introduced into the roads each year will ultimately leads to a parking problem due to limited availability of spaces [2]. A customer satisfaction survey conducted by a car manufacturing company, Bavarian Motor Works (BMW) shows that 20% of the inner traffic of major cities across the nation is in search of parking space [3].

Car parking demand forecasting is a critical part for a business district’s operation management’s function. It has significant influence on achieving optimal operation for the company’s parking management. It can predict the future parking demand and estimate the projection of a company’s parking capacity. This is important because if the parking capacity is inadequate, the resulting shortages can lead to loss of customers and market interest. This will affect the company’s economy, business, loss of time and image. If the demand is over forecasted, it will be a waste of resources as the parking spaces will not be used and wasted [4].

Demand forecasting in operation management is categorized into three forecasting-time-horizon categories which are short-range forecast, medium-range forecast and long-range forecast. Short-range forecast has a time span of up to 1 year but is generally less than 3 months. It is used for capacity, planning purchasing, job scheduling, workforce levels, job assignments and production levels. Meanwhile, medium-range forecasting is usually conducted from three month to three year, and long-range forecasting is normally longer than three year, usually covers from one to ten years ahead [5].

Hence, this project studied the parking space demand pattern by employing the short-range demand forecasting method to predict the future parking spaces demand in Wangsa Walk Mall for 24 consecutive weeks. To achieve this, the Fuzzy Set Theory introduced by Prof. Lotfi A. Zadeh in 1965 was applied as a complement tool to the mathematical approach in solving the parking problems [6].

1.2 PROBLEM STATEMENT

1.2.1 Problem Identification

Kuala Lumpur is the second largest populated city in Malaysia and it is the fastest growing metropolitan region in the country. Throughout 2000 to 2010, the average annual population growth rate of Kuala Lumpur grew 1.1% from 1,379,310 to 1,722,500 [7]. The Wangsa Walk Mall, Wangsa Maju is an excellent example of heavy shopper traffic with a pattern of parking inadequacies. Located in Wangsa Maju, Kuala Lumpur the shopping mall is designed as an outdoor walking mall with a total of 173 tenants of high-end outlets.

Frustration will arises because of the wasted time and increasing fuel consumption from going around the parking building just to find an empty parking slot. This scenario happened in any major shopping mall in Malaysia, where hundreds of customers has voice out their concern and dissatisfaction in finding an available parking

space upon thousands of parked car. Hence this project can determine a business district's parking capacity and help the retail sector to achieve optimal operation on the parking spaces management.

In order to accurately forecast the short-range car parking demand, important variables such as type of day (such as normal working day, public holiday, special occasion) and type of school day (normal school day or school holiday) must be considered as each of it as a significant effects to the parking demand. All of these variables when combined together can create a large influence in the forecast result.

1.2.2 Significance of Project

The importance of parking demand forecasting shall be emphasized at all levels as the consequences of under or over estimation will affect the company's business. If under estimated, the result would be loss of customers and market interest, thus will affect the economy, business, loss of time and image. If the demand is over forecasted, it will be a waste of resources as the parking spaces will not be used and wasted [4]. Through this project, the company should be able to predict the future parking demand by focusing on fuzzy logic as a technique used in the forecasting method.

1.3 OBJECTIVES AND SCOPE OF STUDY

1.3.1 Objectives of the Project

The following objectives were to be achieved from this project:

- To study and analyze the parking space demand in a study area
- To be able to forecast the parking space demand for 7 days ahead and able to revise the forecast based on the input changes.
- To design an accurate Short-ranged Parking Demand Forecasting Model using Fuzzy Logic approach by reducing the percentage error between the forecast and actual value of the parking space occupancy (MAPE).

1.3.2 Scope of Study

Generally, the scope of study will cover the following areas:

- Short-ranged parking demand in Wangsa Walk Mall
- Focus on Fuzzy Logic as a technique that will be used in the forecasting method
- *fuzzyTECH 5.52* software as the parking space forecasting tool
- The simulation of short-ranged parking forecasting model

1.4 RELEVANCY OF THE PROJECT

Forecasting in the retail sector presents some unusual challenge. A major technique in the retail sector is tracking demand by maintaining good short-term records. Due to special demand pattern, many service firms encounter problems in providing the best service for expectant customers. These special demand patterns should be kept in records, noting not only the day of the week but also unusual events such as public holiday or special occasion, so that pattern that influence demand can be developed [4].

As forecasts is a critical part of the operation management's function, this projects aims to predict the future parking spaces demand in order to determine the total amount of visitors during school/public holidays, normal working day or during the company's special event. Therefore, an accurate prediction for the car parking demand is very important as the consequences of under or over estimation will affect the company's business.

The forecast accuracy; Mean Absolute Percentage Error (MAPE) is measured based on the error statistic of forecast between the models for one week step. The overall MAPE is calculated by taking into calculation each week's MAPE. The lower overall MAPE shows a better accuracy of the forecasting model.

1.5 FEASIBILITY OF THE PROJECT

This project will be conducted for two semesters. This includes system research, development and improvement. In this project, the data for parking space occupancy throughout the 24 consecutive weeks were obtained from Wangsa Walk Mall, starting from 03 January 2011 until 19 June 2011 (refer to Appendix A). This project is feasible to be conducted within the time frame and scope.

CHAPTER 2

LITERATURE REVIEW

2.1 BACKGROUND: PARKING LOT DESIGN

Interest in availability of parking lot comes from the problems caused by traffic congestion while trying to find available parking spaces. Traffic congestion has been increasing worldwide as a result of increased motorization, urbanization, population growth, and changes in population density. Congestion reduces efficiency of car parking allocation, and increases travel time, air pollution and fuel consumption. Studies on parking lot must be conducted to collect the required information about the capacity and use of existing parking facilities.

In addition, information about the demand for parking can be obtained. Parking studies will be restricted to a specific study area, such as a central business district. A cordon line is drawn to delineate the study area. It should include traffic generators and a periphery, including all points within an appropriate walking distance. The survey area should also include any area that might be impacted by the parking modifications. The boundary should be drawn to facilitate cordon counts by minimizing the number of entrance and exit points [8].

2.1.1 Parking Lot Design and Concep: Accumulation Counts

It is conducted to obtain data on the number of vehicles parked in a study area during a specific period of time. First, the number of vehicles already in that area are counted or estimated. Then the number of vehicles entering and exiting during that specified period are noted, and added or subtracted from the accumulated number of vehicles. Accumulation data are normally summarized by time period for the entire study area.

The occupancy can be calculated by taking accumulation/total spaces. Peaking characteristics can be determined by graphing the accumulation data by time of day. The accumulation graph usually includes cumulative arrival and cumulative departure graphs as well.

Table 1: Accumulation Table for University of Idaho, United States

Table 2. Accumulation Table

Period	Arrive	Cum. Arr.	Depart	Cum. Dep.	Accumulation	Accumulation Factor
4 - 6 AM	25	25	5	5	20	0.0341
6 - 8	50	75	15	20	55	0.0940
8 - 10	250	325	25	45	280	0.4786
10 - 12 Noon	50	375	30	75	300	0.5128
12 - 2 PM	100	475	60	135	350	0.5292
2 - 4	50	525	75	210	325	0.5535
4 - 6	25	550	220	430	130	0.2222
6 - 8	25	575	100	530	55	0.0240
8 - 10	10	585	65	595	0	0
Total		585		585		

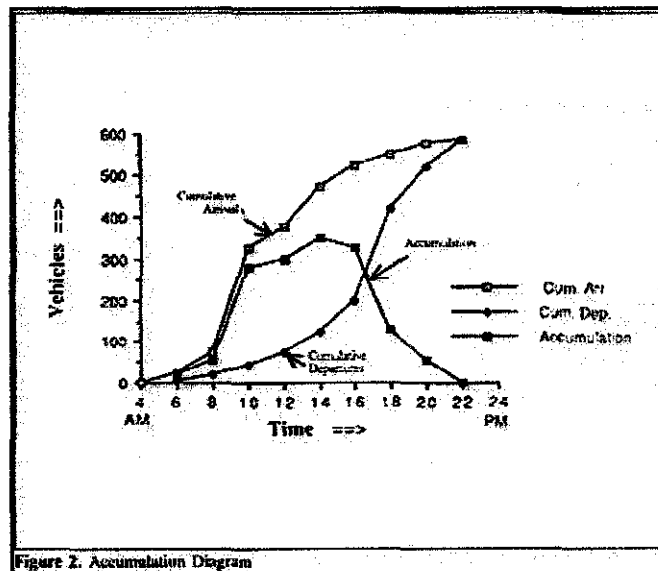


Figure 1: Accumulation Diagram for University of Idaho, United States

2.2 FORECASTING TIME HORIZON

Forecasting is the art and science of predicting future events. Forecasting may involve taking historical data and projecting them into the future with some sort of mathematical model. It may be subjective or intuitive prediction. Or it may involve a combination of these – that is, a mathematical model adjusted by a manager's good judgment. Forecasting are seldom, if ever, perfect. They are also costly and time-consuming to prepare and monitor.

A forecast is usually classified by the future time horizon that it covers. Time horizon fall into three categories which are short-range forecast, medium-range forecast and long-range forecast. Short-range forecast has a time span of up to 1 year but is generally less than 3 months. It is used for capacity, planning purchasing, job scheduling, workforce levels, job assignments and production levels. Medium-range forecast, or intermediate, generally spans from 3 months to 3 year. It is useful in sales planning, production planning and budgeting, cash budgeting, and analysis of various operating plan. Long-range forecast, which is generally 3 years or more in time span, are used in planning for new products, capital expenditures, research and development [4].

Short-range forecast tend to be more accurate than longer-range forecast. Factors that influence demand changes every day. Thus, as the time horizon lengthens, it is likely that forecast accuracy will diminish.

Organizations use three major types of forecasts in planning future operations, which are economic forecast, technological forecast and demand forecast. Economic forecasts address the business cycle by predicting inflation rates, money supplies and housing starts, while technological forecast are concerned with rates of technological progress. Demand forecast which is employed in this project are projections of demand for a company's product, goods or services. These forecasts drive a company's production, capacity and scheduling planning systems and serve as inputs to financial, marketing, management decision and personnel planning [9].

2.3 FUZZY LOGIC THEORY

2.3.1 *The Fuzzy Logic Concept*

Natural language abounds with vague and imprecise concepts, such as “Sally is tall,” or “It is very hot today.” Such statements are difficult to translate into more precise language without losing some of their semantic value: for example, the statement “Sally’s height is 152 cm.” Does not explicitly state that she is tall, and the statement “Sally’s height is 1.2 standard deviations about the mean height for women of her age in her culture” is fraught with difficulties: would a woman 1.1999999 standard deviations above the mean be tall? Which culture does Sally belong to, and how is membership in it defined?

The first publication on fuzzy logic, which also coined its name, dates back to 1965. It was written in the U.S. by Lotfi Zadeh, Professor of Systems Theory at the University of California, Berkeley. From there, the history of fuzzy logic follows the pattern of number recent technologies: invented in the U.S., engineered in Europe, and mass-marketed in Japan [6].

The first industrial applications of fuzzy logic were completed after 1970 in Europe. At Queen Mary College in London, England, Ebrahim Mamdani used fuzzy logic to control a steam generator that he could not get under control with conventional techniques. At the RWTH University of Aachen, Germany, Hans-Jürgen Zimmermann used fuzzy logic for decision support systems. Other industrial applications, such as the control of a cement kiln, followed as a result of this initial work, but fuzzy logic could not get broad acceptance in industry. The few applications that used fuzzy logic hid the fact by circumscribing fuzzy logic with terms such as “multi-valued logic” or “continuous logic.”

Fuzzy logic benefits can be seen in its simplicity and its flexibility. Fuzzy logic can handle problems with imprecise and incomplete data, and it can model nonlinear functions of arbitrary complexity [10]. The basic idea of fuzzy logic is simple: in reality, we cannot define a rule for each possible case. Exact rules that cover a case perfectly can only be defined for a few distinct cases. These rules are discrete points in the continuum of possible cases and humans approximate between them.

2.3.2 *Mathematical Modelling*

Prof. Zadeh contention is that “meaning in natural language is a matter of degree”. If we have a proposition such as “Sally is tall”, then it is not always possible to assert that it is either true or false. When we know that Sally’s height is y , then the “truth”, or more correctly, the “compatibility” of y with “is tall”, is a matter of degree.

If the proposition is “Sally is under 160cm height” and we know Sally’s height, then we can give a yes or no answer to whether the proposition is true or not. In contrast to binary sets that having binary logic (crisp logic), the fuzzy variables may have memberships values of not only 0 or 1. The values can range from 0 to 1. The difference between binary sets and fuzzy logic are as follows:

An ordinary subset A of a set U is determined by its indicator function X_A defined by:

$$X_A(x) = \begin{cases} 1, & x \in A \\ 0, & x \notin A \end{cases}$$

Crisp set A is defined by the characteristic function:

$$A(x): X \rightarrow 0, 1$$

Fuzzy set A is defined by the membership function:

$$A(x): X \rightarrow [0, 1] \quad \text{where: } A(x) = 1 \quad \text{if } x \text{ is totally in } A;$$
$$A(x) = 0 \quad \text{if } x \text{ is not in } A;$$
$$0 < A(x) < 1 \quad \text{if } x \text{ is partly in } A.$$

Or simply,

The previous example shows the fuzzy logic with a numerical value. The fuzzy logic usually is applied for the non numeric linguistic variables. It is often used for facilitating the expression of rules and facts. The Fuzzy Logic uses IF-THEN rules which employ the linguistic variables (fuzzy variables) whose values are in the linguistic terms [10].

The rules are usually expressed in the form of:

IF variable **IS** property **THEN** action

Rule 1:	IF Distance = medium AND Angle = neg_small THEN Power = pos_high
Rule 2:	IF Distance = medium AND Angle = neg_big THEN Power = pos_medium
Rule 3:	IF Distance = far AND Angle = neg_small THEN Power = pos_high

For multiple inputs, the following rule applied:

If X is A, and Y is B **then** Z is C.

2.4 FUZZY LOGIC BASED FORECASTING SYSTEM MODELING

The structure of the fuzzy logic configurations can be represented as below:

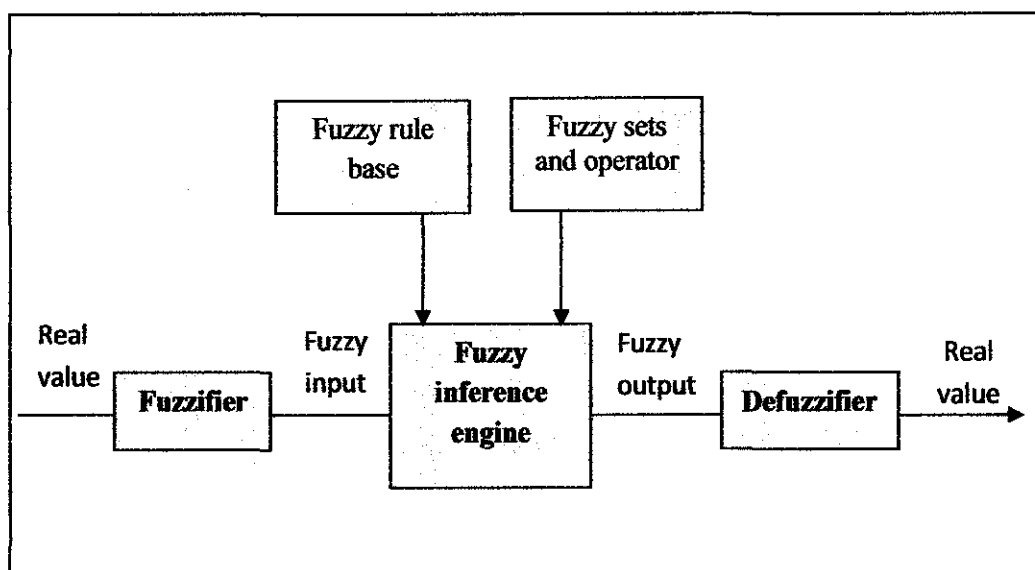


Figure 2: Basic Component of Fuzzy System

2.4.1 Fuzzification

The modelling starts with user entering the inputs. The inputs are in terms of the crisp value. The Fuzzifier will then transform the real valued input variables into fuzzy values.

2.4.2 Fuzzy Inference

The Fuzzy Inference engine will map the fuzzy inputs earlier to the fuzzy output based on the Fuzzy Rules. The Fuzzy Rules and Fuzzy Set Database are entered by the user to enable the inference engine to function properly. Now that all input variables have been converted to linguistic variable values, the fuzzy inference step can identify the rules that apply to the current situation and compute the value of the output linguistic variable.

The fuzzy inference system is used to evaluate the similarity between the previous forecast days and previous similar days resulting in correction factors, used to correct the similar days of the forecast day to obtain the parking demand forecast. The computation of the fuzzy inference consists of three components:

- **Aggregation:**

Computation of the IF part of the rules. This step computes the support of the rule relative to the conditions.

- **Composition:**

Computation of the THEN part of the rules. This step computes the degree of truth for the rule.

- **Result Aggregation:**

After the degrees of truth for the rules are computed, this step determines which rules will contribute to the Defuzzified result

Rule base optimization often consists of arbitrary rule addition/deletion. This method can result in a clumsy trial-and-error approach as the individual importance of a rule can be expressed only as a 0 or 1.

For this reason, most *fuzzyTECH* Editions support an advanced inference method, the Fuzzy Associative Maps inference, or FAM. With FAM, each rule is assigned a degree of support representing the individual importance of the rule. Rules themselves can be “fuzzy” – meaning, with validity between 0 and 1.

The validity of a conclusion is calculated by a linking of the validity of the entire condition with the degree of support by a composition operator. When the product operator is used as the composition operator, the degree of support reflects rule “significance” [6].

2.4.3 Defuzzification

The result produced from the evaluation of fuzzy rules is fuzzy. Membership functions are used to retranslate the fuzzy output into a crisp value. This translation is known as Defuzzification and can be performed using several methods. Thus, the result of the fuzzy inference is retranslated from a linguistic concept to a crisp output value.

2.5 IMPORTANT FACTORS FOR PARKING SPACE DEMAND FORECAST

Generally, there are several factors that may affect the parking space demand forecasting result. Based on the scope of location whereby this project was conducted, which is Wangsa Walk Mall, the following input variables have been taken into account:

2.5.1 School type

The influence of geographical location of Wangsa Walk Mall which is located near several schools around Wangsa Maju and Setapak are one of the determining factors in parking space demand forecasting. There are 43 schools in three kilometre perimeter such as Seri Utama Private School, Kolej Tunku Abdul Rahman, SK/SMK Padang Tembak, SK/SMK Desa Tun Hussein Onn, SK Polis Depot, SK/SMK Taman Setiawangsa, Consist College, SK/SMK Setapak Indah, SK/SMK Wangsa Melawati, SK/SMK Danau Kota, SK/SMK Lembah Keramat, Institute Media Icon, CECE Institute, SK/SMK Wangsa Jaya, Malaysian Institute of Art (MIA), SK/SMK Air Panas, SK Marian Convent, SK/SMK Taman Seri Rampai, SMK Wangsa Maju Seksyen 5 and SMK Zon R1 Wangsa Maju. Parking demand will rises when the school holidays are on and drops during school normal day. Thus, it is necessary to include the type of school day; on and off as the input.

Input Variable: School Type

School Holiday	1
School Day	2

The mall is also participating in 1Malaysia Mega Sale Carnival 2011 (1MMSC) by the Tourism Malaysia. The 1Malaysia Unified Sale offers consumers discounts, incentives and added services throughout all sectors of the economy. Among the sectors that are participating in the Malaysia Unified Sale are retail, food and beverage, hotel, tourist attraction, electronic and electrical. Tourism Malaysia is also adding more excitement to the sale carnival with its Shopping Voucher programme which offers discounts and special sale promotions from various retailers nationwide. This special event has engrossed RM56.5 billion nationwide and from this amount, RM16.2 million was contributed by the shopping sector [11]. Sales and discount promotion has attracted visitors especially during festive season and shopping for the first school of day.

On weekends, the shopping mall will be crowded with people from all walks of life as it is very popular among youngsters as well as families. On weekdays, especially on Wednesday when it is the 'movie day' for TGV Cinemas, visitors will be queuing up for the theatres as all movies will be discounted for the whole day. Hence, it is expected for the increase of parking space demand on Wednesday. An article made by Mandy Barrow, the Britons has made 123 million visits to the cinema in 1998 making it the most popular cultural activity in the UK. In the first half of 2011, the number of people going out for a movie has increased by 17.9% [12].

A survey conducted by Dr Lorrin Koram, a researcher from Stanford University, it is found that one in 20 adults spend their time on Saturday to shop. According to Dr Koram, some women and men has even suffered from compulsive shopping and the disorder continued from the habits of going to shopping mall to shop every Saturday [13].

He also states that compulsive buying is nevertheless a good opportunity for the shopping sector to gain their profits, however it is crucial for the study on human's behaviour and how it can contribute to other aspect. Hence, this finding shows that these shoppers are contributing the numbers of visitors on Saturday and any other day.

Since 2001, eating out has grown in popularity as growing wealth, increasing discretionary spend and a number of social factors have all contributed to the growth in the eating out market. These social factors is particularly the increase in the proportion of working women, the growth in less-traditional family structures and the increase in single households [14]. With great varieties of dining restaurants, eating out has becoming a trend for a handful number of families, as it can be seen every day and every night in Wangsa Walk Mall.

During weekends, it is also the time for families to spend their quality time together. Often the parents are not at home, having to work a five day week from Monday to Friday. This will ultimately lead to the rise on parking demands especially on Saturday and Sunday, if not, on nights of weekdays.

Input Variable: Day Type

Public Holiday	1
Normal Day	2
Special Occasion	3

Special Occasion: Sale, Public Event, Celebrity Appearance, Company Event

2.5.3 Parking Space Demand in Wangsa Walk Mall

The total parking space demand fluctuates depending on the day of the week. Studies on the parking space demand behaviour are required as to understand why certain number is rising and others decreasing.

The table below shows the summary of the minimum and maximum of parking space demand along with the average parking space occupancy on Monday until Friday for the whole 24 weeks (refer Appendix B for details).

Table 2: Parking Space Occupancy Summary

	Minimum		Maximum		Average
	School Day	School Holiday	School Day	School Holiday	
Monday	357	544	692	702	574
Tuesday	375	556	552	593	519
Wednesday	502	574	602	593	568
Thursday	400	431	536	583	489
Friday	459	563	599	664	571
Saturday	616	660	710	709	674
Sunday	625	533	710	708	644

The highest average parking space occupancy was recorded on Saturday and the lowest was on Thursday. The example of parking space occupancy during school day and school holiday are shown in the next graphs.

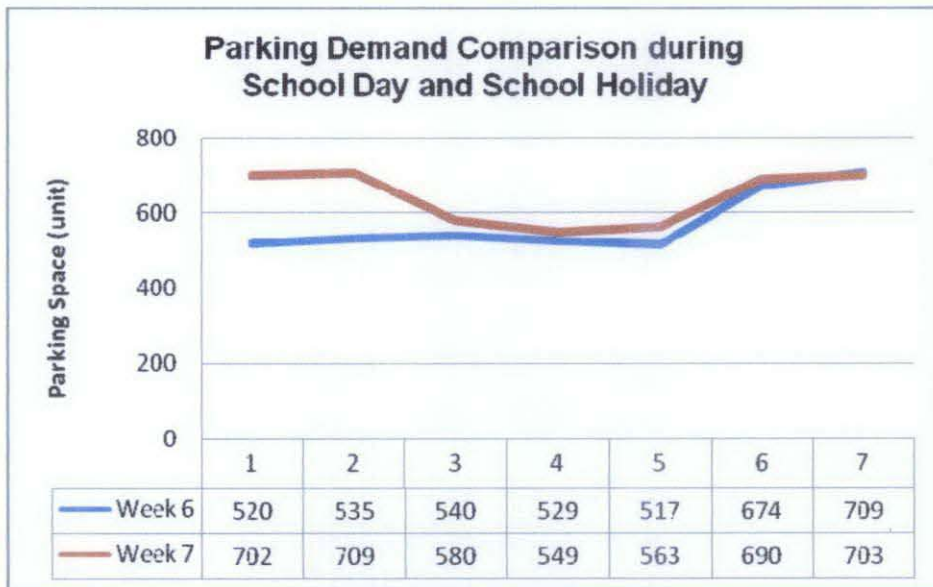


Figure 4: Parking Demand Comparison during School Day and School Holiday

The graph above shows the typical parking space occupancy on 2 consecutive weeks. The normal school day occurs on week 6 and the Chinese New Year holiday occurs in week 7, showing significance increase in the parking demand. From the same graph, it is noticed that the parking space demand on a particular weekday is not largely differs from the day before. However, the demands during weekend were rose drastically.

During week 7 on Wednesday, there are significance drop for parking demand compared to the day before, Monday and Tuesday which is the day of Chinese New Year. It is public holiday for both parents and their children; hence the parking demand for that day is very high. Then, the adults are back to work on Wednesday which ultimately the demand for parking is dropping. However, the number is quiet high compared to other normal day.

CHAPTER 3

METHODOLOGY

RESEARCH METHODOLOGY

This chapter will discuss the detail explanation of methodology that is applied in order to finish the project. This project is divided into a few steps to make sure it flows smoothly.

3.1 PROCEDURE IDENTIFICATION

In order to create the forecasting model, the historical Wangsa Walk Mall daily parking data have been gathered ranging from 3rd January 2011 until 19th June 2011 and further categorized into several types. Then, the input variables which are day type (public holiday, normal working day and special occasion) and school day type (school day and school holiday), and output variables (forecasted parking spaces) were defined. From the data obtained, the trending for parking spaces occupancy are analyzed and grouped into different range.

From the study, the membership functions were established. The membership function of a fuzzy set is a generalization of the indicator function in classical sets. In fuzzy logic, membership function represents the degree of truth as an extension of valuation [15]. The next part is the construction of the Fuzzy IF-THEN Rule Blocks for each forecast output.

During simulation, the forecast output shall be compared with the real parking spaces occupancy of that particular day in order to test the system accuracy. The flow chart in Figure 7 explains the methodology of overall modelling process using *fuzzyTECH* software.

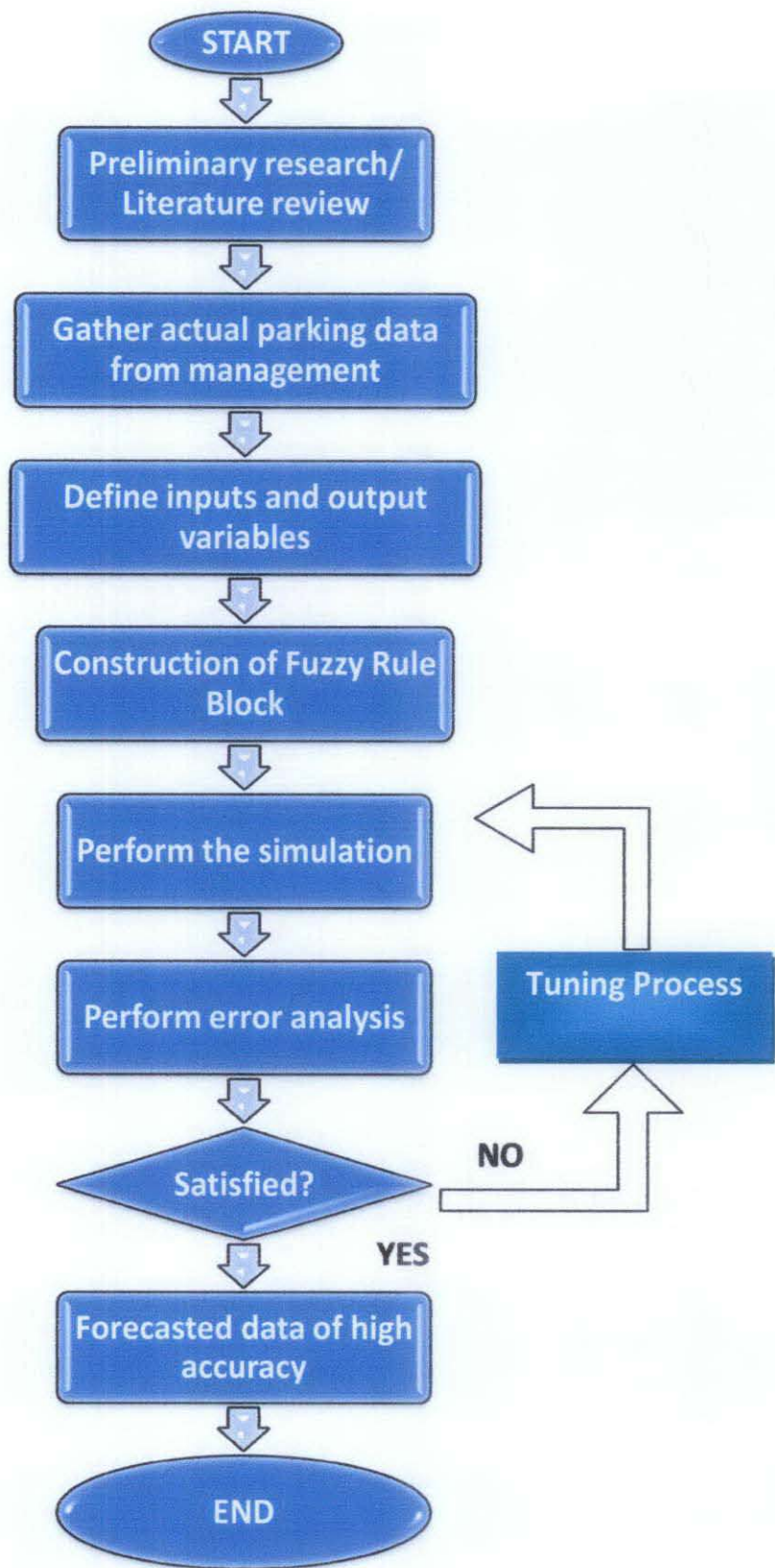


Figure 5: Flowchart for System Design Approach

3.2 SYSTEM DESIGN

The current existing model was developed by using the method shown in Figure 8. Performing error analysis will determine the system accuracy. As the overall mean absolute percentage error (MAPE) for the system was higher than 10%, the process of fine-tuning must be performed to refined the system accuracy.

3.3 ERROR CALCULATION

In order to evaluate the efficiency of the forecast, the performance criteria would be introduced and investigated. The forecast accuracy; Mean Absolute Percentage Error (MAPE) is measured based on the error statistic of forecast between the models for one week step.

The overall MAPE is calculated by taking into calculation each week's MAPE. The lower overall MAPE shows a better accuracy of the forecasting model. The MAPE is computed in terms of weekly MAPE. The MAPE calculation when school day and school holiday can be calculated as follows:

$$\% \text{ Absolute error} = \frac{|\text{Actual parking} - \text{Forecast parking}|}{\text{Actual parking}} \times 100\%$$

$$MAPE = \frac{1}{N} \sum_{i=1}^N \left| \frac{P_A^i - P_F^i}{P_A^i} \right| \times 100\%$$

Where PA is actual parking, PF is the forecasted parking and N is the number of data points.

3.4 PROCESS FINE TUNING

If the forecasted result shows high MAPE, then a process called fine tuning needed to be done. The fine tuning process is a trial and error process, that is repeated during simulation until an optimal result or an accurate model is obtained. The fine tuning involves the process of specifying and editing the IF-THEN rules as well as the Degree of Support (DoS) for the rules.

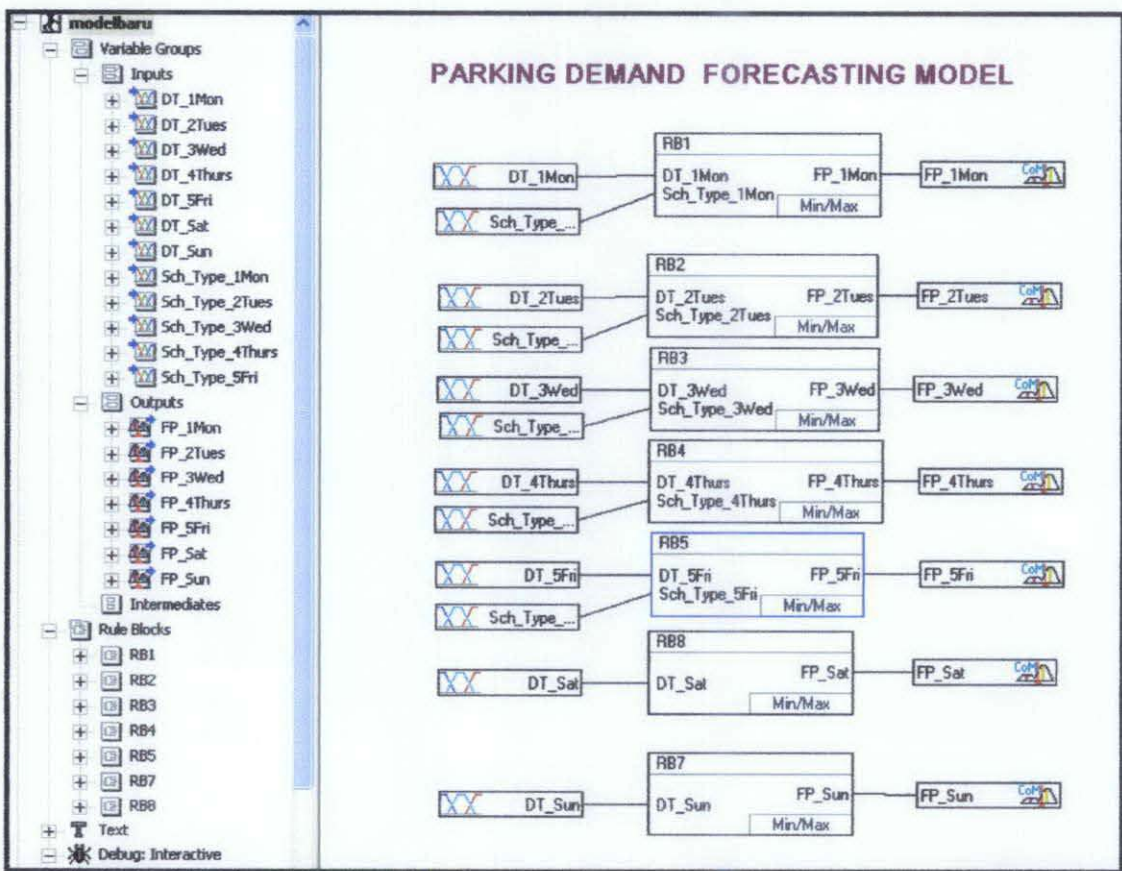


Figure 6: Forecasting Model Project Editor

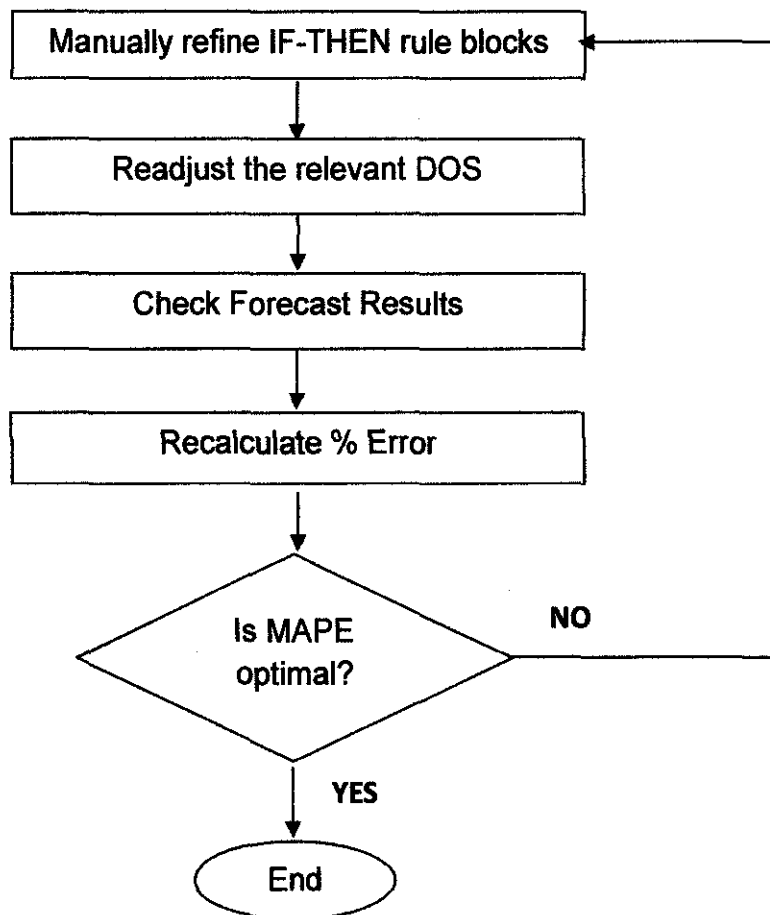


Figure 7: Process of Fine Tuning

First of all, the number of rules in each block, and the configuration the rule were maintained. The manual adjustment will be started from the first rule block, which is the rule block for forecast Monday. In this process, all 24 weeks will be run one by one. In every week that was run, to which rules (in rule block Monday) it was related to will be noted and recorded.

After all the 24 weeks data has been gathered, weeks that may be sharing the same rules are grouped. By doing this, we now know which rules are affected the particular weeks output. Then, adjustment can be made to the Degree of Support of the respective rules, so that the percentage error between forecast value and actual parking demand value on Monday is optimal minimum and balance. The same action was repeated for Tuesday, Wednesday, Thursday, Friday, Saturday and Sunday.

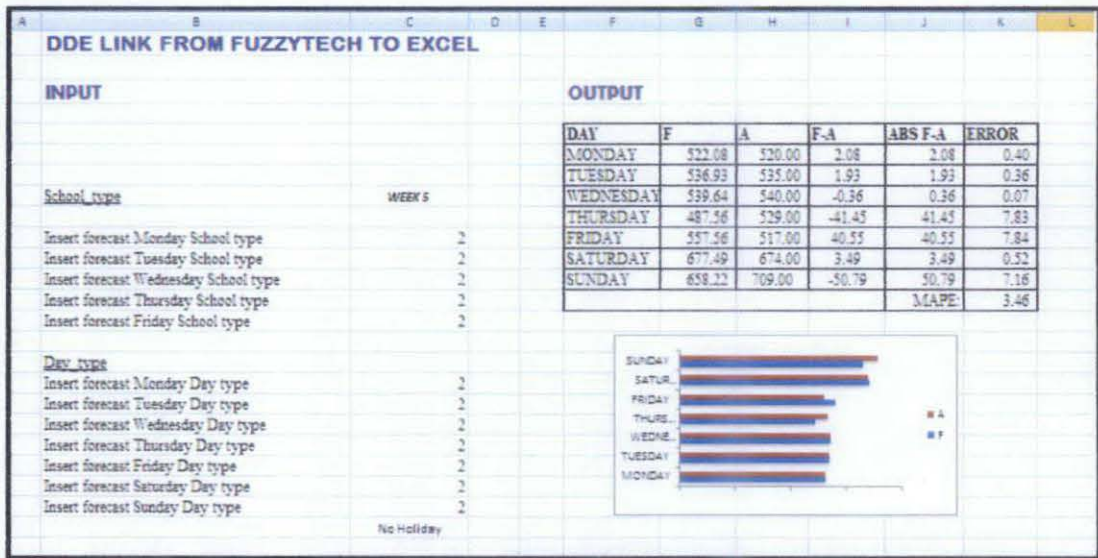


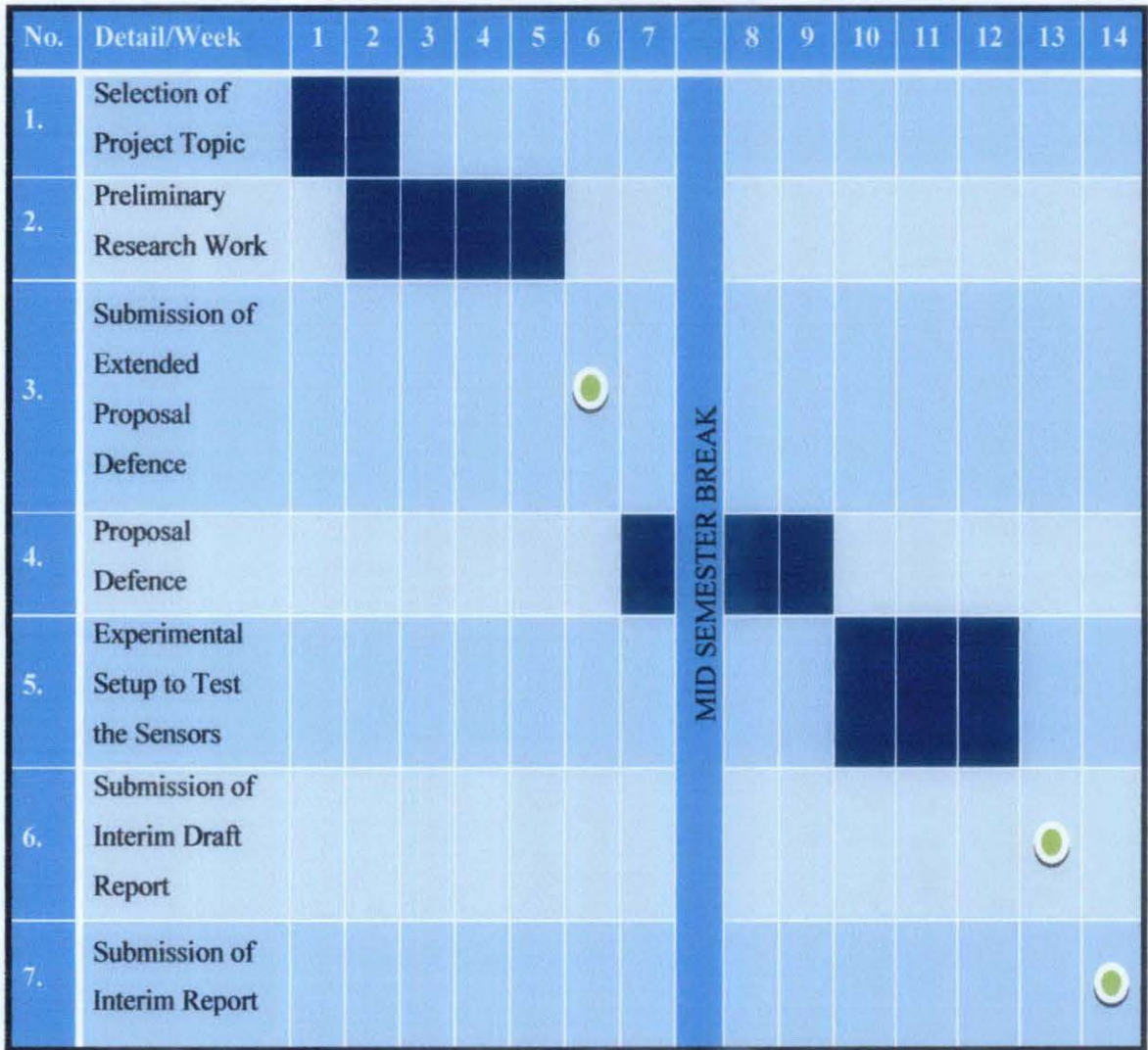
Figure 8: Input Data for Monday, Week 6


3.5 TOOLS

The tools required for this project are *fuzzyTECH 5.52* Professional Edition Software and Microsoft Excel.

3.6 GANTT CHART

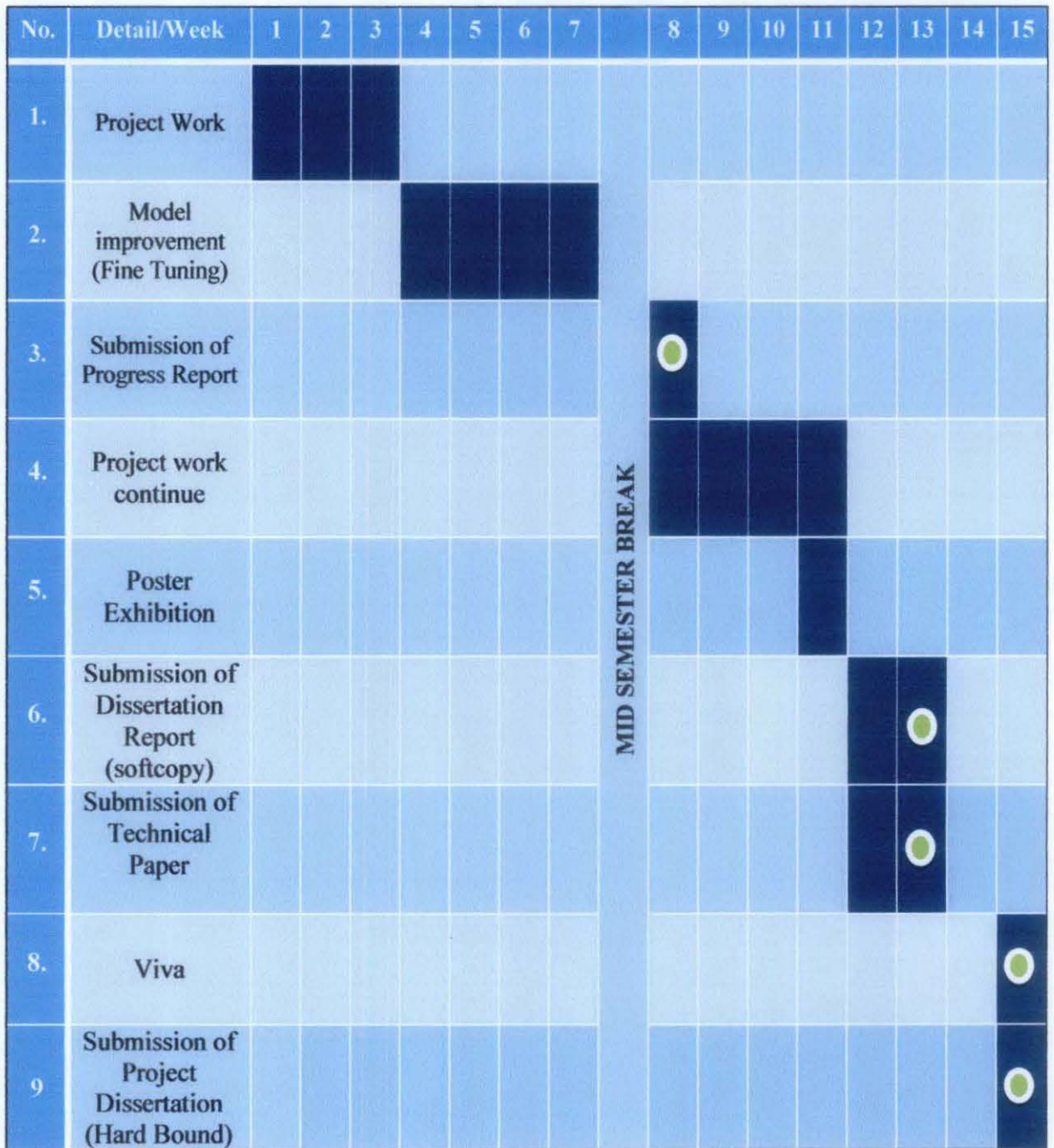
Table 3: Gantt chart for FYP 1



 Suggested Milestone

 Process

Table 4: Gantt chart for FYP 2



 Suggested Milestone
  Process

CHAPTER 4

RESULT & DISCUSSION

4.1 NEW RULE BLOCK

#	IF		THEN	
	DT_1Mon	Sch_Type_1Mon	DoS	FP_1Mon
1	public	Sch_Off	0.10	very_low
2	public	Sch_Off	0.10	low
3	public	Sch_Off	0.10	medium
4	public	Sch_Off	0.30	high
5	public	Sch_Off	0.10	very_high
6	public	Sch_On	1.00	very_low
7	public	Sch_On	1.00	low
8	public	Sch_On	1.00	medium
9	public	Sch_On	1.00	high
10	public	Sch_On	1.00	very_high
11	normal	Sch_Off	0.40	very_low
12	normal	Sch_Off	0.30	low
13	normal	Sch_Off	1.00	medium
14	normal	Sch_Off	1.00	high
15	normal	Sch_Off	1.00	very_high
16	normal	Sch_On	1.00	very_low
17	normal	Sch_On	1.00	low
18	normal	Sch_On	1.00	medium
19	normal	Sch_On	1.00	high
20	normal	Sch_On	1.00	very_high
21	special	Sch_Off	1.00	very_low
22	special	Sch_Off	1.00	low
23	special	Sch_Off	1.00	medium
24	special	Sch_Off	1.00	high
25	special	Sch_Off	1.00	very_high
26	special	Sch_On	1.00	very_low
27	special	Sch_On	1.00	low
28	special	Sch_On	1.00	medium
29	special	Sch_On	1.00	high
30	special	Sch_On	1.00	very_high

Figure 9: Rule Block Monday

#	IF		THEN	
	DT_2Tues	Sch_Type_2Tues	DoS	FP_2Tues
1	public	Sch_Off	0.10	very_low
2	public	Sch_Off	0.20	low
3	public	Sch_Off	0.10	medium
4	public	Sch_Off	1.00	high
5	public	Sch_Off	1.00	very_high
6	public	Sch_On	1.00	very_low
7	public	Sch_On	1.00	low
8	public	Sch_On	1.00	medium
9	public	Sch_On	1.00	high
10	public	Sch_On	1.00	very_high
11	normal	Sch_Off	0.10	very_low
12	normal	Sch_Off	0.20	low
13	normal	Sch_Off	0.10	medium
14	normal	Sch_Off	1.00	high
15	normal	Sch_Off	1.00	very_high
16	normal	Sch_On	1.00	very_low
17	normal	Sch_On	1.00	low
18	normal	Sch_On	1.00	medium
19	normal	Sch_On	1.00	high
20	normal	Sch_On	1.00	very_high
21	special	Sch_Off	1.00	very_low
22	special	Sch_Off	1.00	low
23	special	Sch_Off	1.00	medium
24	special	Sch_Off	1.00	high
25	special	Sch_Off	1.00	very_high
26	special	Sch_On	1.00	very_low
27	special	Sch_On	1.00	low
28	special	Sch_On	1.00	medium
29	special	Sch_On	1.00	high
30	special	Sch_On	1.00	very_high

Figure 10: Rule Block Tuesday

#	IF		THEN	
	DT_3Wed	Sch_Type_3Wed	DoS	FP_3Wed
1	public	Sch_Off	1.00	very_low
2	public	Sch_Off	1.00	low
3	public	Sch_Off	1.00	medium
4	public	Sch_Off	1.00	high
5	public	Sch_Off	1.00	very_high
6	public	Sch_On	1.00	very_low
7	public	Sch_On	1.00	low
8	public	Sch_On	1.00	medium
9	public	Sch_On	1.00	high
10	public	Sch_On	1.00	very_high
11	normal	Sch_Off	0.10	very_low
12	normal	Sch_Off	0.10	low
13	normal	Sch_Off	0.40	medium
14	normal	Sch_Off	1.00	high
15	normal	Sch_Off	1.00	very_high
16	normal	Sch_On	1.00	very_low
17	normal	Sch_On	1.00	low
18	normal	Sch_On	1.00	medium
19	normal	Sch_On	1.00	high
20	normal	Sch_On	1.00	very_high
21	special	Sch_Off	1.00	very_low
22	special	Sch_Off	1.00	low
23	special	Sch_Off	1.00	medium
24	special	Sch_Off	1.00	high
25	special	Sch_Off	1.00	very_high
26	special	Sch_On	1.00	very_low
27	special	Sch_On	1.00	low
28	special	Sch_On	1.00	medium
29	special	Sch_On	1.00	high
30	special	Sch_On	1.00	very_high

Figure 11: Rule Block Wednesday

#	IF		THEN	
	DT_4Thurs	Sch_Type_4Thurs	DoS	FP_4Thurs
1	public	Sch_off	1.00	very_low
2	public	Sch_off	1.00	low
3	public	Sch_off	1.00	medium
4	public	Sch_off	1.00	high
5	public	Sch_off	1.00	very_high
6	public	Sch_On	1.00	very_low
7	public	Sch_On	1.00	low
8	public	Sch_On	1.00	medium
9	public	Sch_On	1.00	high
10	public	Sch_On	1.00	very_high
11	normal	Sch_off	1.00	very_low
12	normal	Sch_off	0.60	low
13	normal	Sch_off	0.10	medium
14	normal	Sch_off	0.10	high
15	normal	Sch_off	0.10	very_high
16	normal	Sch_On	1.00	very_low
17	normal	Sch_On	1.00	low
18	normal	Sch_On	1.00	medium
19	normal	Sch_On	1.00	high
20	normal	Sch_On	1.00	very_high
21	special	Sch_off	1.00	very_low
22	special	Sch_off	1.00	low
23	special	Sch_off	0.70	medium
24	special	Sch_off	0.10	high
25	special	Sch_off	0.20	very_high
26	special	Sch_On	1.00	very_low
27	special	Sch_On	1.00	low
28	special	Sch_On	1.00	medium
29	special	Sch_On	1.00	high
30	special	Sch_On	1.00	very_high

Figure 12: Rule Block Thursday

#	IF		THEN	
	DT_5Fri	Sch_Type_5Fri	DoS	FP_5Fri
1	public	Sch_Off	1.00	very_low
2	public	Sch_Off	1.00	low
3	public	Sch_Off	1.00	medium
4	public	Sch_Off	1.00	high
5	public	Sch_Off	1.00	very_high
6	public	Sch_On	1.00	very_low
7	public	Sch_On	1.00	low
8	public	Sch_On	1.00	medium
9	public	Sch_On	1.00	high
10	public	Sch_On	1.00	very_high
11	normal	Sch_Off	1.00	very_low
12	normal	Sch_Off	1.00	low
13	normal	Sch_Off	1.00	medium
14	normal	Sch_Off	1.00	high
15	normal	Sch_Off	1.00	very_high
16	normal	Sch_On	1.00	very_low
17	normal	Sch_On	1.00	low
18	normal	Sch_On	1.00	medium
19	normal	Sch_On	1.00	high
20	normal	Sch_On	1.00	very_high
21	special	Sch_Off	1.00	very_low
22	special	Sch_Off	1.00	low
23	special	Sch_Off	1.00	medium
24	special	Sch_Off	1.00	high
25	special	Sch_Off	1.00	very_high
26	special	Sch_On	1.00	very_low
27	special	Sch_On	1.00	low
28	special	Sch_On	1.00	medium
29	special	Sch_On	1.00	high
30	special	Sch_On	1.00	very_high

Figure 13: Rule Block Friday

#	IF	THEN	
	DT_Sat	DoS	FP_Sat
1	public	1.00	very_low
2	public	1.00	low
3	public	1.00	medium
4	public	1.00	high
5	public	1.00	very_high
6	normal	0.10	very_low
7	normal	0.10	low
8	normal	0.40	medium
9	normal	0.80	high
10	normal	1.00	very_high
11	special	1.00	very_low
12	special	1.00	low
13	special	1.00	medium
14	special	1.00	high
15	special	1.00	very_high

Figure 14: Rule Block Saturday

#	IF	THEN	
	DT_Sun	DoS	FP_Sun
1	public	1.00	very_low
2	public	1.00	low
3	public	1.00	medium
4	public	1.00	high
5	public	1.00	very_high
6	normal	0.10	very_low
7	normal	0.10	low
8	normal	0.10	medium
9	normal	1.00	high
10	normal	1.00	very_high
11	special	1.00	very_low
12	special	1.00	low
13	special	1.00	medium
14	special	1.00	high
15	special	1.00	very_high
16			

Figure 15: Rule Block Sunday

4.2 SIMULATION RESULT

The process of transferring data from fuzzyTECH to an excel file are called Dynamic Data Exchange (DDE). fuzzyTECH will run and the forecasted output result will be presented in the excel file. All calculations are performed in the table below:

Table 5: MAPE Calculation in Excel for Week 6

DAY	F	A	F-A	ABS F-A	ERROR
MONDAY	522.08	520.00	2.08	2.08	0.40
TUESDAY	536.93	535.00	1.93	1.93	0.36
WEDNESDAY	539.64	540.00	-0.36	0.36	0.07
THURSDAY	487.56	529.00	-41.45	41.45	7.83
FRIDAY	557.56	517.00	40.55	40.55	7.84
SATURDAY	677.49	674.00	3.49	3.49	0.52
SUNDAY	658.22	709.00	-50.79	50.79	7.16
				MAPE:	3.46

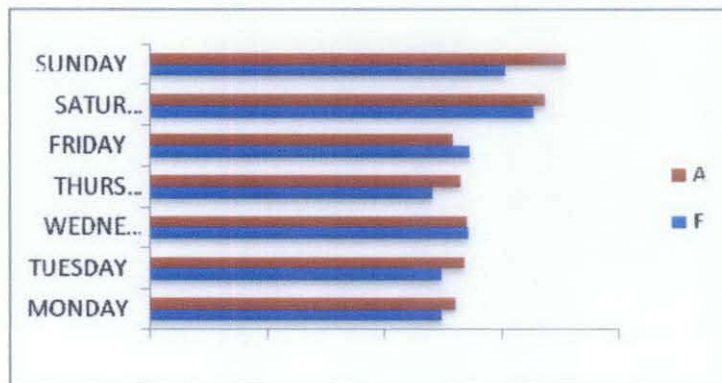


Figure 16: Graph Actual Vs Forecast for Week 6

MAPE values of each school type for the forecast model are shown in Table 6,7 and 8 below:

Table 6: MAPE during School Day

Week	MAPE Value (%)
1	7.50
2	11.49
3	12.41
4	8.43
5	5.61
6	3.46
8	6.33
9	5.88
10	5.73
12	11.82
13	8.36
14	9.47
15	7.32
16	11.56
17	7.88
18	10.89
19	9.85
20	14.84
21	13.71
24	12.43
Total MAPE	9.2535

Table 7: MAPE during School Holiday

Week	MAPE Value (%)
7	7.74
11	6.46
22	6.53
23	8.66
Total MAPE	7.3475

Table 8: Overall MAPE Percentage

Week	MAPE Value (%)
1	7.50
2	11.49
3	12.41
4	8.43
5	5.61
6	3.46
7	7.74
8	6.33
9	5.88
10	5.73
11	6.46
12	11.82
13	8.36
14	9.47
15	7.32
16	11.56
17	7.88
18	10.89
19	9.85
20	14.84
21	13.71
22	6.53
23	8.66
24	12.43
Total MAPE	8.93

The total parking space demand fluctuates depending on the day of the week. Studies on the parking space demand behaviour and its special pattern are required as to understand why certain number is rising and others decreasing. It is noticed that the parking space demand on a particular weekday is not largely differs from the day before. However, the demands during weekend rose drastically.

The graphs showed the forecasted parking versus actual parking data during school day and school holiday in the Figure 17 and 18 below:

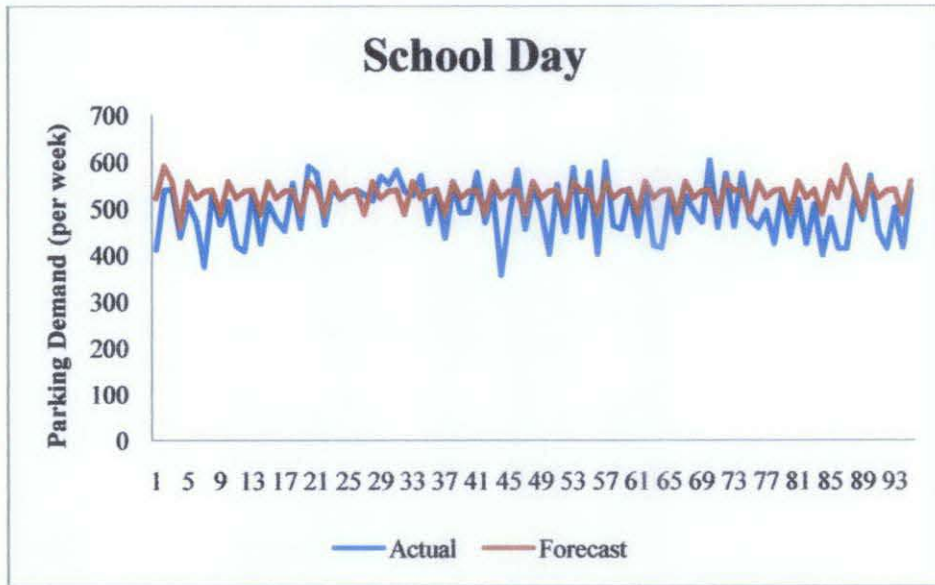


Figure 17: Graph of Actual Vs Forecasted Parking Data (School Day)

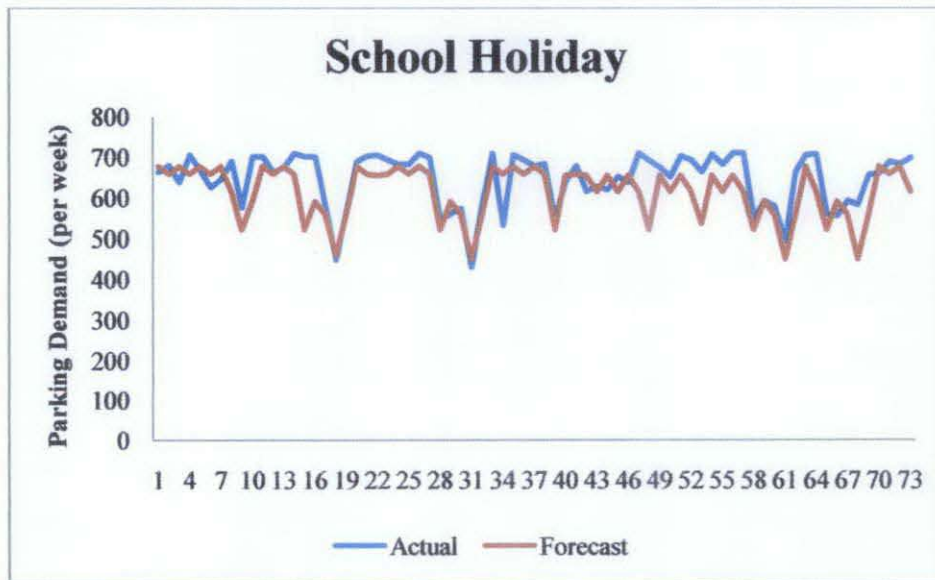


Figure 18: Graph of Actual Vs Forecasted Parking Data (School Holiday)

The graphs showed the fluctuations of actual parking data during weekdays and weekends in the Figure 19, 20 and 21 below:

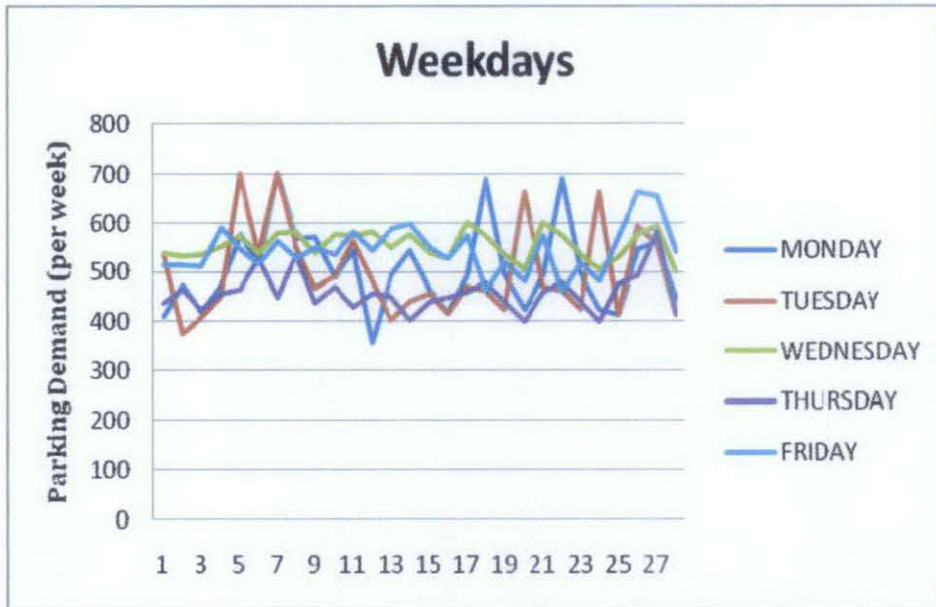


Figure 19: Graph of Actual Parking Data (Weekdays)

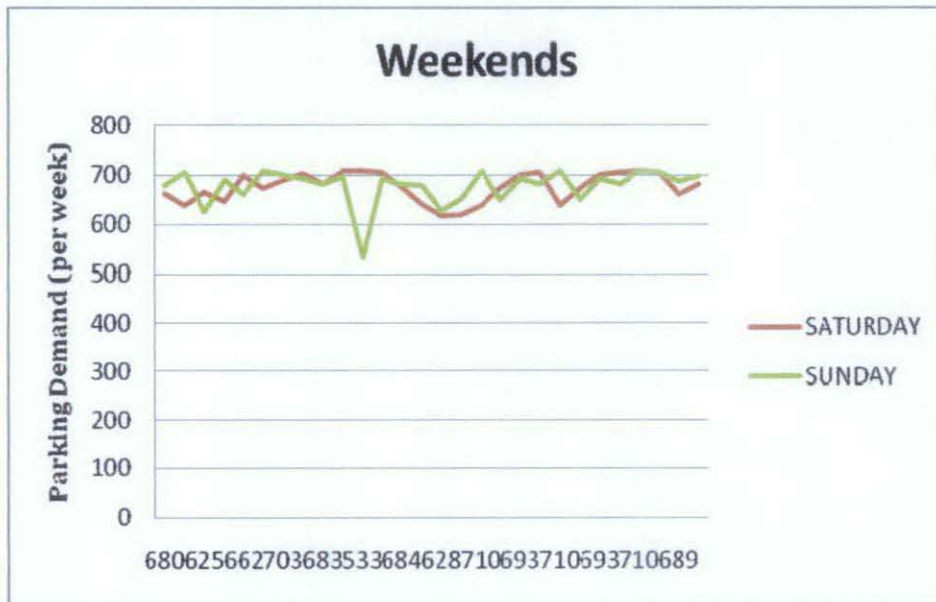


Figure 20: Graph of Actual Parking Data (Weekends)

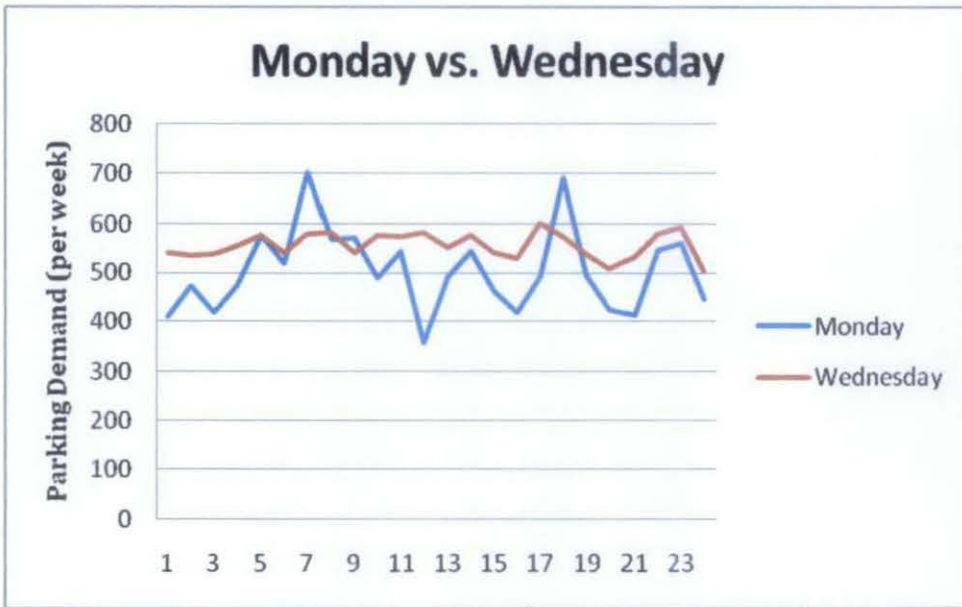


Figure 21: Graph of Actual Parking Data (Monday Vs Wednesday)

4.3 DISCUSSION

Simulations have been carried on for every day of the week on the same forecast models to determine the forecasted parking spaces. The overall MAPE is calculated by taking into calculation each week's MAPE. The overall MAPE result is 8.93, where MAPE during school day is 9.2535 and MAPE during school holidays is 7.3475. The forecasting methodology with fuzzy parameters gives acceptable short-range parking demand forecasting results and they are below the range of 10% MAPE.

From observation, by changing the degree of support for the rules, there are two possible effects. First, either all the week related to the rules are experiencing MAPE decrement or some rules are facing MAPE decrement and some are increased. Thus, it is very crucial to balance the DoS adjustment so as to balance and keep minimum all the MAPE.

From this project, we can conclude that the influence of geographical location of Wangsa Walk Mall is the determining factor in parking space demand forecasting. As

stated before, there are almost 43 schools in three kilometre perimeter such as Seri Utama Private School, Kolej Tunku Abdul Rahman, SK/SMK Padang Tembak, SK/SMK Desa Tun Hussein Onn, SK Polis Depot, SK/SMK Taman Setiawangsa, Consist College, SK/SMK Setapak Indah, SK/SMK Wangsa Melawati, SK/SMK Danau Kota, SK/SMK Lembah Keramat, Institute Media Icon, CECE Institute, SK/SMK Wangsa Jaya, Malaysian Institute of Art (MIA), SK/SMK Air Panas, SK Marian Convent, SK/SMK Taman Seri Rampai, SMK Wangsa Maju Seksyen 5 and SMK Zon R1 Wangsa Maju.

The map in Figure 22 shows the location of the Wangsa Walk Mall with the surrounding schools in three kilometre perimeter below:



Figure 22: Wangsa Walk Mall Location

Another contributing factor would be the day type which is public holiday, normal working day and special event that are conducted by the shopping mall. We can see that during public holiday, the normal parking occupancy rose drastically, as drastic as during special occasion.

Special events that are conducted at Wangsa Walk Mall for the past six months between 3 January 2011 and 19 June 2011 are Superstar Karaoke Season 2 @WangsaWalk, Crime Prevention Awareness Day, Fun Chess Move @WM, 1Malaysia Mega Sales and Chinese New Year Sales. These events attracted thousands of visitors around the Federal Territory of Kuala Lumpur where parking demand will rise drastically which up to the point the parking spaces could not occupy all the expected visitors.

In addition, on Monday the parking demand is low compared to other days of the week as less people came due to it is the first day of the week. During public holiday, demand rose drastically on any day of the week. As example on Monday, there are Thaipusam, KL Day and Labour Day on week 5, 6 and 18 respectively. On Wednesday which is TGV Cinemas Movie Day, parking demand is steadily high throughout 24 weeks as every Wednesday the movie is sold out.

Results of the forecasting show that the number of parking spaces is inadequate. Maximum occupancy for Wangsa Walk Mall's parking spaces during a period of time is 670. However, forecasting result shown over 700 spaces is needed. The excess number is contributed by the cars that are going around Wangsa Walk Mall in search of parking space even though it is already full. Wangsa Walk Mall should have future plans to add more parking to cater more visitors.

CHAPTER 5

CONCLUSION AND RECOMMENDATION

5.1 CONCLUSION

Short-range parking demand forecasting was performed for the month of January until June in a data set of 24 weeks. The forecasting methodology with fuzzy parameters gives acceptable short-range parking demand forecasting results and they are below the range of 10% MAPE. The modification on DoS and IF-THEN rules on each rule blocks show significant decrement in the MAPE.

5.2 RECOMMENDATION

For the next semester, it is recommended to expand the membership function and to increase the number of rules. If the numbers of rules are increased, this may improve the system model accuracy. It is recommended to add the rule of population growth rate of Wangsa Maju residential area as the input variable to the forecasting model.

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APPENDIX

APPENDIX A

WANGSA WALK PARKING DEMAND FOR YEAR 2011 (3Jan2011-19June2011)

Week No.	1	2	3	4
MONDAY	412	473	420	475
TUESDAY	539	375	408	452
WEDNESDAY	541	534	537	554
THURSDAY	438	466	425	458
FRIDAY	514	515	512	590
SATURDAY	663	638	667	646
SUNDAY	680	706	625	690
Average Weekday Parking	488.8	472.6	460.4	505.8
Comment(s)				Thaipusam

Week No.	5	6	7	8
MONDAY	576	520	702	569
TUESDAY	701	535	701	552
WEDNESDAY	576	540	580	582
THURSDAY	465	529	449	536
FRIDAY	546	517	563	528
SATURDAY	701	674	690	705
SUNDAY	662	709	703	693
Average Weekday Parking	573.4	528.2	600.6	553.4
Comment(s)	Thaipusam	Chinese New Year		Maulidur
	KL Day	Chap Goh Mei		Rasul

Week No.	9	10	11	12
MONDAY	571	491	544	357
TUESDAY	468	492	562	484
WEDNESDAY	541	577	574	582
THURSDAY	437	470	431	456
FRIDAY	548	536	582	545
SATURDAY	682	710	709	706
SUNDAY	683	699	533	693
Average Weekday Parking	513	513.2	538.6	484.8
Comment(s)			School Holiday	

Week No.	13	14	15	16
MONDAY	493	544	463	420
TUESDAY	402	439	456	416
WEDNESDAY	551	577	541	530
THURSDAY	451	402	441	449
FRIDAY	587	599	549	527
SATURDAY	678	641	616	620
SUNDAY	684	679	628	652
Average Weekday Parking	496.8	512.2	490	468.4
Comment(s)		Hindu New Year Crime Prevention Awareness Day	Fun Moves Chess @WM	Fun Moves Chess @WM Easter Day

Week No.	17	18	19	20
MONDAY	493	692	495	424
TUESDAY	471	462	424	664
WEDNESDAY	602	574	537	508
THURSDAY	460	478	440	400
FRIDAY	574	459	519	480
SATURDAY	638	675	703	708
SUNDAY	710	651	693	682
Average Weekday Parking	520	533	483	495.2
Comment(s)	Superstar Karaoke Labour Day	Labour Day Superstar Karaoke	Superstar Karaoke	Wesak Day Superstar Karaoke

Week No.	21	22	23	24
MONDAY	414	548	560	448
TUESDAY	414	592	556	414
WEDNESDAY	532	579	593	502
THURSDAY	476	494	583	416
FRIDAY	570	664	657	540
SATURDAY	710	706	660	683
SUNDAY	710	708	689	699
Average Weekday Parking	481.2	575.4	589.8	464
Comment(s)	Superstar Karaoke (Finale)	School Holiday	School Holiday	Father's Day

APPENDIX B
PARKING DEMAND SUMMARY

MONDAY	TUESDAY	WEDNESDAY	THURSDAY	FRIDAY	SATURDAY	SUNDAY
412	539	541	438	514	663	680
473	375	534	466	515	638	706
420	408	537	425	512	667	625
475	452	554	458	590	646	690
576	701	576	465	546	701	662
520	535	540	529	517	674	709
702	701	580	449	563	690	703
569	552	582	536	528	705	693
571	468	541	437	548	682	683
491	492	577	470	536	710	699
544	562	574	431	582	709	533
357	484	582	456	545	706	693
493	402	551	451	587	678	684
544	439	577	402	599	641	679
463	456	541	441	549	616	628
420	416	530	449	527	620	652
493	471	602	460	574	638	710
692	462	574	478	459	675	651
495	424	537	440	519	703	693
424	664	508	400	480	708	682

414	414	532	476	570	710	710
548	592	579	494	664	706	708
560	556	593	583	657	660	689
448	414	502	416	540	683	699
504.3	499.6	556	460.4	550.9	676.2	677.5
MAX 702	701	602	583	664	710	710
MIN 357	375	502	400	459	616	533
SUM 12104	11990	13344	11050	13221	16229	16261

APPENDIX C
MEMBERSHIP FUNCTIONS

Input Variables

Day Type

Public Holiday	1
Normal Day	2
Special Occasion	3

Special Occasion: Sale, Public Event, Celebrity Appearance, Company Event

School Type

School Holiday	1
School Day	2

Output Variables

Forecasted Parking Demand: Monday

Very Low	300-437 units
Low	437-471 units
Medium	471-573 units
High	573-642 units
Very High	642-702 units

Forecasted Parking Demand: Tuesday

Very Low	300-437 units
Low	437-471 units
Medium	471-573 units
High	573-642 units
Very High	642-702 units

Forecasted Parking Demand: Wednesday

Very Low	440-510 units
Low	510-545 units
Medium	545-580 units
High	580-615 units
Very High	615-650 units

Forecasted Parking Demand: Thursday

Very Low	350-410 units
Low	410-450 units
Medium	450-530 units
High	530-575 units
Very High	575-620 units

Forecasted Parking Demand: Friday

Very Low	400-500 units
Low	500-535 units
Medium	535-600 units
High	600-650 units
Very High	650-700 units

Forecasted Parking Demand: Saturday

Very Low	600-625 units
Low	625-640 units
Medium	640-670 units
High	670-693 units
Very High	693-710 units

Forecasted Parking Demand: Sunday

Very Low	500-570 units
Low	570-600 units
Medium	600-640 units
High	640-675 units
Very High	675-710 units

APPENDIX D
WANGSA WALK MALL FIGURES



Figure A: Wangsa Walk Mall Parking Space (North)



Figure B: Wangsa Walk Mall Parking Space (South)



Figure C: Wangsa Walk Mall Parking Payment Counter (Front)



Figure D: Wangsa Walk Mall Parking Management Department (Back)