

A NEW HYBRID MODEL OF DENGUE INCIDENCE RATE USING NEGATIVE
BINOMIAL GENERALISED ADDITIVE MODEL AND FUZZY C-MEANS
MODEL: A CASE STUDY IN SELANGOR

NAZEERA BINTI MOHAMAD

A thesis submitted in
fulfilment of the requirement for the award of the
Degree of Master of Science



Faculty of Applied Sciences and Technology
Universiti Tun Hussein Onn Malaysia

AUGUST 2018

DEDICATION

To my beloved parents

Encik Mohamad bin Chokro and Puan Nur Sa'adah binti Md. Yatim

and

To my beloved siblings

Naily, Nadzreen and Nuratiqah

This humble work is a sign of my love to you!

Lots of love,
Nazeera Mohamad



PTTA UTHM
PERPUSTAKAAN TUNKU TUN AMINAH

ACKNOWLEDGEMENT

“In the name of Allah, Most Gracious, Most Merciful”

Firstly, I have been extremely fortunate to be supervised by an excellent and dedicated supervisor and co-supervisor; Dr Norziha binti Che Him and Dr Mohd. Saifullah bin Rusiman. I am really grateful to both of them for their sound advice, encouragement and enthusiasm in my work throughout completing this study. Besides, I would like to express my greatest gratitude and sincere appreciation for their guidance, ideas, patience and support that has been influential in the process of researching and writing this study.

Besides, I've received excellent support and guidance from my friends and my greatest gratitude to my dear parent and siblings for their infinite support. This study has been greatly strengthened by the guidance of these individuals, as well as other individuals who are not mentioned here. Thank you for all the support and may Allah bless all of you.

Finally, I would like to thank all of the lecturers in Universiti Tun Hussein Onn Malaysia (UTHM) especially from the Department of Mathematics and Statistics, Faculty of Applied Sciences and Technology (FAST) for their support and help throughout completing this study. Besides, I would like to thank UTHM for providing the fund which allowed me to undertake this study through Geran Penyelidikan Pascasiswazah (GPPS). I am furthermore appreciative to Department of Irrigation and Drainage Malaysia (DIDM) and Department of Statistics Malaysia (DoSM) for their cooperation in supplying data which used throughout this study. I am also grateful to the funding received through the Ministry of Higher Education (MoHE) under MyBrain15 Program to support my postgraduate studies. Thank you for making it possible for me to complete this study.

ABSTRACT

Dengue is one of the top reason for illness and mortality in the world with beyond one-third of the world's population living in the risk areas of dengue infection. In this study, there are five stages to achieve the research objectives. Firstly, the verification of predetermined variables. Secondly, the identification of new datasets after clustered by district and Fuzzy C-Means Model (FCM). Thirdly, the development of models using the existing dataset and the new datasets which clustered by the two different clustering categories. Then, to assess the models developed by using three measurement methods which are deviance (D), Akaike Information Criteria (AIC) and Bayesian Information Criteria (BIC). Lastly, the validation of model developed by comparing the value of D , AIC and BIC between the existing model and the new models developed which used the new datasets. There are two different clustering techniques applied which are clustering the data by district and by FCM. This study proposed a new modelling hybrid framework by using two statistical models which are FCM and negative binomial Generalised Additive Model (GAM). This study successfully presents the significant difference in the climatic and non-climatic factors that influenced dengue incidence rate (DIR) in Selangor, Malaysia. Results show that the climatic factors such as rainfall with current month up to 3 months and number of rainy days with current month up to lag 3 months are significant to DIR. Besides, the interaction between rainfall and number of rainy days also shows strong positive relationship to DIR. Meanwhile, non-climatic variables such as population density, number of locality and lag DIR from 1 month until 3 months also show significant relationship towards DIR. For both clustering techniques, there are two clusters formed and there are four new models developed in this study. After comparing the values of D , AIC and BIC between the existing model and the new models, this study concluded that four new models recorded lower values compared to the existing model. Therefore, the four new models are selected to present the dengue incidence in Selangor.

ABSTRAK

Denggi adalah salah satu sebab utama penyakit dan kematian di dunia dengan lebih satu pertiga daripada populasi dunia yang hidup di kawasan risiko jangkitan denggi. Dalam kajian ini, terdapat lima peringkat untuk mencapai matlamat penyelidikan. Pertama, pengesahan pembolehubah yang telah ditetapkan. Kedua, pengenalpastian dataset baru selepas menggunakan teknik pengelompokan mengikut daerah dan *Fuzzy C-Means Model* (FCM). Ketiga, pembinaan model menggunakan dataset asal dan dataset baru yang dikelompokkan melalui dua kategori teknik pengelompokan yang berbeza. Kemudian, penilaian model yang dibina dengan menggunakan tiga kaedah pengukuran iaitu nilai penyimpangan terkecil (D), *Akaike Information Criteria* (AIC) dan *Bayesian Information Criteria* (BIC). Akhir sekali, pengesahan model yang dibina dengan membandingkan nilai D , AIC dan BIC antara model asal dan model baru yang dibina menggunakan dataset yang baru. Terdapat dua teknik pengelompokan yang berlainan iaitu mengikut daerah dan FCM. Kajian ini mencadangkan rangka kerja baru melalui pemodelan hibrid dengan dua model statistik iaitu FCM dan negatif binomial *Generalised Additive Model* (GAM). Kajian ini berjaya membuktikan faktor iklim dan bukan iklim yang signifikan mempengaruhi kadar kejadian denggi (DIR) di Selangor, Malaysia. Hasil daripada analisis untuk daerah dan FCM mendapati faktor iklim seperti hujan dengan bulan semasa sehingga 3 bulan sebelumnya dan jumlah hari hujan dengan bulan semasa sehingga 3 bulan seterusnya adalah penting kepada DIR. Selain itu, interaksi antara hujan dengan bilangan hari hujan juga menunjukkan hubungan positif yang kuat dengan DIR. Pembolehubah bukan iklim seperti kepadatan penduduk, bilangan lokaliti dan lag DIR dari 1 bulan hingga 3 bulan juga menunjukkan hubungan yang signifikan terhadap DIR. Kedua-dua teknik pengelompokan membentuk dua kelompok dan terdapat empat model baru yang dibina dalam kajian ini. Setelah membandingkan nilai D , AIC dan BIC antara model asal dan model-model baru, kajian ini mendapati bahawa empat model baru mencatatkan nilai yang lebih rendah

berbanding model asal. Oleh itu, empat model baru tersebut dipilih untuk mewakili kejadian denggi di Selangor.



TABLE OF CONTENTS

DECLARATION	ii
DEDICATION	iii
ACKNOWLEDGEMENT	iv
ABSTRACT	v
TABLE OF CONTENTS	viii
LIST OF TABLES	xi
LIST OF FIGURES	xiii
LIST OF SYMBOLS AND ABBREVIATIONS	xvii
CHAPTER 1 INTRODUCTION	1
1.1 Motivations	1
1.2 Problem statement	4
1.3 Research objectives	4
1.4 Significance of study	5
1.5 Scope of study	6
1.6 Summary	6
CHAPTER 2 LITERATURE REVIEW	8
2.1 Introduction to Malaysia	8
2.2 Climate in Malaysia	9
2.3 Dengue fever and its transmission	10
2.4 Dengue fever worldwide	13
2.5 Dengue fever in Malaysia	16
2.5.1 Dengue distribution in Selangor	17
2.6 Surveillance and control	19
2.7 Climate and dengue	20
2.8 Non-climate and dengue	23
2.9 Application of Generalised Additive Model in modelling dengue cases	26

2.10 Application of Fuzzy C-Means Model in in modelling disease cases	27
2.11 Application of statistical methods in modelling dengue cases	28
2.12 Generalised Linear Model development	30
2.12.1 Poisson model	32
2.12.2 Negative Binomial model	34
2.13 Overdispersion	35
2.14 Goodness-of-fit	36
2.15 Generalised Additive Model	36
2.16 Fuzzy C-Means Model	37
2.17 Summary	40
CHAPTER 3 METHODOLOGY	41
3.1 Description of collated datasets	41
3.2 Dengue Incidence Rate (DIR)	43
3.3 Research procedure	43
3.4 Summary	48
CHAPTER 4 DATA ANALYSIS	49
4.1 Dengue incidence pattern	49
4.1.1 Annual dengue incidence pattern	51
4.1.2 Monthly dengue incidence pattern	52
4.2 Verification of predetermined variables using original dataset	53
4.2.1 Examine the relationship between variables	54
4.3 Identification of new datasets	59
4.3.1 Data clustering by district	59
4.3.2 Data clustering by Fuzzy C-Means Model (FCM)	66
4.4 Verification of predetermined variables using new datasets	68
4.4.1 Examining the relationship between variables using new dataset clustered by district	69



4.4.2 Examining the relationship between variables using new dataset clustered by FCM	77
4.5 Summary	85
CHAPTER 5 MODELLING FRAMEWORK	86
5.1 Model development using original dataset	86
5.2 Model development using dataset clustered by district	89
5.3 Model development using dataset clustered by FCM	95
5.4 Model comparison	100
5.5 Summary	103
CHAPTER 6 SUMMARY AND CONCLUSIONS	104
6.1 Contributions	104
6.2 Limitations of study	105
6.3 Recommendations	105
REFERENCES	107
VITA	



LIST OF TABLES

2.1	Summary of the literature review on climatic factors that influenced dengue incidence worldwide	22
2.2	Summary of the literature review on non-climatic factors that influenced dengue incidence worldwide	25
2.3	Summary of the literature review on the application of statistical methods in modelling diseases rate	31
3.1	Source and original resolution of variables	42
3.2	Summary of variables description	42
4.1	Total number of dengue cases and average DIR in Selangor from 2010 - 2015	49
4.2	Average monthly DIR for 6 districts in Cluster 1	61
4.3	Average monthly DIR for 3 districts in Cluster 2	61
4.4	The division of the districts in Selangor by clustering	62
4.5	Total number of dengue cases and average DIR in each cluster from January 2010 to August 2015 in Selangor	65
4.6	District with the highest annual DIR from January 2010 to August 2015	65
4.7	The value of c and F values for DIR	66
4.8	Descriptive values of DIR in cluster 1 and cluster 2	67
5.1	Comparison values of Deviance (D), Akaike Information Criterion (AIC) and Bayesian Information Criterion (BIC) between a Poisson GLM and a negative binomial GLM for original dataset	87
5.2	Coefficient values of negative binomial GLM for original data	87

5.3	Coefficient values of Model A using original data	88
5.4	Comparison values of Deviance (D), Akaike Information Criterion (AIC) and Bayesian Information Criterion (BIC) between a Poisson GLM and a negative binomial GLM for data clustered by district	90
5.5	Coefficient values of negative binomial GLM for Cluster 1 by district	90
5.6	Coefficient values of negative binomial GLM for Cluster 2 by district	91
5.7	Coefficient values of Model B for Cluster 1 by district	93
5.8	Coefficient values of Model C for Cluster 2 by district	94
5.9	Comparison values of Deviance (D), Akaike Information Criterion (AIC) and Bayesian Information Criterion (BIC) between a Poisson GLM and a negative binomial GLM for data clustered by FCM	95
5.10	Coefficient values of negative binomial GLM for Cluster 1 by FCM	96
5.11	Coefficient values of negative binomial GLM for Cluster 2 by FCM	97
5.12	Coefficient values of Model D for Cluster 1 by FCM	98
5.13	Coefficient values of Model E for Cluster 2 by FCM	100
5.14	Comparison values of Deviance (D), Akaike Information Criterion (AIC) and Bayesian Information Criterion (BIC) by using a negative binomial GAM clustering method	101
5.15	The best potential models that explained the dengue incidence rate in Selangor from January 2010 – August 2015	107



LIST OF FIGURES

1.1	Dengue cases reported in Malaysia 1995 – 2015	2
2.1	Location of Malaysia	9
2.2	Dengue transmission cycle	11
2.3	Dengue fever risk map 2015	13
2.4	Actual number of dengue cases in Selangor	18
3.1	Flow chart of research framework	47
4.1	Population estimation by district in Selangor from 2010 - 2015	50
4.2	Annual DIR per 100,000 population for 9 districts of Selangor from January 2010 to August 2015	51
4.3	Average monthly DIR per 100,000 population of Selangor from January 2010 to August 2015	53
4.4	Relationship between log monthly DIR and population density	54
4.5	Relationship between log monthly DIR and number of locality	55
4.6	Relationship between log monthly DIR and (a) log monthly DIR lag 1 month, (b) log monthly DIR lag 2 months, (c) log monthly DIR lag 3 months	56
4.7	Relationship between log monthly DIR and (a) rainfall current, (b) rainfall lag 1 month, (c) rainfall lag 2 months, (d) rainfall lag 3 months	57
4.8	Relationship between log monthly DIR and (a) number of rainy days current month, (b) number of rainy days lag 1 month, (c) number of rainy days lag 2 months, (d) number of rainy days lag 3 months	58

4.9	Relationship between rainfall and number of rainy days	59
4.10	Mean monthly DIR by district for Cluster 1 in Selangor	63
4.11	Mean monthly DIR by district for Cluster 2 in Selangor	64
4.12	Plot for DIR based on cluster 1 and cluster 2	67
4.13	Plot for DIR membership function (FCM Clustering)	67
4.14	Triangular membership function for DIR	68
4.15	Relationship between log monthly DIR and population density for clustering data by district	69
4.16	Relationship between log monthly DIR and number of locality for clustering data by district	70
4.17	Relationship between log monthly DIR for Cluster 1 and (a) log monthly DIR lag 1 month, (b) log monthly DIR lag 2 monthly DIR lag 2 months, (c) log monthly DIR lag 3 months for clustering data by district	71
4.18	Relationship between log monthly DIR for Cluster 2 and (a) log monthly DIR lag 1 month, (b) log monthly DIR lag 2 monthly DIR lag 2 months, (c) log monthly DIR lag 3 months for clustering data by district	72
4.19	Relationship between log monthly DIR for Cluster 1 and (a) rainfall current month, (b) rainfall lag 1 month, (b) rainfall lag 2months, (d) rainfall lag 3 months for clustering data by district	73
4.20	Relationship between log monthly DIR for Cluster 2 and (a) rainfall current month, (b) rainfall lag 1 month, (b) rainfall lag 2 months, (d) rainfall lag 3 months by district	74
4.21	Relationship between log monthly DIR for Cluster 1 and (a) number of rainy days current month, (b) number of rainy days lag 1 month, (c) number of rainy days lag 2 months, (d) number of rainy days lag 3 months for clustering data by district	75
4.22	Relationship between log monthly DIR for Cluster 2 and (a) number of rainy days current month, (b) number of rainy days lag 1 month, (c) number of rainy days lag 2	

	months, (d) number of rainy days lag 3 months by district	76
4.23	Relationship between rainfall and number of rainy days for Cluster 1 and Cluster 2 from January 2010 – August 2015 for clustering data by district	77
4.24	Relationship between log monthly DIR and population density for clustering data by FCM	78
4.25	Relationship between log monthly DIR and the number Of locality for clustering data by FCM	79
4.26	Relationship between log monthly DIR for Cluster 1 and (a) log monthly DIR lag 1 month, (b) log monthly DIR lag 2 monthly DIR lag 2 months, (c) log monthly DIR lag 3 months for clustering data by FCM	79
4.27	Relationship between log monthly DIR for Cluster 2 and (a) log monthly DIR lag 1 month, (b) log monthly DIR lag 2 monthly DIR lag 2 months, (c) log monthly DIR lag 3 months for clustering data by FCM	80
4.28	Relationship between log monthly DIR for Cluster 1 and (a) rainfall current month, (b) rainfall lag 1 month, (b) rainfall lag 2months, (d) rainfall lag 3 months for clustering data by FCM	81
4.29	Relationship between log monthly DIR for Cluster 2 and (a) rainfall current month, (b) rainfall lag 1 month, (c) rainfall lag 2 months , (d) rainfall lag 3 months for clustering data by FCM	82
4.30	Relationship between log monthly DIR for Cluster 1 and (a) number of rainy days current month, (b) number of rainy days lag 1 month, (c) number of rainy days lag 2 months, (d) number of rainy days lag 3 months for clustering data by FCM	83
4.31	Relationship between log monthly DIR for Cluster 2 and (a) number of rainy days current month, (b) number of rainy days lag 1 month, (c) number of rainy days lag 2 months, (d) number of rainy days lag 3 months for clustering data by FCM	84

- 4.32 Relationship between rainfall and number of rainy days for Cluster 1 and Cluster 2 from January 2010 – August 2015 for clustering data by FCM

85



LIST OF SYMBOLS AND ABBREVIATIONS

<i>AIC</i>	-	Akaike Information Criterion
<i>BIC</i>	-	Bayesian Information Criterion
<i>z</i>	-	Climatic factors
<i>r</i>	-	Correlation coefficient
<i>d</i>	-	District
<i>p</i>	-	Estimation of the total population
<i>e</i>	-	Expected number of dengue cases
<i>L</i>	-	Likelihood
β	-	Linear predictor
μ	-	Mean
<i>m</i>	-	Month
<i>x</i>	-	Non-climatic factors
<i>y</i>	-	Number of new dengue cases
<i>i</i>	-	Number of observations
ϕ	-	Scale parameter
<i>f</i>	-	Smooth function
<i>v</i>	-	Unknown relative dengue factor
<i>t</i>	-	Year
BI	-	Breteau Index
CI	-	Container Index
COMBI	-	Communication for Behavioral Impact
DENV-1	-	Dengue Type 1 Virus
DENV-2	-	Dengue Type 2 Virus
DENV-3	-	Dengue Type 3 Virus
DENV-4	-	Dengue Type 4 Virus

DHF	-	Dengue Hemorrhagic Fever
DIR	-	Dengue Incidence Rate
DoSM	-	Department of Statistics Malaysia
EWS	-	Early Warning System
FCM	-	Fuzzy C-Means Model
GAM	-	Generalised Additive Model
GLM	-	Generalised Linear Model
GLMM	-	Generalised Linear Mixed Model
GSV	-	Generalised Cross Validation
HI	-	House Index
MOH	-	Ministry of Health
ONI	-	Oceanic Niño Index
UBRE	-	Un-Biased Risked Estimator
WHO	-	World Health Organisation



PTTHM
PERPUSTAKAAN TUNKU TUN AMINAH

CHAPTER 1

INTRODUCTION

This chapter introduces the main research objectives and the motivations of this study. Exploration of the relationship between climatic and non-climatic factors and the incidence rate of dengue is the major focus of the study and also for using the relationships which existed to provide prompt notification of dengue incidence. The chapter begins by outlining the factors associated with dengue disease generally and to be specific, in Selangor as well as Malaysia. It then moves to determine the research objectives and concludes by embarking the framework for the following chapters of the study.

1.1 Motivations

Infectious disease is a leading cause of death worldwide. Early exposure is important to control the emergence of disease, whether occurred naturally or deliberately introduced. Besides that, surveillance of infectious disease is conducted with the main intention to notify the epidemics in the early stage and necessary to control the epidemics of infectious disease. Around the world, every health agency is struggling to put an enormous effort in controlling the epidemics, unfortunately, the disease still continues to spread.

Meanwhile, the dengue incidence has grown severely in Malaysia and now significantly as a major public health concern (Mohd-Zaki *et al.*, 2014; Juni *et al.*, 2015). Public health agencies in Malaysia need to spend more attention and priority in reducing the number of dengue cases reported. Lin *et al.* (2016) proved the establishment of the integrated community-based control strategy as the new mechanism that could be applied to reduce the dengue activity.

However, the symptoms were slightly identical with other infectious disease and the situation made the disease difficult to confirm. In general, the dengue cases reported keep on increasing because the geographical distribution of dengue expands year by year. This leads to the detection of new dengue risk area in rural and also in urban areas (Aloka *et al.*, 2013; Juni *et al.*, 2015). Figure 1.1 presents the increasing number of dengue cases which peak point was in 2015 compared lower cases recorded in 2011.

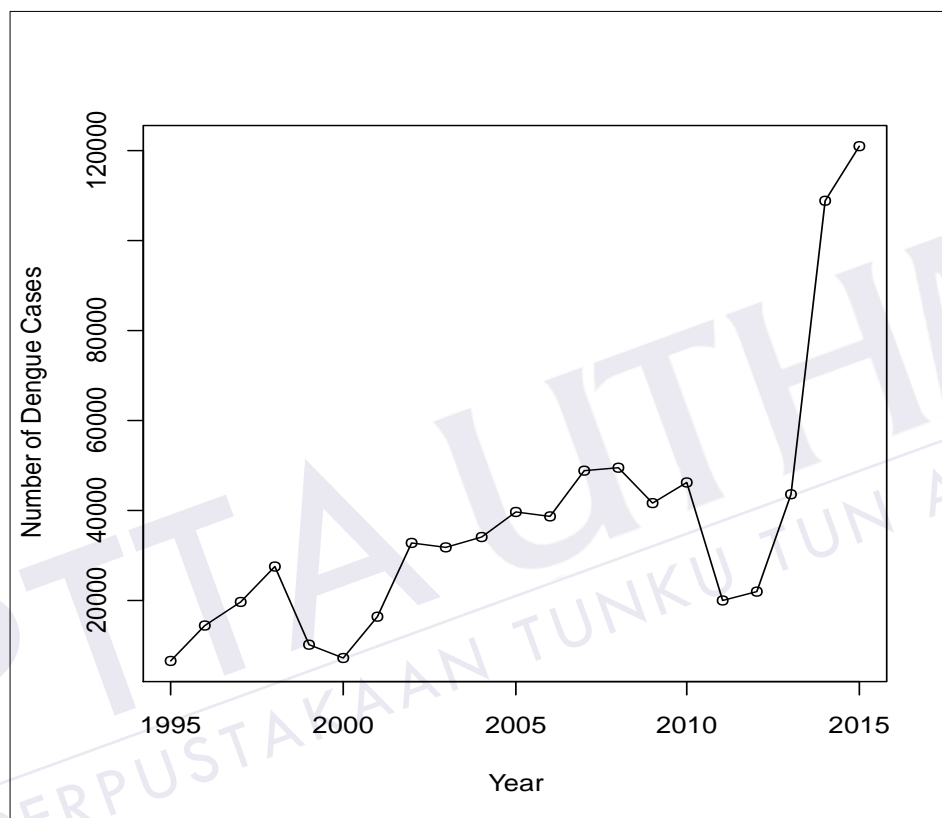


Figure 1.1: Dengue cases reported in Malaysia 1995-2015
(Ministry of Health Malaysia, 2015)

Dengue is the main health problem in Malaysia where Ministry of Health (MOH) Malaysia reported dengue is endemic since 1971. Since then, Malaysia was known as one of the worst affected countries and the dengue incidence rate (DIR) has fluctuated from 5.4 cases per 100,000 population in 1974 and the cases reported double to 10.4 cases per 100,000 population in 1987 (Shehkar & Huat, 1992). Meanwhile, the rate continues to increase four times higher between the year 1999 to the next eight years (MOH Malaysia, 2010). The frequency of dengue outbreaks has occurred relatively every 2 years (Guzman & Kouri, 2008). The number of dengue reported keep increasing and numerous programs have been introduced involving community

and health authorities since the 1970s (Che Him *et al.*, 2012). However, there has been slightly little research on modelling dengue by using climate and other covariates across the state that is recorded as the hotspot area in Malaysia.

A major problem arises in pursuing this study is the availability of information regarding non-climatic factors in Malaysia. This problem also occurred in many different countries, for example, in Vietnam, Schmidt *et al.* (2011) used population density and water supply as the main variables in their study. However, they justified that population density measure is imprecise because of the insufficient information such as information for death, migration and travel are not account in their study. But, there are some researchers found the relationship between non-climatic factors and dengue fever. For example, Stewart-Ibarra & Lowe (2013) investigated the importance of climatic and non-climatic factors in dengue fever in Ecuador and found that climatic and non-climatic factors gave important role in dengue transmission. Therefore, the results from the previous research could be the benchmark for further study to plan the exploration in finding the relationship between climatic and non-climatic factors and dengue incidence in Malaysia.

The application of statistical analysis in modelling dengue cases has been applied worldwide. Che Him *et al.* (2012) used Negative Binomial Generalised Additive Models (GAM) to explore the potential for using climatic covariates in modelling dengue in Malaysia. They found that both numbers of rainy days and rainfall with zero and three months, temperature lagged zero month and sea surface temperature lagged six months as the best predictor in a dengue prediction model in Malaysia. Even clustering technique is quite new in modelling dengue fever. Shaukat *et al.* (2015) able to apply clustering technique to different areas of the district in Jhelum, Pakistan. They strongly suggested to adopt clustering technique in future research since this study has proved better visualization in determining the highest potential dengue risk area.

Therefore, there has been only a few research on modelling dengue by using climatic and non-climatic covariates across the state that is recorded as the “hotspot” area in Malaysia. There are a few questions that need to be explored throughout this study; to what extent can relationships established, to what extent are there hotspot area affected and to what extent do any of potential factors incorporate in order to develop sensible models for dengue disease in Malaysia. The next section focuses on more specific research aims.

1.2 Problem statement

The distribution of dengue epidemics is growing exponentially while the distribution in terms of number and severity cases and the new areas of localities infected by dengue virus infection are found. One of the major problem endured in dengue epidemiology is the poor knowledge of the dengue risk factors including both climatic and non-climatic factors and the relationship among them. This problem usually occurs in rural areas as many dengue cases were not reported. Besides, there are existing studies reported on the relationship between dengue incidence and climatic factors, such as a number of rainy days and amount of rainfall. However, in Malaysia, there has been a limited amount of research on the relationship between those climatic factors and DIR. Besides, there are some limitations to the non-climatic confounding factors such as number of localities and population density due to unavailability of data which lead to misleading the climate-disease relationship.

Next, dengue fever is contagious in Malaysia since the early 1970s. Therefore, it is necessary to identify the high-risk areas of dengue fever. So far, in Malaysia, there is no author used clustering areas of localities in their study. Therefore, by applying clustering technique in this study, the health authorities and professionals can determine and focus on the source of dengue attack in any area especially the high-risk areas. Lastly, in modelling DIR, there are non-linear relationships with independent variables potentially to occur (Che Him *et al.*, 2012). For this reason, GAM is the most appropriate statistical model to identify the non-linear relationships. In addition, in Malaysia, due to data limitations, there is a lack of modelling dengue studies has been conducted on large enough data sets to predict the incorporation of climatic and non-climatic factors to provide an early alarm of future dengue outbreak. Therefore, this study comprises of monthly dengue data from the year 2010 until 2015 and hopefully, the outcome can help health workers and stakeholders to speed up the development of dengue program in future dengue outbreaks.

1.3 Research objectives

This study embarks on the following objectives:

1. To verify pre-determined factors to be adopted in the new modelling framework.

2. To identify datasets by clustering based on district and Fuzzy C-Means Model (FCM) categories.
3. To propose the new modelling framework by using a combination of the Negative Binomial Generalised Additive Model (GAM) and Fuzzy C-Mean Model (FCM) for dengue fever in Malaysia.
4. To assess the performance of new models which is a combination of Negative Binomial Generalised Additive Model (GAM) and Fuzzy C-Mean Model (FCM) by comparing value of Deviance (D), Akaike Information Criteria (AIC) and Bayesian Information Criteria (BIC).
5. To validate models by comparing the proposed model with existing model.

1.4 Significance of study

This study has diversified unique features that can contribute to minimise the identified dengue risk problem in Malaysia. Firstly, this study aimed to develop modelling dengue cases which focused on monthly data of dengue cases in Selangor and related confounding factors from January 2010 to August 2015 (68 months). This study is different from the other research because this study considered a long amount of time on a monthly basis of dengue data in Selangor. Secondly, in the modelling, this study undergo two different clustering processes in which the data set clustered by district and the application of FCM. After the clustering process, the clustered data set then adopted by Negative Binomial GAM to develop model represent dengue incidence in Selangor. So far, this is the first study that has considered the clustering approach especially the application of FCM with the combination of Negative Binomial GAM.

Next, another major contribution of this study is the application of GAM in modelling dengue cases in Selangor. Due to the small scale of year data, this study used the smooth function for month to narrow down the impact of analysis for GAM. Therefore, one of the most important findings in this study is the ability to deal with the next dengue outbreak by made several months ahead of preparation by referring to the potential model developed in this study.

1.5 Scope of study

The data that has been used in this study is dengue data. The dengue data was obtained from Vector Unit of Operational Dengue, Public Health Division at Ministry of Health Malaysia. This data was collected from nine administrative districts in Selangor which included Gombak, Hulu Langat, Hulu Selangor, Klang, Kuala Langat, Kuala Selangor, Petaling, Sabak Bernam and Sepang were used in this study. The collated data consists of monthly confirmed dengue cases in nine districts in Selangor. This data was recorded from January 1st, 2010 until August 31st, 2015, which consists of 68 months. In addition, a number of locality in each district of Selangor that has been affected by dengue virus also provided in the datasets besides the dengue data. This study choose Selangor as a case study because it has recorded the highest number of dengue cases, which accounted for 50% (50,000 cases) of total dengue cases (100,028 cases) reported in Malaysia during 2016. Besides that, Selangor was one of the highest population density of all states in Malaysia, which estimated 674 people per square kilometre (Department of Statistics Malaysia, 2011).

The second variable that used in this study is rainfall. The rainfall data was obtained from Department of Irrigation and Drainage Malaysia. This study collated rainfall data from 22 rainfall stations in Selangor. All the rainfall stations that used in this study have to represent each district in Selangor. From the data that has been provided, the average amount of rainfall (in millimetre) and the average number of rainy days (in days) were used to modelling dengue cases in Selangor.

1.6 Summary

This chapter has outlined the motivation for the research considered in the next chapters of this study. Therefore, the outline of next chapters is as follows. Chapter 2 include a literature review of confounding factors related to the DIR including climatic and non-climatic factors. Dengue fever and its transmission both global and specific in Malaysia and Selangor, together with an extra review of related studies using statistical techniques to develop dengue model is included. A brief introduction to Malaysia and its climate as well as dengue surveillance practiced are also discussed. Then, Chapter 3 is an introduction of GLM and GAM to explain monthly dengue cases in Selangor from January 2010-August 2015. FCM also introduced to discuss the

clustering process of the datasets. Subsequently, Chapter 4 explores datasets particularly to dengue in Selangor and potential climatic and non-climatic covariates relevant to DIR. Selection of factors is used to identify climatic and non-climatic factors that have a relationship with DIR in Selangor. The used of clustering approach in the datasets is also explained. In Chapter 5, the application of FCM and GAM to develop model represent dengue incidence in Selangor. The potential models for Selangor are then tested by comparing the values of Akaike Information Criteria (*AIC*), Bayesian Information Criteria (*BIC*) and deviance, in order to choose the best models, represent DIR in Selangor. The final chapter of this study summarises the conclusions, results and recommendations for future modelling dengue research.



PTTA UTHM
PERPUSTAKAAN TUNKU TUN AMINAH

REFERENCES

- Abdul Mohit, M. (2013). Objective analysis of variation in the regional quality of life in Malaysia and its policy implications. *Procedia-Social and Behavioral Sciences*, 101, 454-464.
- Ab-Fatah, M., Subenthiran, S., Abdul-Rahman, P.S., Saat, Z. & Thayan, R. (2015). Dengue serotypes surveillance among patients admitted for dengue in two major hospitals in Selangor, Malaysia, 2010 – 2011. *Tropical Biomedicine*, 32(1), 187-191.
- Ahmad Nizal, M.G., Rozita, H., Mazrura, S., Hidayatulfathi, O., Faridah, M.A., Noor Artika, I. & Er, A.C. (2012). Dengue infections and circulating serotypes in Negeri Sembilan, Malaysia. *Malaysian Journal of Public Health Medicine* 2012, 12(1), 21-30.
- Ahmed, M.M. (2010). Clinical profile of dengue fever infection in King Abdul Aziz University Hospital Saudi Arabia. *The Journal of Infection in Developing Countries*, 4(8), 503-510.
- Ahmed, S.A., Siddiqi, J.S., Quaiser, S. & Kamal, S. (2015). Using PCA, Poisson and Negative Binomial Model to study the climatic factor and dengue fever outbreak in Lahore. *Journal of Basic & Applied Sciences*, 11, 8-16.
- Akaike, H. (1974). A new look at the statistical model identification. *Institute of Electrical and Electronics Engineers (IEEE) Transactions on Automatic Control*, 19(6), 716-723.
- Alhaeli, A., Bahkali, S., Ali, A., Househ, M.S. & El-Metwally, A.A. (2016). The epidemiology of dengue fever in Saudi Arabia: A systematic review. *Journal of Infection and Public Health*, 9(2), 117-124.

- Alkhalidy, I. (2017). Modelling the association of dengue fever cases with temperature and relative humidity in Jeddah, Saudi Arabia – A generalised linear model with break-point analysis. *Acta Tropica*, 168, 9-15.
- Aloka, M., Premaratne, H.L. & Fernando, M.G.N.A.S. (2013). Towards an early warning system to combat dengue. *International Journal of Computer Science and Electronics Engineering*, 1(2), 252-256.
- Alshehri, M.S.A. (2013). Dengue fever outburst and its relationship with climatic factors. *World Applied Sciences Journal*, 22(4), 506-515.
- Alto, B.W. & Bettinardi, D. (2013). Temperature and dengue virus infection in mosquitoes: Independent effects on the immature and adult stages. *The American Society of Tropical Medicine and Hygiene*, 88(3), 497-505.
- Amirabadizadeh, M., Huang, Y.F. & Lee, T.S. (2015). Recent trends in temperature and precipitation in the Langat River Basin, Malaysia. *Advances in Meteorology*, 2015.
- A-Nuegoonpipat, A., Berlioz-Arthaud, A., Chow, V., Endy, T., Lowry, K., Mai Ie, Q., Ninh, T.U., Pyke, A., Reid, M., Reynes, J.M., Su Yun, S.T., Thu, H.M., Wong, S.S., Holmes, E.C. & Aaskov, J. (2004). Sustained transmission of dengue virus type 1 in the Pacific due to repeated introductions of different Asian stains. *Virology*, 329(2), 505-512.
- Aura, D.H.M. & Alfonso, J.R.M. (2010). Potential influence of climate variability on dengue incidence registered in a western pediatric Hospital of Venezuela. *Tropical Biomedicine*, 27(2), 280-286.
- Awan, S. (2015). Association of climate factors with outbreak of dengue fever – An eco-epidemiological analysis. *International Journal of Epidemiology*, 44(suppl_1), i45-i46.
- Awang, N.A. & Hamid, M.A. (2013). Sea level rise in Malaysia. *HydroLink-Special Issue on Sea Level Rise-Adaptation Measures*, 2, 47-49.
- Aziz, A.T., Dieng, H., Abu Hassan, A., Mahyoub, J.A., Turkistani, A.M., Mesed, H., Koshike, S., Satho, T., Salmah, M.R. & Hamdan, A. (2012). A household survey

of container-breeding mosquitoes and climatic factors influencing the prevalence of *Aedes aegypti* (Diptera: Culicidae) in Makkah City, Saudi Arabia. *Asian Pacific Journal of Tropical Biomedicine*, 2(11), 849-857.

Azmawati, M.N., Aniza, I. & Ali, M. (2013). Evaluation of Communication for Behavioral Impact (COMBI) program in dengue prevention: a qualitative and quantitative study in Selangor, Malaysia. *Iranian Journal of Public Health*, 42(5), 538-539.

Back, A.T. & Lundkvist, A. (2013). Dengue viruses- an overview. *Infection Ecology & Epidemiology*, 3(1), 1-21.

Barbosa, G.L., Donalisio, M.R., Stephan, C., Lourenco, R.W., Andrade, R., Arduino, M.d.B. & Lima, V.L.C. (2014). Spatial distribution of the risk of dengue and the entomological indicators in Sumaré, State of São Paulo, Brazil. *Public Library of Science Neglected Tropical Disease*, 8(5), 1-9.

Behnam, M.A.M, Nitsche, C., Boldescu, V. & Klein, C.D. (2016). The medical chemistry of dengue virus. *Journal of Medical Chemistry*, 59, 5622-5649.

Bezdek, J.C. (1981). *Pattern recognition with fuzzy objective function algorithm*. 1st ed. New York: Plenum Press.

Bhatt, S., Gething, P.W., Brady, O.J., Messina, J.P., Farlow, A.W., Moyes, C.L., Drake, J.M., Brownstein, J.S., Hoen, A.G., Sankoh, O., Myers, M.F., George, D.B., Jaenisch, T., Wint, G.R.W., Simmons, C.P., Scott, T.W., Farrar, J.J. & Hay, S.I. (2013). The global distribution and burden of dengue. *Nature*, 496(7446), 504-507.

Bithell, J.F. (1990). An application of density estimation to geographical epidemiology. *Statistics in Medicine*, 9(6), 691-701.

Bozza, F.A., Cruz, O.G., Zagne, S.M.O., Azeredo, E.L., Nogueira, R.M.R., Assis, E.F., Bozza, P.T. & Kubelka, C.F. (2008). Multiple cytokine profile from dengue patients: MIP-1beta and IFN-gamma as predictive factors for severity. *BioMed Central Infectious Diseases*, 8(1), 86.

- Breslow, N.E. (1984). Extra-Poisson variation in log-linear models. *Journal Royal Statistical Society*, 33(1), 38-44.
- Butterworth, M.K., Morin, C.W. & Comrie, A.C. (2017). An analysis of the potential impact of climate change on dengue transmission in the Southeastern United States. *Environmental Health Perspectives*, 125(4), 579-585.
- Cameron, A.C. & Trivedi, P.K. (1998). *Regression Analysis of Count Data*. Cambridge: Cambridge University Press.
- Cao-Lormeau, V.M. (2009). Dengue viruses binding proteins from *Aedes aegypti* and *Aedes polynesiensis* salivary glands. *Virology Journal*, 6(1), 1-4.
- Centers for Disease Control and Prevention (2014). *Dengue*. United States of America: U.S. Department of Health & Human Services.
- Chan, T.C., Hu, T.H. & Hwang, T.S. (2015). Daily forecast of dengue fever incidents for urban villages in city. *International Journal of Health Geographics*, 14(9), 1-11.
- Chandy, S., Ramanathan, K., Manoharan, A., Mathai, D. & Baruah, K. (2013). Assessing effect of climate on the incidence of dengue in Tamil Nadu. *Indian Journal of Medical Microbiology*, 31(3), 283-286.
- Che Dom, N., Ahmad, A.H., Ishak, A.R. & Ismail, R. (2013). Assessing the risk of dengue fever based on the epidemiological, environmental and entomological variables. *Asia Pasific International Conference on Environment-Behaviour Studies*, 105, 183-194.
- Che Him, N., Bailey, T.C. & Stephenson, D.B. (2012). Climate variability and dengue incidence in Malaysia. *27th International Workshop on Statistical Modelling IWSM'27*. Prague, Czech Republic: The Statistical Modelling Society. Volume 2.
- Cheah, W.K., Ng, K.S., Marzilawati, A.R. & Lum, L.C. (2014). A review of dengue research in Malaysia. *Medical Journal of Malaysia*, 69, 59-67.
- Chen, T.C., Tsay, J.D., Yen, M.C. & Matsumoto, J. (2013). The winter rainfall of Malaysia. *Journal of Climate*, 26(3), 936-958.

- Cheng, Y.L., Lin, Y.S., Chen, C.L., Wan, S.W., Ou, Y.D., Yu, C.Y., Tsai, T.T., Tseng, P.C. & Lin, C.F. (2015). Dengue virus infection causes the activation of distinct NF- κ B pathways for inducible nitric oxide synthase and TNF- α expression in RAW264.7 cells. *Mediators of Inflammation*, 2015, 274025.
- Cheong, Y.L. (2015). *Dengue disease in Malaysia: Vulnerability mapping and environmental risk assessment*. Humboldt-Universität zu Berlin – Geographisches Institut: Ph.D. Thesis.
- Cheong, Y.L., Burkart, K., Leitão, P.J. & Lakes, T. (2013). Assessing weather effects of dengue disease in Malaysia. *International Journal of Environmental Research and Public Health*, 10, 6319-6334.
- Chew, M.H., Rahman, M., Jelip, J. & Isahak, I. (2012). All serotypes of dengue viruses circulating in Kuala Lumpur, Malaysia. *Current Research Journal of Biological Sciences*, 4(2), 229-234.
- Choi, Y., Tang, C.S., McIver, L., Hashizume, M., Chan, V., Abeyasinghe, R.R., Iddings, S. & Huy, R. (2016). Effects of weather factors on dengue fever incidence and implications for interventions in Cambodia. *BioMed Central Public Health*, 16, 241.
- Clapham, H.E., Quyen, T.H., Hue Kien, D.T., Dorigatti, I., Simmons, C.P. & Ferguson, N.M. (2016). Modelling virus and antibody dynamics during dengue virus infection suggests a role of antibody in virus clearance. *Public Library of Science Computational Biology*, 12(5), e1004951.
- Colón-González, F.J., Fezzi, C., Lake, I.R. & Hunter, P.R. (2013). The effects of weather and climate change on dengue. *Public Library of Science Neglected Tropical Diseases*, 7(11), e2503.
- Cordeiro, G.M. & McCullagh, P. (1991). Bias correction in Generalized Linear Models. *Royal Statistics Society*, 53(3), 629-643.
- Cramer, D. & Howitt, D.L. (2004). *The Sage Dictionary of Statistics: A Practical Resource for Students in the Social Sciences*. London: SAGE Publications Ltd.

- Das, S., Sarfraz, A., Jaiswal, N. & Das, P. (2017). Impediments of reporting dengue cases in India. *Journal of Infection and Public Health*. 1-5.
- Dean, C. & Lawless, J.F. (1987). Tests for detecting overdispersion in Poisson regression model. *Journal of the American Statistical Association*, 84(406), 467-472.
- Deen, J.L., Harris, E., Wills, B., Balmaseda, A., Hammond, S.N., Rocha, C., Dung, N.M., Hung, N.T., Hien, T.T. & Farrar, J.J. (2006). The WHO dengue classification and case definitions: time for a reassessment. *Lancet*, 368(9530), 170-173.
- Delatte, H., Toty, C., Boyer, S., Bouetard, A., Bastien, F. & Fontenille, D. (2013). Evidence of habitat structuring *Aedes albopictus* populations in Réunion Island. *Public Library of Science Neglected Tropical Disease*, 7(3), e2111.
- Department of Statistics Malaysia (2011). *Population Distribution and Basic Demographic Characteristics Report 2010 (Updated: 05/08/2011)*. Putrajaya.
- Dhang, C.C., Benjamin, S., Saranum, M.M., Fook, C.Y., Lim, L.H., Ahmad, N.W. & Sofian-Azirun, M. (2005). Dengue vector surveillance in urban residential and settlement areas in Selangor, Malaysia. *Tropical Biomedicine*, 22(1), 39-43.
- Diaz-Quijano, F.A. (2015). Dengue severity: a key determinant of underreporting. *Tropical Medicine & International Health*, 20(10), 1403.
- Dunn, J.C. (1973). A fuzzy relative of the ISODATA process and its use in detecting compact, well-separated clusters. *Journal of Cybernetics*, 3, 32-57.
- Ebi, K.L. & Nealon, J. (2016). Dengue in a changing climate. *Environmental Research*, 151, 115-123.
- Falope, O., Hanson, K. & Azizan, A. (2015). Dengue and mosquito control programs: a comparative analysis. *Journal of Applied Life Sciences International*, 2(1), 35-48.
- Favier, C., Degallier, N., Rosa-Freitas, M.G., Boulanger, J.P., Costa Lima, J.R., Luitgards-Moura, J.F., Menkès, C.E., Mondet, B., Oliveira, C., Weimann, E.T.S. & Tsouris, P. (2006). Early determination of the reproductive number for vector-

borne diseases: the case of dengue in Brazil. *Tropical Medicine & International Health*, 11(3), 332-340.

Getachew, D., Tekie, H., Gebre-Michael, T., Balkew, M. & Mesfin, A. (2015). Breeding sites of *Aedes aegypti*: Potential dengue vectors in Dire Dawa, East Ethiopia. *Interdisciplinary Perspectives on Infectious Diseases*, 2015.

Gubler, D.J. (1997). Dengue and dengue hemorrhagic fever. *Seminars in Pediatric Infectious Diseases*, 8(1), 3-9.

Gubler, D.J. (1998). Dengue and dengue hemorrhagic fever. *Clinical Microbiology Reviews*, 11(3), 480-496.

Gubler, D.J. (2002). Epidemic dengue/dengue hemorrhagic fever as a public health, social and economic problem in the 21st century. *Trends Microbiology*, 10(2), 100-103.

Gubler, D.J. (2011). Dengue, urbanization and globalization: the unholy trinity of the 21st century. *Tropical Medical and Health*, 39(4SUPPLEMENT), S3-S11.

Gubler, D.J., Suharyono, W., Tan, R., Abidin, M. & Sie, A. (1981). Viremia in patients with naturally acquired dengue infection. *Bulletin of the World Health Organization*, 59(4), 623.

Guzman, M.G. & Kouri, G. (2008). Dengue haemorrhagic fever integral hypothesis: confirming observations, 1987-2007. *Transactions of the Royal Society of Tropical Medicine and Hygiene*, 102(6), 522-523.

Guzman, M.G., Halstead, S.B., Artsob, H., Buchy, P., Farrar, J., Gubler, D.J., Hunsperger, E., Kroeger, A., Margolis, H.S., Martinez, E., Nathan, M.B., Pelegrino, J.L., Simmons, C., Yoksan, S. & Peeling, R.W. (2010). Dengue: A continuing global threat. *Nature Reviews Microbiology*, 8, S7-S16.

Hales, S., Edwards, S.J. & Kovats, R.S. Impacts on health climate extremes. *Climate change and health: risks and responses*. Geneva: World Health Organization. 2002. pp. 79-102.

- Hasan, T. & Bambrick, H. (2013). The effects of climate variables on the outbreak of dengue in Queensland (2008-2009). *Southeast Asian Journal Tropical Medicine Public Health*, 44(4), 613-622.
- Hassan, H., Shohaimi, S. & Hashim, N.R. (2012). Risk mapping of dengue in Selangor and Kuala Lumpur, Malaysia. *Geospatial Health*, 7(1), 21-25.
- Hastie, T. & Tibshirani, R. (1986). Generalized additive models. *Statistical Science*, 3, 297-310.
- Herrera-Martinez, A.D. & Rodriguez-Morales, A.J. (2010). Potential influence of climate variability on dengue incidence registered in a western pediatric Hospital of Venezuela. *Tropical Biomedicine*, 27(2), 280-286.
- Hii, Y.L., Ahmad Zaki, R., Aghamohammadi, N. & Rocklöv, J. (2016). Research on climate and dengue in Malaysia: a systematic review. *Current Environmental Health Reports*, 3, 81-90.
- Hii, Y.L., Rocklöv, J., Ng, N., Tang, C.S., Pang, F.Y. & Sauerborn, R. (2009). Climate variability and increase in intensity and magnitude of dengue incidence in Singapore. *Global Health Action*, 2(1), 2036.
- Hilbe, J.M. (2007). *Negative Binomial Regression*. Cambridge: Cambridge University Press.
- Hock, S.S. (2007). *The Population of Peninsular Malaysia*. Singapore: Singapore University Press.
- Holmes, E.C., Tio, P.H., Perera, D., Muhi, J. & Cardoso, J. (2009). Importation and co-circulation of multiple serotypes of dengue virus in Sarawak, Malaysia. *Virus Research*, 143(1), 1-5.
- Honório, N.A., Nogueira, R.M.R., Codeço, C.T., Carvalho, M.S., Cruz, O.G., Magalhães, M.d.A.F.M., de Araújo, J.M.G., de Araújo, E.S.M., Gomes, M.Q., Pinheiro, L.S., Pinel, C.d.S. & Lourenço-de-Oliveira, R. (2009). Spatial evaluation and modeling of dengue seroprevalence and vector density in Rio de Janeiro, Brazil. *Public Library of Science Neglected Tropical Diseases*, 3(11), e545.

- Islam, M.A., Hasanuzzaman, M., Rahim, N.A., Nahar, A. & Hosenuzzaman, M. (2014). Global renewable energy- based electricity generation and smart grid system for energy security. *The Scientific World Journal*, 2014.
- Johansson, M.A., Dominici, F. & Glass, G.E. (2009). Local and global effects of climate on dengue transmission in Puerto Rico. *Public Library of Science Neglected Tropical Diseases*, 3(2), e382.
- Juni, M.H., Hayati, K.S., Cheng, C.M., Pyang, G.S., Abd Samad, N.H. & Zainal Abidin, Z.S. (2015). Risk behaviour associated with dengue fever among rural population in Malaysia. *International Journal of Public Health and Clinical Sciences*, 2(1), 114-127.
- Karim, M.N., Munshi, S.U., Anwar, N. & Alam, M.S. (2012). Climatic factors influenceing dengue cases in Dhaka city: A model for dengue prediction. *Indian Journal of Medical Research*, 136(1), 32-39.
- Kok, P.H., Mohd Akhir, M.F., Tangang, F. & Husain, M.L. (2017). Spatiotemporal trends in the southwest monsoon wind-driven upwelling in the southwestern part of the South China Sea. *Public Library of Science One*, 12(2), e0171979.
- Lam, S.K. (1993). Strategies for dengue control in Malaysia. *Tropical Medicine*, 35(4), 303-307.
- Lau, S.M., Vythilingam, I., Doss, J.I., Sekaran, S.D., Chua, T.H., Wan Sulaiman, W.Y., Chinna, K., Lim, Y.A.L. & Venugopalan, B. (2015). Surveillance of adult *Aedes* mosquitoes in Selangor, Malaysia. *Tropical Medicine and International Health*, 20(10), 1272-1280.
- Lawless, J.F. (1987). Negative binomial and mixed Poisson regression. *Canadian Journal of Statistics*, 15, 209-225.
- Lee, H.L., Rohani, A., Khadri, M.S., Nazni, W.A., Rozilawati, H., Nurulhusna, A.H., Nor Afizah, A.H., Roziah, A., Rosilawati, R. & Teh, C.H. (2015). Dengue vector control in Malaysia- challenges and recent advances. *The International Medical Journal Malaysia*, 14(1), 11-16.

- Lee, H.S., Nguyen-Viet, H., Nam, V.S., Lee, M., Won, S., Duc, P.P. & Grace, D. (2017). Seasonal patterns of dengue fever and associated climate factors in 4 provinces in Vietnam from 1994 to 2013. *BioMed Central Infectious Diseases*, 17, 218.
- Lekdee, K. & Ingsrisawang, L. (2013). Generalized linear mixed models with spatial random effects for spatio-temporal data: An application to dengue fever mapping. *Journal of Mathematics and Statistics*, 9(2), 137-143.
- Li, Y., Kamara, F., Zhou, G., Puthiyakunnon, S., Li, C., Liu, Y., Zhou, Y., Yao, L., Yan, G. & Chen, X.G. (2014). Urbanization increases *Aedes albopictus* larval habitats and accelerates mosquito development and survivorship. *Public Library of Science Neglected Tropical Diseases*, 8(11), e3301.
- Lin, H., Liu, T., Song, T., Lin, L., Xiao, J., Lin, J., He, J., Zhong, H., Hu, W., Deng, A., Peng, Z., Ma, W. & Zhang, Y. (2016). Community involvement in dengue outbreak control: An integrated rigorous intervention strategy. *Public Library of Science Neglected Tropical Diseases*, 10(8), e0004919.
- Loader, C. (1999). *Local Regression and Likelihood*. USA: Springer.
- Loo, Y.Y., Billa, L. & Singh, A. (2015). Effect of climate change on seasonal monsoon in Asia and its impact on the variability of monsoon rainfall in Southeast Asia. *Geoscience Frontiers*, 6(6), 817-823.
- Lowe, R., Bailey, T.C., Stephenson, D.B., Graham, R.J., Coelho, C.A., Carvalho, M.S. & Barcellos, C. (2011). Spatio-temporal modelling of climate-sensitive disease risk: towards an early warning system for dengue in Brazil. *Computers & Geosciences*, 37(3), 371-381.
- Malavige, G.N., Fernando, S., Fernando, D.J. & Seneviratne, S.L. (2004). Dengue viral infections. *British Medical Journals*, 80(948), 588-601.
- Mammen Jr, M.P., Pimgate, C., Koenraadt, J.M., Rothman, A.L., Aldstadt, J., Nisalak, A., Jarman, R.G., Jones, J.W., Srikiatkachorn, A., Ypil-Butac, C.A., Getis, A., Thammapalo, S., Marrison, A.C. & Scott, T.W. (2008). Spatial and temporal clustering of dengue virus transmission in Thai Villages. *Public Library of Science Neglected Tropical Diseases Medicine*, 5(11), e205.

- Masron, T., Yaakob, U., Mohd Ayob, N. & Mokhtar, A.S. (2012). Population and spatial distribution of urbanisation in Peninsular Malaysia 1957 – 2000. *Malaysia Journal of Society and Space*, 8(2), 20-29.
- McCullagh, P. & Nelder, J.A. (1989). *Generalized Linear Models*. 2nd ed. (Vol. 37). London: Chapman and Hall.
- Meltzer, M.I., Rigau-Pérez, J.G., Clark, G.G., Reiter, P. & Gubler, D.J. (1998). Using disability-adjusted life years to assess the economic impact of dengue in Puerto Rico: 1984-1994. *The American Society of Tropical Medicine and Hygiene*, 59(2), 265-271.
- Ministry of Health Malaysia (2010). *Management of Dengue Infection in Adults (Revised 2nd Edition)*. Putrajaya: Malaysia Health Technology Assessment Section (MaHTAS).
- Mohd-Zaki, A.H., Brett, J., Ismail, E. & L'Azou, M. (2014). Epidemiology of dengue disease in Malaysia (2000-2012): A systematic literature review. *Public Library of Science Neglected Tropical Diseases*, 8(11), e3159.
- Morales, I., Salje, H., Saha, S. & Gurley, E.S. (2016). Seasonal distribution and climatic correlates of dengue disease in Dhaka, Bangladesh. *The American Journal of Tropical Medicine and Hygiene*, 94(6), 1359-1361.
- Morin, C.W., Comrie, A.C. & Ernst, K. (2013). Climate and dengue transmission: Evidence and implications. *Environmental Health Perspectives*, 121(11-12), 1264-1272.
- Morrison, A.C., Getis, A., Santiago, M., Rigau-Perez, J.G. & Reiter, P. (1998). Exploratory space-time analysis of reported dengue cases during an outbreak in Florida, Puerto Rico, 1991 – 1992. *The American Journal of Tropical Medicine and Hygiene*, 58, 287-298.
- Mudin, R.N. (2015). Dengue incidence and the prevention and control program in Malaysia. *The International Medical Journal Malaysia*, 14(1), 5-9.

- Muhammad Azami, N.A., Salleh, S.A., Neoh, H.M., Syed Zakaria, S.Z. & Jamal, R. (2011). Dengue epidemic in Malaysia: Not a predominantly urban disease anymore. *BioMed Central Research Notes*, 4, 216.
- Murray, N.E.A., Quam, M.B. & Wilder-Smith, A. (2013). Epidemiology of dengue: past, present and future prospects. *Clinical Epidemiology*, 5, 299-309.
- Naim, M.R., Sahani, M., Hod, R., Hidayatulfathi, O., Idrus, S., Norzawati, Y., Hazrin, H., Tahir, A., Wen, T.H., King, C.C. & Zainudin, M.A. (2014). Spatial-temporal analysis for identification of vulnerability to dengue in Seremban district, Malaysia. *International Journal of Geoinformatics*, 10(1), 31-38.
- Neiderud, C.J. (2015). How urbanisation affects the epidemiology of emerging infectious diseases. *Infection Ecology & Epidemiology*, 5(1), 27060.
- Nelder, J.A. & Wedderburn, R.W.M. (1972). Generalized linear models. *Journal of the Royal Statistical Society*, 135(3), 370-384.
- Ng, L.C., Chem, Y.K., Koo, C., Mudin, R.N., Mohd Amin, F., Lee, K.S. & Kheong, C.C. (2015). 2013 Dengue outbreaks in Singapore and Malaysia caused by different viral strains. *The American Journal of Tropical Medicine and Hygiene*, 92(6), 1150-1155.
- Ong, S.Q. (2017). Dengue vector control in Malaysia: a review for current and alternative strategies. *Sains Malaysiana*, 45(5), 777-785.
- Ooi, E.E. & Gubler, D.J. (2009). Dengue in Southeast Asia: epidemiological characteristics and strategic challenges in disease prevention. *Cadernos de Saude Publica*, 25, S115-S124.
- Pang, E.L. & Loh, H.S. (2016). Current perspectives on dengue episode in Malaysia. *Asian Pacific Journal of Tropical Medicine*, 9(4), 395-401.
- Paul, B. & Tham, W.L. (2015). Interrelation between climate and dengue in Malaysia. *Health*, 7, 672-678.
- Peña-García, V.H., Triana-Chávez, O., Mejía-Jaramillo, A.M., Díaz, F.J., Gómez-Palacio, A. & Arboleda-Sánchez, S. (2016). Infection rates by dengue virus in mosquitoes and the influence of temperature may be related to different endemicity

- patterns in three Colombian cities. *International Journal of Environmental Research and Public Health*, 13, 734.
- Peng, T.N. (2012). Internal migration in the Klang Valley of Malaysia: issues and implications. *Malaysian Journal of Chinese Studies*, 1, 40-59.
- Petry, F.E., Robinson, V.B. & Cobb, M.A. (2005). *Fuzzy Modeling with Spatial Information for Geographic Problems*. Berlin: Springer.
- Pinheiro, F.P. & Corber, S.J. (1997). Global situation of dengue and dengue haemorrhagic fever, and its emergence in the Americas. *World Health Statistics Quarterly*, 50(3-4), 161-169.
- Polwiang, S. (2016). The correlation of climate factors on dengue transmission in urban area: Bangkok and Singapore cases. *PeerJ Preprints*, 4, e2322v1.
- Poonaveswari, S. (1993). Dengue situation in Malaysia. *Malaysia Journal of Pathology*, 15, 3-7.
- Premaratne, M.K., Perera, S.S.N., Malavige, G.N. & Jayasinghe, S. (2017). Mathematical modelling of immune parameters in the evolution of severe dengue. *Computational and Mathematical Methods in Medicine*, 2017.
- Promprou, S., Jaroensutasinee, M., & Jaroensurasinee, K. (2005). Climatic factors affecting dengue haemorrhagic fever incidence in Southern Thailand. *Dengue Bulletin*, 29, 41-48.
- Qi, X., Weng, Y., Li, Y., Meng, Y., Chen, Q., Ma, J. & Gao, G.F. (2015). The effects of socioeconomic and environmental factors on the incidence of dengue fever in the Pearl River Delta, China, 2013. *Public Library of Science Neglected Tropical Diseases*, 9(10), e0004159.
- Ramachandran, V.G., Roy, P., Das, S., Mogha, N.S. & Bansal, A.K. (2016). Empirical model for estimating dengue incidence using temperature, rainfall and relative humidity: a 19-year retrospective analysis in East Delhi. *Epidemiology Health*, 38, e2016052.

- Ramasamy, R. & Surendran, S.N. (2012). Global climate change and its potential impact on disease transmission by solinity-tolerant mosquito vectors in Coastal zones. *Frontiers in Physiology*, 3, 198.
- Ratner, B. (2009). The correlation coefficient: Its values range between +1/-1, or do they? *Journal of Targeting, Measurement and Analysis of Marketing*, 17(2), 139-142.
- Reiter, P. (2001). Climate change and mosquito-borne disease. *Environmental Health Perspectives*, 109(Suppl 1), 141-161.
- Ross, T.J. (2010). *Fuzzy Logic with Engineering Applications*. 3rd edition. USA: John Wiley & Sons.
- Rozhan, S., Jamsiah, M., Rahimah, A. & Ang, K.T. (2006). The COMBI (Communication for Behavioural Impact) program in the prevention and control of dengue – The Hulu Langat experience. *Jurnal Kesehatan Masyarakat*, 12(1), 19-32.
- Rusiman, M.S., Adnan, R., Nasibov, E. & Jacob, K. (2012). Adjustment of an intensive care unit (ICU) data in fuzzy c-regression models. *Journal of Science and Technology*, 4(2), 99-108.
- Russell, R.C., Currie, B.J., Lindsay, M.D., Mackenzie, J.S., Ritchie, S.A. & Whelan, P.I. (2009). Dengue and climate change in Australia: predictions for the future should incorporate knowledge from the past. *The Medical Journal of Australia*, 190(5), 265-268.
- Rustempasic, I. & Can, M. (2013). Diagnosis of Parkinson's disease using Fuzzy C-Means clustering and pattern recognition. *Southeast Europe Journal of Soft Computing*, 2(1), 42-49.
- Saikia, D. & Dutta, J.C. (2017). Early diagnosis of dengue disease using fuzzy inference system. *Microelectronics, Computing and Communications (MicroCom), 2016 International Conference*. Durgapur. India. pp. 1-6.
- Sarti, E., L'Azou, M., Mercado, M., Kuri, P., Siqueira, J.B., Solis, E., Noriega, F. & Ochiai, R.L. (2016). A comparative study on active and passive epidemiological

surveillance for dengue in five countries of Latin America. *International Journal of Infectious Diseases*, 44, 44-49.

Schmidt, W.P., Suzuki, M., Thiem, V.D., White, R.G., Tsuzuki, A., Yoshida, L.M., Yanai, H., Haque, U., Tho, L.H., Anh, D.D. & Ariyoshi, K. (2011). Population density, water supply and the risk of dengue fever in Vietnam: Cohort study and spatial analysis. *Public Library of Science Medicine*, 8(8), e1001082.

Schwämmle, V. & Jensen, O.N. (2010). A simple and fast method to determine parameters for fuzzy c-means cluster analysis. *Bioinformatics*, 26(22), 2841-2848.

Schwarz, G. (1978). Estimating the dimension of a model. *The Annals of Statistics*, 6(2), 461-464.

Shaukat, K., Masood, N., Shafaat, A., Jabbar, K., Shabbir, H. & Shabbir, S. (2015). Dengue fever in perspective of clustering algorithms. *Data Mining in Genomics & Proteomics*, 6(3), 1-5.

Shehkar, K.C. & Huat, O.L. (1992). Epidemiology of dengue/dengue hemorrhagic fever in Malaysia – A retrospective epidemiological study 1973 – 1987. Part I: Dengue hemorrhagic fever (DHF). *Asia Pacific Journal of Public Health*, 6(2), 15-25.

Shepard, D.S., Undurraga, E.A. & Halasa, Y.A. (2013). Economic and disease burden of dengue in Southeast Asia. *Public Library of Science Neglected Tropical Diseases*, 7(2), e2055.

Silva, F.D., dos Santos, A.M., Corrêa, R.D.F.C.F. & Caldas, A.D.J.M. (2016). Temporal relationship between rainfall, temperature and occurrence of dengue cases in São Luís, Maranhão, Brazil. *Ciencia & Saude Coletiva*, 21(2), 641-646.

Singh, M.S.J., Syed Hassan, S.I. & Ain, M.F. (2007). Rainfall attenuation and rainfall rate measurement in Malaysia comparison with prediction models. *American Journal of Applied Sciences*, 4(1), 5-7.

Siqueira-Junior, J.B., Maciel, I.J., Barcellos, C., Souza, W.V., Carvalho, M.S., Nascimento, N.E., Oliveira, R.M., Morais-Neto, O. & Martelli, C.M.T. (2008).

Spatial point analysis based on dengue on dengue surveys at household level in central Brazil. *BioMed Central Public Health*, 8, 361.

Sirisena, P.D.N.N., Noordeen, F., Kurukulasuriya, H., Romesh, T.A.L.A.R. & Fernando, L.K. (2017). Effect of climatic factors and population density on the distribution of dengue in Sri Lanka: A GIS based evaluation for prediction of outbreaks. *Public Library of Science one*, 12(1), e0166806.

Sittisombut, N., Sistayanarain, A., Cardoso, M.J., Salminen, M., Damrongdachakul, S., Kalayanaroj, S., Rojanasuphot, S., Supawadee, J. & Maneekam, N. (1997). Possible occurrence of a genetic bottleneck in dengue serotype 2 viruses between the 1980 and 1987 epidemic seasons in Bangkok, Thailand. *The American Journal of Tropical Medicine and Hygiene*, 57, 100-108.

Skae, F.M.T. (1902). Dengue fever in Penang. *The British Medical Journal*, 2(2185), 1581-1582.

Sriprom, M., Chalvet-Monfray, K., Chaimane, T., Vongsawat, K. & Bicout, D.J. (2010). Monthly district level risk of dengue occurrences in Sakon Nakhon Province, Thailand. *Science Total Environment*, 408(22), 5521-5528.

Standish, K., Kuan, G., Avilés, W., Balmaseda, A. & Harris, E. (2010). High dengue case capture rate in four years of cohort study in Nicaragua compared to National Surveillance data. *Public Library of Science Neglected Tropical Diseases*, 4(3), e633.

Stewart-Ibarra, A. & Lowe, R. (2013). Climate and non-climate drivers of dengue epidemics in southern coastal Ecuador. *The American Society of Tropical Medicine and Hygiene*, 88(5), 971-981.

Stewart-Ibarra, A.M., Muñoz, A.G., Ryan, S.J., Ayala, E.B., Borbor-Cordova, M.J., Finkelstein, J.L., Mejia, R., Ordoñez, T., Recalde-Coronel, G.C. & Rivero, K. (2014). Spatiotemporal clustering, climate periodicity and social-ecological risk factors for dengue during an outbreak in Machala, Ecuador, in 2010. *BioMed Central Infectious Diseases*, 14(1), 610.

- Suhaila, J., Deni, S.M., Zin, W.Z.W. & Jemain, A.A. (2010). Trends in Peninsular Malaysia rainfall data during the southeast monsoon and northeast monsoons seasons: 1975-2004. *Sains Malaysiana*, 39(4), 533-542.
- Suhaili, M.R., Hosein, E., Mokhtar, Z., Ali, N., Palmer, K. & Md. Isa, M. (2004). Applying Communication-for-Behavioral-Impact (COMBI) in the prevention and control of dengue in Johor Bahru, Johore, Malaysia. *Dengue Bulletin*, 28(SUPPLEMENT), 39-43.
- Sumanasinghe, N., Mikler, A., Tiwari, C. & Muthukudage, J. (2016). Geo-statistical dengue risk model using GIS techniques to identify the risk prone areas by linking rainfall and population density factors in Sri Lanka. *Ceylon Journal of Science*, 45(3), 39-46.
- Syafrina, A.H., Zalina, M.D. & Juneng, L. (2015). Historical trend of hourly extreme rainfall in Peninsular Malaysia. *Theoretical and Applied Climatology*, 120(1), 259-285.
- Taksande, A. & Lakhkar, B. (2013). Knowledge, attitude and practice (KAP) of dengue fever in the rural area of Central India. *Shiraz E-Medical Journal*, 13(4), 146-157.
- Tantawichien, T. (2012). Dengue fever and dengue haemorrhagic fever in adolescents and adults. *Paediatrics and International Child Health*, 32(s1), 22-27.
- Teixeira, T.T.D.A & Cruz, O.G. (2011). Spatial modeling of dengue and socio-environmental indicators in the city of Rio de Janeiro, Brazil. *Cadernos de Saúde Pública*, 27(3), 591-602.
- Telle, O., Vaguet, A., Yadav, N.K., Lefebvre, B., Daudé, E., Paul, R.E., Cebeillac, A. & Nagpal, B.N. (2016). The spread of dengue in an endemic urban milieu- the case of Delhi, India. *Public Library of Science One*, 11(1), e0146539.
- Teurlai, M., Menkès, C.E., Cavarero, V., Degallier, N., Descloux, E., Grangeon, J.P., Guillaumot, L., Libourel, T., Lucio, P.S., Mathieu-Daudé, F. & Mangeas, M. (2015). Socio-economic and climate factors associated with dengue fever spatial heterogeneity: A worked example in New Caledonia. *Public Library of Science Neglected Tropical Diseases*, 9(12), e0004211.

- Thammapalo, S., Chongsuvivatwong, V., Geater, A. & Dueravee, M. (2008). Environmental factors and incidence of dengue fever and dengue haemorrhagic fever in an urban area, Southern Thailand. *Epidemiology and Infection*, 136(1), 135-143.
- Tiensuwan, M. & O'Brien, T. (2013). Modeling dengue virus infection patients for each severity of dengue disease in Thailand. *Far East Journal of Mathematical Science*, 1, 1-20.
- Torres, C., Barquil, S., Melgarejo, M. & Olarte, A. (2014). Fuzzy model identification of dengue epidemic in Colombia based on multiresolution analysis. *Artificial Intelligence in Medicine*, 60(1), 41-51.
- Torres, J.R., Orduna, T.A., Piña-Pozas, M., Vázquez-Vega, D. & Sarti, E. (2017). Epidemiological characteristics of dengue disease in Latin America and in the Caribbean: a systematic review of the literature. *Journal of Tropical Medicine*, 2017, 1-18.
- Turkington, C. & Ashby, B. (2007). *The Encycloëdia of Infectious Diseases*. 3rd ed. New York: Infobase Publishing.
- Wade, G. (2009). The origins and evolution of ethnocracy in Malaysia. *The Asia-Pacific Journal*, 7(47), 3-11.
- Wahba, G. (1985). A comparison of GCV and GML for choosing the smoothness parameter in the Generalized Spline Smoothing Problem. *The Annals of Statistics*, 13, 1378-1402.
- Wallace, H.G., Lim, T.W., Rudnick, A., Knudsen, A.B., Cheong, W.H. & Chew, V. (1980). Dengue hemorrhagic fever in Malaysia: the 1973 epidemic. *Southeast Asian Journal Tropical Medicine Public Health*, 11(1), 1-13.
- Wan Fairos, W.Y., Wan Azaki, W.H., Mohamad Alias, L. & Bee Wah, Y. (2010). Modelling dengue fever (DF) and dengue haemorrhagic fever (DHF) outbreak using Poisson and Negative Binomial model. *International Journal of Mathematical, Computational, Physical, Electrical and Computer Engineering*, 4(2), 1-6.

- Wang, S., Geng, Z., Zhang, J., Chen, Y. & Wang, J. (2014). A fuzzy c-means model based on the spatial structural information for brain MRI segmentation. *International Journal of Signal Processing, Image Processing and Pattern Recognition*, 7(1), 313-322.
- Wilder-Smith, A. (2012). Dengue infections in travelers. *Paediatrics and International Child Health*, 32(s1), 28-32.
- Wong, C.L., Liew, J., Yusop, Z., Ismail, T., Venneker, R. & Uhlenbrook, S. (2016). Rainfall characteristics and regionalization in Peninsular Malaysia based on a high resolution gridded data set. *Water*, 8(11), 500.
- Wong, C.L., Venneker, R., Uhlenbrook, S., Jamil, A.B.M. & Zhou, Y. (2009). Variability of rainfall in Peninsular Malaysia. *Hydrology and Earth System Sciences Discussions*, 6(4), 5471-5503.
- World Health Organization (2015). *Dengue control*. Retrieved October 18, 2016, from <http://www.who.int/denguecontrol/epidemiology>
- Wu, P.C., Guo, H.R., Lung, S.C., Lin, C.Y. & Su, H.J. (2007). Weather as an effective predictor for occurrence of dengue fever in Taiwan. *Acta Tropica*, 103, 50-57.
- Xiang, J., Hanse, A., Liu, Q., Liu, X., Tong, M.X., Sun, Y., Cameron, S., Hanson-Easey, S., Han, G.S., Williams, C., Weinstein, P. & Bi, P. (2017). Association between dengue fever incidence and meteorological factors in Guangzhou, China, 2005-2014. *Environmental Research*, 153, 17-26.
- Xie, X.L. & Beni, G. (1991). A validity measure for clustering. *Institute of Electrical and Electronics Engineers (IEEE) on Pattern Analysis and Machine Intelligence*, 13(4), 841-846.
- Xu, L., Stige, L.C., Chan, K.S., Zhou, J., Yang, J., Sang, S., Wang, M., Yang, Z., Yan, Z., Jiang, T., Lu, L., Yue, Y., Liu, X., Lin, H., Xu, J., Liu, Q. & Stenseth, N.C. (2017). Climate variation drives dengue dynamics. *Proceedings of the National Academy of Sciences of the United States of America*, 114(1), 113-118.
- Yung, C.F., Lee, K.S., Thein, T.L., Tan, L.K., Gan, V.C., Wong, J.G.X., Lye, D.C., Ng, L.C. & Leo, Y.S. (2015). Dengue serotypes-specific differences in clinical

manifestations, laboratory parameters and risk of severe disease in adults, Singapore. *The American Journal of Tropical Medicine and Hygiene*, 92(5), 999-1005.

Zhang, Y., Wang, T., Liu, K., Xia, Y., Lu, Y., Jing, Q., Yang, Z., Hu, W. & Lu, J. (2016). Developing a time series predictive model for dengue in Zhongshan, China based on weather and Guangzhou dengue surveillance data. *Public Library of Science Neglected Tropical Diseases*, 10(2), e0004473.



PTTHM
PERPUSTAKAAN TUNKU TUN AMINAH