

**A STUDY ON THE CHARACTERISTICS OF RAINFALL DATA AND ITS  
PARAMETER ESTIMATES**

**JAYANTI A/P ARUMUGAM**

**UNIVERSITI TEKNOLOGI MALAYSIA**

A STUDY ON THE CHARACTERISTICS OF RAINFALL DATA AND ITS  
PARAMETER ESTIMATES

JAYANTI A/P ARUMUGAM

A dissertation submitted in fulfillment of the  
requirements for the award of the degree of  
Master of Science (Mathematics)

Faculty of Science  
Universiti Teknologi Malaysia

JANUARY 2013

Especially dedicated to my parents

Arumugam Munusamy

&

Sukla Arumugam,

and my sister, Zena

for their continuous support and encouragement

at all times,

also to

all who have been a source of guidance and inspiration to me

throughout this journey.

## ACKNOWLEDGEMENT

First and foremost I would like to thank the Almighty for always guiding me in the right direction and fulfilling my ardent desire to complete this wonderful and educational journey smoothly.

I would also like to express my sincere appreciation and gratitude to my supervisor, Dr. Shariffah Suhaila Syed Jamaludin, for the continuous encouragement, guidance, and critics throughout the journey towards the completion of my thesis. Without whose continued support and interest, this thesis would not have been the same as presented here. For patiently guiding me to complete this thesis smoothly and successfully I shall always be grateful to Dr. Shariffah.

Last, but not the least I would also like to thank my parents and sister from the bottom of my heart for all their encouragement throughout this journey and for just being there when I needed them during the difficult times. Without them this journey would not have been as enjoyable as it has been for me.

## ABSTRACT

The modeling of the rainfall process has been of interest in simulation studies to assess its impact in the fields of agriculture, water management and others. The rainfall data series used has been obtained from the Malaysian Drainage and Irrigation Department for a 33 year period from 1975 to 2007. The purpose of this study is to investigate the Tweedie family of distributions and determine the appropriate distribution to model the rainfall data using the parameter estimated. Based on the estimated parameter, it was suggested that the Gamma distribution which is a special case of the Tweedie family of distributions is suitable to model the rainfall data instead of the Poisson-gamma distribution. Consequently, a combination of a first order Markov chain and gamma distribution function is identified to model the rainfall process of occurrence and amount separately on a monthly timescale for the ten selected rain gauge stations across Peninsular Malaysia. These model parameter estimates were obtained using the method of maximum likelihood. Conversely, during the estimation of these model parameters certain general characteristics were revealed. Firstly, the transitional probability of a wet day to a wet day was higher but parallel to the transition from a dry day to a wet day. This characteristic revealed the linear relationship between the transitional probabilities and the monthly fraction of wet days. Secondly, the  $\beta$  parameter in the gamma distribution functions used to describe the amount of rainfall, is related to the monthly amount of rain per wet day. Therefore, a short method is proposed using the regression technique to estimate the model parameters from these monthly summaries. The relative error analysis revealed that there was no significant difference between the long and short method parameter estimates. Hence, this short method would be very useful in cases where there is a lack of detailed daily rainfall data available.

## ABSTRAK

Kajian simulasi amat menitikberatkan pemodelan proses hujan bagi menilai kesannya terhadap bidang pertanian, pengurusan air dan lain-lain. Rekod data curahan hujan sepanjang tempoh 33 tahun dari tahun 1975 hingga 2007 yang diperolehi daripada Jabatan Pengairan dan Saliran Malaysia digunakan dalam kajian ini. Tujuan kajian ini adalah untuk menyiasat taburan keluarga Tweedie bagi menentukan taburan yang paling sesuai bagi memodelkan data curahan hujan menerusi anggaran parameter taburan. Berdasarkan parameter yang dianggarkan, didapati bahawa taburan gamma yang merupakan kes khas taburan keluarga Tweedie merupakan taburan paling sesuai bagi memodelkan data curahan hujan berbanding taburan Poisson-gamma. Maka, gabungan rantai Markov order satu dan taburan gamma dikenal pasti bagi memodelkan proses hujan iaitu kejadian hujan serta jumlah hujan secara berasingan untuk setiap bulan bagi sepuluh stesen curahan hujan yang dipilih dari seluruh semenanjung Malaysia. Kaedah “maximum likelihood” digunakan bagi memperolehi anggaran parameter model. Proses anggaran parameter model ini telah mendedahkan beberapa ciri-ciri umum. Pertama sekali, didapati bahawa kebarangkalian peralihan dari hari hujan ke hari hujan lebih tinggi tetapi selari dengan kebarangkalian peralihan dari hari kering ke hari hujan. Daripada ciri ini telah dapat disimpulkan bahawa wujudnya hubungan linear diantara anggaran kebarangkalian peralihan dan pecahan hari hujan bulanan. Seterusnya, parameter  $\beta$  daripada taburan gamma yang menggambarkan jumlah hujan, mempunyai hubung kait dengan min jumlah hujan pada hari-hari yang hujan. Maka, dengan mengaplikasikan teknik regression, satu kaedah pendek dicadangkan bagi menganggarkan parameter model dari statistik data curahan bulanan dan bukannya data curahan harian. Selepas itu, ralat relatif bagi anggaran parameter menggunakan kaedah panjang serta pendek dikira, membuktikan bahawa tidak terdapat perbezaan yang signifikan diantara kedua kaedah tersebut. Maka secara umumnya, kaedah pendek ini amat berguna bagi kes dimana data harian curahan didapati tidak lengkap.

**TABLE OF CONTENTS**

<b>CHAPTER</b>	<b>TITLE</b>	<b>PAGE</b>
	<b>DECLARATION</b>	ii
	<b>DEDICATION</b>	iii
	<b>ACKNOWLEDGEMENT</b>	iv
	<b>ABSTRACT</b>	v
	<b>ABSTRAK</b>	vi
	<b>TABLE OF CONTENTS</b>	vii
	<b>LIST OF TABLES</b>	x
	<b>LIST OF FIGURES</b>	xii
	<b>LIST OF ABBREVIATIONS</b>	xiv
	<b>LIST OF SYMBOLS</b>	xv
	<b>LIST OF APPENDICES</b>	xvi
<b>1</b>	<b>INTRODUCTION</b>	
	1.1 Introduction	1
	1.2 Background of Study	2
	1.3 Statement of Problem	4
	1.4 Objectives of The Study	5
	1.5 Scope of Study	6
	1.6 Significance of The Study	6

<b>2</b>	<b>LITERATURE REVIEW</b>	
2.1	Introduction	8
2.2	Overview on Rainfall Modeling	9
2.2.1	Modeling rainfall Occurrence	10
2.2.2	Modeling Rainfall Amount	12
2.2.3	Modeling Rainfall Occurrence and Amount	13
2.3	Tweedie family of Distributions	16
2.4	Applications of rainfall Modeling	17
<b>3</b>	<b>RESEARCH METHODOLOGY</b>	
3.1	Introduction	20
3.2	Data	21
3.3	Descriptive Statistics	23
3.4	Types of Distributions	26
3.4.1	The Tweedie family of Distributions	26
3.4.2	Gamma Distribution	28
3.4.3	Markov Chain Model	30
3.4.3.1	Discreet Time Markov Chain	30
3.4.3.2	First Order Markov Chain	31
3.4.4	Regression Analysis	33
3.4.4.1	Simple Linear Regression	34
3.5	Parameter Estimation	35
3.5.1	Maximum likelihood Estimate for Tweedie Distribution	35
3.5.2	Maximum likelihood Estimate for Gamma Distribution	36
3.6	Research Framework	39
3.7	Gantt Chart	40



<b>4</b>	<b>RESULTS AND DISCUSSIONS</b>	
4.1	Introduction	41
4.2	Summary Statistics by Station	41
4.3	Tweedie Family of Distributions parameter estimates	47
4.4	<b>Parameter estimates of Gamma Distribution</b>	50
4.5	<b>Transition Probability Estimates for First order Markov chain Model</b>	55
4.6	Proposed Short Method Using Regression Analysis	62
4.6.1	Estimation of Transition Probabilities	62
4.6.2	Estimation of Alpha and Beta parameters	69
4.7	Conclusion	76
<b>5</b>	<b>CONCLUSIONS AND RECOMMENDATIONS</b>	
5.1	Introduction	77
5.2	Conclusions	77
5.3	Recommendations	81
	<b>REFERENCES</b>	82 - 86
	<b>APPENDICES</b>	87 - 99

## LIST OF TABLES

<b>TABLE NO.</b>	<b>TITLE</b>	<b>PAGE</b>
<b>3.1</b>	The 10 chosen rain gauge stations with their latitude and longitude	21
<b>4.1</b>	The names of the stations and descriptive statistics	44
<b>4.2</b>	Daily Rainfall Summary of statistics	45
<b>4.3</b>	The corresponding distributions based on the index parameter, $p$	47
<b>4.4</b>	Parameter estimates and the values of log likelihood for the Gamma distribution of station E06 in the eastern region	51
<b>4.5</b>	Parameter estimates and the values of log likelihood for the Gamma distribution of station NW03 in the northwest region	52
<b>4.6</b>	Parameter estimates and the values of log likelihood for the Gamma distribution of station SW02 in the southwest region	53
<b>4.7</b>	Parameter estimates and the values of log likelihood for the Gamma distribution of station W04 in the western region	54
<b>4.8</b>	The frequency distribution for the first order transition counts for the month of January at station E06	56
<b>4.9</b>	The maximum likelihood estimates for the transition probabilities for station E06 in the eastern region	57
<b>4.10</b>	The maximum likelihood estimates for the transition probabilities for station NW03 in the northwest region	58

<b>4.11</b>	The maximum likelihood estimates for the transition probabilities for station SW01 in the southwest region	59
<b>4.12</b>	The maximum likelihood estimates for the transition probabilities for station W14 in the western region	60
<b>4.13</b>	Parameter estimates for the simple linear regression model of regressing the transitional probabilities of a dry day to a wet day on the fractions of monthly wet days	66
<b>4.14</b>	Relative error calculated for the transition probabilities estimated using the short method	69
<b>4.15</b>	Parameter estimates for the simple linear regression model of regressing the beta parameter on the monthly amount of rain (mm) per wet day	72
<b>4.16</b>	Relative error for the $\alpha$ and $\beta$ parameter estimated using the short method	75

## LIST OF FIGURES

<b>FIGURE NO.</b>	<b>TITLE</b>	<b>PAGE</b>
3.1	Locations of the 10 rain gauge stations that have been selected across Peninsular Malaysia	22
3.2	Gamma distribution frequency distributions for different values of the shape parameter, $\alpha$ when scale parameter, $\beta$ , is kept constant at 1 (left) and for values of the scale parameter, $\beta$ , when shape parameter $\alpha$ is kept constant at 2 (right).	29
4.1	The map of Peninsular Malaysia and the monthly mean rainfall for all the selected stations.	42
4.2	The monthly index parameter estimates obtained for the stations on the East, Northwest and Southwest region	48
4.3	shows the index parameter estimates obtained for the stations on the Western regions	49
4.4	Transitional probabilities and the fractions of wet days for each month in Padang Katong, Kangar station	63
4.5	Transitional probabilities and the fractions of wet days for each month in Malacca station	64
4.6	Transitional probabilities of a dry day to a wet day of the fractions of wet days for selected 10 stations	65
4.7	Transitional probability of a dry day to a wet day against monthly fraction of wet days	67
4.8	Transitional probability of a wet day to a wet day against monthly fraction of wet days	67

<b>4.9</b>	Alpha and beta parameters and the amount of rainfall per wet day at the Dispensari Kroh station	70
<b>4.10</b>	Alpha and beta parameters and the amount of rainfall per wet day at Gua Musang station	71
<b>4.11</b>	Beta parameters and the amount of rainfall per wet day for all ten stations	72
<b>4.12</b>	Alpha estimates using short method against monthly amount of rain per wet day	74
<b>4.13</b>	Beta estimates using short method against monthly amount of rain per wet day	74

**LIST OF ABBREVIATIONS**

CV	-	Coefficient of Variation
AIC	-	Akaike's Information Criterion
BIC	-	Bayesian Information Criterion
CGF	-	Cumulant Distribution Function
NSRP	-	Neyman–Scott Rectangular Pulses
CERES	-	Crop Estimation through Resource and Environmental Synthesis
LAD	-	Least Absolute Deviation
SPI	-	Standardized Precipitation Index
NDVI	-	Normalized Difference Vegetation Index
EDM	-	Exponential Dispersion Models
GLM	-	Generalized Linear Models
MLE	-	Maximum Likelihood Estimate
pdf	-	Probability density function

**LIST OF SYMBOLS**

$\mu$	–	Mean
$\sigma$	–	Standard Deviation
$k$	–	Kurtosis
$p$	–	Index parameter
$\alpha$	–	Shape parameter for gamma distribution
$\beta$	–	Scale parameter for gamma distribution
$p_{00}$	–	Transition probability from a dry day to a dry day
$P(W D), p_{01}$	–	Transition probability from a dry day to a wet day
$p_{10}$	–	Transition probability from a wet day to a dry day
$P(W W), p_{11}$	–	Transition probability from a wet day to a wet day
$a$	–	Intercept parameter for simple linear regression method
$b$	–	Slope parameter for simple linear regression method

**LIST OF APPENDICES**

<b>APPENDIX</b>	<b>TITLE</b>	<b>PAGE</b>
A	Transition probabilities estimates for first order Markov chain model	87
B	Parameter estimates and the values of log likelihood for the Gamma distribution	90
C	Error obtained for the transition probability estimated using the short method by stations	93
D	Error obtained for the gamma distribution parameters estimated using the short method by stations	95
E	R Programming scripts	97



## **CHAPTER 1**

### **INTRODUCTION**

#### **1.1 Introduction**

Malaysia is a Southeast Asian country straddling the South China Sea. Due to Malaysia's proximity to the equator, it has a tropical rainforest climate. Thus, Peninsular Malaysia experiences summer all year round including two monsoon seasons which largely contribute to the amount of rainfall observed in Peninsular Malaysia. The country experiences the Northeast monsoon during November to February and the Southwest monsoon in the months from May till August. The Northeast monsoon brings with it heavy rain spells along the east coast of Peninsular Malaysia while the Southwest monsoon leaves most parts of the Peninsula dry.

Due to Malaysia's tropical rainforest climate there isn't an extremely dry season and every month has a mean precipitation amount of at least 60 millimeters. It is suggested that rainfall is frequent in this region. Therefore, it is necessary to study the rainfall process by modeling the rainfall data. Rainfall modeling has been widely studied in various disciplines such as climatology, meteorology and hydrology towards facilitating a better understanding of rainfall pattern and its characteristics. The models obtained have been useful in various fields. For example, water resource planning, water management and, agricultural planning. It has also been studied to aid in drought and flood prediction, rainfall runoff modeling, crop growth studies, impact of climate change

studies and become beneficial to numerous other fields of study. Thus, rainfall modeling plays a crucial role in many fields that are either directly or indirectly impacted by the rainfall process.

The rainfall data is a unique data series since it consist of zeroes and non-zero values. Zero values indicate non-rainy days while the non-zero values represent the rainy days. Due to the characteristic of the rainfall data, two different models have been used in modeling rainfall data separately. The first model is denoted as rainfall occurrences which are used to model the sequence of wet and dry days while the rainfall amount is used to describe the amount of rainfall observed during a wet day. Recently it was discovered that the Tweedie family distribution which contains the Poisson-gamma distribution with a fixed range of the index parameter value, could model the rainfall process of occurrence and amount simultaneously. Hence this study will investigate the Tweedie family distributions to identify the best distribution that can be used to model the monthly rainfall data series at a few selected rain gauge stations in Peninsular Malaysia.

## **1.2 Background of Study**

The modeling of monthly rainfall data can be performed via two known approaches firstly the modeling of rainfall occurrence and rainfall amount simultaneously. Secondly the modeling of rainfall occurrence and rainfall amount using two separate models for each process. Modeling rainfall occurrence and amount can be carried out on numerous time scales ranging from hourly to daily to monthly or even seasonally based on the rainfall data in hand. In this study, the approach shall be to model rainfall occurrence and amount on the monthly timescale at selected rain gauge stations in Peninsular Malaysia.

Commonly, there have been numerous studies to model rainfall occurrence (Gabriel and Neumann, 1962; Stern and Coe, 1984; Deni et al., 2009) and there have also been various studies on modeling rainfall amount (Aksoy, 2000; Srikanthan and Pegram, 2009; Hussain et al., 2010). Then there have been research on the modeling of both rainfall occurrence and also amount using separate models. And in the recent years studies have suggested the modeling the rainfall process of occurrence and amount simultaneously using the Poisson-gamma distribution which belongs to the Tweedie family distributions (Dunn, 2004).

Accordingly the Poisson-gamma distribution that belongs to the Tweedie family distributions has been capable of modeling rainfall occurrence and amount simultaneously when the index parameter estimate for the rainfall data analyzed fell between (1,2) with a 95 percent confidence interval. Exploring the Tweedie family of distributions in studies has been due to its unique characteristics of assigning the appropriate distribution based on the value of its index parameter estimate. This special characteristic makes the Tweedie family of distributions interesting to be studied in order to obtain the parameter estimate and determine the appropriate distribution suitable to model the rainfall data. The Tweedie distribution contains the Poisson-gamma (Tweedie) distribution and also a few distributions that are known as its special cases.

The special cases of the Tweedie family of distributions are dependent on the value of the index parameter estimate of the data series. Hence these special cases include the Normal distribution (index parameter equals to 0), Poisson distribution (index parameter equals to 1), Gamma distribution (index parameter equals or greater than 2), the inverse Gaussian distribution (index parameter equals to 3) and the Tweedie distribution better known as the Poisson-gamma distribution (index parameter values between 1 and 2). Meanwhile there isn't any distribution that exists for index parameter values between 0 and 1 (Jorgensen, 1987).

Consequently, the concept of this study is to investigate the rainfall data through the estimation of the index parameter for the Tweedie family distributions. Followed by the fitting the rainfall data to an appropriate distribution within the Tweedie family of distributions based on the index parameter estimates obtained. Thus in the nutshell, this study shall model the rainfall process of occurrence and amount for the rainfall data. So the study will investigate the characteristic of the rainfall data series in order to determine the type of distribution from the Tweedie family distributions which is suitable to model the rainfall data based on the index parameter estimate obtained.

Conversely, during the evaluation of the model parameters certain general characteristics were revealed. Firstly the transitional probability of a wet day to a wet day is higher but parallel to the transition from a dry day to a wet day. This lead to a linear relationship discovered between the transitional probabilities and the monthly fraction of wet days. Secondly, the  $\beta$  parameter in the gamma distribution functions used to describe the amount of rainfall, is related to the monthly amount of rain per wet day. Therefore, a short method is proposed using the regression technique to estimate the model parameters from these monthly summaries.

### **1.3 Statement of Problem**

The Tweedie family of distributions contains the Poisson-gamma distribution that is able to model rainfall occurrence and amount simultaneously. In recent years, the Poisson-gamma or Tweedie distribution has been used to model the rainfall process successfully in various studies (Dunn, 2004). However the suitability of the application of the Poisson-gamma distribution to the Malaysia rainfall data series is yet to be studied in detail. Thus, the purpose of this study is to investigate the Tweedie family of distributions to determine if the rainfall data being studied can be modeled using the Poisson-gamma (Tweedie) distribution contained within. Hence, the estimation of the index parameter will allow for the identification of the suitable distribution from the

Tweedie family of distributions that can be used to model the monthly rainfall data. So we can determine based on the index parameter estimate whether the Poisson-gamma (Tweedie) distribution can fit the rainfall data.

Since one of the loopholes on modeling rainfall data is the unavailability of long series of historical daily rainfall data thus a method to estimate the parameter estimates using the monthly averages is very much beneficial to research on rainfall process modeling. Hence, based on certain general characteristics that appear to exist between model parameters and the monthly statistics of the rainfall data reveals a simple method for estimating model parameters. This proposed simple method will be known as the short method. The short method will be beneficial to those areas with restricted historical rainfall data series.

#### **1.4 Research Objectives**

The objectives of this study are :

1. To determine the characteristics of rainfall occurrence and rainfall amount.
2. To model the rainfall data series using the appropriate distribution from the Tweedie family of distributions
3. To propose an alternative method to obtain the model parameter of rainfall occurrence and amount using the technique of regression.

## **1.5 Scope of Study**

This study will model the rainfall occurrence and amount for the rainfall data being studied. Thus the study will investigate the rainfall data to determine the index parameter estimate so as to determine the appropriate distribution that fits the data from the Tweedie family of distributions. Also the analysis will reveal whether the rainfall data can be modeled using the Poisson-gamma (Tweedie) distribution in the Tweedie family of distributions.

Monthly rainfall data from 10 selected rain gauge stations across Peninsular Malaysia over a period of 33 years (1975-2007) will be used in this study to obtain an appropriate monthly rainfall model at each station for rainfall occurrence and rainfall amount. A threshold value of 1mm shall be used in the study to categorize a dry or wet day as any amount of rainfall below the threshold value will be classified as a dry day with no rainfall and vice versa. This study shall identify a rainfall model, based on the Tweedie family of distribution that shall be used to model the monthly rainfall process to resemble the actual rainfall process as closely as possible.

## **1.6 Significance of the Study**

In this study the rainfall process shall be modeled based on the appropriate distribution identified after the parameter estimate is obtained from the Tweedie family of distributions. Thus rainfall occurrence and rainfall amount shall be modeled simultaneously if the parameter estimate obtained belongs to the Poisson-gamma (Tweedie) distribution otherwise rainfall occurrence and amount shall be modeled separately using appropriate distributions. Hence through this study the appropriateness of modeling the rainfall data using the Tweedie distribution shall be investigated. Besides that through this study a comprehensive model for rainfall occurrence and amount shall be obtained.

The short method proposed using the linear relationship between the model parameters and the monthly summaries of the rainfall data is a method that can aid in the modeling of rainfall data where there are scarce resources of long series of daily rainfall data. Besides that, through this short method weather simulation models can be applied to those areas that were previously untouchable as there weren't enough rainfall data series to obtain the model parameter and consequently obtain the model and apply these results on simulation studies.

Water resources around the globe are being severely depleted in recent years due to climate change, population growth, pollution and so on. Thus being a form of freshwater, rainwater replenishes these depleting water resources around the globe that can be harvested for use in many ways. Hence rainfall modeling is a part of the water resource management initiative (Al-Qinna et al., 2011), can aid in water cycle planning and management (Wilks, 1999). Besides that, modeling has been known to have frequent inputs for agriculture (Semenov and Porter, 1995), hydrological models and is also essential to climate change studies (Wilks, 1999; Fowler et al., 2005) and it can also provide clues towards discerning climate change (Brissette et al., 2007) at the present moment and also for future events. Hence, this study will contribute to these fields in the future.

## REFERENCES

- Aksoy, H. (2000). Use of gamma distribution in hydrological analysis. *Turkish Journal of Engineering Environmental Science*, 24, 419–428.
- Allan, D. M., Haan, C. T., (1975). *Stochastic Simulation of Daily Rainfall*. Technical Report 82. Water Resources Institute, University of Kentucky.
- Al-Qinna, M. I., Nezar, A. H., Mutewekil M.O. and Fayez Y.A. (2011). Drought analysis in Jordan under current and future climates. *Climatic Change*. 106,421-440.
- Brissette, F., Khalili M. and Leconte, R. (2007). Efficient stochastic generation of multi-site synthetic precipitation data. *Journal Hydrology*. 345(3–4),121–133.
- Buishand, T. A., Shabalova, M. V. and Brandsma, T. (2004). On the choice of the temporal aggregation level for statistical downscaling of precipitation. *Journal of Climate*. 17, 1816-1827.
- Chandler, R. E. and Wheeler, H. S. (2002). Analysis of rainfall variability using generalized linear models: a case study from the west of Ireland. *Water Resources Research*. 38(10), 1192–1202.
- Coe, R. and Stern, R. D. (1982). The use of rainfall models in agricultural planning. *Journal of Applied Meteorology*. 21, 1024-1031.
- Coe, R. and Stern, R. D. (1984). A model fitting analysis of daily rainfall data. *Journal of Royal Statistical. Society Series A*. 147,Part1,1-34.
- DeCarlo, L. T. (1997). Signal detection theory and generalized linear models. *Psychological Methods*. 3, 186-205.
- Deni, S. M., Jemain, A. A. and Ibrahim K. (2009). Fitting optimum order Markov chain models for daily rainfall occurrences in peninsular Malaysia. *Theoretical and Applied Climatology*. 97, 109–121.



- Dobson, A. J. (2002). *An Introduction to Generalized Linear Models*. (3<sup>rd</sup> edition). Chapman and Hall, London.
- Dunn P. K. (2001). *Likelihood-based inference for tweedie exponential dispersion models*. Unpublished PhD Thesis, University of Queensland.
- Dunn, P. K. (2004). Occurrence and quantity of precipitation can be modelled simultaneously. *International Journal of Climatology*. 24, 1231–1239.
- Dunn, P. K. (2009). *Tweedie: Tweedie Exponential Family Models*. R Package, Vienna, Austria, R Package Version 2.0.2.
- Dunn, P. K. and White, N. (2005). Power-variance models for modelling rainfall. *Proceedings of the 20th International Workshop on Statistical Modelling*. 10-15 July 2005. Sydney, Australia.
- Dunn, P. K. and Smyth, G. K. (2001). Tweedie family densities: methods of evaluation. *Proceedings of the 16<sup>th</sup> International Workshop on Statistical Modelling*, 2-6<sup>th</sup> July Odessa, Denmark.
- Dunn, P. K. and Smyth, G. K. (2005). Series evaluation of Tweedie exponential dispersion model densities. *Statistics and Computing*. 15, 267–280.
- Dunn, P. K. and Smyth, G. K. (2008). Evaluation of Tweedie exponential dispersion model densities by Fourier inversion. *Statistics and Computing*, 18, 73–86.
- Durban, M. and Glasbey, C. A. (2001). Weather modelling using a multivariate latent Gaussian model. *Agricultural and Forest Meteorology*. 109, 187-201.
- Fowler, H. J., Kilsby, C. G., O’Connell, P. E. and Burton A. (2000). A stochastic rainfall model for the assessment of regional water resource systems under changed climatic conditions. *Hydrology and Earth System Sciences*. 4, 261-280.
- Fowler, H. J., Kilsby, C. G., O’Connell, P. E. and Burton A. (2005). A weather-type conditioned multi-site stochastic rainfall model for the generation of scenarios of climatic variability and change. *Journal of Hydrology*. 308, 50-56.
- Gabriel, K. R. and Neumann, J. (1962). A Markov chain model for daily rainfall occurrences at Tel Aviv. *Quarterly Journal of Royal Meteorological Society* 88, 90-95.
- Gates, P. and Tong, H. (1976). On Markov Chain modeling to some weather data. *Journal of Applied Meteorology*. 15, 1145-1151.

- Geng, S., Penning de Vries, F.W.T. and Supit, I., (1986). A simple method for generating rainfall data. *Agricultural and Forest Meteorology*. 36, 363-376.
- Glasbey, C. A. and Nevison, I. M. (1997). *Rainfall modelling using a latent Gaussian variable*. In *Modelling Longitudinal and Spatially Correlated Data: Methods, Applications, and Future Directions*, number 122 in *Lecture Notes in Statistics*, pages 233-242. Springer, New York.
- Hamlin, M. J. and Rees, D. H. (1987). The use of rainfall forecasts in the optimal management of small-holder rice irrigation: a case study. *Hydrological Sciences*. 32 (1), 15–29.
- Hansen, J. W. and Ines, A. V. M. (2005). Stochastic disaggregation of monthly rainfall data for crop simulation studies. *Agricultural and Forest Meteorology* 131, 233–246.
- Hansen, J. W., Mishra, A., Rao, K. P. C., Indeje, M. and Ngugi, R. K. (2009). Potential value of GCM-based seasonal rainfall forecasts for maize management in semi-arid Kenya. *Agricultural Systems*. 101, 80–90.
- Hansen, J. W., Robertson, A. W. and Ines, A. V. M. (2007). Downscaling of seasonal Precipitation for Crop Simulation. *Journal of Applied Meteorology and Climatology* 46.
- Hasan, Md. M. and Dunn, P. K. (2010). A simple Poisson-Gamma model for modeling rainfall occurrence and amount simultaneously. *Agricultural and Forest Meteorology* 150, 1319-1330.
- Hussain, Z., Mahmood, Z. and Hayat, Y. (2010). Modeling the daily rainfall amounts of north-west Pakistan for agricultural planning, Sarhad. *Journal of Agriculture*. 27(2), 313-321.
- Jamaludin, S. and Jemain, A. A. (2007). Fitting Daily Rainfall Amount in Peninsular Malaysia Using Several Types of Exponential Distributions. *Journal of Applied Sciences Research*., 3(10), 1027-1036.
- Jamaludin, S. and Jemain, A. A. (2008). Fitting the statistical distribution for daily rainfall in Peninsular Malaysia based on the AIC criterion. *Journal of Applied Sciences Research*. 4, 1846-1857.
- Jørgensen B. (1987). Exponential dispersion models (with discussion). *Journal of Royal Statistical. Society Series B Statistical Methodology*. 49, 127–162.

- Lennox, S. M., Dunn, P. K., Power, B. D. and Devoil, P. (2004). A statistical distribution for modeling rainfall with promising applications in crop science. Technical report in: Fischer, T. et al.(Eds.), *New Directions for a Diverse Planet: Proceedings for the 4<sup>th</sup> International Crop Science Congress*. Brisbane, Australia.
- McCullagh, P. and Nelder, J. A. (1989). *Generalized Linear Model*. (2<sup>nd</sup> edition). Chapman and Hall, London.
- Mimikou, M. (1983). Daily precipitation occurrence modelling with Markov chain of seasonal order. *Water Resources Research*. 10(2), 223 – 236.
- model. *Journal of Hydrology*. 210, 178–191
- Nasseri, M. and Zahraie, B. (2011). Application of simple clustering on space-time mapping of mean monthly rainfall patterns. *International Journal of Climatology*. 31(5), 732-741. Published online in Wiley InterScience.
- Nnaji, A. O. (2001). Forecasting seasonal rainfall for agricultural decision-making in northern Nigeria. *Agricultural and Forest Meteorology* 107, 193-205.
- R Development Core Team (2009). *R: A Language and Environment for Statistical Computing*, R Foundation for Statistical Computing, Vienna, Austria, ISBN 3-900051-07-0, URL: <http://www.R-project.org>.
- Richardson, C.W. and Wright, D.A., (1984). WGEN: A model for generating daily weather variables. U.S. Department of Agriculture, *Agricultural Research Service, ARS-8*, 88.
- Semenov, M. A. and Porter, J. R. (1995). Climatic variability and the modelling of crop yields. *Agricultural Forest Meteorology*. 73, 265–283.
- Sharda, V. N. and Das, P. K. (2005). Modelling weekly rainfall data for crop planning in the sub-humid climate of India. *Agricultural Water Management*. 76,120-138.
- Smyth, G. K. (1996). Regression analysis of quantity data with exact zeroes. *Proceedings of the Second Australia–Japan Workshop on Stochastic Models in Engineering, Technology and Management*. Technical report. Technology Management Centre, University of Queensland, pp. 572–580.
- Smyth, G. K., with contributions from Hu, Y., Dunn, P. K. (2009). *Statmod: Statistical Modeling*. R Package Version 1.4.1.

- Srikanthan, R. and McMahon, T. A. (2001). Stochastic generation of annual, monthly and daily climate data: A review. *Hydrology and Earth System Sciences*. 5(4), 653-670.
- Srikanthan, R. and Pegram, G. G. S. (2009). A nested multisite daily rainfall stochastic generation model. *Journal of Hydrology*. 371, 142–153
- Stern, R. D. and Coe, R. (1984). A model fitting analysis of daily rainfall data (with discussion). *Journal of the Royal Statistical Society, Series A* 147, 1–34.
- Tilahun, K. (2006). The characterization of rainfall in the arid and semi-arid regions of Ethiopia. *Water South Africa*. 32, 429-436.
- Tweedie, M. C. K., (1984). An index which distinguishes between some important exponential families. Statistics: applications and new directions. *Proceedings of the Indian Statistical Institute Golden Jubilee International Conference*. Technical Report. Indian Statistical Institute, Calcutta.
- Wilks D. S. (1998). Multisite generalization of a daily stochastic precipitation generation
- Wilks, D. S. (1999). Multisite downscaling of daily precipitation with a stochastic weather generator. *Climate Research* 11:125-136.
- Wilks, D. S. and Wilby, R. L. (1999). *The weather generation game: a review of stochastic weather models*. *Progress in Physical Geography*. 23(3), 329-357.