

ESTIMATING VALUE AT RISK FOR SUKUK MARKET USING
GENERALIZED AUTOREGRESSIVE CONDITIONAL
HETEROSKEDASTICITY MODELS

PANTEA HAFEZIAN

UNIVERSITI TEKNOLOGI MALAYSIA

SELF- ESTIMATING VALUE AT RISK FOR SUKUK MARKET USING
GENERALIZED AUTOREGRESSIVE CONDITIONAL HETEROSKEDASTICITY
MODELS

PANTEA HAFEZIAN

A thesis submitted in fulfilment of the
requirements for the award of the degree of
Doctor of Philosophy (Fiqh Science and Technology)

Faculty of Islamic Civilization
Universiti Teknologi Malaysia

FEBRUARY 2017

Specially dedicated to

*My beloved mother and father who have always encouraged me
throughout my journey of education*

ACKNOWLEDGEMENT

First and foremost, I would like to express my deepest gratitude to Allah Almighty for helping me this thesis has successfully been completed through all the difficulties.

Second, I would like to express the deepest appreciation to my supervisor Dr. Hussin Salamon for his constant support during my study at UTM. He inspired me greatly to work on this project; willingness to motivate me contributed tremendously to this project. Besides, I would like to thank my co-supervisor Dr. Mahendran Shitan. I have learned a lot from him and I am fortunate to have him as my mentor and supervisor.

ABSTRACT

Islamic Finance has experienced unsurpassed growth over the past ten years. A major reason for this accelerated growth is the wide issuance of Sukuk. The structure of Sukuk are akin to conventional bonds and allow sovereign and corporate entities raising funds in capital markets in compliance with the Sharia philosophy of Islam. This study offers an overview of the Sukuk data time series, and a description of the statistical and distributional features related to its operation as a liquidity instrument in the secondary market. Subsequently a wide-ranging list comprising both symmetric and asymmetric GARCH models such as GARCH, EGARCH, GJR-GARCH, IGARCH and asymmetric power GARCH were considered for modelling the volatility of the Sukuk market. Concisely, through this investigation, the researcher determined if Sukuk are financial tools with the exact characteristics of conventional bonds, or a separate financial instrument with features of their own. The significant of this matter is linked to the fact that in several developing countries with pronounced Muslim populations, thus the study maintained that traditional debt markets cannot thrive if the availability of Sukuk is beyond reach. Ultimately, this study discovered that while an obvious theoretical disparity exists between Sukuk and conventional bonds, there are similarities in their performances where the secondary market is concerned. This study also determined that for predicting capability in the Sukuk market, the performance of asymmetric GARCH models is superior to that of symmetric models. Finally, this study demonstrated that the student-t distribution is more favorable than normal or generalized error distribution.

ABSTRAK

Kewangan Islam telah mengalami pertumbuhan yang tiada tandingannya dalam tempoh sepuluh tahun yang lalu. Sebab utama berlakunya pertumbuhan ekonomi yang pesat ialah terbitan pelbagai jenis Sukuk. Struktur Sukuk adalah serupa dengan ikatan konvensional dan merupakan entiti korporat yang diiktiraf untuk mengumpul dana dalam pasaran modal selaras dengan falsafah Syariah Islam. Kajian ini menjelaskan gambaran keseluruhan siri masa data Sukuk dan huraian ciri-ciri statistik dan pengagihan yang berkaitan dengan operasi sebagai instrumen tunai dalam pasaran sekunder. Senarai menyeluruh yang terdiri daripada kedua-dua model GARCH simetri dan tidak simetri seperti GARCH, EGARCH, GJR-GARCH, IGARCH, dan GARCH kuasa simetri digunakan untuk pemodelan turun naik pasaran Sukuk. Ringkasnya, melalui kajian ini, penyelidik menentukan adakah Sukuk merupakan instrumen kewangan dengan ciri-ciri sebenar bon konvensional, atau instrumen kewangan berasingan yang mempunyai ciri-cirinya sendiri. Kepentingan perkara ini dikaitkan dengan keadaan di beberapa negara membangun dengan populasi penduduk Islam yang tinggi, justeru, kajian ini menetapkan bahawa pasaran hutang tradisional tidak mampu berkembang maju jika kewujudan Sukuk sukar untuk dimanfaatkan. Secara muktamadnya, kajian mendapati walaupun perbezaan teori jelas wujud antara Sukuk dan ikatan konvensional, namun terdapat persamaan sumbangan antara kedua-duanya di pasaran sekunder. Pengkaji juga mendapati bahawa dari segi keupayaan meramal dalam pasaran Sukuk, prestasi model GARCH tidak simetri adalah lebih unggul daripada model simetri. Akhirnya, pengkaji telah membuktikan bahawa pengagihan pelajar-t adalah lebih baik daripada pengagihan biasa atau pengagihan ralat umum.

TABLE OF CONTENTS

CHAPTER	TITLE	PAGE
	DECLARATION	ii
	DEDICATION	iii
	ACKNOWLEDGEMENT	iv
	ABSTRACT	v
	ABSTRAK	vi
	TABLE OF CONTENTS	vii
	LIST OF TABLES	xi
	LIST OF FIGURES	xiii
	LIST OF ABBREVIATION	xv
1	INTRODUCTION	1
	1.1 Introdguction	1
	1.2 Background of Study	5
	1.3 Problem Statement	7
	1.4 Objectives of the Study	8
	1.5 Research Questions	8
	1.6 Scope of the Study	9
	1.7 Structure of Thesis	12
	1.8 Definition of Terminologies	12
	1.9 Summary	13
2	LITERATURE REVIEW	15
	2.1 Overview of Islamic Finance	15
	2.2 Characteristics of Islamic Finance	18
	2.2.1 Prohibition of Interest (Riba) and Usury	18

2.2.2	Transaction on Sinful (Haram) Activities	21
2.2.3	Prohibition of Exclusive Uncertainty (Gharar)	21
2.2.4	Profit and Loss Sharing:	23
2.2.5	Asset Baking in Islamic Finance	24
2.2.6	The structure of Islamic financial contracts	25
2.2.7	Securitization in Islam	29
2.3	Overview of Sukuk	30
2.3.1	Different Types of Sukuk	32
2.3.2	Tradable and non-Tradable Sukuk	39
2.3.3	Differences between Sukuk and conventional bonds	39
2.3.4	Risk exposures	43
2.4	Financial Time Series	45
2.4.1	Persistence	46
2.4.2	Volatility Clustering	46
2.4.3	Heavy Tail Data Distribution	47
2.4.4	Asymmetric Volatility	47
2.5	Value at risk	47
2.5.1	Mathematical concept of value at risk	49
2.5.2	Overview of VaR methodologies	50
2.6	GARCH methods	52
2.7	Previous Studies	53
2.8	Summary	60
3	METHODOLOGY AND DATA COLLECTION	61
3.1	Introduction	61
3.2	Research design	61
3.2.1	Quantitative research	61
3.2.2	Qualitative research	62
3.2.3	Mixed methods approach:	62
3.3	Region of Study	63
3.4	Data Collection	64
3.5	Data Analysis	64
3.6	Fitting a Time Series Model	65

3.6.1	Normality	65
3.6.2	Volatility Clustering and Dependency	66
3.6.3	Stationary test for Sukuk distribution:	67
3.7	Estimating VaR	68
3.7.1	GARCH (1, 1)	68
3.7.2	EGARCH (1,1)	69
3.7.3	GJR-GARCH (1,1)	69
3.7.4	IGARCH (1,1)	70
3.7.5	Asymmetric Power GARCH (1, 1)	70
3.8	Evaluation performance of GARCH methods	71
3.8.1	In-sample evaluation	71
3.8.2	Out-of-sample Evaluation Framework	72
3.9	Summary:	75
4	RESULTS AND DISCUSSIONS	76
4.1	Introduction	76
4.2	Data analysis	76
4.3	Data and Descriptive Statistics	79
4.4	Graphical analysis	80
4.5	Inferential analysis	88
4.5.1	Preliminary and Inferential analysis for Malaysia	88
4.5.2	Preliminary and Inferential analysis for IDB (Saudi)	96
4.5.3	Preliminary and Inferential analysis for UAE	102
4.5.4	Preliminary and Inferential analysis for Turkey	108
4.5.5	Preliminary and Inferential analysis for Bahrain	114
4.5.6	Preliminary and Inferential analysis for Indonesia	120
5	CONCLUSIONS	127
5.1	Introduction	127
5.2	Findings	128
5.2.1	Descriptive Findings	128

5.2.2	The Behaviour of Sukuk Returns in the Secondary Market	129
5.2.3	The best GARCH method to fit for modelling Sukuk return	130
5.3	Conclusions and Implications	131
5.4	Discussion	133
5.5	Limitations	135
5.6	Suggestions for Future Research	135
5.7	Summary	136
REFERENCES		137

LIST OF TABLES

TABLE NO.	TITLE	PAGE
1.1	Global Sukuk Issuances From 2001-2014	3
1.2	International Sukuk issuances country wise break-up (2001- July 2014)	10
3.1	Percentage Points of the Chi-Square Distribution	66
4.1	Selected issues of sukuk	78
4.2	Summary of Descriptive Statistics and Inferential Analysis of Malaysia Sukuk Returns	90
4.3	In- sample evaluation and parameters estimates of all GARCH models for Malaysian Sukuk return series, using the entire dataset and assuming three different distributions for the residuals.	92
4.4	Out of sample forecasting performance of GARCH-class models with normal, student-t and ged distributions for Malaysian Sukuk returns in 99%, 97.5% and 95% confidence interval ($\alpha= 1\%$, $\alpha= 2.5\%$ and $\alpha= 5\%$)	95
4.5	Summary of Descriptive Statistics and Inferential Analysis of IDB Sukuk Returns	97
4.6	In- sample evaluation and parameters estimates of all GARCH models for IDB Sukuk return series, using the entire dataset and assuming three different distributions for the residuals.	99
4.7	Out of sample forecasting performance of GARCH-class models with normal, student-t and ged distributions for IDB Sukuk returns in 99%, 97.5% and 95% confidence interval ($\alpha= 1\%$, $\alpha= 2.5\%$ and $\alpha= 5\%$)	101
4.8	Summary of Descriptive Statistics and Inferential Analysis of UAE Sukuk Returns	103
4.9	In- sample evaluation and parameters estimates of all GARCH models for UEA Sukuk return series, using the entire dataset and assuming three different distributions for the residuals.	105

4.10	Out of sample forecasting performance of GARCH-class models with normal, student-t and ged distributions for UAE Sukuk returns in 99%, 97.5% and 95% confidence interval ($\alpha= 1\%$, $\alpha= 2.5\%$ and $\alpha= 5\%$)	107
4.11	Summary of Descriptive Statistics and Inferential Analysis of Turkey Sukuk returns	109
4.12	In- sample evaluation and parameters estimates of all GARCH models for Turkey Sukuk return series, using the entire dataset and assuming three different distributions for the residuals.	111
4.13	Out of sample forecasting performance of GARCH-class models with normal, student-t and ged distributions for Turkey Sukuk returns in 99%, 97.5% and 95% confidence interval ($\alpha= 1\%$, $\alpha= 2.5\%$ and $\alpha= 5\%$)	113
4.14	Summary of Descriptive Statistics and Inferential Analysis of Bahrain Sukuk Returns	115
4.15	In- sample evaluation and parameters estimates of all GARCH models for Bahrain Sukuk return series, using the entire dataset and assuming three different distributions for the residuals.	117
4.16	Out of sample forecasting performance of GARCH-class models with normal, student-t and ged distributions for Bahrain Sukuk returns in 99%, 97.5% and 95% confidence interval ($\alpha= 1\%$, $\alpha= 2.5\%$ and $\alpha= 5\%$)	119
4.17	Summary of Descriptive Statistics and Inferential Analysis of Indonesia Sukuk returns	121
4.18	In- sample evaluation and parameters estimates of all GARCH models for Indonesia Sukuk return series, using the entire dataset and assuming three different distributions for the residuals.	123
4.19	Out of sample forecasting performance of GARCH-class models with normal, student-t and ged distributions for Indonesia Sukuk returns in 99%, 97.5% and 95% confidence interval ($\alpha= 1\%$, $\alpha= 2.5\%$ and $\alpha= 5\%$)	125
5.1	The Best Fitted GARCH models based on in-sample evaluation	130
5.2	The best GARCH model to forecast and estimate VaR of Sukuk	131

LIST OF FIGURES

FIGURE NO.	TITLE	PAGE
1.1	Trend of Global Sukuk Issuances (From 2001-2014)	3
1.2	International Sukuk Issuance (Jan 2001- March 2015, USD Millions)	3
1.3	Sukuk Issuance in 2014 by issuer type	4
1.4	International Sukuk issuances country wise break- up (2001-July 2014)	10
4.1	Sukuk Data Set Price Movement for Bahrain, Indonesia, Malaysia, Saudi Arabia, Turkey and UAE	80
4.2	Continuously Compounded Daily Returns of Sukuk for Bahrain, Indonesia, Malaysia, Saudi Arabia, Turkey and UAE calculated by $rt = \ln(xtxt - 1)$ where xt is the value of the sukuk at t.	81
4.3	Histogram and density plot of daily returns and the best fitted normal distribution, $N(\mu, \sigma)$ for Sukuk returns in Malaysia. The horizontal axis displays to return, while the vertical axis correspond the value of the density. The green plot indicates the normal distribution and the red one is density plot for Malaysia Sukuk returns. The reveals that the Malaysian Sukuk time series are leptokurtic which means it is heavier in the tail and more peaked around the mean.	82
4.4	:Histogram and density plot of daily returns and the best fitted normal distribution, $N(\mu, \sigma)$ for Sukuk returns in Saudi Arabia. The horizontal axis displays to return, while the vertical axis correspond the value of the density. The green plot indicates the normal distribution and the red one is density plot for Sukuk returns in Saudi Arabia. The shows that the Sukuk time series in Saudi Arabia is leptokurtic which means it is heavier in the tail and more peaked around the mean.	83
4.5	Histogram and density plot of daily returns and the best fitted normal distribution, $N(\mu, \sigma)$ for Sukuk returns in Turkey. The horizontal axis displays to return, while the vertical axis correspond the value of the density. The green plot	

- indicates the normal distribution and the red one is density plot for Sukuk returns in Turkey. The reveals that the Sukuk time series in Turkey is leptokurtic which means it is heavier in the tail and more peaked around the mean. 84
- 4.6 Histogram and density plot of daily returns and the best fitted normal distribution, $N(\mu, \sigma)$ for Sukuk returns in Bahrain. The horizontal axis displays to return, while the vertical axis correspond the value of the density. The green plot indicates the normal distribution and the red one is density plot for Sukuk returns in Bahrain. The reveals that the Sukuk time series in Bahrain is leptokurtic which means it is heavier in the tail and more peaked around the mean. 85
- 4.7 Histogram and density plot of daily returns and the best fitted normal distribution, $N(\mu, \sigma)$ for Sukuk returns in Indonesia. The horizontal axis displays to return, while the vertical axis correspond the value of the density. The green plot indicates the normal distribution and the red one is density plot for Sukuk returns in Indonesia. The reveals that the Sukuk time series in Indonesia is leptokurtic which means it is heavier in the tail and more peaked around the mean. 86
- 4.8 Histogram and density plot of daily returns and the best fitted normal distribution, $N(\mu, \sigma)$ for Sukuk returns in UAE. The horizontal axis displays to return, while the vertical axis correspond the value of the density. The green plot indicates the normal distribution and the red one is density plot for Sukuk returns in UAE. The reveals that the Sukuk time series in UAE is leptokurtic which means it is heavier in the tail and more peaked around the mean. 87
- 4.9 Scattered Plot for Sample Mean Against Standard Deviation of the Sukuk Returns in Studies Countries. 88

LIST OF ABBREVIATIONS

AAOIFI	-	Accounting and Auditing Organization for Islamic Financial Institutions
AIC	-	Akaike Information Criterion
AP	-	Asymmetric Power
BIC	-	Bayesian Information Criterion
EGARCH	-	Exponential Generalized Autoregressive Conditional Heteroskedastic
GARCH	-	Generalized Autoregressive Conditional Heteroskedasticity
GCC	-	Gulf Cooperation Council
GJR	-	Glosten-Jagannathan-Runkle
HQC	-	Hannan-Quinn Information Criteria
IGARCH	-	Integrated Generalized Autoregressive Conditional Heteroskedastic
UAE	-	United Arab Emirate
VaR	-	Value at Risk

CHAPTER 1

INTRODUCTION

1.1 Introduction

In order to endure in the competitive global market, financial managers and policy makers need to improve the forecasting and estimating abilities of methods to evaluate the financial risk. Over the last ten years, the unpredictability of financial markets worldwide has elevated the significance of financial risk management (Yamai & Yoshiba, 2005).

The occurrence of several financial crises during the last fifteen years has exposed some weaknesses in current risk management methods. This situation has driven risk managers and policy makers to seek improved approaches to estimate market risk in the financial industry as well as to formulate and redefine risk management schemes (Yamai & Yoshiba, 2005).

While much emphasis has been placed on the prediction and estimation of risk for conventional finance, Islamic markets have been somewhat neglected in this area. The development of Islamic finance is particularly vital for markets in countries with elevated Muslim populations as their progress is hindered by their sole reliance on a conventional scheme (Tariq, 2004).

Among all the areas of Islamic finance development, the growth pace of *Sukuk* or Islamic bonds stands head and shoulders above the rest (Hesse, Jobst, & Solé, 2008). *Sukuk* are long term financial tools that come with the features of conventional bonds while retaining the rules of Shariah (Islamic principles).

Although the Sukuk markets are in their infancy and tracing its origins to Bahrain in 2001, but the market of Sukuk has grown significantly in recent years. However, of late, these markets have undergone rapid development in all aspects, including those of magnitude, quantity and refinement (Ab Majid, Shahimi, & Bangaan, 2010).

The brisk development of the Sukuk market on an international level has culminated in its yearly issuances virtually tripling from USD 45 bln in 2011 to USD 118.8 bln in 2014. Notably, this speedy growth was stimulated by the key fundamental markets such as Malaysia, Saudi Arabia and the United Arab Emirates (UAE), as well as up-and-coming players such as Turkey and Indonesia. Ground-breaking issuances were documented in 2014 from the UK, Hong Kong, Senegal, South Africa and Luxembourg. This served to fortify the standing of *Sukuk* markets as practical and assertive providers of financial support (Zawya, 2013).

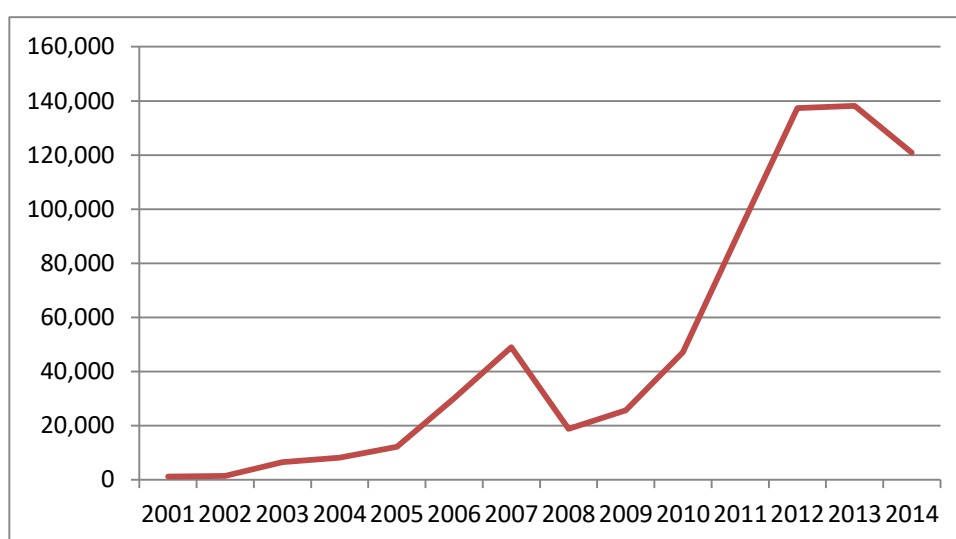
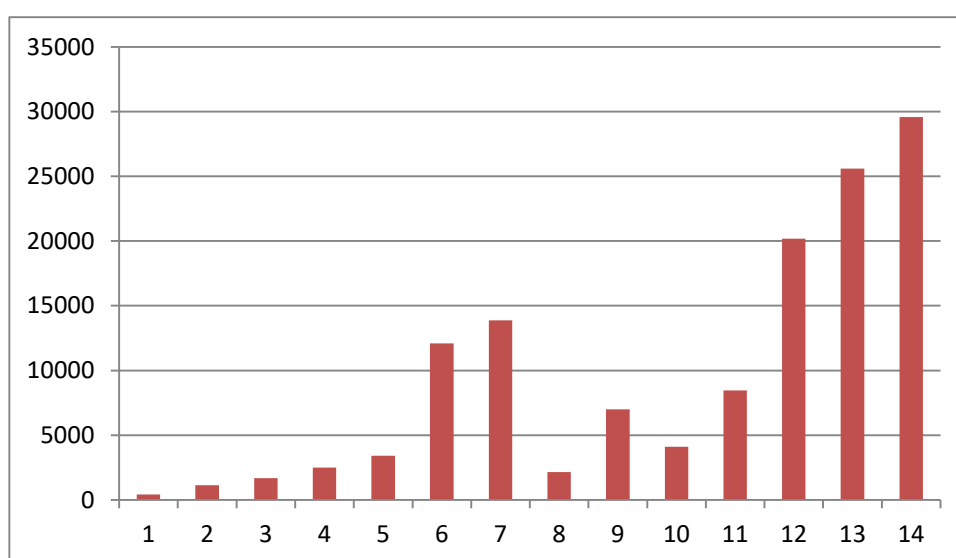
As illustrated in Figure1.1 and Table1.1, the worldwide issuance of Sukuk experienced a steep climb over recent years (Figure 1.1 and Table 1.1 are derived from IIFM¹ data and all amounts are based on million USD). The international issuance of *Sukuk*, amounting to USD 100 million, was initiated by Bahrain in 2001. And by the close of 2007, it had risen to USD 49 billion. *Sukuk* issuance was expected to swell in 2008, but the global crises of 2007 applied the brakes to this development (Cakir & Raei, 2007).

After an initial shortfall, issuances began to gain momentum from 2009 to record USD140 billion at the close of 2012 (POOR'S, 2012) and USD120, 854 billion at the close of 2014. As revealed by Standard and Poor's rating service, despite the issuance of *Sukuk* is presently minor in comparison to global conventional bonds, it has the potential to develop and rub shoulders with the big players in this domain and join to the mainstream.

¹ International Islamic Financial Market

Table 1.1: Global Sukuk Issuances From 2001-2014

Year	Issue	Year	Issue
2001	1,172	2008	18,752
2002	1,372	2009	25,571
2003	6,410	2010	47,081
2004	8,140	2011	92,403
2005	12,180	2012	137,310
2006	29,992	2013	138,170
2007	48,929	2014	120,854

**Figure 1.1** Trend of Global Sukuk Issuances (From 2001-2014)**Figure 1.2** International Sukuk Issuance (Jan 2001- March 2015, USD Millions)

Over the past year, encouraging developments in the Sukuk market have led to an elevated level of cross-border activities in this market. The market saw a rise of Sukuk issuances in foreign currency, especially in USD; while Malaysia drew issuances from other regions including Singapore and Turkey.

Other than issuances, Sukuk listings also experienced increased cross-border activities as issuers raised listings especially on major European stock exchanges, specifically, the London, Irish and Luxembourg Stock Exchanges. Another significant development between 2013 and 2014 was the conformity of Islamic financial institutions to Basel III liquidity and capital requirements (Global Sukuk Report, Q1 2015).

As portrayed in Figure 1.3, sovereign issuers generally preside over the market. This is evident as they are credited with more than 50% of the overall issuances. Meanwhile, supra-national organizations that include the Islamic Development Bank (IDB) also have a share in the market with several hefty issuances that amounted to 15.2% (Report, Q1 2015)

Based on (Report, Q1 2015) the main part of Sukuk corporate issuers are from the financial sectors of Turkey and Malaysia. It is important to mention that two financial institutions from Turkey issued MYR-denominated Sukuk in Malaysia, and announced plans for more similar cross-border issuances in the future.

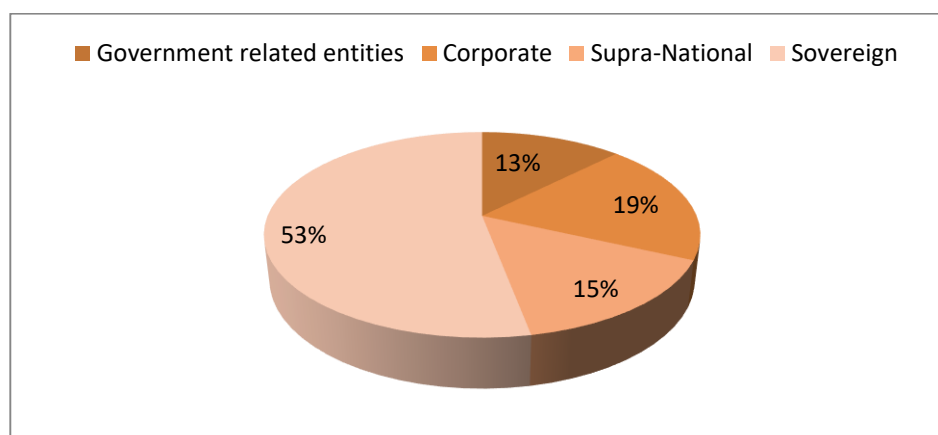


Figure 1.3 Sukuk Issuance in 2014 by issuer type

1.2 Background of Study

The many investigations conducted on Sukuk are mostly concept-based and very few were focused on the evaluation of the Sukuk market as a financial tool. In other words, in spite of several studies that took place about Sukuk recently, most of them are conceptual and theoretical and not attention has been paid to the evaluation of the Sukuk market as a financial instrument. With this in mind, this study sets out to determine if there are any disparities between Sukuk and conventional bonds in the secondary market. During this endeavour, all conceptual as well as constructional differences and similarities are set aside. We also explored the most favourable procedure for approximating downside market risk in the Sukuk market.

Forecasting and estimating financial loss is the main task of financial managers and policy makers to survive in the competitive global market. The importance of financial risk management has increased in the last decade mainly due to increased volatility in financial market all over the world. Considering the importance of this risk, its evaluation is necessary for all banks and financial markets. In this context, we will present the measurement methods VaR in order to evaluate risk of Sukuk (Alexander, 2009).

The chosen VaR methodology for evaluating risk is widely applied in the area of finance. The VaR approach measures the downside risk of a portfolio position as the maximum loss that can be materialize at a future prescribed date with a given probability due to adverse changes in relevant asset and liability prices (Cakir & Raei, 2007).

Value at risk (VaR) is deemed the most effective and prominent instrument for gauging downside market risk. VaR became the basic market management tool for all financial institutions when the Basel Committee on banking supervision ruled that these institutions are obliged to satisfy capital requirements based on the VaR estimate (Alexander, 2009).

VaR is a mathematical approach that portrays the highest degree of expected loss for a portfolio over a given holding period at a specific confidence level (Jorion, 2007). VaR can be regarded as an envelope for all volatility and correlations among separate risk variables over time (Cakir & Raei, 2007).

Although the general usage of VaR is deemed easy, significant and extensive, its utilization as a procedure for estimation and prediction financial risk remains demanding. The main challenge to VaR application is the lack of unique accepted method for its computation so that considerably distinct results can be realized through the implementation of different approaches. The disparities in results can be put down to the varying assumptions taken into consideration by each process (Kuester, Mittnik, & Paolella, 2006; D. G. McMillan & Kambouroudis, 2009).

It is on record that several common traits exist in the stock return time series. These traits, which include asymmetry, fat-tail and volatility clustering, are known as stylized facts. As such, an appropriate approach would be one that considers these stylized facts with regard to financial returns (Mills and Markellos, 2008).

Previously-developed VaR estimation procedures including Variance-Covariance and Historical Simulation have been mostly discarded. This is attributable to their unacceptable assumptions regarding normal distribution for asset return in parametric approaches, and constant variance in cases of non-parametric approaches (Abad, Benito, & del Rey, 2009).

Bollerslev (1986) came up with Generalized Autoregressive Conditional Heteroscedasticity (GARCH) for the modelling of time-varying volatility data in financial markets. This procedure is held in high esteem as it considers significant features of the financial time series. During this thesis, a wide range of both symmetric and asymmetric GARCH methods (including GARCH, EGARCH, GJR-GARCH, IGARCH and Asymmetric power GARCH) were considered for modelling volatility in the Sukuk market.

Bearing in mind the significance of Sukuk, this study focused on providing a synopsis on Sukuk data time series as well as a description on the features of its statistical distribution function. To the best of our knowledge, studies on the behaviour of Sukuk as a liquidity instrument in the secondary market are few and far between. This investigation also applied a variety of GARCH techniques to approximate the VaR of Sukuk for both in-sample and out-sample performances. This was to determine the most suitable model for the Sukuk market.

1.3 Problem Statement

As already mentioned Sukuk refers to bonds issued in accordance with Islamic law. The rapid spread of Sukuk, primarily in the Middle East and Islamic countries in Asia, has made it an important tool for raising funds and asset management. Sukuk provide government and corporations with access to the huge and growing Islamic Liquidity pool, in addition to the conventional investor base (El Qorchi 2005).

The most outstanding feature of Sukuk to assess is that it can be considered as an alternative to conventional bonds to have more diversification in portfolio. At the other side, it is important to examine if the secondary market behaviour of Sukuk and conventional bonds are so distinguished that there is significant value in issuing Sukuk instead of bonds.

Obtaining positive answer to these questions raises Sukuk capacity to decrease risk to the portfolio and promote Sukuk as a long term alternative to conventional bonds. Thus, for Muslims and non-Muslims alike who are concerned about reducing risks to their investment, Sukuk would appear to be a wise investment (Cakir & Raei, 2007).

At the other side, Sukuk provide sovereign governments and corporations with access to the huge and growing Islamic liquidity pool, in addition to the conventional investor base. Particularly, the initial step for evaluating Sukuk as an

alternative to conventional bonds is the assessment of its performance in secondary markets. Subsequently, an improved procedure for modelling volatility as well as for approximating and predicting VaR in the Sukuk market needs to be realized.

It is our opinion that no attention has been paid to approximation and prediction of VaR in the Sukuk market.

The theme of my thesis covers an area which is relatively new and under-developed. The many studies took place on Sukuk are mostly concept- based and very few were focused on the evaluation of the Sukuk market as a financial tool.

The main contribution of this research is that in some respects this study will set a precedent for a financial market, which is still evolving in terms of regulatory and legal framework relating to the Islamic finance industry. In other words, this research aims to contribute to the debate on the issuance of Sukuk as an alternative investment/financial instruments.

1.4 Objectives of the Study

1. Describe the performance of Sukuk returns in the secondary market and provide an overview about modelling of Sukuk returns.
2. Apply a variety of GARCH procedures to Sukuk return series and uncover more appropriate procedures for the modelling of Sukuk returns based on in-sample criterion.
3. Compare the forecasting capability of various GARCH procedures to estimate the VaR of Sukuk based on out-of-sample evaluation.

1.5 Research Questions

- i. How is the performance of Sukuk returns in the secondary market?

- ii. Which GARCH procedure is most appropriate for the modelling of Sukuk returns?
- iii. Which GARCH model is superior for forecasting and estimating the VaR of Sukuk?

1.6 Scope of the Study

To analyse the risk of Sukuk we consider six Sukuk issued by Malaysia, UAE (Dubai) and Saudi Arabia (IDB), Bahrain, Indonesia and Turkey. These countries are ideal for this research since they are the most active countries in terms of Sukuk issuance. Figure 1.4 shows the global Sukuk issuances by each country for the time period of 2001 till July of 2014.

As it can be seen the major part of Sukuk issuance was from Malaysia (42%) and in to lesser extend the Sukuk issuance was from Gulf Cooperation Council (GCC) countries. In GCC countries the most active countries in Sukuk market are UAE and Saudi Arabia which have 18% and 7% of the market respectively. Since the dominant part of Sukuk market in the world is in these three countries and generally they are considered as leader of global Sukuk market, we choose them for this research.

However we didn't restrict our research only to these three countries. In order to avoid dependent on a specific financial market and obtain more robustness results we also consider three more countries including Bahrain, Indonesia and Turkey.

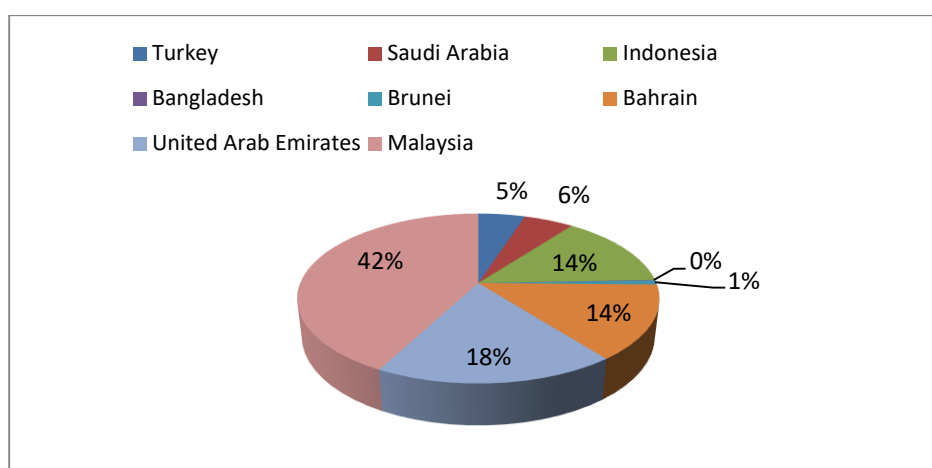


Figure 1.4 International Sukuk issuances country wise break- up (2001-July 2014)

Table 1.2 represent the international Sukuk issuances country wise break- up from 2001 till July 2014. As can be seen from this table, in terms of number of issues the first place is related to Bahrain and United Arab Emirate, respectively. However, in terms of USD million, the first place is related to Saudi Arabia with USD 21,542 million which consist 18.43 percent of total value and the second step is related to Malaysia with USD 16,339 million and 13.98 percent of total value.

Table 1.2: International Sukuk issuances country wise break-up (2001-July 2014)

ASIA & FAR EAST	NUMBER OF ISSUES	AMOUNT USD MILLIONS	% OF TOTAL VALUE
Malaysia	32	16,339	13.98%
Indonesia	7	5,782	4.95%
Singapore	4	711	0.61%
China	1	97	0.08%
Hong Kong	2	196	0.17%
Japan	3	190	0.16%
Pakistan	1	600	0.51%
Total	50	23,917	20.47%
GCC & MIDDLE EAST	NUMBER OF ISSUES	AMOUNT USD MILLIONS	% OF TOTAL VALUE
Bahrain	94	6,830	5.84%
Kuwait	13	2,127	1.82%
Qatar	10	8,935	7.65%

Saudi Arabia	30	21,542	18.43%
United Arab Emirates	68	45,986	39.35%
Total	215	85,420	73.09%
ARFICA	NUMBER OF ISSUES	AMOUNT USD MILLIONS	% OF TOTAL VALUE
Sudan	1	130	0.11%
Total	1	130	0.11%
EUROPE & OTHERS	NUMBER OF ISSUES	AMOUNT USD MILLIONS	% OF TOTAL VALUE
Kazakhstan	1	77	0.07%
France	1	0.65	0.001%
Germany	1	55	0.05%
Luxembourg	2	60	0.05%
Turkey	12	5753	4.92%
United Kingdom	8	687	0.59%
USA	3	767	0.66%
Total	28	7,400	6.33%
Grand Total	294	116,867	100.00%

It should be noted that in this study, the application is restricted to international issuances of Sukuk by the government of Malaysia, Saudi Arabia, United Arab Emirates, Turkey, Bahrain and Indonesia. The lack of accessible data, especially data on secondary market trading, hindered the selection of issuers as well as the investigation in general.

As stated previously, Sukuk markets are in their formative years and for the most part trading is currently confined to primary markets. Although secondary market data are accessible in some areas, the time series is inadequate for the realization of an acceptable outcome. These limitations have forced us to confine our investigation to the sovereign issuance of *susuk* in the global markets, an area where secondary market data are at hand. Local Sukuk are left out as the lack of secondary market price data renders the assessment of volatility and risk unreliable.

1.7 Structure of Thesis

The remainder of this thesis is arranged in the following order: The opening section of chapter two provides the theoretical concepts and features related to Islamic finance and Sukuk. This is followed by a description of VaR, its concept, as well as various statistical and econometrical procedures for its approximation. This chapter also includes a re-evaluation of previous studies. These studies are separated into two categories. The first category focuses on previous studies related to Sukuk risk assessment, while the second category concentrates on previous studies linked to VaR estimation.

Chapter three provides a description on the methodology, data and scope of this study. Chapter four involves the application of various GARCH procedures to Sukuk returns, as well as examinations on the in-sample and out-sample performances of GARCH models. Chapter five opens with an in-depth discussion on the outcomes, followed by the revelation of findings and an account on the limitations of this investigation as well suggestions on ways to overcome them.

1.8 Definition of Terminologies

Islamic Banking and Finance: Islamic banking and finance is based on the principles of Islamic law. This mode of banking is ruled by two main decrees; the sharing of profit and loss, and more importantly, the ruling out of collection and payment of interest.

Sukuk: Similar to a bond in western finance, Sukuk is an Islamic financial certificate which meets the terms of Islamic religious laws. As the conventional interest paying bond arrangement is not permitted in Islam, the Sukuk issuer sells the certificate to an investor, who then rents it back to the issuer for a predetermined rental fee. Through a contractual agreement, the investor is assured that at a future date, the bonds will be bought back by the issuer at par value.

Risk: This refers to the possibility of the actual return to an investment varying from the anticipated return. Risk involves the chance of suffering partial or total loss of the primary investment. For the most part, the various forms of risk are gauged through a calculation of the standard deviation of the historical returns, or average returns of a particular investment. An elevated standard deviation points to a raised risk level.

Value at risk: This is a mathematical procedure utilized for measuring and quantifying the degree of financial risk for a given duration. This procedure is employed by risk administrators and policy makers to gauge and manage the risks taken on by organizations. The computation of value at risk involves three variables; the likely amount of use, the probability of loss related to that amount, and the time frame. The main task of risk managers is to make certain that the risk taken on by an organization does not exceed its risk tolerance capacity.

Riba: Riba, or interest in conventional economic vocabulary, denotes “an excess” and is described as “any unjustifiable increase of capital whether through loans or sales.” Riba also refers to any positive, fixed and predetermined rate.

1.9 Summary

Forecasting and estimating financial loss is the main task of financial managers and policy makers to survive in the competitive global market. The importance of financial risk management has increased in the last decade mainly due to increased volatility in financial market all over the world. In spite of extensive research on forecasting and estimating value at risk in the conventional financial markets, less attention has been paid to the Sukuk markets.

Sukuk has established itself as the fastest developing aspect of Islamic banking. The market of these long-term Islamic compliant financial tools has grown from \$1,172 billion in 2001, to USD 118.8bln in 2014. Currently, Sukuk are also issued in other regions by sovereign, corporate and international and global

organizations including the Saxony-Anhalt German State, GE Capital, and the International Finance Corporation (IFC).

While the significance of forecasting and estimating risk in the financial market is clear, we are of the opinion that the *Sukuk* market has been overlooked in these areas. The focus of this study is on identifying, if any, the technical disparities between Sukuk and conventional bonds. For this analysis, we deliberated on six Sukuk issued by Malaysia, UAE, Saudi Arabia, Bahrain, Indonesia and Turkey.

REFERENCES

- Ab Majid, H., Shahimi, S., and Bangaan, M. H. S. (2010). "Sukuk Defaults and its Implication: A Case Study of Malaysian Capital Market." In 8th International Conference on Islamic Economics and Finance.
- Abad, P. and Benito, S. (2013). "A Detailed Comparison of Value at Risk Estimates." *Mathematics and Computers in Simulation* 94: 258–276.
- Abad, P., et al. (2009). "A Detailed Comparison of Value at Risk in International Stock Exchanges", FUNCAS Working Paper, 452.
- Afshar, T. A. (2013). Compare and Contrast Sukuk (Islamic Bonds) with Conventional Bonds, Are they Compatible? *Journal of Global Business Management*, 9(1), 44.
- Ahmad, A. (2009). Riba, its economic rationale and implications. Institute of Islamic Banking and Insurance [www.islamic-banking.com].
- Ainley, M., Mashayekhi, A., Hicks, R., Rahman, A., and Ravalia, A. (2007). "Islamic Finance in the UK: Regulation and Challenges", Financial Services Authority London.
- Akaike, H. (1974). "A New Look at the Statistical Model Identification." *Automatic Control, IEEE Transactions on* 19(6): 716-723.
- Al-Amine, M. A. B. M. (2008). "Sukuk Market: Innovations and Challenges." *Islamic Capital Markets*: 33.
- Alam, N., Kabir Hassan, M., and Haque, M. A. (2013). "Are Islamic Bonds Different from Conventional Bonds? International Evidence from Capital Market Tests." *Borsa Istanbul Review* 13 (3): 22–29.
- Alexander, C. (2009). *Market Risk Analysis, Value at Risk Models*, Wiley. com.
- Andersen, T. G. and Bollerslev, T. (1998). "Answering the Skeptics: Yes, Standard Volatility Models do Provide Accurate Forecasts." *International Economic Review* 39(4):885-905.

- Angelidis, T., Benos, A., and Degiannakis, S. (2004). "The Use of GARCH Models in VaR Estimation." *Statistical Methodology* 1(1): 105-128.
- Anita, H. (2012). *When Two World Collide*. Global Finance, NY, 24(1), 29.
- Artzner, P., Delbaen, F., Eber, J. M., and Heath, D. (1999). "Coherent Measures of Risk." *Mathematical Finance* 9(3): 203-228.
- Awartani, B. and Corradi, V. (2005). "Predicting the Volatility of the S&P-500 Stock Index Via GARCH Models: the Role of Asymmetries." *International Journal of Forecasting* 21(1): 167-183.
- Ayub, M. (2005). "Securitization, sukuk and Fund Management Potential to be Realized by Islamic Financial Institutions." *Sixth International Conference on Islamic Economics, Banking and Finance*.
- Ayub, M. (2009). *Understanding Islamic Finance (Vol. 462)*: John Wiley & Sons.
- Bao, Y., Lee, T., and Salto, L. B. (2006). "Evaluating Predictive Performance of Value at Risk Models in Emerging Markets: A Reality Check." *Journal of Forecasting* 25(2): 101-128.
- Barbour, D., et al. (1997). "Asset Securitisation in Emerging Market Economies: Fundamental Considerations." *YB Int'l Fin. & Econ. L.* 2: 281.
- Bekiros, S. D., and Georgoutsos, D. A. (2005). "Estimation of Value-at-Risk by extreme Value And Conventional Methods: A Comparative Evaluation Of Their Predictive Performance." *Journal of International Financial Markets, Institutions and Money* 15(3): 209-228.
- Bensalah, Y. (2000). *Steps in Applying Extreme Value Theory to Finance: A Review*, Bank of Canada.
- Bhattacharyya, M., Chaudhary, A., and Yadav, G. (2008). "Conditional VaR Estimation Using Pearson's Type IV Distribution." *European Journal of Operational Research* 191(2): 386-397.
- Bollerslev, T. (1986). "Generalized Autoregressive Conditional Heteroskedasticity." *Journal Of Econometrics* 31(3): 307-327.
- Cakir, S. and Raei, F. (2007). "Sukuk vs. Eurobonds: Is there a Difference in Value-at-Risk?" *IMF Working Papers*: 1-20.
- Campbell, J. Y., Andrew, W. L., and Mackinlay, A. C. (1997). *The Econometrics of Financial Markets*, Princeton University Press.

- Chin, W. C. (2008). "Heavy-tailed Value-at-Risk Analysis for Malaysian Stock Exchange." *Physica A: Statistical Mechanics and its Applications* 387(16): 4285-4298.
- Chong, C. W., Ahmad, M. I., and Abdullah, M. Y. (1999). "Performance of GARCH Models in Forecasting Stock Market Volatility." *Journal of Forecasting* 18(5): 333-343.
- Christoffersen, P. F. (1998). "Evaluating Interval Forecasts." *International Economic Review* 39 (4): 841-862.
- Chuang, I. Y., Lu, J. R., and Lee, P. H. (2007). "Forecasting Volatility in the Financial Markets: a Comparison of Alternative Distributional Assumptions." *Applied Financial Economics* 17(13): 1051-1060.
- Consigli, G. (2002). "Tail Estimation and Mean-VaR Portfolio Selection in Markets Subject to Financial Instability." *Journal of Banking & Finance* 26(7): 1355-1382.
- Cont, R. (2007). Volatility Clustering in Financial Markets: Empirical Facts and Agent-based Models. *Long Memory in Economics* 289-309.
- Danielsson, J. (2002). "The Emperor has no Clothes: Limits to Risk Modelling." *Journal of Banking & Finance* 26(7): 1273-1296.
- Danielsson, J., and De Vries, C. G. (2000). "Value-at-Risk and Extreme Returns." *Annales d'Economie et de Statistique*: 239-270.
- David Cabedo, J., and Moya, I. (2003). "Estimating Oil Price 'Value at Risk' using the Historical Simulation Approach." *Energy Economics* 25(3): 239-253.
- Dimitrakopoulos, D. N., Kavussanos, M. G., and Spyrou, S. I. (2010). "Value at Risk Models for Volatile Emerging Markets Equity Portfolios." *The Quarterly Review of Economics and Finance* 50(4): 515-526.
- El Qorchi, M. (2005). "Islamic Finance Gears up." *Finance and Development* 42(4): 46-50.
- Engle, R. F., and Bollerslev, T. (1986). "Modelling the Persistence of Conditional Variances." *Econometric Reviews* 5(1): 1-50.
- Evans, T., and McMillan, D. G. (2007). "Volatility Forecasts: the Role of Asymmetric and Long-memory Dynamics and Regional Evidence." *Applied Financial Economics* 17(17): 1421-1430.

- Fan, Y., et al. (2008). "Estimating 'Value at Risk' of Crude Oil Price and its Spillover Effect Using the GED-GARCH Approach." *Energy Economics* 30(6): 3156-3171.
- Gabriel, A. S. (2012). "Evaluating the Forecasting Performance of GARCH Models. Evidence from Romania." *Procedia-Social and Behavioral Sciences* 62: 1006-1010.
- Gencay, R. and F. Selcuk (2004). "Extreme Value Theory and Value-at-Risk: Relative Performance in Emerging Markets." *International Journal of Forecasting* 20(2): 287-303.
- Gencay, R., Selcuk, F., and Ulugülyagci, A. (2003). "High Volatility, Thick Tails and Extreme Value Theory in Value-at-Risk Estimation." *Insurance: Mathematics and Economics* 33(2): 337-356.
- Giannopoulos, K., and Tunaru, R. (2005). "Coherent Risk Measures under Filtered Historical Simulation." *Journal of Banking and Finance* 29(4): 979-996.
- Gilli, M. (2006). "An Application of Extreme Value Theory for Measuring Financial Risk." *Computational Economics* 27(2-3): 207-228.
- Gilli, M., and Kellezi, E. (2006). "An Application of Extreme Value Theory for Measuring Financial Risk." *Computational Economics* 27(2): 207-228.
- Global Sukuk Report. (Q1 2015).
- Glosten, L. R., Jagannathan, R., and Runkle, D. E. (1993). "On the Relation between the Expected Value and the Volatility of the Nominal Excess Return on Stocks." *The Journal of Finance* 48(5): 1779-1801.
- Godlewski, C. J., Turk-Ariss, R., and Weill, L. (2011). "Do Markets Perceive Sukuk and Conventional Bonds as Different Financing Instruments?" *Bank of Finland (BOFIT) Discussion Papers* 6: 2011. ISSN 1456-5889.
- Gokcan, S. (2000). "Forecasting Volatility of Emerging Stock Markets: Linear Versus Non-Linear GARCH Models." *Journal of Forecasting* 19(6): 499-504.
- Haas, M., and Pigorsch, C. (2011). *Financial Economics, Fat-Tailed Distributions. Complex Systems in Finance and Econometrics* 308-339.
- Hannan, E. J., and Quinn, B. G. (1979). "The Determination of The order of an Autoregression." *Journal of the Royal Statistical Society. Series B (Methodological)*: 190-195.

- Hartz, C., Mittnik, S., and Paolella, M. (2006). "Accurate Value-at-Risk Forecasting Based on the Normal-GARCH Model." *Computational Statistics and Data Analysis* 51(4): 2295-2312.
- Hassan, K. A. (2012). "Comparison between Sukuk and Conventional Bonds: Value at Risk Approach." Master Thesis, Westminster University, UK.
- Hendricks, D. (1996). "Evaluation of Value-at-Risk Models Using Historical Data." *Economic Policy Review* 2(1).
- Hesse, H., Andreas, A. J., and Juan, S. (2008). "Trends and Challenges in Islamic Finance." *World Economics* 9(2): 175-193.
- Hung, J.C., Lee, M. C., and Liu, H. C. (2008). "Estimation of Value-at-Risk for Energy Commodities Via Fat-tailed GARCH Models." *Energy Economics* 30(3): 1173-1191.
- Ilias, S. (2008). *Islamic Finance: overview and Policy Concerns*, DTIC Document.
- Iqbal, M., and Llewellyn, D. T. (2002). "Islamic Banking and Finance: New Perspectives on Profit-Sharing and Risk", Edward Elgar Publishing.
- Iqbal, Z., and Mirakhor, A. (2011). *An Introduction to Islamic Finance: Theory and Practice*, Wiley. com.
- Iqbal, Z., and Tsubota, H. (2006). "Emerging Islamic Capital Markets." *Islamic Finance Review*, Euromoney Handbook, and Euromoney Institutional Investor PLC, London: 5-11.
- Jobst, A. (2007). *The Economics of Islamic Finance and Securitization* (EPub), International Monetary Fund.
- Jorion, P. (2007). "Value at Risk: the New Benchmark for Managing Financial Risk", McGraw-Hill New York.
- Kassberger, S., and Kiesel, R. (2006). "A Fully Parametric Approach to Return Modelling and Risk Management of Hedge Funds." *Financial Markets and Portfolio Management* 20(4): 472-491.
- Koksal, B., and Orhan, M. (2012). "A Comparison of GARCH Models for VAR Estimation." *Expert Systems with Applications* 39: 3582-3592.
- Kuester, K., Mittnik, S., and Paolella, M. S. (2006). "Value-at-Risk Prediction: A Comparison of Alternative Strategies." *Journal of Financial Econometrics* 4(1): 53-89.
- Kupiec, P. H. (1995). "Techniques for verifying the accuracy of risk Measurement Models." *The Journal of Derivatives* 3(2): 73-84.

- Kwiatkowski, D., Phillips, P. C. B., Schmidt, P., and Shin, Y. (1992). "Testing the Null Hypothesis of Stationarity Against the Alternative of a unit Root: How Sure are we that Economic Time Series have a unit Root?" *Journal of Econometrics* 54(1): 159-178.
- Levich, R. M. (1985). "Empirical Studies of Exchange Rates: Price Behavior, Rate Determination and Market Efficiency." *Handbook of International Economics* 2: 979-1040.
- Lim, C. M., and Sek, S. K. (2013). "Comparing the Performances of GARCH-Type Models in Capturing the Stock Market Volatility in Malaysia." *International Conference On Applied Economics (ICOAE)* 5: 478-487.
- Liu, H. C., and Hung, J. C (2010). "Forecasting S & P-100 Stock Index Volatility: The Role of Volatility Asymmetry and Distributional Assumption in GARCH Models." *Expert Systems with Applications* 37(7): 4928-4934.
- Ljung, G. M., and Box, G. E. P. (1978). "On a Measure of Lack of Fit in Time Series Models." *Biometrika* 65(2): 297-303.
- Loudon, G. F., watt, W. H., and Yadav, P. K. (2000). "An Empirical Analysis of Alternative Parametric ARCH Models." *Journal of Applied Econometrics* 15(2): 117-136.
- Maghyereh, A. I., and Al-Zoubi, H. A. (2006). "Value-at-Risk under Extreme Values: the Relative Performance in MENA Emerging Stock Markets." *International Journal of Managerial Finance* 2(2): 154-172.
- Mandelbrot, B. (1967). "The Variation of Some other Speculative Prices." *The Journal of Business* 40(4): 393-413.
- Manganelli, S., and Engle, R. (2001). *Value at Risk Models in Finance*, European Central Bank Working Paper No.75.
- Marimoutou, V., Raggad, B., and Trabelsi, A. (2009). "Extreme Value Theory and Value at Risk: Application to Oil Market." *Energy Economics* 31(4): 519-530.
- McMillan, D., Speight, A., and Apgwilym, O. (2000). "Forecasting UK Stock Market Volatility." *Applied Financial Economics* 10(4): 435-448.
- McMillan, D. G., and Kambouroudis, D. (2009). "Are Risk Metrics Forecasts Good Enough? Evidence from 31 Stock Markets." *International Review of Financial Analysis* 18(3): 117-124.

- McNeil, A. J., and Frey, R. (2000). "Estimation of Tail-Related Risk Measures for Heteroscedastic Financial Time Series: an Extreme Value Approach." *Journal of Empirical Finance* 7(3): 271-300.
- Meera, M., Kameel, A., and Abdul Razak, D. (2005). "Islamic Home Financing through Musharakah Mutanaqisah and al-Bay' Bithaman Ajil Contracts: A Comparative Analysis." *Review of Islamic Economics* 9(2): 5-30.
- Miller, N. D., et al. (2007). "UK Welcomes the Sukuk-How the UK Finance Bill Should Stimulate Islamic Finance in London, Much to the Delight of the City's Banks." *Int'l Fin. L. Rev.* 26: 24.
- Mirakhor, A. (1995). "Theory of an Islamic Financial System." *Encyclopaedia of Islamic Banking*.
- Mirakhor, A., and Zaidi, I. (1988). "Stabilization and Growth in an Open Islamic Economy." *IMF Working Paper No. 88/22*
- Mussa, M. (1979). *Empirical Regularities in the Behavior of Exchange Rates and Theories of the Foreign Exchange Market*. Carnegie-Rochester Conference Series on Public Policy, North-Holland.
- Nelson, D. B. (1991). "Conditional Heteroskedasticity in Asset Returns: A new Approach." *Econometrica: Journal of the Econometric Society* 59(2): 347-370.
- Nelson, D. B. (1990). "Stationarity and Persistence in the GARCH (1,1) Model." *Econometric Theory* 6(03): 318-334.
- Newey, W. K., and West, K. D. (1986). "A Simple, Positive Semi-definite, Heteroskedasticity and Autocorrelationconsistent Covariance Matrix." *National Bureau of Economic Research Cambridge, Mass., USA*.
- Ozun, A., Cifter, A., and Yilmazer, S. (2007). "Filtered Extreme Value Theory for Value-At-Risk Estimation." *MPRA Paper, Muhich Personal RePEc Archive*.
- Politis, D. N. (2004). "A Heavy-tailed Distribution for ARCH Residuals with Application to Volatility Prediction."
- Poon, S. H. and Granger, C. W. J. (2003). "Forecasting Volatility in Financial Markets: A Review." *Journal of Economic Literature* 41(2): 478-539.
- Presley, J. R., and Sessions, J. G. (1994). "Islamic Economics: the Emergence of a New Paradigm." *The Economic Journal* 104:584-596.

- Rosly, S. A. (2005). "Critical Issues on Islamic Banking and Financial Markets: Islamic Economics, Banking and Finance, Investments, Takaful and Financial Planning." Dinamas Publishing. ISBN 9834248903.
- Sarma, M., Thomas, S., and Shah, A. (2003). "Selection of Value-at-Risk Models." *Journal of Forecasting* 22(4): 337-358.
- Schaumburg, J. (2012). "Predicting Extreme Value at Risk: Nonparametric Quantile Regression with Refinements from Extreme Value Theory." *Computational Statistics and Data Analysis* 56(12): 4081-4096.
- Schwarz, G. (1978). "Estimating the Dimension of a Model." *The Annals of Statistics* 6(2): 461-464.
- Shaikh, S., and Saeed, S. (2010). "Sukuk Bond: The Global Islamic Financial Instrument." MPRA Paper 26700.
- Siddiqi, M. N. (2005). *Riba, Bank Interest and the Rationale of its Prohibition*, Markazi Maktaba Islami Publishers.
- Spierdijk, L. (2016). "Confidence intervals for ARMA–GARCH value-at-risk: The case of heavy tails and skewness." *Computational Statistics & Data Analysis*, 100, 545-559.
- Tariq, A. A. (2004). "Managing Financial Risks of Sukuk Structures." Loughborough University, UK, (mimeo).
- Theodossiou, P. (1998). "Financial Data and the Skewed Generalized T Distribution." *Management Science* 44(12-Part-1): 1650-1661.
- Tsay, R. S. (2005). *Analysis of Financial Time Series*, Wiley. com.
- Usmani, M. T. (2008). "Sukuk and their Contemporary applications." Internet Download 1-16.
- Uusmani, M. T., and Taqi Usmani, M. (2002). *An Introduction to Islamic finance*, Brill.
- Verhoeven, P., and McAleer, M. (2004). "Fat Tails and Asymmetry in Financial Volatility Models." *Mathematics and Computers in Simulation* 64(3): 351-361.
- Vishwanath, S. R., and Azmi, S. (2009). "An Overview of Islamic Sukuk Bonds." *The Journal of Structured Finance* 14(4): 58-67.
- Wilhelmsson, A. (2006). "Garch Forecasting Performance under Different Distribution Assumptions." *Journal of Forecasting* 25(8): 561-578.

- Wilson, R. (2008). "Innovation in the Structuring of Islamic Sukuk Securities." *Humanomics* 24(3): 170-181.
- Wilson, R. (2004). "Overview of the Sukuk Market." *Islamic Bonds: Your Guide to Issuing, Structuring and Investing in Sukuk*, Euromoney Books, London: 6-7.
- Wirch, J. L. (1997). "Value-at-Risk for Risk Portfolios." Unpublished Working Paper.
- Wu, G. (2001). "The Determinants of Asymmetric Volatility." *Review of Financial Studies* 14(3): 837-859.
- Yamai, Y., and Yoshida, T. (2005). "Value-at-Risk Versus Expected Shortfall: A Practical Perspective." *Journal of Banking and Finance* 29(4): 997-1015.
- Yang, L., & Hamori, S. (2013). Dependence structure among international stock markets: a GARCH–copula analysis. *Applied Financial Economics*, 23(23), 1805-1817.
- Zamir, I., and Tsubota, H. (2009). "The World Bank, Emerging Islamic Capital Markets—A Quickening Pace and a New potential." *World bank Report by World bank*.
- Zikovic, S., and Aktan, B. (2009). "Global Financial Crisis and VaR Performance in Emerging Markets: A Case of EU Candidate States-Turkey and Croatia." *Zbornik radova Ekonomskog fakulteta u Rijeci* 27(1): 149-170.
- Ab Majid, H., Shahimi, S., & Bangaan, M. H. S. (2010). Sukuk Defaults and its Implication: A Case Study of Malaysian Capital Market. Paper presented at the 8th International Conference on Islamic Economics and Finance.
- Abad, P., & Benito, S. (2012). A detailed comparison of value at risk estimates. *Mathematics and Computers in Simulation*.
- Abad, P., & Benito, S. (2013). A detailed comparison of value at risk estimates. *Mathematics and Computers in Simulation*, 94, 258-276.
- Abad, P., Benito, S., & del Rey, S. (2009). A Detailed Comparison of Value at Risk in International Stock Exchanges: *FUNCAS Working Paper*, 452.
- Ahmad, A. (2009). Riba, its economic rationale and implications. *Institute of Islamic Banking and Insurance* [www. islamic-banking. com].
- Ahmed, H. (2014). Islamic banking and Shari'ah compliance: a product development perspective. *Journal of Islamic finance.*, 3(2), 15-29.

- Ainley, M., Mashayekhi, A., Hicks, R., Rahman, A., & Ravalia, A. (2007). *Islamic finance in the UK: regulation and challenges* (Vol. 9): Financial Services Authority London.
- Akaike, H. (1974). A new look at the statistical model identification. *Automatic Control, IEEE Transactions on*, 19(6), 716-723.
- Al-Amine, M. A.-B. M. (2008a). Sukuk market: Innovations and challenges. *ISLAMIC CAPITAL MARKETS*, 33.
- Al-Amine, M. A.-B. M. (2008b). Sukuk market: innovations and challenges. *ISLAMIC CAPITAL MARKETS*, 33.
- Alam, N., Hassan, M. K., & Haque, M. A. (2013). Are Islamic bonds different from conventional bonds? International evidence from capital market tests. *Borsa Istanbul Review*.
- Alexander, C. (2009). *Market Risk Analysis, Value at Risk Models* (Vol. 4): Wiley.com.
- Andersen, T. G., & Bollerslev, T. (1998). Answering the skeptics: Yes, standard volatility models do provide accurate forecasts. *International economic review*, 885-905.
- Angelidis, T., Benos, A., & Degiannakis, S. (2004). The use of GARCH models in VaR estimation. *Statistical methodology*, 1(1), 105-128.
- Artzner, P., Delbaen, F., Eber, J. M., & Heath, D. (1999). Coherent measures of risk. *Mathematical finance*, 9(3), 203-228.
- Awartani, B., & Corradi, V. (2005). Predicting the volatility of the S&P-500 stock index via GARCH models: the role of asymmetries. *International Journal of Forecasting*, 21(1), 167-183.
- Ayub, M. (2005). Securitization, sukuk and fund management potential to be realized by Islamic Financial Institutions. *Sixth International Conference on Islamic Economics, Banking and Finance*.
- Ayub, M. (2009). *Understanding Islamic Finance* (Vol. 462): John Wiley & Sons.
- Bao, Y., Lee, T. H., & Saltoglu, B. (2006). Evaluating predictive performance of value at risk models in emerging markets: a reality check. *Journal of Forecasting*, 25(2), 101-128.
- Barbour, D., Norton, J., & Slover, T. (1997). Asset Securitisation in Emerging Market Economies: Fundamental Considerations. *YB Int'l Fin. & Econ. L.*, 2, 281.

- Bekiros, S. D., & Georgoutsos, D. A. (2005). Estimation of Value-at-Risk by extreme value and conventional methods: a comparative evaluation of their predictive performance. *Journal of International Financial Markets, Institutions and Money*, 15(3), 209-228.
- Bensalah, Y. (2000). Steps in applying extreme value theory to finance: A review: Bank of Canada.
- Bhattacharyya, M., Chaudhary, A., & Yadav, G. (2008). Conditional VaR estimation using Pearson's type IV distribution. *European Journal of Operational Research*, 191(2), 386-397.
- Bollerslev, T. (1986). Generalized autoregressive conditional heteroskedasticity. *Journal of econometrics*, 31(3), 307-327.
- Cakir, S., & Raei, F. (2007). Sukuk vs. Eurobonds: Is there a difference in Value-at-Risk? IMF Working Papers, 1-20.
- Campbell, J. Y., & Andrew, W. (1997). Lo, and A. Craig Mackinlay, 1997, *The econometrics of financial markets*: Princeton University Press.
- Chin, W. C. (2008). Heavy-tailed value-at-risk analysis for Malaysian stock exchange. *Physica A: Statistical Mechanics and its Applications*, 387(16), 4285-4298.
- Chong, C. W., Ahmad, M. I., & Abdullah, M. Y. (1999). Performance of GARCH models in forecasting stock market volatility. *Journal of Forecasting*, 18(5), 333-343.
- Christoffersen, P. F. (1998). Evaluating interval forecasts. *International economic review*, 841-862.
- Chuang, I.-Y., Lu, J.-R., & Lee, P.-H. (2007). Forecasting volatility in the financial markets: a comparison of alternative distributional assumptions. *Applied Financial Economics*, 17(13), 1051-1060.
- Consigli, G. (2002). Tail estimation and mean-VaR portfolio selection in markets subject to financial instability. *Journal of Banking & Finance*, 26(7), 1355-1382.
- Cont, R. (2007). Volatility clustering in financial markets: empirical facts and agent-based models *Long memory in economics* (pp. 289-309): Springer.
- Danielsson, J. (2002). The emperor has no clothes: Limits to risk modelling. *Journal of Banking & Finance*, 26(7), 1273-1296.

- Danielsson, J., & De Vries, C. G. (2000). Value-at-risk and extreme returns. *Annales d'Economie et de Statistique*, 239-270.
- David Cabedo, J., & Moya, I. (2003). Estimating oil price 'Value at Risk' using the historical simulation approach. *Energy Economics*, 25(3), 239-253.
- Dimitrakopoulos, D. N., Kavussanos, M. G., & Spyrou, S. I. (2010). Value at risk models for volatile emerging markets equity portfolios. *The Quarterly Review of Economics and Finance*, 50(4), 515-526.
- Ding, Z., Granger, C. W., & Engle, R. F. (1993). A long memory property of stock market returns and a new model. *Journal of empirical finance*, 1(1), 83-106.
- El Qorchi, M. (2005). Islamic finance gears up. *Finance and Development*, 42(4), 46.
- Engle, R. F., & Bollerslev, T. (1986). Modelling the persistence of conditional variances. *Econometric reviews*, 5(1), 1-50.
- Evans, T., & McMillan, D. G. (2007). Volatility forecasts: the role of asymmetric and long-memory dynamics and regional evidence. *Applied Financial Economics*, 17(17), 1421-1430.
- Fan, Y., Zhang, Y.-J., Tsai, H.-T., & Wei, Y.-M. (2008). Estimating 'Value at Risk' of crude oil price and its spillover effect using the GED-GARCH approach. *Energy Economics*, 30(6), 3156-3171.
- Gabriel, A. S. (2012). Evaluating the Forecasting Performance of GARCH Models. Evidence from Romania. *Procedia-Social and Behavioral Sciences*, 62, 1006-1010.
- Gencay, R., & Selcuk, F. (2004). Extreme value theory and value-at-risk: relative performance in emerging markets. *International Journal of Forecasting*, 20(2), 287-303.
- Gençay, R., Selçuk, F., & Ulugülyağci, A. (2003). High volatility, thick tails and extreme value theory in value-at-risk estimation. *Insurance: Mathematics and Economics*, 33(2), 337-356.
- Giannopoulos, K., & Tunaru, R. (2005). Coherent risk measures under filtered historical simulation. *Journal of Banking & Finance*, 29(4), 979-996.
- Gilli, M. (2006). An application of extreme value theory for measuring financial risk. *Computational Economics*, 27(2-3), 207-228.
- Gilli, M., & Kellezi, E. (2006). An application of extreme value theory for measuring financial risk. *Computational Economics*, 27(2), 207-228.

- Glosten, L. R., Jagannathan, R., & Runkle, D. E. (1993). On the relation between the expected value and the volatility of the nominal excess return on stocks. *The journal of finance*, 48(5), 1779-1801.
- Godlewski, C. J., Turk-Ariss, R., & Weill, L. (2013). Sukuk vs. conventional bonds: A stock market perspective. *Journal of Comparative Economics*, 41(3), 745-761.
- Godlewski, C. J., Turk-Ariss, R., Weill, L., Goel, R. K., & Mehrotra, A. (2011). Do markets perceive sukuk and conventional bonds as different financing instruments? *Bank of Finland (BOFIT) Discussion Papers*, 6, 2011.
- Gokcan, S. (2000). Forecasting volatility of emerging stock markets: linear versus non-linear GARCH models. *Journal of Forecasting*, 19(6), 499-504.
- Haas, M., & Pigorsch, C. (2011). *Financial Economics, Fat-Tailed Distributions Complex Systems in Finance and Econometrics* (pp. 308-339): Springer.
- Hannan, E. J., & Quinn, B. G. (1979). The determination of the order of an autoregression. *Journal of the Royal Statistical Society. Series B (Methodological)*, 190-195.
- Hartz, C., Mittnik, S., & Paoletta, M. (2006). Accurate value-at-risk forecasting based on the normal-GARCH model. *Computational Statistics & Data Analysis*, 51(4), 2295-2312.
- Hassan, K. A. (2012). *Comparison between Sukuk and Conventional Bonds: Value at Risk Approach*.
- Hendricks, D. (1996). Evaluation of value-at-risk models using historical data. *Economic Policy Review*, 2(1).
- Hesse, H., Jobst, A., & Solé, J. (2008). Trends and challenges in Islamic finance. *World Economics*, 9(2), 175-193.
- Hung, J.-C., Lee, M.-C., & Liu, H.-C. (2008). Estimation of value-at-risk for energy commodities via fat-tailed GARCH models. *Energy Economics*, 30(3), 1173-1191.
- Ilias, S. (2008). *Islamic finance: overview and policy concerns*.
- Iqbal, M., & Llewellyn, D. T. (2002). *Islamic banking and finance: New perspectives on profit-sharing and risk*: Edward Elgar Publishing.
- Iqbal, Z., & Mirakhor, A. (2011). *An introduction to Islamic finance: theory and practice (Vol. 687)*: Wiley. com.

- Iqbal, Z., & Tsubota, H. (2006). Emerging Islamic capital markets. *Islamic Finance Review, Euromoney Handbook, and Euromoney Institutional Investor PLC*, London, 5-11.
- Jobst, A. (2007). *The Economics of Islamic Finance and Securitization* (EPub): International Monetary Fund.
- Jorion, P. (2007). *Value at risk: the new benchmark for managing financial risk* (Vol. 3): McGraw-Hill New York.
- Kassberger, S., & Kiesel, R. (2006). A fully parametric approach to return modelling and risk management of hedge funds. *Financial markets and portfolio management*, 20(4), 472-491.
- Köksal, B., & Orhan, M. (2011). A Comparison of GARCH Models for VAR Estimation. *Expert Systems with Applications*, Forthcoming.
- Kuester, K., Mittnik, S., & Paolella, M. S. (2006). Value-at-risk prediction: A comparison of alternative strategies. *Journal of Financial Econometrics*, 4(1), 53-89.
- Kupiec, P. H. (1995). Techniques for verifying the accuracy of risk measurement models. *THE J. OF DERIVATIVES*, 3(2).
- Kwiatkowski, D., Phillips, P. C. B., Schmidt, P., & Shin, Y. (1992). Testing the null hypothesis of stationarity against the alternative of a unit root: How sure are we that economic time series have a unit root? *Journal of Econometrics*, 54(1), 159-178.
- Levich, R. M. (1985). Empirical studies of exchange rates: price behavior, rate determination and market efficiency. *Handbook of international economics*, 2, 979-1040.
- Lim, C. M., & Sek, S. K. (2013). Comparing the performances of GARCH-type models in capturing the stock market volatility in Malaysia.
- Liu, H.-C., & Hung, J.-C. (2010). Forecasting S&P-100 stock index volatility: The role of volatility asymmetry and distributional assumption in GARCH models. *Expert Systems with Applications*, 37(7), 4928-4934.
- Ljung, G. M., & Box, G. E. P. (1978). On a measure of lack of fit in time series models. *Biometrika*, 65(2), 297-303.
- Loudon, G. F., Watt, W. H., & Yadav, P. K. (2000). An empirical analysis of alternative parametric ARCH models. *Journal of Applied Econometrics*, 15(2), 117-136.

- Maghyereh, A. I., & Al-Zoubi, H. A. (2006). Value-at-risk under extreme values: the relative performance in MENA emerging stock markets. *International Journal of Managerial Finance*, 2(2), 154-172.
- Mandelbrot, B. (1967). The variation of some other speculative prices. *The Journal of Business*, 40(4), 393-413.
- Manganelli, S., & Engle, R. (2001). Value at risk models in finance: European Central Bank Working Paper Series.
- Marimoutou, V., Raggad, B., & Trabelsi, A. (2009). Extreme value theory and value at risk: application to oil market. *Energy Economics*, 31(4), 519-530.
- Maurer, B. (2005). *Mutual life, limited: Islamic banking, alternative currencies, lateral reason*: Princeton University Press.
- McMillan, D., Speight, A., & Apgwilym, O. (2000). Forecasting UK stock market volatility. *Applied Financial Economics*, 10(4), 435-448.
- McMillan, D. G., & Kambouroudis, D. (2009). Are RiskMetrics forecasts good enough? Evidence from 31 stock markets. *International Review of Financial Analysis*, 18(3), 117-124.
- McNeil, A. J., & Frey, R. (2000). Estimation of tail-related risk measures for heteroscedastic financial time series: an extreme value approach. *Journal of empirical finance*, 7(3), 271-300.
- Meera, M., Kameel, A., & Abdul Razak, D. (2005). Islamic Home Financing through Musharakah Mutanaqisah and al-Bay' Bithaman Ajil Contracts: A Comparative Analysis. *Review of Islamic Economics*, 9(2), 5-30.
- Miller, N. D., Challoner, J., & Atta, A. (2007). UK Welcomes the Sukuk-How the UK Finance Bill Should Stimulate Islamic Finance in London, Much to the Delight of the City's Banks. *Int'l Fin. L. Rev.*, 26, 24.
- Mirakhor, A. (1995). Theory of an Islamic Financial System. *Encyclopaedia of Islamic Banking*.
- Mirakhor, A., & Zaidi, I. (1988). *Stabilization and Growth in an Open Islamic Economy*.
- Mussa, M. (1979). Empirical regularities in the behavior of exchange rates and theories of the foreign exchange market. Paper presented at the Carnegie-Rochester Conference Series on Public Policy.
- Nelson, D. B. (1990). Stationarity and persistence in the GARCH (1, 1) model. *Econometric theory*, 6(03), 318-334.

- Nelson, D. B. (1991). Conditional heteroskedasticity in asset returns: A new approach. *Econometrica: Journal of the Econometric Society*, 347-370.
- Newey, W. K., & West, K. D. (1986). A simple, positive semi-definite, heteroskedasticity and autocorrelation consistent covariance matrix: National Bureau of Economic Research Cambridge, Mass., USA.
- Nikolaev, N. Y., Boshnakov, G. N., & Zimmer, R. (2013). Heavy-tailed mixture GARCH volatility modeling and Value-at-Risk estimation. *Expert Systems with Applications*, 40(6), 2233-2243.
- Orhan, M., & Köksal, B. (2012). A comparison of GARCH models for VaR estimation. *Expert Systems with Applications*, 39(3), 3582-3592.
- Ozun, A., Cifter, A., & Yilmazer, S. (2007). Filtered Extreme Value Theory for Value-At-Risk Estimation.
- P. Abad, S. B. (2012). A detailed comparison of value at risk estimates. *Mathematics and Computers in Simulation*.
- Politis, D. N. (2004). A heavy-tailed distribution for ARCH residuals with application to volatility prediction.
- Poon, S. H., & Granger, C. W. J. (2003). Forecasting volatility in financial markets: A review. *Journal of Economic Literature*, 41(2), 478-539.
- Presley, J. R., & Sessions, J. G. (1994). Islamic economics: the emergence of a new paradigm. *The Economic Journal*, 584-596.
- Report, G. S. (Q1 2015).
- Ro'i, Y. (2015). *The USSR and the Muslim world: Issues in domestic and foreign policy*: Routledge.
- Rosly, S. A. (2005). *Critical issues on Islamic banking and financial markets: Islamic economics, banking and finance, investments, Takaful and Financial Planning*. Dinamas Publishing.
- Sarma, M., Thomas, S., & Shah, A. (2003). Selection of Value-at-Risk models. *Journal of Forecasting*, 22(4), 337-358.
- Schaumburg, J. (2012). Predicting extreme value at risk: Nonparametric quantile regression with refinements from extreme value theory. *computational statistics and data analysis*.
- Schwarz, G. (1978). Estimating the dimension of a model. *the Annals of Statistics*, 6(2), 461-464.

- Shaikh, S., & Saeed, S. (2010). Sukuk bond: The global Islamic financial instrument. MPRA Paper, 26700.
- Siddiqi, M. N. (2005). Riba, Bank interest and the rationale of its prohibition: Markazi Maktaba Islami Publishers.
- Tariq, A. A. (2004). Managing financial risks of sukuk structures. Loughborough University, UK, September (mimeo).
- Theodossiou, P. (1998). Financial data and the skewed generalized t distribution. *Management Science*, 44(12-Part-1), 1650-1661.
- Tsay, R. S. (2005). *Analysis of financial time series* (Vol. 543): Wiley. com.
- Usmani, M. T. (2008). Sukuk and their contemporary applications. Internet download, 1-16.
- Usmani, T. (2007). *ṣukūk and their Contemporary Applications*. Retrieved from the internet.
- Verhoeven, P., & McAleer, M. (2004). Fat tails and asymmetry in financial volatility models. *Mathematics and Computers in Simulation*, 64(3), 351-361.
- Vishwanath, S. R., & Azmi, S. . (2009). An overview of Islamic Sukuk bonds. *The Journal of Structured Finance*, 14(4), 58-67.
- Wilhelmsson, A. (2006). Garch forecasting performance under different distribution assumptions. *Journal of Forecasting*, 25(8), 561-578.
- Williams, C. (2011). Research methods. *Journal of Business & Economics Research (JBER)*, 5(3).
- Wilson, R. (2004). Overview of the sukuk market. *Islamic Bonds: Your Guide to Issuing, Structuring and Investing in Sukuk*, Euromoney Books, London, 6-7.
- Wilson, R. (2008). Innovation in the structuring of Islamic sukuk securities. *Humanomics*, 24(3), 170-181.
- Wilson, R. (2008). Innovation in the structuring of Islamic< IT> sukuk</IT> securities. *Humanomics*, 24(3), 170-181.
- Wirch, J. L. (1997). Value-at-Risk for Risk Portfolios. Unpublished Working Paper.
- Wu, G. (2001). The determinants of asymmetric volatility. *Review of Financial Studies*, 14(3), 837-859.
- Yamai, Y., & Yoshihara, T. (2005). Value-at-risk versus expected shortfall: A practical perspective. *Journal of Banking & Finance*, 29(4), 997-1015.

- Zamir, I., & Tsubota, H. (2009). The World Bank, Emerging Islamic Capital Markets–A Quickening Pace and a New potential. World bank Report by World bank.
- Zikovic, S., & Aktan, B. (2009). Global financial crisis and VaR performance in emerging markets: A case of EU candidate states-Turkey and Croatia. Zbornik radova Ekonomskog fakulteta u Rijeci, 27(1), 149-170.