Modelling Volatility of Daily Stock Returns: Evidence from Select Banking and Finance Companies in India

Tapas Kumar Tripathy

Assistant Professor, Department of Commerce
Prabhat Kumar College, Contai
Email ID: tapastripathy2007@mail.com

Abstract

This study empirically explores the volatility pattern of NSE listed banking and finance companies based on time series dataset taking into consideration of daily closing adjusted stock price from 2001-02 to 2015-16. The objective of this paper is to study volatility design of daily stock returns. The application of GARCH, T-GARCH models provides the evidence of the persistence of time varying asymmetric volatility. Time based volatility behaviour of Indian stock market happened due to recent global financial meltdown which is originated from US sub-prime crisis. Also, effect captured by different models show that negative shocks have significant effect on conditional volatility.

Key Words: Asymmetric volatility, Conditional volatility, financial meltdown

JEL Classification: C32, C53, G28

1. Introduction

Volatility measures the changing amount of uncertainty or risk in securities price. It is becoming crucial and important due to growing linkages of national market in currency, different economic events, commodity and stocks with rest of the world markets. Exotic of capital market always needs with high macro fundamentals like relevant corporate information. After liberalization, privatization and globalization, financial markets have entered into a new segment of global integration and liberalization with lots of new and innovative financial instruments like derivatives, etc. Derivative trading in India have started to control over the fluctuations in the stock and commodity prices. It also regarded that now a days the trading volume in the stock market and cash flows in India are increasing (Yadav, 2016). After the liberalization, privatization and globalization, volatility has become an important area and it has gained attention from academicians, researchers and others stakeholders. Stock return volatility is variability in stock price returns which is usually measured by standard deviation

(S.D.) of the stock price returns. For the benefit of research, we have done log transformation of stock price return series. Volatility is considered superior returns and higher volatility due to risk (Kumar, 2016). Stock price returns are essentially non-linear system in nature, because it derives from a stochastic process. However, past researchers have used both linear and nonlinear model. Modelling volatility are beneficial due to several reasons such as investment decision making, assets pricing calculation, expected returns calculation etc. There are different divisional properties in case of financial time series, such as leptokurtosis and volatility (Tsay, 2016). As volatility is not apparent, it throws a challenge both for Autoregressive conditional heteroskedasticity (ARCH) as well as Generalized Autoregressive conditional heteroskedasticity (GARCH) model and searching right model of ARCH and GARCH model become matter of crucial significance. Searching exact set of reasons of stock returns volatility out of huge no of observable and as well as unobservable variables namely e.g., changes in economic policy situation, different shocks in the capital market, psychological facts and different rumors which are not easily observable. Against this backdrop, the paper seeks to quantify stock returns volatility of select stock of NSE listed banking sector companies using daily adjusted closing stock price return.

2. Past studies and Research Gap

Researchers and scholars across the globe have contributed in the study of measurement of stock returns volatility of select companies long time since. A few of such representative works are reviewed in this segment. In one of the earlier studies, Ali (2016) in his research paper 'Stock market volatility and returns: A study of NSE & BSE in India' has examined relationship between returns and volatility and persistence of volatility in Indian stock market. This study shows that there exists a significant presence of volatility clustering and degree of volatility in daily return series and existence of leverage effect indicating that negative shocks or bad news have more impact on volatility than the positive shocks or good news this study also exhibits that the relation between returns and volatility at the return series were statistically insignificant. Bhowmik (2013) in the research paper 'Stock market volatility: An evaluation' has proposed a descriptive study on stock market volatility in Indian stock exchanges and used seven case studies. He emphasized on political factors behind volatility and attempted to relate economic growth with stock market volatility in the long run process after reviewing a few econometric models and concludes that political instability and depression catapulted the stock market volatility. Banumathy et al. (2015) in the research paper 'Modelling stock market volatility: Evidence from India' have studied volatility pattern of Indian stock market based on index data using both symmetric and asymmetric models. The study has also argued that

increased risk did not increase the returns. Eryilmaz (2015) in the research paper 'Modelling stock market volatility: The case of BIST – 100' has investigated stock return volatility for BIST –100 indices. He has divided study period into three different time periods using ARCH, GARCH, EGARCH and TGARCH models. According to this study, most suitable volatility model for return series is EGARCH (1,1) and negative news have more effect to the stock market. Kumar (2009) in the research paper 'Volatility in the Indian Stock Market: A Case of Individual Securities' has investigated volatility in the individual stocks listed at NSE using daily closing prices of 29 selected companies from the S&P NIFTY from 1996- 97 to 2006 -07. Major findings are negative stock returns during 2000-01 followed by 2002-03, 2006-07 and 1998-99. Particularly 1999-00 has been highly volatile year for 20 securities out of 29 securities followed by 1998-99, 2000-01 and 1997-98. Kumar & Singh (2014) in the research paper 'Volatility modeling, seasonality and risk return relationship in GARCH- in- mean framework: the case of Indian stock and commodity markets' have examined volatility and seasonality in Indian stock and commodity markets. Their study concluded that stock returns have shown persistence in the volatility clustering and asymmetric properties. Also, concluded that conditional volatility structure GARCH (1, 1) is found to be more appropriate. Padhi (2006) in the research paper 'Stock Market Volatility in India: A Case of Select Scripts' has investigated that market volatility at the individual script level and at the indices level to know how volatility changes in the same trend or it varies across the sectors and conducted that LM test is using to confirm the presence of ARCH effect. Different ARCH coefficients are found for different indices at different lag values and argued that many sectors showing the same trend for volatility characteristics. Naveen & Mallikarjunappa (2016) in the research paper 'A study on comparative analysis of risk and return with reference to stocks of CNX BANK NIFTY' have examined the banking sector securities risk and return and their volatility. The study has initiate that some stocks move in the opposite direction to the market, some stocks move along with the market, some stock are less risky compared to the market and some stocks are riskier compared to the market. A few studies identify the reasons and factors for volatility in Indian stock market as well as different risk factors associated with sectorial indices and such volatility is how much important in Indian future stock indices market. Other studies have explored the relationship between return of the securities and market return like different indices and their co – movement.

In addition to previous studies, there is a need for further study on measurement of stock returns volatility in order to have perception of different companies listed in major stock exchanges in India like NSE and to explore how volatility of individual script changes with respect to

different time period in respect to different economic policies, incident, etc. and underlying different factors and shocks which can affect individual securities. Keeping in mind of this research gap, specific objectives of the current study are set.

3. Objectives

The objectives of the present study are:

- (i) To study the stock returns volatility property of select stock of NSE listed banking sector companies using descriptive statistics;
- (ii) To know the presence of volatility in sample companies daily return series using ARCH (1) model;
- (iii) To analyse volatility in select NSE listed companies using GARCH and TGARCH Model.

4. Research Methodology

Sample Design and Sources of data

This study is considering fifteen years secondary data (daily closing adjusted prices) acquire from Capitaline corporate database from 2001-02 to 2015-16. Judgemental sampling has been used to investigate the volatility of select companies as sample for the study. With a focus on designing sample, at first market capitalisation has taken as parameter for sample selection and one sector namely banking and finance include of 18 companies are taken into reflection as final sample for this study.

Period of Study

In our study, we have considered major financial changes in the global financial meltdown phenomena such as e.g., global financial crisis, rupee value uncertainty, US sub-prime crisis, borrowing price bubble in US financial markets etc. The total study period has been divided into two parts: pre-recession period (2001-2002 to 2006-2007) and post-recession period (2007-2008 to 2015-2016). The entire analysis has been conducted during 2001-2002 to 2006-2007, 2007-2008 to 2015-2016 and 2001-02 to 2015-16.

Statistical and Econometrics tools

Descriptive statistics: Within the sphere of descriptive statistics, we have used mean, S.D., variance, skewness and kurtosis. The Jarque-Bera test is usually used to know whether normality exists in the data set.

Test of Stationarity: In order to persuade the fundamental situation time series econometrics, there is always need for testing unit root. In this paper Augmented Dickey-Fuller (ADF) test is used to know the Stationarity level. In general, the approaches to unit root test are

$$Y_t = D_t + Z_t + \varepsilon_t$$

Where, D_t is known the deterministic component; Z_t is also known the stochastic component and ε_t meant stationary error process. ADF test consists following regression equation:

$$\nabla Y_{t} = \beta_{1} + \beta_{2}t + \delta Y_{t-1} + \sum_{i=1}^{m} \alpha_{i} \nabla Y_{t-1} + \varepsilon_{t}$$

Where ε_t are used as a pure noise error with the number of lagged differences. The main summary based on this test is known the ADF is statistically significant at 5% level or not. This specifies to reject null hypothesis and accept that the data series returns are stationery; hence, it is mean reverting and confirms the non-existence of autocorrelation.

Test for Heteroscedasticity: In order to know heteroscedasticity of our dataset, we have used ARCH model. ARCH model: $\sigma_t^2 = a_0 + \sum_{t=1}^q a_t U_{t-1}^2$ (Where a_0 is mean and a_1 is conditional volatility and Ut - 1 is white noise representing residual of time series).

Volatility Measurement Technique:

GARCH Model

In this study, we have used GARCH (1, 1) model which is more superior to ARCH model. It happens both lagged squared residuals and lagged variances. This model is considered as generalized ARCH exercise to see the squared volatility (σ_t^2) of the concerned time period is allowed to depend on previous squared volatilities, as well as previous squared values of the process.

Mean equation: $r_t = \mu + \varepsilon_t$ and Variance equation: $\sigma_t = \alpha + \alpha \varepsilon_{t-1}^2 + \beta \alpha_{t-1}^2$

Where $\alpha > 0$, $\alpha_1 > 0$, $\beta_1 > 0$, and r_t is expressed as the return of the asset at the time t, μ is the average return, and ε_t is the residual return and estimate (β) shows the persistence of volatility of a shock.

TGARCH model

Again, TGARCH model is used to express positive and negative shocks of equal magnitude have a different impact on stock returns volatility, which may be describe to a 'leverage effect'. The conditional variance for the simple TGARCH model is defined follows-

$$\sigma_{t}^{2} = \alpha_{0} + \sum_{i=1}^{s} (\alpha_{i} + \gamma_{i} N_{t-1}) a_{t-1}^{2} + \sum_{i=1}^{m} \beta_{j} \sigma_{t-j}^{2}$$

Where N_{t-1} is known as negative indicator, a_{t-1} and $\alpha_i, \gamma_{i,}$ and β_j are used nonnegative parameter fulfilling conditions similar to those of GARCH model. This model also considers zero as its threshold to separate the impact of past shocks.

5. Results and Discussion

An important observation in our article is in relation to changes squeeze from pre to post global financial recession. Mean returns of selected stock price are establish almost low in all three study periods. S.D. and variances of stock price return series are beginning to be low but significant in pre-global financial recession period (shown in Table – 1.11)and the standard deviation of return is found to be significant (3.4% S.D. in case of Kotak Mahindra Bank) during the pre-global financial meltdown period, which is comparatively lower than other study periods (Post global financial recession period and overall period). The null hypothesis (Ho: Return series are normal) of JB test was rejected for all cases and does not follow a normal distribution. Normality test results exhibited in Table 1.21.

Test of Stationarity

The summary of ADF tests at level are shown in Table- 1.31. We have studied the complete dataset are Stationary for all select companies return data series as test statistics is lower than all critical values.

ARCH test

In our investigation about heteroscedasticity of stock price return series of eighteen stocks, we found ARCH effect in all stock during pre global financial recession time period shown in Table- 1.41. However, the same test is conducted during post global recession period exhibited non existence of heteroscedasticity only in State Bank of India (SBIN) and Bajaj Holdings stock price return series, which supports presence of ARCH effects in the residuals.

GARCH test

During pre-global financial recession period, the highest and lowest ARCH and GARCH combined effect was found to be .984 (Federal Bank script return series) and .757 (ICICI Bank script return series), while such values of seven companies in banking and financial sector are near to one during post global financial recession period. Again, in overall study period, average value of ARCH and GARCH are ranges between .992 and .798. Out of eighteen companies seventeen companies' average value of ARCH and GARCH is unity shown in 1.71. This result apparently points out that global financial meltdown shocks have persistent for future periods.

T-GARCH test

T-GARCH model is run on daily stock return series to understand the asymmetric coefficient effect. After conducting TGARCH test during different study period and in different dataset. Test results concluded that coefficient of leverage (γ) is positive in maximum scripts returns series. It is also noticed that negative shocks or bad news have a greater effect on the conditional variance than the positive shocks or good news;

6. Conclusions

Housing price bubble in USA in 2008 was one of the major findings for financial meltdown around the world over time. Its effect has been found not only in all segment of financial markets but also in every outlook of socio-economic life. Measurement of volatility is important for determining cost of capital for financial assets, also for leverage and investment. There has been a matter of time varying volatility from stock price return of select companies that exhibited the sign of clustering, high persistence, and predictability in India stock market. The other important result is that returns acknowledge differently to the arrival of negative and positive shocks. The primary empirical findings of bank and financial stock price return series are far from normality. In maximum cases in GARCH (1, 1) model, the sum of the coefficient ($\alpha + \beta$) is near to one, which suggest that the volatility is highly persistent. The coefficient (γ) measures TGARCH asymmetry or leverage parameter has exhibited that only four companies under banking and financial sector are found negative effect. According to TGARCH model in financial time series, it is concluded that good news has an impact on ARCH term, while bad news has impact on ARCH as well as GARCH term.

References

- Ali, I. (2016) Stock market volatility and returns: A study of NSE & BSE in India, *IJHSSS*,
 Vol. II(IV), pp. 202 211
- Banumathy, K., Azhagaiah, R. (2015). Modelling stock market volatility: Evidence from India, *Managing Global Transition*, Vol. 13, pp. 27 -42
- Bhowmik, D. (2013). Stock Market Volatility: An Evaluation, *IJSRP*, Vol. 3(10), pp. 1-18
- Eryilmaz, F.(2015). Modelling stock market volatility: the case of BIST 100, *ABP*, Vol. 1(5), pp. 37 47
- Joshi, P. (2014). Forecasting Volatility of Bombay Stock Exchange, *IJCRAR*, *Vol.* 2, No. 7,
 pp. 222- 230

- John, A., Oduro, F. T., Prah, J. A. (2016). Modelling volatility and the risk return relationship of some stocks on the Ghana stock exchange, *American Journal of Economics*, Vol. 6, No. 6,pp. 281 299
- Kumar, N. S. (2016). Study on volatility of stock market data of selected companies engaging in retail finance in India, *PIJMIS*, Vol. 5(1), pp. 1 13
- Kumar, R. (2009). Volatility in the Indian Stock Market: A Case of Individual Securities, *JARE*, Vol. 1(1), 2009, pp. 43 – 53
- Kumar, B., Singh, P., (2008). Volatility modeling, seasonality and risk return relationship in GARCH- in- mean framework: the case of Indian stock and commodity markets, *Indian Institute of Management*, Ahmedabad Working Paper, pp. 01 31
- Mallikarjunappa, T., Afsal, E. M. (2008). The Impact of Derivatives on stock Market Volatility: A Study of the Nifty Index, *AAMJAF*, Vol. 4(2), pp. 43-65
- Najjar, D. A. L. (2016). Modelling and estimation of volatility using ARCH/GARCH models in Jordan's stock market, *Asian Journal of Finance & Accounting*, Vol. 8 (1), pp. 152 167
- Padhi, P. (2006). Stock returns Volatility: A case study of Pharmaceuticals and Computer software in India, *Journal of Academic Research in Economics*, Vol. 1(1), pp. 1-13
- Padhi, P. (2006). Stock Market Volatility in India: A Case of Select Scripts, *Journal of Academic Research in Economics*, Vol. 1(2), pp. 1-54
- Raju, M. T. & Ghosh, A. (2008). Stock market volatility- An international comparison, SEBI working paper series no. 8, pp. 1-19
- Shankar, C., Ramula, K. (2014). Volatility and Correlation of Stock Indices on Indian Stock Market, *IJRBM*, Vol. 2(4), pp. 17- 26
- Tripathy, Tapas Kumar (2020). Modelling volatility of daily stock returns: Evidence from NSE listed computer software & multimedia scripts, *IJARCMSS*, Vol. 03(3), pp. 141 148
- Vijayalakshmi, S., Gaur, S. (2013). Modeling Volatility: Indian Stock and Foreign Exchange Markets, *JEIEFB*, *Vol.* 2 (1), pp. 583-178
- Yadav, S. (2016). Impact of derivatives trading on the volatility of stock market in India: A review, *Asian Journal of Management Research*, Vol. 6(3), pp. 567 579
- Zakaria, Z., Muhammad, J., Zulkifli, A. H. (2012). The impact of dividend policy on the share price volatility: Malaysian construction and materials companies, *IJEMS*, *Vol.* 2(5), pp. 1-8

Tables

Table – 1.11

Descriptive Statistics results of Daily Adjusted Return

Company Name	Pre	-Globa	l recession	Period	Pos	st -Global	recession P	eriod
	Mean	S. D.	Kurtosis	Skewness	Mean	S. D.	Kurtosis	Skewness
SBIN	0.001	0.022	6.13	-0.27	0.0042	0.19	2112.02	45.27
Kotak Mah. Bank	0.002	0.034	9.03	0.86	0.0011	0.027	8.35	0.075
Federal Bank	0.002	0.03	9.79	0.52	0.0007	0.023	5.82	0.2
Oriental Bank	0.001	0.029	9.93	-0.016	0.0001	0.03	5.23	0.14
HDFC Bank	0.001	0.021	19.38	0.81	0.0009	0.02	8.16	0.45
ICICI Bank	0.001	0.025	6.07	0.18	0.0005	0.028	8.84	0.26
IDBI Bank	0.001	0.035	8.62	0.52	0.0003	0.02	7.5	0.25
Corporation Bank	0.0009	0.028	7.67	0.43	6.47	0.022	7.9	0.4
Bank of Baroda	0.0011	0.031	7.87	0.09	0.0009	0.026	8.65	0.71
Axis Bank	0.0022	0.03	8.24	0.6	0.001	0.027	6.62	0.35
Syndicate Bank	0.0014	0.029	9.53	0.41	0.0004	0.026	5.81	0.02
Bank of India	0.0022	0.032	6.61	0.35	0.0002	0.03	7.38	0.357
Bajaj Holdings	0.0007	0.026	176.55	-7.37	0.0002	0.026	149.6	-5.94
HDFC	0.0013	0.022	6.67	0.33	0.0008	0.024	9.58	0.68
Max Financial	0.0017	0.033	9.73	1.21	0.0005	0.027	10.08	0.1
Sundaram								
Finance	0.0013	0.025	10.86	1.09	0.0013	0.02	7.46	0.53
Reliance Capital	0.002	0.033	8.92	0.21	0.0003	0.034	7.81	0.22
Bajaj Fin.	0.0013	0.029	9.93	0.82	0.0016	0.027	13.33	1.11

Table- 1.21
Descriptive Statistics and Normality Test results of Daily Adjusted Return

Company Name		Overal	ll Period			Norma	ality Test Res	ult
							Decision	
							Rule – Ho	
							Rejected	
							when P	
							value <5%	
							(Ho: Return	
			Kurt-				series are	Data series
	Mean	S. D.	osis	Skewness	J-B	Prob	normal)	normality
SBIN	0.003	0.147	3477	57.86	1.88	.000	Rejected	Not normal
Kotak Mah. Bank	0.0019	0.03	10.03	0.48	7984.7	.000	Rejected	Not normal
Federal Bank	0.0013	0.027	9.4	0.48	6411.1	.000	Rejected	Not normal
Oriental Bank	0.0006	0.029	7.48	0.091	3137	.000	Rejected	Not normal
HDFC Bank	0.001	0.0202	16.12	0.65	3.94	.000	Rejected	Not normal
ICICI Bank	0.0008	0.026	8.22	0.26	4295.6	.000	Rejected	Not normal
IDBI Bank	0.0007	0.031	8.6	0.52	5070.4	.000	Rejected	Not normal
Corporation							Rejected	
Bank	0.0004	0.024	8.56	0.49	4961.6	.000		Not normal
Bank of Baroda	0.001	0.028	8.95	0.38174	5626.8	.000	Rejected	Not normal
Axis Bank	0.0016	0.028	8.05	0.51	4140.7	.000	Rejected	Not normal
Syndicate Bank	0.0009	0.027	8.08	0.2	4311.1	.000	Rejected	Not normal
Bank of India	0.001	0.3	6.79	0.31	2308.5	.000	Rejected	Not normal
Bajaj Holdings	0.0008	0.024	130.8	-4.62	255651	.000	Rejected	Not normal
HDFC	0.001	0.022	9.19	0.63	6217.6	.000	Rejected	Not normal
					10413.		Rejected	
Max Financial	0.0012	0.03	10.98	0.87	6	.000		Not normal
Sundaram					10266.		Rejected	
Finance	0.0013	0.022	10.88	0.98	8	.000		Not normal
Reliance Capital	0.001	0.033	8.55	0.27	4844.4	.000	Rejected	Not normal
		_			13069.		Rejected	
Bajaj Fin.	0.0019	0.028	11.92	1.07	9	.000		Not normal

Table- 1.31
Stationarity Test Results using Augmented Dickey Fuller (At Level)

Compa	Intercept		Interce	pt &	None		Decisio	n Rule	Decision on	Data series
nies			Trend				(Test	statistics	Ho (Ho: Unit	stationarity
Name	Test	C.V.	Test	C.V.	Test	C.V.	value	is more	root in Data	
	Statistics	(5%)	Statis	(5%)	Statis	(5%)	negativ	e than	series)	
			tics		tics		critical	value,		
							then v	we reject		
							Null H	ypothesis)		
		-2.86		-3.41		-1.94	More	negative	Ho rejected &	Stationary
							test	statistics	Return series	series
							(P=0.00	00)	do not have	
SBIN	-67.48		-67.48		-67.46				any unit root	
		-2.86		-3.41		-1.94	More	negative	Ho rejected &	Stationary
Kotak							test	statistics	Return series	series
Mah.							(P=0.00	00)	do not have	
Bank	-26.96		-27.04		-56.04				any unit root	
		-2.86		-3.41		-1.94	More	negative	Ho rejected &	Stationary
							test	statistics	Return series	series
Federal							(P=0.00	00)	do not have	
Bank	-18.09		-18.21		-17.85				any unit root	
		-2.86		-3.41		-1.94	More	negative	Ho rejected &	Stationary
							test	statistics	Return series	series
Oriental							(P=0.00	00)	do not have	
Bank	-19.23		-19.34		-19.19				any unit root	
		-2.86		-3.41		-1.94	More	negative	Ho rejected &	Stationary
							test	statistics	Return series	series
HDFC							(P=0.00	00)	do not have	
Bank	-27.78		-27.78		-30.06				any unit root	
		-2.86		-3.41		-1.94	More	negative	Ho rejected &	Stationary
							test	statistics	Return series	series
ICICI							(P=0.00	00)	do not have	
Bank	-27.78		-27.29		-27.19				any unit root	
		-2.86		-3.41		-1.94	More	negative	Ho rejected &	Stationary
IDBI							test	statistics	Return series	series
Bank	-19.09		-19.11		-19.05		(P=0.00	00)		

								do not have	
								any unit root	
		-2.86		-3.41		-1.94	More negative	Ho rejected &	Stationary
Corpora							test statistics	Return series	series
tion							(P=0.000)	do not have	
Bank	-56.89		-56.93		-56.88			any unit root	
		-2.86		-3.41		-1.94	More negative	Ho rejected &	Stationary
							test statistics	Return series	series
Bank of							(P=0.000)	do not have	
Baroda	-18.68		-18.71		-19.03			any unit root	
		-2.86		-3.41	-27.00	-1.94	More negative	Ho rejected &	Stationary
							test statistics	Return series	series
Axis							(P=0.000)	do not have	501105
Bank	-44.75		-44.55		-59.47		(any unit root	
		-2.86		-3.41		-1.94	More negative	Ho rejected &	Stationary
							test statistics	Return series	series
Syndicat							(P=0.000)	do not have	
e Bank	-30.21		-43.89		-30.15			any unit root	
		-2.86		-3.41		-1.94	More negative	Ho rejected &	Stationary
							test statistics	Return series	series
Bank of							(P=0.000)	do not have	
India	-28.35		-26.45		-28.27			any unit root	
		-2.86		-3.41		-1.94	More negative	Ho rejected &	Stationary
Bajaj							test statistics	Return series	series
Holding							(P=0.000)	do not have	
S	-14.21		-14.25		-14.06			any unit root	
		-2.86		-3.41		-1.94	More negative	Ho rejected &	Stationary
							test statistics	Return series	series
							(P=0.000)	do not have	
HDFC	-26.96		-26.98		-26.66			any unit root	
		-2.86		-3.41		-1.94	More negative	Ho rejected &	Stationary
Max							test statistics	Return series	series
Financi							(P=0.000)	do not have	
al	-22.6		-22.62		-22.47			any unit root	

		-2.86		-3.41		-1.94	More	negative	Ho rejected &	Stationary
Sundara							test	statistics	Return series	series
m							(P=0.0	00)	do not have	
Finance	-61.01		-61.01		-60.8				any unit root	
		-2.86		-3.41		-1.94	More	negative	Ho rejected &	Stationary
							test	statistics	Return series	series
Reliance							(P=0.0	00)	do not have	
Capital	-33.99		-34.02		-33.94				any unit root	
		-2.86		-3.41		-1.94	More	negative	Ho rejected &	Stationary
							test	statistics	Return series	series
Bajaj							(P=0.0	00)	do not have	
Fin.	-59.9		-59.89		-59.64				any unit root	

Table- 1.41
ARCH Test for Heteroskedasticity (Pre -Global recession Period)

Banking	F-statistic	Prob.	Obs* R-	Prob.	Decision on Ho	ARCH effects a	are
and Finance		F	squared	Chi-	(Null hypothesis	present or not	
Companies				Square	of		
					homoskedasticit		
					y of residuals is		
					rejected if P-		
					Value<0.05)		
		0.000			Rejected	ARCH effects a	are
SBIN	222.6		194.09	0.000		present	
Kotak Mah.		0.000			Rejected	ARCH effects a	are
Bank	162.53		146.83	0.000		present	
Federal		0.000			Rejected	ARCH effects a	are
Bank	232.78		200.63	0.000		present	
Oriental		0.000			Rejected	ARCH effects a	are
Bank	596.03		427.11	0.000		present	
		0.000			Rejected	ARCH effects a	are
HDFC Bank	382.1		304.93	0.000		present	
		0.000			Rejected	ARCH effects a	are
ICICI Bank	169.94		152.85	0.000		present	

Tapas Kumar Tripathy

		0.000			Rejected	ARCH	effects	are
IDBI Bank	152.48		138.59	0.000		present		
Corporation		0.000			Rejected	ARCH	effects	are
Bank	68.57		65.66	0.000		present		
Bank of		0.000			Rejected	ARCH	effects	are
Baroda	448.17		345.53	0.000		present		
		0.000			Rejected	ARCH	effects	are
Axis Bank	92.27		87.04	0.000		present		
Syndicate		0.000			Rejected	ARCH	effects	are
Bank	90.82		85.75	0.000		present		
Bank of		0.000			Rejected	ARCH	effects	are
India	124.02		114.7	0.000		present		
Bajaj		0.000			Rejected	ARCH	effects	are
Holdings	23.93		23.59	0.000		present		
		0.000			Rejected	ARCH	effects	are
HDFC	55.32		55.43	0.000		present		
Max		0.000			Rejected	ARCH	effects	are
Financial	74.46		71.03	0.000		present		
Sundaram		0.000			Rejected	ARCH	effects	are
Finance	166.12		149.67	0.000		present		
Reliance		0.000			Rejected	ARCH	effects	are
Capital	52.56		50.85	0.000		present		
		0.000			Rejected	ARCH	effects	are
Bajaj Fin.	36.78		35.95	0.000		present		

Table- 1.51

ARCH Test for Heteroskedasticity (Post -Global recession Period)

Banking	F-	Prob. F	Obs* R-	Prob.	Decision on Ho	ARCH effects
and Finance	statistic		squared	Chi-	(Null hypothesis of	are present or
Companies				Square	homoskedasticity	not
					of residuals is	
					rejected if P-	
					Value<0.05)	
ann.	00044	002	00044	002	Accepted	No ARCH
SBIN	.00044	.983	.00044	.983		effects

Kotak Mah.					Rejected	ARCH effects
Bank	120.13	.000	114.09	.000		are present
Federal					Rejected	ARCH effects
Bank	39.19	.000	38.55	.000		are present
Oriental					Rejected	ARCH effects
Bank	36.76	.000	36.20	.000		are present
HDFC Bank	34.75	.000	34.25	.000	Rejected	ARCH effects are present
ICICI Bank	164.72	.000	15.51	.000	Rejected	ARCH effects are present
IDBI Bank	213.08	.000	194.65	.000	Rejected	ARCH effects are present
Corporation					Rejected	ARCH effects
Bank	33.68	.000	33.21	.000		are present
Bank of					Rejected	ARCH effects
Baroda	58.80	.000	57.34	.000		are present
Axis Bank	130.50	.000	123.39	.000	Rejected	ARCH effects are present
Syndicate					Rejected	ARCH effects
Bank	36.11	.000	35.62	.000		are present
Bank of					Rejected	ARCH effects
India	29.47	.000	29.11	.000		are present
Bajaj					Accepted	No ARCH
Holdings	1.85	.1736	1.85	.173		effects
HDFC	37.05	.000	36.48	.000	Rejected	ARCH effects are present
Max					Rejected	ARCH effects
Financial	27.33	.000	27.03	.000		are present
Sundaram					Rejected	ARCH effects
Finance	164.57	.000	153.38	.000		are present
Reliance					Rejected	ARCH effects
Capital	40.61	.000	39.92	.000		are present
Bajaj Fin.	146.60	.000	137.67	.000	Rejected	ARCH effects are present

Table- 1.61
ARCH Test for Heteroskedasticity (Overall Period)

Banking	F-	Prob. F	Obs* R-	Prob.	Decision on Ho	ARCH effects
and	statistic		squared	Chi-	(Null hypothesis	are present or
Finance				Square	of	not
Companies					homoskedasticity	
					of residuals is	
					rejected if P-	
					Value<0.05)	
SBIN	.00025	.98	.000256	.98	Accepted	No ARCH effects
Kotak Mah.					Rejected	ARCH effects
Bank	331.17	0.000	304.34	0.000	· ·	are present
Federal					Rejected	ARCH effects
Bank	438.65	0.000	392.02	0.000		are present
Oriental					Rejected	ARCH effects
Bank	496.03	0.000	438.05	0.000		are present
HDFC Bank	507.99	0.000	447.35	0.000	Rejected	ARCH effects are present
ICICI Bank	306.28	0.000	283.20	0.000	Rejected	ARCH effects are present
IDBI Bank	370.28	0.000	337.03	0.000	Rejected	ARCH effects are present
Corporation					Rejected	ARCH effects
Bank	131.50	0.000	127.09	0.000		are present
Bank of					Rejected	ARCH effects
Baroda	488.39	0.000	432.09	0.000		are present
Axis Bank	232.30	0.000	218.80	0.000	Rejected	ARCH effects are present
Syndicate					Rejected	ARCH effects
Bank	173.29	0.000	166.15	0.000		are present
Bank of					Rejected	ARCH effects
India	119.40	0.000	115.76	0.000		are present
Bajaj					Accepted	No ARCH
Holdings	4.75	.029	4.75	.029		effects
HDFC	77.76	0.000	76.22	0.000	Rejected	ARCH effects are present

Max					Rejected	ARCH effects
Financial	134.13	0.000	129.54	0.000		are present
Sundaram					Rejected	ARCH effects
Finance	386.04	0.000	349.94	0.000		are present
Reliance					Rejected	ARCH effects
Capital	202.49	0.000	192.17	0.000		are present
Bajaj Fin.	80.77	0.000	79.09	0.000	Rejected	ARCH effects are present

Table-1.71 (GARCH Model)

Company Name	Esti	mated	Model	with	AIC	SIC	Log	Decision					
		val	lues				Likelihood	(Decision Rule:					
								Volatility of					
T	irct Pari	od - Co	officia	nts - GAF	<u>РСИ (1</u>	1)		shocks is highly persistence when					
r	iist i ei	ou - C	Jemene	iis - GAr	CII (1,	1)		$\alpha_i + \beta_i = 1$					
Banking and	$lpha_{\scriptscriptstyle 0}$	$\alpha_{_{1}}$	$eta_{_{1}}$	$\alpha_j + \beta_i$									
Finance Sector													
State Bank of	1.74	.088	.875	.963	-4.96	-4.94	3734.3	Very high					
India								persistence value					
	.0001	.227	.617	.844	-4.14	-4.12	3118.7	Comparatively					
Kotak Mah. Bank								low persistence					
Kotak Man. Dank								value					
	.0001	.163	.736	.899	-4.24	-4.22	3064.6	Comparatively					
Federal Bank								low persistence value					
	2.22	.164	.820	.984	-4.58	-4.56	3451.7	Variue Very high					
Oriental Bank	2.22	.104	.020	.704	-4.50	-4.50	3431.7	persistence value					
	4.44	.150	.748	.898	-5.02	-5.00	3780.4	Comparatively					
								low persistence					
HDFC Bank								value					
	.0001	.178	.579	.757	-4.64	-4.62	3495.6	Comparatively					
ICICI Bank								low persistence					
Terer Bunk	0001	170	(75	051	4.05	4.02	3050.4	value					
	.0001	.179	.675	.854	-4.05	-4.03	3030.4	Comparatively low persistence					
IDBI Bank								value					
	6.11	.161	.776	.937	-4.42	-4.40	3327.8	Comparatively					
Composition Darie								low persistence					
Corporation Bank								value					

	4.50	100	004	0.70	4.20		2220.4	
Bank of Baroda	4.68	.132	.821	.953	-4.29	-4.27	3229.4	Very high
	0001	1.42	606	920	4.27	4.25	2214.4	persistence value
	.0001	.143	.686	.829	-4.27	-4.25	3214.4	Comparatively
Axis Bank								low persistence value
	7.24	.188	.739	.927	-4.43	-4.41	3335.9	Comparatively
	7.24	.100	.139	.921	-4.43	-4.41	3333.9	low persistence
Syndicate Bank								value
	.0001	.192	.645	.837	-4.18	-4.16	3148.2	Comparatively
	.0001	.172		1007			51.0.2	low persistence
Bank of India								value
	7.62	.121	.691	.812	-5.01	-5.00	3775.8	Comparatively
								low persistence
Bajaj Holdings								value
	4.44	.150	.748	898	-5.02	-5.00	3780.4	Comparatively
								low persistence
HDFC								value
	6.90	.149	.802	.951	-4.13	-4.11	3113.8	Very high
Max Financial								persistence value
	.0001	.209	.624	.833	-4.63	-4.61	3463.9	Comparatively
C I E:								low persistence
Sundaram Finance								value
	.0001	.180	.682	.862	-4.29	-4.28	3236.0	Comparatively
Reliance Capital								low persistence
Kettance Capitat	0001							value
	.0001	.223	.554	.777	-4.39	-4.34	3291.8	Comparatively
Bajaj Fin.								low persistence
	aard Da		Ta a ff ica	in to C	DCII (1	1)		value
Se	1			ents - GA	ARCH (1	., 1)		
Kotak Mah. Bank	5.39	.063	.927	.99	-4.71	-4.70	5260.0	Very high
Kotak Maii. Daiik								persistence value
Federal Bank	1.04	.054	.927	.981	-4.73	-4.72	5287.4	Very high
rederal Bank								persistence value
	.0001	.131	.685	.816	-4.21	-4.20	4702.5	Comparatively
Oriental Bank								low persistence
Oriental Bank	1.62	0.46	0.40	005	7.00	7.2 0	50065	value
HDFC Bank	1.63	.046	.949	.995	-5.29	-5.28	5906.7	Very high
1121 0 2 4 4 1	0.05	067	010	006	1.62	1.60	£150 0	persistence value
ICICI Bank	9.05	.067	.919	.986	-4.62	-4.60	5158.8	Very high
	1 5 1	000	905	004	4 20	-4.37	4905.4	persistence value
IDBI Bank	1.51	.089	.895	.984	-4.38	-4.37	4895.4	Very high persistence value
	9.47	.080	.903	.983	-4.95	-4.94	5532.1	Very high
Corporation Bank	∂. ₩/	.000	.503	.303	-4.33	-4.74	3334.1	persistence value
	2.66	.112	.856	.968	-4.54	-4.53	5071.2	Very high
Bank of Baroda	۷.00	.112	.050	.500	-4.34	-4.33	50/1.2	persistence value
		<u> </u>			<u> </u>			persistence value

IndusInd Bank	5.72	.073	.920	.993	-4.54	-4.53	5070.4	Very high persistence value
Axis Bank	1.23	.088	.894	.982	-4.58	-4.57	5118.3	Very high persistence value
Syndicate Bank	3.31	.069	.885	.954	-4.45	-4.44	5530.0	Very high
	3.07	.123	.848	.971	-4.30	-4.29	4808.5	very high
Bank of India								persistence value
HDFC	4.47	.056	.934	.99	-4.90	-4.89	5475.6	Very high persistence value
Max Financial	3.35	.109	.851	.96	-4.51	-4.50	5039.0	Very high persistence value
	6.56	.188	.649	.837	-5.12	-5.11	5715.1	Comparatively
Sundaram Finance								low persistence value
Reliance Capital	2.00	.096	.890	.986	-4.11	-4.11	4598.4	Very high persistence value
Bajaj Fin.	1.91	.099	.880	.979	-4.57	-4.56	5106.0	Very high persistence value
Ov	erall Pe	riod - (Coeffici	ients - GA	RCH (1	., 1)		
Banking and	α_0	α_1	β_1	$\alpha_i + \beta_i$				
Finance Sector	0	1	7 -1	<i>,</i> , ,				
Kotak Mah. Bank	7.02	.058	.934	.992	-4.46	-4.47	8344.0	Very high persistence value
Federal Bank	5.23	.119	.810	.929	-4.53	-4.52	8323.3	Very high persistence value
Oriental Bank	2.43	.107	.869	.976	-4.34	-4.33	8122.0	Very high persistence value
HDFC Bank	1.77	.142	.819	.961	-5.19	-5.18	9704.5	Very high persistence value
ICICI Bank	1.77	.082	.891	.973	-4.62	-4.61	8631.4	Very high persistence value
	5.84	.115	.826	.941	-4.23	-4.22	7911.3	Comparatively
IDBI Bank								low persistence value
Corporation Bank	1.37	.099	.884	.983	-4.72	-4.72	8835.2	Very high persistence value
Bank of Baroda	3.32	.124	.840	.964	-4.44	-4.43	8297.0	Very high persistence value
Axis Bank	2.47	.090	.879	.969	-4.44	-4.43	8299.6	Very high persistence value
	4.17	.099	.848	.947	-4.44	-4.43	8855.4	Comparatively
Syndicate Bank								low persistence value
Bank of India	5.14	.136	.815	.951	-4.25	-4.24	7944.6	Very high persistence value

Tapas Kumar Tripathy

W.D. F. C	1.28	.089	.884	.973	-4.94	-4.93	9238.1	Very hig	gh
HDFC								persistence valu	ıe
16 5	4.41	.123	.834	.957	-4.35	-4.34	8139.1	Very hig	gh
Max Financial								persistence valu	ıe
	7.47	.192	.666	.858	-4.91	-4.90	9153.9	Comparatively	
								low persistence	ce
Sundaram Finance								value	
	3.14	.101	.872	.973	-4.18	-4.17	7817.8	Very hig	gh
Reliance Capital								persistence valu	ıe
Bajaj Fin.	6.34	.148	.777	.925	-4.49	-4.48	8372.2		

(Compilation of Stock price returns data using EViews 8.0 where α_0 : constant in the model represents a long-run average; α_1 : The ARCH term which is the lag of the squared residuals from the mean equation, represents news about volatility from the previous period; β_1 : The GARCH term is the last period's forecast variance; $\alpha_j + \beta_i$ is close to one which means any shock to volatility is permanent.)

Table-1.81 (T- GARCH Model)

Company	Esti	mated 1	Model w	vith val	ues	AIC	SIC	Log	Decision			
Name								Likelih	(Decision Rule: If γ			
								ood	is significant &			
									positive, negative			
									shocks have a larger			
									effect on σ_t^2 than			
									the positive shocks)			
First Period - Coefficients - GARCH (1, 1)												
Banking and	$\alpha_{\scriptscriptstyle 0}$	$\alpha_{\scriptscriptstyle 1}$	γ	β_{1}	$\alpha_j + \beta_i$							
Finance												
Sector												
	2.44	.069	.054	.851	.92	-4.96	-4.94	3736.55	Positive γ &			
State Bank of									negative shocks is			
									larger effect on			
India									volatility			
	.001	.158	.135	.637	.795	-4.14	-4.12	3122.82	Positive γ &			
Kotak Mah.									negative shocks is			
									larger effect on			
Bank									volatility			
	.001	.099	.157	.722	.821	-4.25	-4.23	3072.66	Positive γ &			
Federal Bank									negative shocks is			

									larger effect on volatility
	2.23	.163	.002	.819	.982	-4.58	-4.56	3451.79	Positive (lower) γ &
									negative shocks is
									lower effect on
Oriental Bank									volatility
	4.13	.109	.058	.767	.876	-5.02	-5.00	3781.90	Positive γ &
									negative shocks is
HDEC D1-									larger effect on
HDFC Bank									volatility
	.0001	.117	.103	.630	.747	-4.64	-4.62	3498.75	Positive γ &
									negative shocks is
ICICI Bank									larger effect on
TOTOT Built	.0001	.140	.085	.665	905	-4.05	4.02	3052.78	volatility
	.0001	.140	.085	.003	.805	-4.05	-4.03	3032.78	Positive γ &
									negative shocks is larger effect on
IDBI Bank									larger effect on volatility
	6.17	.122	.083	.776	.898	-4.42	-4.40	3330.68	Positive γ &
									negative shocks is
Corporation									larger effect on
Bank									volatility
	4.82	.108	.056	.817	.925	-4.29	-4.27	3231.19	Positive γ &
Bank of									negative shocks is
									larger effect on
Baroda									volatility
	.0001	.130	.027	.687	.817	-4.26	-4.24	3214.75	Positive γ &
									negative shocks is
Axis Bank									larger effect on
7 AXIS Builk	6.65	201	022	740	0.5	4.42	4.41	2226.22	volatility
Syndicate	6.65	.201	032	.749	.95	-4.43	-4.41	3336.23	Negative γ &
Bank									volatility is not affected
	.0001	.133	.106	.635	.768	-4.18	-4.16	3150.86	
	.0001	.133	.100	.033	.708	-4.10	-4.10	3130.60	, α
									negative shocks is larger effect on
Bank of India									volatility
	7.34	.079	.082	.701	.78	-5.01	-4.99	3778.34	Positive γ &
n · ·									negative shocks is
Bajaj									larger effect on
Holdings									volatility
	4.13	.109	.058	.767	.876	-5.02	-5.00	3781.90	Positive γ &
									negative shocks is
up E C									larger effect on
HDFC									volatility

	7.01	.096	.153	.791	.887	-4.15	-4.12	3125.09	Positive γ &
M									negative shocks is
Max									larger effect on
Financial									volatility
Sundaram	.0001	.237	052	.612	.849	-4.62	-4.60	3464.41	Negative γ &
Finance									volatility is not
Tinance	0001	110	150			4.20	4.20	22.12.11	affected
	.0001	.113	.178	.631	.744	-4.30	-4.28	3242.41	Positive γ &
Reliance									negative shocks is
Capital									larger effect on volatility
-	.0001	.231	.012	.550	.781	-4.39	-4.37	3291.83	Positive γ &
	.0001	.231	.012		.,01	1.57	1.57	3271.03	negative shocks is
									larger effect on
Bajaj Fin.									volatility
			Second	Period	- Coef	ficients	- GAR	CH (1, 1)	
Banking and	α_0	α_1	γ	β_1	$\alpha_j + \beta_i$				
Finance		1							
Sector									
	6.79	.038	.063	.920	.958	-4.71	-4.70	5268.73	Positive γ &
Kotak Mah.									negative shocks is
									larger effect on
Bank									volatility
	7.95	.026	.045	.938	.964	-4.74	-4.72	5294.96	Positive $\gamma \&$
									negative shocks is
Federal Bank									larger effect on
Tederar Bank	0001	000	.071	602	702	4.21	4.10	4705.29	volatility
	.0001	.099	.0/1	.693	.792	-4.21	-4.19	4705.38	Positive γ &
									negative shocks is larger effect on
Oriental Bank									volatility
	1.50	.025	.043	.950	.975	-5.29	-5.28	5913.97	Positive γ &
									negative shocks is
									larger effect on
HDFC Bank									volatility
	7.82	.017	.100	.924	.941	-4.64	-4.62	5180.35	Positive γ &
									negative shocks is
ICICI Bank									larger effect on
TCTCT Dalik	1.54	070	022	905	074	4.20	4.27	4906.27	volatility
	1.54	.079	.022	.895	.974	-4.38	-4.37	4896.37	Positive (lower) γ &
									negative shocks is lower effect on
IDBI Bank									lower effect on volatility
		L						<u> </u>	voiatility

	9.26	.083	008	.904	.987	-4.95	-4.94	5532.28	Negative γ &
Corporation									volatility is not
Bank									affected
	2.39	.069	.080	.864	.933	-4.55	-4.53	5079.90	Positive γ &
Bank of									negative shocks is
Baroda									larger effect on volatility
	1.19	.047	.078	.898	.945	-4.59	-4.57	5129.72	Positive γ &
	1.17	.047	.070	.070	.,,,,	7.57	4.57	3129.72	negative shocks is
									larger effect on
Axis Bank									volatility
	3.49	.037	.060	.885	.922	-4.46	-4.45	5538.45	Positive γ &
Syndicate									negative shocks is
Bank									larger effect on
Bank	3.18	.090	.065	.847	.937	-4.31	-4.29	4812.66	volatility Positive γ &
	3.10	.090	.003	.647	.937	-4.31	-4.29	4812.00	, α
									negative shocks is larger effect on
Bank of India									volatility
	4.53	.027	.064	.934	.961	-4.91	-4.89	5486.41	Positive γ &
									negative shocks is
HD E C									larger effect on
HDFC									volatility
	3.37	.098	.024	.850	.948	-4.51	-4.49	5039.67	Positive γ &
Max									negative shocks is
Financial									larger effect on volatility
	6.64	.197	016	.646	.843	-5.12	-5.10	5715.28	Negative $\gamma_{\&}$
Sundaram									volatility is not
Finance									affected
	1.93	.066	.062	.890	.956	-4.12	-4.11	4605.86	Positive γ &
Reliance									negative shocks is
Capital									larger effect on
	1.59	.064	.058	.892	.956	-4.58	-4.56	5112.01	volatility Positive γ &
	1.07	.504	.555	.572	.,,,,	1.50	1.50	0112.01	negative shocks is
									larger effect on
Bajaj Fin.									volatility
			Overall	Period	l - Coef	fficients	- GAR	CH (1, 1)	
Banking and	$lpha_0$	α_1	γ	β_1	$\alpha_j + \beta_i$				
Finance									
Sector									

	6.73	.045	.030	.933	.978	-4.46	-4.45	8350.34	Positive γ &
Kotak Mah.									negative shocks is
									larger effect on
Bank									volatility
	5.34	.083	.080	.807	.89	-4.53	-4.52	8332.85	Positive γ &
									negative shocks is
									larger effect on
Federal Bank									volatility
	2.78	.093	.038	.861	.954	-4.34	-4.33	8125.22	Positive γ &
									negative shocks is
									larger effect on
Oriental Bank									volatility
	1.54	.079	.093	.842	.921	-5.20	-5.19	9717.59	Positive $\gamma_{\&}$
									negative shocks is
									larger effect on
HDFC Bank									volatility
	1.48	.034	.094	.899	.933	-4.63	-4.62	8659.02	Positive γ &
									negative shocks is
									larger effect on
ICICI Bank									volatility
	5.95	.100	.033	.825	.925	-4.23	-4.22	7913.38	Positive γ &
									negative shocks is
10010 1									larger effect on
IDBI Bank									volatility
	1.43	.096	.010	.882	.978	-4.72	-4.71	8835.55	Positive γ &
Corporation									negative shocks is
_									larger effect on
Bank									volatility
	3.31	.092	.071	.838	.93	-4.44	-4.43	8305.64	Positive γ &
Bank of									negative shocks is
									larger effect on
Baroda									volatility
	2.52	.069	.052	.875	.944	-4.44	-4.43	8306.62	Positive γ &
									negative shocks is
Awis D 1									larger effect on
Axis Bank									volatility
	4.53	.085	.031	.843	.928	-4.44	-4.43	8857.57	Positive γ &
Syndicate									negative shocks is
									larger effect on
Bank									volatility
	5.65	.112	.053	.808	.92	-4.25	-4.24	7948.27	Positive γ &
									negative shocks is
Don't - CY 1									larger effect on
Bank of India									volatility

	1.20	.049	.070	.893	.942	-4.95	4.94	9251.39	Positive γ &
									negative shocks is
HDFC									larger effect on
пргс									volatility
	4.64	.096	.075	.826	.922	-4.36	-4.35	8147.66	Positive γ &
Max									negative shocks is
Financial									larger effect on
Tinanciai									volatility
Sundaram	7.59	.201	017	.663	.864	-4.91	-4.90	9154.07	Negative γ &
									volatility is not
Finance									affected
	3.35	.075	.054	.869	.944	-4.18	-4.17	7825.15	Positive γ &
Reliance									negative shocks is
									larger effect on
Capital									volatility
	5.67	.120	.042	.794	.914	-4.49	-4.48	8374.38	Positive γ &
									negative shocks is
D E.									larger effect on
Bajaj Fin.									volatility