"Derivative trading and structural breaks in volatility in India: an ICSS approach"

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DERIVATIVE TRADING AND STRUCTURAL BREAKS IN VOLATILITY IN INDIA: AN ICSS APPROACH

Abstract

Researchers argue that ignoring the structural breaks in the time-series variance can cause significant upward biases in the degree of persistence in estimated GARCH models. Against this backdrop, the present study empirically examines the effect of stock futures on the underlying stock's volatility in India by incorporating the structural breaks with the help of ICSS test and AR (1)-GARCH (1, 1) model for 30 most liquid and actively traded underlying stocks and their associated futures contracts. The study period ranges from the 1st January 2000 or the listing date of the particular stock (whichever is prior) till 31st March 2019. The study contributes to the on-going debate regarding the effect of derivatives on the underlying stock market's volatility in two ways. Firstly, by taking into consideration the breaks in the volatility and, secondly, studying the effect of single stock futures will allow us to evaluate company-specific response to futures trading directly. The study offers a mixed outcome for the stocks under consideration. However, there is evidence of a decline in unconditional volatility for the majority of the stocks. The overall findings indicate that trading in stock futures may not have any detrimental effect on the underlying stock's volatility.

Keywords

stock futures, volatility, AR (1)-GARCH (1, 1), underlying, cash market, futures market

JEL Classification G11, G14

INTRODUCTION

Volatility modeling of the financial asset is one of the critical aspects of economic research as it guides the investors on the risk associated with the investment. In India, derivatives trading started in the year 2000 with the launch of futures contracts on the Nifty Index of NSE and Sensex Index of Bombay Stock Exchange (BSE). Options trading began in the Indian markets in June 2001. Ever since then F&O segment is surging in terms of the number of contracts traded, volume, and offering of new products. The F&O segment of the NSE surpassed the equity market from 2019 to 2020, with an average daily turnover of ₹ 2,37,590,973 Cr, as compared to ₹ 6,81,983 Cr in the cash segment (derivatives updates on NSE website, www.nseindia.com, 2019). NSE outperformed the US-based CME group to claim its No.1 ranking in terms of derivatives trading with more than 6 billion contracts traded volume in 2019 (Das & Sahgal, 2020). Derivatives were introduced to offer a hedging mechanism and enhance the liquidity, thereby increasing the market's overall efficiency. The effect of the listing of derivatives on the underlying market's volatility and, thus, its job in increasing or decreasing the underlying markets' volatility has remained an intense subject of empirical and analytical interest.

Questions about the effect of derivatives trading on underlying market volatility have been empirically addressed in two ways. Firstly, by analyzing variation in volatility over the pre- and post-derivatives trading phases and, secondly, by measuring the effect of derivatives trading on the behavior of the underlying markets by comparing the performance with proxies. Moreover, most studies examining the effect of derivatives on the underlying market volatility used some type of GARCH model with dummy variable regressors¹. However, this approach is based on the underlying presumption that any changes detected during the post-derivatives phase are caused by derivatives trading alone.

An increase in volatility could be the outcome of various other events, such as the initiation of a rolling settlement system, circuit breakers, and changes in regulations, and so on. If the structural breaks in variances of the examined time-series are ignored, the degree of persistence of the GARCH model estimate may be significantly biased. Several studies, such as Diebold (1986), Granger and Hyung (1999), Mikosch and Starica (2000), Diebold and Inoue (2001), have stated that neglecting the structural breaks can lead to spurious GARCH model estimation. The primary reasons for such structural breaks could be the changes in the mechanism of exchange rate systems, global financial markets crisis, or the evolution of the stock markets. The shocks produced by these significant economic or political events may cause a deviation in the financial time-series (Andreou & Ghysels, 2002; Wang & Moore, 2009).

1. LITERATURE REVIEW

The derivatives market and its effect on the underlying market volatility are debated again and again with supporting and countering theories.

1.1. Increased volatility due to futures trading

Wats (2017) examined the effect of the derivatives contracts' expiration on the underlying market's volatility using the GARCH family models. He concluded that spot market volatility has increased during the expiry days and week after the listing of the derivatives. Other studies that find a significant increase in the Index return volatility following the listing of futures include Harris (1989), Brorsen (1991), Lee and Ohk (1992), Antoniou and Holmes (1995), Yao (2016).

1.2. Decreased volatility due to futures trading

Others argue that futures' listing potentially reduces the spot market's volatility, thus stabilizing the market. One of the clarifications for the destabilizing theory is that trades in the derivatives market destabilize the underlying market by providing an alternative route for the transmission and reflection of data in the cash market (Cox & Ross, 1976; Ross, 1989). Gulen and Mayhew (2000) studied the effect of index futures on the volatility of the international equity markets by taking the sample of 21 European nations by applying the BEKK model and GJR-GARCH. They found that the volatility of the underlying market has declined for most of the countries under study.

Similarly, Yilgor, Lidvine, and Mebounou (2016), Chiraz (2016), Bhaumik, Karanasos, and Kartsaklas (2016) used different GARCH family models such as Markow-Switching GARCH, ARFI GARCH, EGARCH, and TGARCH. They found the evidence indicating the decline in the underlying market volatility after listing the derivatives trading. Several studies like Pilar and Rafael (2002), Bandivadekar and Ghosh (2003), Thenmozhi (2002), Raju and Karande (2003), Sarangi and Patnaik (2007) have reported a significant decline in the underlying market volatility in India.

1.3. Mixed evidence/no impact of futures trading

Using the GARCH (1, 1) model, Rahman (2001) investigated the effect of trading in index futures on the volatility of Dow Jones Industrial Average (DJIA) component stocks and observed no variation in conditional volatility. Mallikajunappa and Afzal (2008), Thenmozhi (2002), Kavussanos, Visvikis, and Alexakis (2008) argued that improvement in the volatility cycle was not due to the listing

¹ See Chan (1991), Reyes (1996), Pericli and Koutmos (1997), Mckenzie, Brailsford, and Faff (2001), Tse (1999), Rahman (2001), Gulen and Mayhew (2000), Bandivadekar and Ghosh (2003), Pok and Poshakwale (2006), and Ryoo and Smith (2006).

of derivatives, but due to many other factors, such as the improved distribution of information and greater transparency.

Due to the increase in the speed of information flow, the stock prices have become more sensitive to the recent innovation in the post-derivative period. Bohl, Salm, and Wilfling (2009), C. Lee, Stevenson, and M. Lee (2014) employed Markow-Switching GARCH, which endogenously identified the distinct volatility regimes, to analyze the effect of derivatives on the volatility and found no influence on the spot market. Mallikajunappa and Afzal (2008), Sarangi and Patnaik (2007) applied the GARCH model with dummy variables and did not find any significant effect of derivatives on the underlying market volatility.

Moreover, the literature is inconclusive about whether the listing of derivatives leads to an increase or decrease in the underlying market's volatility. The vast majority of the studies, which are found in the arena of derivative segments, are concentrated on the effect of index futures on the underlying market. A limited number of studies have been undertaken in the area of single stock futures. Indian studies based on stock futures focus on conceptual clarity or cover only a short period. Research focusing on the index analysis does not consider the stock-specific characteristics, which could also play a significant role in the formation of the volatility. The present study empirically examines the effect of stock futures on India's underlying stock's volatility by incorporating structural breaks. The study contributes to the on-going debate regarding the effect of derivatives on the underlying stock market's volatility in two ways. Firstly, by reinvestigating the issue by applying a distinct analytical technique, which is based on the methodology used by Aggarwal, Inclan, and Leal (1999), Andreou and Ghysels (2002), Malik and Hassan (2004), Kang, Jung, Park, and Yoon (2007). The study attempts to model the underlying stock's volatility with stock futures by considering the breaks in the volatility. It aims at identifying the structural breaks, if any, in the stock prices by applying the ICSS test of Inclan and Tiao (1994). Secondly, studying the effect of single stock futures will allow us to directly evaluate company-specific responses to futures trading, in contrast to the market-wide effect gained from research with index futures.

2. METHOD

The Individual Stock Futures (ISF) has proved to be a hugely successful financial instrument on Indian bourses, and NSE has continued to account for the majority of total volumes traded in the ISF segment all over the world. The resulting sample for this study comprises 30 most liquid and actively traded underlying stocks on which futures contracts are available. These 30 stock futures contribute to around

		Volume			Volume				
Stock ASHOKLEY		Contracts		Stock	Contracts				
	Futures	Options	Total		Futures	Options	Total		
ASHOKLEY	928	1,673	2,601	INFRATEL	3,358	1,383	4,741	Î	
AUROPHARMA	1,674	2,003	3,677	INFY	1,759	2,441	4,200		
AXISBANK	3,645	3,697	7,342	ITC	1,050	1,287	2,337		
BHARTIARTL	11,550	13,551	25,101	JUSTDIAL	1,371	1,179	2,550		
CIPLA	1,236	1,025	2,261	KOTAKBANK	1,261	803	2,064		
GLENMARK	4,891	5,215	10,106	LT	1,309	1,322	2,631		
GRASIM	2,895	1,610	4,505	MARUTI	1,729	3,598	5,327		
HDFC	1,469	1,418	2,887	RELIANCE	7,010	15,910	22,920		
HDFCBANK	2,234	1,423	3,657	SBIN	3,579	4,940	8,519		
HEROMOTOCO	1,865	2,769	4,634	TATAELXSI	1,447	610	2,057		
IBULHSGFIN	3,827	1,527	5,354	TATASTEEL	3,581	4,489	8,070		
ICICIBANK	3,781	3,536	7,317	TCS	1,451	2,877	4,328		
IDEA	4,913	7,224	12,137	VEDL	1,796	1,860	3,656		
INDIGO	1,369	853	2,222	YESBANK	10,764	6,420	17,184		
INDUSINDBK	1,759	1,049	2,808	ZEEL	2,053	1,602	3,655		

Table 1. List of selected stocks and their volume

Source: https://www1.nseindia.com

70–80% of the total trading volume of the F&O segment of NSE, excluding the index futures. The majority of them are also part of the S&P Nifty Index, the Benchmark Index of NSE. The data extracted for 30 stocks have been procured from the Bloomberg database. The study period will range from the 1st January 2000 or the listing date of the particular stock (whichever is prior) till 31st March 2019.

2.1. Testing for ARCH effect

The ARCH test involves testing the existence of heteroscedasticity in the time-series data. Lagrange multiplier (LM) test by Engle helps in checking for ARCH effect. Let $\varepsilon_t = y_t - u_t$ be the residual series. The squared series ε_t^2 is utilized to implement the LM test for checking conditional heteroscedasticity. The null hypothesis is stated as follows:

$$H_{0}: \quad \alpha_{i} = 0, \ i = 1, 2, \dots, q$$

versus

$$H_i: \quad \alpha_i \neq 0$$
, for at least one *i*.

In the linear regression

$$\varepsilon_t^2 = \omega + \alpha_1 \varepsilon_{t-1}^2 + \ldots + \alpha_q \varepsilon_{t-q}^2, \ t = q+1, \ldots, N,$$

where q is the length of ARCH lags, and N is the number of observations used in the regression equation.

The test statistic for LM test is defined by:

$$LM = NR^2$$
,

where R^2 is the *R*-squared from the regression of ε_t^2 in the equation and defined by:

$$R^{2} = \frac{Regression \ sum \ of \ squares}{Total \ sum \ of \ squares}.$$

2.2. Testing for multiple structural breaks (Iterated Cumulative Sums of Squares (ICSS) algorithm of Inclan and Tiao (1994))

Iterative Cumulative Sums of Squares (ICSS) algorithm proposed by Inclan and Tiao (1994) allows for detecting multiple breakpoints in the variance in a time series. The idea behind the ICSS algorithm of Inclan and Tiao can be outlined in sequential steps. The unconditional variance of financial time-series is stationary until a sudden break is observed. After that, until the occurrence of the next structural break, the unconditional variance is stationary. This process repeats through time, generating multiple numbers of structural breaks in the unconditional variance in n observations:

$$\sigma_t^2 = \begin{cases} \tau_0^2 \ 1 < t < i_1 \\ \tau_1^2 \ i_1 < t < i_1 \\ \tau_M^2 \ i_M < t < i_1 \end{cases}$$

To estimate the number of changes and the point in time of variance shifts, a cumulative sum of squared residuals is used, $C_k = \sum \varepsilon_t^2$, k = 1, 2, ..., n, where $\{\varepsilon_t\}$ is a series of 'uncorrelated random variables with zero mean and unconditional variance σ_t^2 . Inclan and Tiao define the statistic:

$$D_k = \frac{C_k}{C_n} - \frac{tk}{n}, \ k = 1, 2, \dots, n, \ D_0 = D_n = 0.$$

If no sudden changes occur during the entire sampling duration in the variance of the sequence, D_k oscillates about zero. If there are one or more sudden shifts in variance, then the D_k statistics will drift either above or below the zero. The ICSS algorithm helps in identifying breaks in variance of the time-series at different points in time.

2.3. Linking the structural breaks in volatility with trading in stock futures

First, the dates for the structural breaks in the stocks will be estimated. Later, these structural breaks were matched with the dates of the listing of stock futures on the individual stocks. If a structural break is found within six months of the listing of stock futures, it has been attributed as likely to derivative trading.

AR (1)-GARCH (1, 1) is a GARCH family model, in which the mean is determined by a first-order auto-regressive AR (1), with a GARCH (1, 1) error:

$$x_t = u_t + \sigma_t \varepsilon_t$$
, $\mathbf{E}[\varepsilon_t] = 0$, $\mathbf{E}[\varepsilon_t^2] = 1$, ε_t i.i.d.,

$$\mu_{t} = \lambda X_{t-1},$$

$$\sigma_{t}^{2} = a_{0} + a \left(X_{t-1} - \mu_{t-1} \right)^{2} + b \sigma_{t-1}^{2}.$$

Once all the structural breakpoints are identified, dummy variables are created for each detected break. Each dummy variable is denoted with value one onwards from the identified location until the end of the data series and 0 elsewhere.

3. RESULTS

Table 2 displays the result of the ADF unit root test. All the variables are non-stationary at the lev-

el as the *p*-value is more than 0.05%. Therefore, the Unit Root Test is conducted in the first difference for all the variables. All the series are stationary at the first difference at 1% level of significance. The results of the ADF test indicate that all variables are integrated of the same order.

Table 3 depicts the ARCH test results for all 30 stocks traded at the cash segment of NSE. The standard diagnostic test of the Residuals from the model confirms the presence of ARCH effect. There is a presence of the ARCH effect in the closing return series of all the variables.

After detecting the structural breaks in the return series of selected highly traded 30 stocks, an attempt has been made to associate these structural

	9	Spot	Fu	itures		9	pot	Futures	
Stock	ADF at level	ADF at first difference	ADF at level	ADF at first difference	Stock	ADF at level	ADF at first difference	ADF at level	ADF at first difference
ASHOKLEY	-2.669	-77.9823	-1.8401	-25.085	INFRATEI	-1.903	-252.625	-1.840	-251.084
	(-0.079)	(-0.00)	(-0.361)	(-0.00)		(-0.330)	(-0.000)	(-0.361)	(-0.000)
	-3.075	-14.3857	-3.067	-14.026	INEV	-2.8434	-264.133	-2.696	-264.045
AUNOFHANIMA	(-0.112)	(-0.000)	(-0.114)	(-0.000)		(–0.052)	(-0.000)	(-0.074)	(-0.000)
AVISDANIK	-2.407	-216.989	-2.473	-218.720	ITC	-1.793	-435.009	-1.887	-297.513
ANISDAINK	(–0.139)	(-0.000)	(-0.121)	(-0.000)	пс	(–0.389)	(-0.000)	(–0.333)	(-0.000)
	-2.496	(–240.736)	-420.76	-420.769		-1.436	-169.532	-1.450	-218.429
DHANHANTE	(0.116)	(-0.000)	(-0.000)	(0.000)	JUSTDIAL	(–0.565)	(-0.000)	(–0.558)	(-0.000)
	-1.471	-305.751	-1.505	-189.269		-2.743	-254.133	-2.596	-254.045
CIPLA	(-0.548)	(-0.000)	(-0.531)	(-0.000)	KUTAKBAINK	(-0.072)	(-0.000)	(-0.064)	(-0.000)
	-1.476	-296.195	-1.189	-186.673	1.7	-2.496	-240.736	-420.769	-420.769
GLEINIVIARK	(-0.546)	(-0.000)	(-0.681)	(-0.000)	LI	(0.116)	(-0.000)	(-0.000)	(0.000)
CDACINA	-1.903	-252.625	-1.840	-251.084	MADUTI	-1.683	-435.009	-1.797	-298.513
GRASIIVI	(-0.330)	(-0.000)	(-0.361)	(-0.000)	IVIARUTI	(-0.389)	(-0.000)	(-0.333)	(-0.000)
	-2.843	-264.133	-2.696	-264.045	DELIANCE	-1.803	-242.625	-1.740	-241.084
HDFC	(-0.052)	(-0.000)	(-0.074)	(-0.000)	RELIANCE	(-0.320)	(-0.000)	(-0.351)	(-0.000)
	-2.283	-174.557	-2.256	-169.916	CDIN	-2.496	-240.736	-420.769	-420.769
HDECRANK	(-0.177)	(-0.000)	(-0.186)	(-0.000)	SRIN	(0.116)	(-0.000)	(-0.000)	(0.000)
	-1.219	-116.178	-1.098	-128.5		-1.471	-305.751	-1.505	-189.269
HEROMOTOCO	(-0.668)	(-0.000)	(-0.718)	-0.0001	TATAELXSI	(-0.548)	(-0.000)	(-0.531)	(-0.000)
	-0.992	-161.898	-0.692	-162.498	TATACTEEL	-1.476	-296.195	-1.189	-186.673
IBOTH2GEIN	(-0.758)	(-0.000)	(-0.846)	(-0.000)	TATASTEEL	(-0.546)	(-0.000)	(-0.681)	(-0.000)
	-1.783	-425.009	-1.897	-298.513	TCC	-1.903	-252.625	-1.840	-251.084
ICICIBANK	(-0.389)	(-0.000)	(-0.333)	(-0.000)	ICS	(-0.330)	(-0.000)	(-0.361)	(-0.000)
	-2.843	-264.133	-2.696	-264.045	VEDI	-2.283	-174.557	-2.256	-169.916
IDEA	(-0.052)	(-0.000)	(-0.074)	(-0.000)	VEDL	(-0.177)	(-0.000)	(-0.186)	(-0.000)
INDICO	0.895	-195.973	0.538	-277.977		-1.211	-116.178	-1.098	-128.5
INDIGO	(-0.995)	(-0.000)	(-0.988)	(-0.000)	YESBANK	(-0.668)	(-0.000)	(-0.718)	(-0.000)
	-1.360	-216.679	-1.370	-216.604	7551	-0.992	-161.898	-0.692	-162.498
INDOZINDRK	(-0.603)	(-0.000)	(–0.598)	(-0.000)	ZEEL	(–0.758)	(-0.000)	(-0.846)	(-0.000)

Table 2. Unit root test (augmented Dickey-Fuller test)

Note: () denote p-value.

Variables	<i>p</i> -value	Inference	Variables	<i>p</i> -value	Inference
ASHOKLEY	0.000	Present	INFRATEL	0.000	Present
AUROPHARMA	0.000	Present	INFY	0.000	Present
AXISBANK	0.000	Present	ITC	0.000	Present
BHARTIARTL	0.000	Present	JUSTDIAL	0.000	Present
CIPLA	0.000	Present	KOTAKBANK	0.000	Present
GLENMARK	0.000	Present	LT	0.000	Present
GRASIM	0.000	Present	MARUTI	0.000	Present
HDFC	0.000	Present	RELIANCE	0.000	Present
HDFCBANK	0.000	Present	SBIN	0.000	Present
HEROMOTOCO	0.000	Present	TATAELXSI	0.000	Present
IBULHSGFIN	0.000	Present	TATASTEEL	0.000	Present
ICICIBANK	0.000	Present	TCS	0.000	Present
IDEA	0.000	Present	VEDL	0.000	Present
INDIGO	0.000	Present	YESBANK	0.000	Present
INDUSINDBK	0.000	Present	ZEEL	0.000	Present

Table 3. Results of ARCH test

breaks with the listing dates of stock futures on individual stocks. The stocks, which have displayed similar patterns in terms of changes in persistence in volatility, unconditional volatility, and rate of adjustment to new information (measured by α), have been grouped. The detailed analysis after incorporating detected structural breaks into the AR (1)-GARCH (1, 1) Model is presented in Appendix. Stocks were divided into seven categories, viz. Panel A, Panel B, Panel C, Panel D, Panel E, Panel F, and Panel G. This classification is based on the influence of stock futures on the underlying volatility. If a structural break is observed six months after listing the stock futures, it is associated with trading in futures. After this structural break date, the change in persistence of volatility, unconditional volatility, and rate of adjustment to new information (denoted by α) is observed and reported in Table 4. In the case of AUROPHARMA, ICICIBANK, and JUSTDIAL, the total persistence increases, while α and unconditional volatility declined for the period after this break (Panel A). On the contrary, ASHOKLEY, AXISBANK, HDFCBANK, INDUSINDBK, INFRATEL, RELIANCE, and TCS have shown a downfall in

Table 4. Impact of stock futures on the volatility of underlying stocks

	Im	Impact on the volatility					
Stock	16 - 4	Direction of impact					
Stock	stock futures trading	Persistence	α	Unconditional Volatility			
	Panel	4					
AUROPHARMA	Yes	\downarrow	\uparrow	\uparrow			
ICICIBANK	Yes	\downarrow	\uparrow	\uparrow			
JUSTDIAL	Yes	\downarrow	\uparrow	\uparrow			
	Panel	В					
ASHOKLEY	Yes	\downarrow	\downarrow	\downarrow			
AXISBANK	Yes	\downarrow	\downarrow	\downarrow			
HDFCBANK	Yes	\downarrow	\downarrow	\downarrow			
INDUSINDBK	Yes	\downarrow	\downarrow	\downarrow			
INFRATEL	Yes	\downarrow	\downarrow	\downarrow			
RELIANCE	Yes	\downarrow	\downarrow	\downarrow			
TCS	Yes	\downarrow	\downarrow	\downarrow			
	Panel	C					
BHARTIARTL	Yes	\uparrow	\uparrow	\downarrow			
INFY	Yes	\uparrow	\uparrow	\downarrow			
MARUTI	Yes	\uparrow	\uparrow	\downarrow			

	Im	pact on the volati	lity			
Stock		Direction of impact				
Stock	If structural break associated with stock futures trading	Persistence	α	Unconditional Volatility		
	Panel	D		<u>.</u>		
CIPLA	Yes	\downarrow	\uparrow	\downarrow		
HDFC	Yes	\downarrow	\uparrow	\downarrow		
INDIGO	Yes	\downarrow	\uparrow	\downarrow		
LT	Yes	\downarrow	\uparrow	\downarrow		
	Panel	E				
GRASIM	Yes	\uparrow	\downarrow	\downarrow		
HEROMOTOCO	Yes	\uparrow	\downarrow	\downarrow		
ITC	Yes	\uparrow	\downarrow	\downarrow		
SBIN	Yes	\uparrow	\downarrow	\downarrow		
TATASTEEL	Yes	\uparrow	\downarrow	\checkmark		
ZEEL	Yes	\uparrow	\downarrow	\checkmark		
	Panel	F				
IDEA	Yes	\uparrow	\uparrow	\uparrow		
	Panel	G				
GLENMARK	No	-	-	_		
IBULHSGFIN	No	-	-	-		
KOTAKBANK	No	-	-	-		
TATAELXSI	No	-	-	-		
VEDL	No	-	-	-		
YESBANK	No	-	-	-		
Total - 20	Yes = 24	Increased = 10	Increased = 11	Increased = 4		
10tai = 30	No = 6	Decreased = 14	Decreased = 13	Decreased = 20		

Table 4 (cont.). Impact of stock futures on the volatility of underlying stocks

the total persistence, unconditional volatility, and α (Panel B).

Panel C comprises of BHARTIARTL, INFY, and MARUTI. There is a decline in the unconditional volatility, but its persistence, as well as α , has increased after the occurrence of the structural break. Panel D consists of CIPLA, HDFC, INDIGO, and LT for which the total persistence and unconditional volatility have declined, but α has increased during the observed structural break in volatility. There is observed an increase in α , and reduction in the total persistence and the unconditional volatility of GRASIM, HEROMOTOCO, ITC, SBIN, TATASTEEL, and ZEEL for the period after the listing of stock futures (Panel E). However, no structural break is observed within six months after the listing of stock futures for GLENMARK, IBULHSGFIN, KOTAKBANK, TATAELXSI, and VEDL, and YESBANK (Panel G).

No structural break was observed within six months after listing the stock futures for six out

of thirteen stocks. Unconditional volatility has declined for twenty out of twenty-four stocks for which structural breaks were observed within six months after the listing of stock futures. It is noted that the unconditional volatility has declined for the majority of the stocks after the listing of futures contracts. Total persistence has risen for ten stocks while declined for fourteen stocks. On the other hand, α has increased for eleven stocks, while it has decreased for thirteen stocks.

4. DISCUSSION

Through this study, an attempt has been made to model the underlying stock's volatility with stock futures by taking into consideration the breaks in the volatility. Several studies, such as Diebold (1986), Granger and Hyung (1999), Mikosch and Starica (2000), Diebold and Inoue (2001), have stated that neglecting the structural breaks can lead to spurious GARCH model estimation. Therefore, Iterated Cumulative Sums of Squares (ICSS) algorithm of Inclan and Tiao (1994) was applied for detecting the multiple structural breaks for 30 highly traded and liquid stocks.

If a break is observed within six months after the listing of stock futures, then unconditional volatility, the nature of changes in total persistence, and α have been examined. Reduction in the unconditional volatility was observed for twenty

out of thirty stocks after incorporating detected structural breaks into the AR(1)-GARCH(1,1) model. It is noted that the unconditional volatility has declined for the majority of the stocks after the listing of futures contracts. Total persistence has risen for ten stocks while declined for fourteen stocks. On the other hand, α has increased for eleven stocks, while it has decreased for thirteen stocks.

CONCLUSION

Through this analysis, any consistent patterns were not found in terms of changes in total persistence, unconditional volatility, and a for the underlying stocks for the period after the relevant breaks. The mixed outcome could be due to stock-specific characteristics, which could also play a significant role in the formation of the volatility. Consequently, the listing of stock futures may not have any clear effect on the underlying stock's volatility. The findings of the study reveal that the unconditional volatility has declined for the majority of the stocks after the listing of futures contracts. The analysis findings suggest that trading in stock futures may not have any detrimental effect on the underlying stock's volatility. These findings are in line with conclusions drawn by Badhani, Harish, and Chauhan (2008), Malik and Shah (2016).

AUTHOR CONTRIBUTIONS

Conceptualization: Sanjeeta Shirodkar. Data curation: Sanjeeta Shirodkar. Formal analysis: Sanjeeta Shirodkar. Investigation: Guntur Anjana Raju, Sanjeeta Shirodkar. Methodology: Guntur Anjana Raju, Sanjeeta Shirodkar Project administration: Sanjeeta Shirodkar. Resources: Guntur Anjana Raju. Software: Guntur Anjana Raju. Software: Guntur Anjana Raju. Validation: Guntur Anjana Raju. Visualization: Sanjeeta Shirodkar. Writing – original draft: Sanjeeta Shirodkar. Writing – review & editing: Guntur Anjana Raju.

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Period	ω	α	в	Total Persistence: (α+β)	Unconditional Volatility: $\omega/(1-lpha-eta)$							
Volatility Breaks in ASHOKLEY												
Date of commencement of Derivative trading: 20-04-2005												
05/07/2002_30/12/2004	1.311	0.194	0.669	0.864	9.611							
31/12/2004_17/11/2005	0.717	0.165	0.764	0.928	9.994							
18/11/2005_29/09/2008	1.830	0.157	0.630	0.786	8.553							
30/09/2008_02/11/2009	-0.021	0.109	0.898	1.007	3.062							
03/11/2009_10/02/2010	0.513	0.038	0.868	0.906	5.461							
11/02/2010_20/05/2011	0.732	0.155	0.718	0.872	5.719							
21/05/2011_16/02/2016	0.684	0.060	0.762	0.823	3.856							
17/02/2016_02/06/2017	0.954	0.366	0.397	0.763	4.021							
03/06/2017_29/03/2019	1.270	0.067	0.449	0.516	2.623							
		Volatilit	y Breaks in	n AUROPHARMA								
	Date of co	mmencer	nent of De	rivative trading: 12-05-200	5							
05/01/2000_30/04/2001	12.687	0.411	-0.084	0.327	18.856							
10/05/2001_17/09/2003	0.064	0.087	0.910	0.997	22.097							
18/09/2003_11/05/2004	4.754	0.043	0.425	0.468	8.938							
12/05/2004_05/08/2005	0.071	-0.076	1.063	0.987	5.376							
06/08/2005_09/01/2006	3.321	0.076	0.353	0.429	5.818							
10/01/2006_15/01/2008	0.492	0.113	0.761	0.874	3.899							
16/01/2008_04/08/2009	0.962	0.057	0.882	0.940	15.928							
05/08/2009_09/02/2010	2.761	-0.091	0.581	0.490	5.409							
10/02/2010_09/02/2011	1.707	0.043	0.341	0.384	2.770							

Period	ω	α	в	Total Persistence: (α+β)	Unconditional Volatility: ω/(1-α-β)
10/02/2011_07/03/2012	1.148	0.152	0.741	0.893	10.707
08/03/2012_28/08/2015	1.129	0.042	0.790	0.831	6.690
29/08/2015_29/03/2019	0.550	0.076	0.784	0.861	3.951
		Volat	ility Breaks	s in AXISBANK	`
	Date of co	mmencer	nent of Dei	rivative trading: 20-04-200	5
05/01/2000_16/11/2001	3.256	0.310	0.540	0.850	21.713
17/11/2001_01/01/2003	0.142	0.266	0.784	1.051	-2.803
02/01/2003 06/06/2005	0.172	0.096	0.888	0.984	10.853
07/06/2005 18/01/2006	3.323	0.085	0.411	0.497	6.601
19/01/2005 18/01/2008	2.728	0.259	0.453	0.712	9.478
19/01/2008 18/08/2009	2.281	0.079	0.815	0.894	21.560
19/08/2009 07/06/2012	1.175	0.146	0.558	0.704	3.962
08/06/2012 20/11/2014	0.056	0.039	0.940	0.979	2.657
21/11/2014 24/09/2015	0.840	0.032	0.703	0.735	3.169
25/09/2015 31/01/2017	1.287	-0.019	0.264	0.245	1.705
01/02/2017 29/03/2019	1.037	0.276	0.123	0.400	1.726
		Volatil	ity Breaks i	in BHARTIARTL	
	Date of co	mmencen	nent of Der	rivative trading : 20-04-200)5
05/01/2000_22/11/2001	3.995	0.197	0.462	0.658	11.698
23/11/2001_20/01/2003	1.738	0.367	0.372	0.739	6.649
21/01/2003_14/06/2004	1.967	0.023	0.808	0.831	11.613
15/06/2004_22/07/2005	2.967	0.059	0.465	0.524	6.234
23/07/2005_19/09/2007	0.620	0.134	0.710	0.843	3.959
20/09/2007_10/10/2008	2.520	0.004	0.678	0.682	7.925
11/10/2008_27/05/2009	1.247	0.444	0.575	1.019	-64.730
28/05/2009_12/07/2010	3.565	0.257	0.166	0.423	6.175
13/07/2010_10/01/2013	2.566	0.110	0.274	0.384	4.166
11/01/2013_29/03/2019	2.790	0.058	0.089	0.147	3.272
		Vo	atility Brea	aks in CIPLA	
	Date of co	mmencer	ment of De	rivative trading: 02-07-200	1
05/01/2000_19/07/2000	13.355	0.229	0.021	0.249	17.791
20/07/2000_23/10/2001	1.187	0.049	0.772	0.820	6.605
24/10/2001_28/04/2003	0.779	0.046	0.466	0.513	1.599
29/04/2003_06/07/2004	1.756	0.187	0.476	0.663	5.214
07/07/2004_02/02/2006	1.546	0.100	0.384	0.484	2.994
03/02/2006_18/08/2009	0.745	0.135	0.729	0.864	5.466
19/08/2009_15/08/2014	0.946	0.014	0.549	0.562	2.162
16/08/2014_03/09/2015	0.217	0.011	0.930	0.941	3.664
04/09/2015_28/12/2016	1.343	0.252	0.138	0.390	2.201
29/12/2016_23/05/2017	0.210	0.197	0.547	0.744	0.818
24/05/2017_29/03/2019	0.530	0.144	0.646	0.790	2.527
	Data of co	Volati	lity Breaks	IN GLENIMARK	2
25/02/2000 11/12/2001	0 396	0.058		0 965	11 238
12/12/2001 05/07/2002	0.390	-0.120	1.060	0.940	8,007
06/07/2002 30/12/2004	1 311	0 194	0.669	0.864	9 611
31/12/2004 17/11/2006	0 717	0.165	0.764	0.928	9,994
18/11/2006 29/09/2008	1.830	0.157	0,630	0.786	8.553
30/09/2008 02/11/2009	-0.021	0.109	0.898	1.007	3.062
03/11/2009 10/02/2010	0.513	0.038	0.868	0.906	5.461
11/02/2010 20/05/2011	0.732	0.155	0.718	0.872	5.719
21/05/2011 16/02/2016	0.684	0.060	0.762	0.823	3.856
17/02/2016 02/06/2017	0.954	0.366	0.397	0.763	4.021
03/06/2017_29/03/2019	1.270	0.067	0.449	0.516	2.623
		· · · · · · · · · · · · · · · · · · ·	· · · · ·		4

Period	ω	α	в	Total Persistence: (α+β)	Unconditional Volatility: $\omega/(1-\alpha-\beta)$
		Vola	tility Break	s in GRASIM	
	Date of co	mmencer	nent of De	rivative trading: 02-07-200	1
05/01/2000_04/10/2001	5.439	0.159	0.458	0.617	14.200
05/10/2001_17/06/2003	0.006	-0.021	1.017	0.996	1.761
18/06/2003_16/07/2004	0.793	0.060	0.815	0.875	6.353
17/07/2004_12/09/2005	0.480	0.033	0.692	0.725	1.749
13/09/2005_13/03/2007	0.331	0.224	0.736	0.960	8.348
14/03/2007_21/01/2008	0.720	0.038	0.748	0.786	3.368
22/01/2008_06/10/2009	1.128	0.096	0.776	0.872	8.822
07/10/2009_03/07/2012	1.592	0.241	0.100	0.341	2.415
04/07/2012_25/07/2013	1.019	0.167	0.085	0.252	1.362
26/07/2013_10/03/2015	0.166	0.087	0.861	0.947	3.148
11/03/2015_05/08/2016	0.354	0.104	0.623	0.728	1.299
06/08/2016_29/03/2019	0.513	0.022	0.807	0.829	3.004
		Volati	lity Breaks	in HDFCBANK	
	Date of co	mmencer	nent of Dei	rivative trading: 29-08-200	3
05/01/2000_05/01/2001	1.467	0.188	0.651	0.839	9.129
06/01/2001_09/10/2003	0.336	0.187	0.744	0.931	4.841
10/10/2003_11/05/2004	0.968	-0.108	0.862	0.754	3.933
12/05/2004_18/05/2006	0.416	0.081	0.799	0.881	3.488
19/05/2006_27/06/2008	0.160	0.056	0.921	0.976	6.773
28/06/2008_22/12/2011	0.050	0.055	0.934	0.990	4.850
23/12/2011_06/08/2013	0.904	0.023	0.553	0.576	2.133
07/08/2013_06/10/2015	0.178	0.054	0.890	0.944	3.172
07/10/2015_29/03/2019	0.216	0.052	0.833	0.885	1.872
		Vo	latility Brea	iks in HDFC	
	Date of co	mmencer	ment of De	rivative trading: 02-07-200	1
05/01/2000_05/02/2001	0.602	0.241	0.724	0.965	17.222
06/02/2001_16/10/2001	1.111	0.386	0.474	0.860	7.939
17/10/2001_22/05/2003	0.501	0.306	0.484	0.791	2.393
23/05/2003_14/05/2004	1.487	0.145	0.448	0.593	3.656
15/05/2004_30/03/2006	0.548	0.028	0.770	0.798	2.712
31/03/2006_28/11/2008	0.352	0.103	0.855	0.958	8.337
29/11/2008_08/11/2010	0.036	0.045	0.941	0.985	2.477
09/11/2010_04/01/2012	3.047	-0.064	0.009	-0.055	2.889
05/01/2012_03/06/2014	0.032	0.039	0.948	0.986	2.355
04/06/2014_07/10/2015	0.598	0.024	0.521	0.545	1.314
08/10/2015_29/03/2019	0.407	0.057	0.461	0.517	0.843
		Volatilit	y Breaks in	HEROMOTOCO	
	Date of co	mmencer	ment of De	rivative trading: 31-01-200	3
05/01/2000_15/03/2001	0.263	0.071	0.893	0.964	7.290
16/03/2001_25/04/2003	0.427	0.266	0.707	0.973	15.935
26/04/2003_27/04/2004	0.073	0.082	0.900	0.981	3.916
28/04/2004_26/07/2005	0.149	0.047	0.919	0.966	4.339
27/07/2005_15/05/2006	0.767	0.074	0.639	0.713	2.671
16/05/2006_08/10/2007	0.305	0.015	0.919	0.935	4.669
09/10/2007_31/07/2009	0.569	0.079	0.875	0.954	12.340
01/08/2009_01/08/2011	0.271	0.060	0.861	0.921	3.418
02/08/2011_24/10/2017	0.215	0.071	0.874	0.946	3.953
25/10/2017_08/06/2018	0.484	-0.111	0.976	0.865	3.582
09/06/2018_29/03/2019	0.179	0.081	0.869	0.950	3.598

Period	ω	α	в	Total Persistence: (α+β)	Unconditional Volatility: ω/(1–α–β)
		Volat	ility Breaks	in ICICIBANK	
	Date of co	mmencer	nent of De	rivative trading: 31-01-200	3
05/01/2000_11/04/2000	0.748	-0.161	1.133	0.972	26.559
12/04/2000_01/10/2001	8.059	0.329	0.179	0.508	16.380
02/10/2001_05/07/2002	1.035	0.153	0.716	0.869	7.885
06/07/2002_21/03/2003	0.198	0.054	0.897	0.950	3.989
22/03/2003_11/05/2004	2.431	0.143	0.498	0.641	6.771
12/05/2004_28/02/2005	0.043	-0.054	1.041	0.987	3.219
29/02/2005_12/10/2007	0.526	0.062	0.822	0.884	4.531
13/10/2007_18/08/2009	0.778	0.109	0.857	0.966	22.936
19/08/2009_16/07/2013	0.106	0.042	0.930	0.973	3.868
17/07/2013_24/07/2015	0.034	0.003	0.985	0.988	2.970
25/07/2015_16/11/2016	3.588	0.050	0.198	0.249	4.776
17/11/2016_29/03/2019	2.071	-0.008	0.357	0.349	3.181
		Vo	latility Brea	aks in IDEA	
	Date of co	mmencer	nent of De	rivative trading: 08-03-200	7
13/03/2007_10/08/2007	1.942	-0.050	0.589	0.540	4.219
11/08/2007_15/05/2009	0.982	0.093	0.831	0.923	12.803
16/05/2009_06/08/2010	0.384	0.004	0.935	0.940	6.357
07/08/2010_04/06/2013	3.145	0.198	0.007	0.205	3.958
05/06/2013_12/01/2017	3.275	0.150	0.600	0.750	13.101
13/01/2017_29/03/2019	7.650	0.306	-0.082	0.224	9.861
		Volatil	ity Breaks	in IBULHSGFIN	
	Date of co	mmencer	nent of De	rivative trading: 30-09-201	0
26/07/2013_19/09/2013	10.009	0.113	-0.079	0.034	10.358
20/09/2013_02/06/2014	1.584	0.032	0.771	0.803	8.039
03/06/2014_22/03/2016	1.929	0.037	0.589	0.626	5.158
23/03/2016_01/11/2018	0.271	0.086	0.855	0.942	4.641
02/11/2018_29/03/2019	3.118	-0.063	0.716	0.652	8.969
		Volatil	ty Breaks i	n INDUSINDBK	
	Date of co	mmencer	nent of De	rivative trading: 29-10-201	0
05/01/2000_15/03/2001	2.278	0.189	0.655	0.844	14.638
16/03/2001_06/02/2002	2.859	0.356	0.041	0.397	4.743
07/02/2002_05/05/2003	0.372	0.081	0.855	0.937	5.862
06/05/2003_07/12/2006	1.365	0.118	0.754	0.872	10.630
08/12/2006_09/03/2007	0.969	-0.211	1.177	0.966	28.330
10/03/2007_22/07/2009	0.736	0.094	0.872	0.966	21.552
23/07/2009_02/06/2010	3.850	0.260	0.223	0.483	7.450
03/06/2010_02/04/2012	5.351	0.184	-0.181	0.002	5.364
03/04/2012_20/06/2014	0.049	0.057	0.933	0.989	4.644
21/06/2014_16/11/2016	0.362	0.034	0.808	0.842	2.292
16/11/2016_29/03/2019	0.127	0.101	0.833	0.935	1.957
		Volat	ility Break	s in INFRATEL	
	Date of co	mmencer	nent of De	rivative trading: 24-09-201	5
28/12/2012_07/06/2013	1.175	0.146	0.558	0.704	3.962
08/06/2013_20/11/2014	0.056	0.039	0.940	0.979	2.657
21/11/2014_24/09/2015	0.840	0.032	0.703	0./35	3.169
25/09/2015_31/01/2017	1.287	-0.019	0.264	0.245	1./05
01/02/201/_29/03/2019	1.037	0.276	0.123	0.400	1.726
	Detf	Vola	TIITY Break	is in INDIGO	
12/11/2015 22/01/2016	11 C10	nmencer		0 106	11 426
23/01/2016 16/02/2016	10 775	_0.304 _0.122	0.103	0.150	14.430
17/02/2016 10/02/2016	J E 3 3	-0.123	0.003	0.340	<u>کې.ب</u> ک ۲ دع
±1/02/2010_10/2010	2.000	0.050	0.000	0.000	J.UJZ

Period	ω	α	в	Total Persistence: (α+β)	Unconditional Volatility: $\omega/(1-\alpha-\beta)$
20/08/2016_27/04/2017	2.401	0.212	-0.098	0.114	2.711
28/04/2017_29/03/2019	0.977	0.024	0.828	0.852	6.613
	<u>`````````````````````````````````````</u>	Vo	latility Brea	aks in INFY	
	Date of co	mmencei	ment of De	rivative trading: 02-07-200	1
05/01/2000_27/02/2001	0.710	0.100	0.852	0.951	14.612
28/02/2001_03/11/2001	6.251	-0.196	1.046	0.850	41.605
05/11/2001_17/05/2004	1.033	0.108	0.778	0.886	9.097
18/05/2004 28/03/2006	0.447	0.005	0.810	0.814	2.408
29/03/2006_24/07/2006	5.691	0.336	-0.126	0.210	7.206
25/07/2006_01/05/2009	0.053	0.061	0.931	0.992	6.624
02/05/2009_12/07/2012	0.628	0.230	0.598	0.828	3.650
13/07/2012_11/01/2013	0.360	0.030	0.737	0.767	1.545
12/01/2013_13/03/2014	0.560	1.277	0.205	1.482	-1.163
14/03/2014_18/07/2016	0.850	-0.018	0.699	0.681	2.661
19/07/2016_3/29/2019	1.292	0.170	0.137	0.307	1.864
	,	V	olatility Bre	aks in ITC	
	Date of co	mmencei	ment of De	rivative trading: 02-07-200	1
05/01/2000 07/02/2001	-0.038	-0.023	1.026	1.003	11.359
08/02/2001 09/11/2001	3.004	0.577	0.097	0.674	9.213
10/11/2001 21/05/2002	0.579	0.078	0.662	0.740	2.226
22/05/2002_01/02/2005	0.387	0.111	0.759	0.869	2.959
02/02/2005_05/09/2005	0.744	0.001	0.565	0.566	1.715
06/09/2005 12/11/2007	0.357	0.076	0.836	0.912	4.045
13/11/2007_25/08/2009	0.459	0.084	0.845	0.928	6.416
26/08/2009_10/11/2014	0.561	0.096	0.629	0.725	2.039
11/11/2014_20/07/2017	2.498	0.272	-0.110	0.162	2.980
21/07/2017_29/03/2019	0.202	0.015	0.842	0.857	1.409
		Volat	tility Breaks	s in JUSTDIAL	
	Date of co	mmencer	nent of Dei	rivative trading: 02-04-201	3
12/07/2012 06/06/2013	0.360	0.030	0.737	0.767	1.545
07/06/2013 04/10/2013	0.296	-0.186	1.113	0.927	4.057
05/10/2013 29/05/2014	1.163	0.056	0.877	0.933	17.414
30/05/2014 28/01/2016	5.033	0.366	-0.084	0.282	7.013
29/01/2016 09/03/2016	3.886	-0.478	1.427	0.950	77.137
10/10/2017 29/05/2018	2.735	0.242	0.632	0.873	21.582
30/05/2018_29/03/2019	2.368	0.094	0.608	0.702	7.942
	· ·	Volatil	ity Breaks i	in KOTAKBANK	
	Date of co	mmencer	nent of De	rivative trading: 29-12-200	5
05/01/2000 16/11/2001	3.256	0.310	0.540	0.850	21,713
17/11/2001 01/01/2003	0.142	0.266	0.784	1.051	-2.803
02/01/2003 18/11/2004	0.172	0.096	0.888	0.984	10.853
19/11/2004 04/05/2006	3.323	0.085	0.411	0.497	6.601
05/05/2006 18/01/2008	2.728	0.259	0.453	0.712	9.478
19/01/2008 18/08/2009	2.281	0.079	0.815	0.894	21.560
19/08/2009 07/06/2012	1.175	0.146	0.558	0.704	3.962
08/06/2012 20/11/2014	0.056	0.039	0.940	0.979	2.657
21/11/2014 24/09/2015	0.840	0.032	0.703	0.735	3.169
25/09/2015 31/01/2017	1.287	-0.019	0.264	0.245	1.705
01/02/2017 29/03/2019	1.037	0.276	0.123	0.400	1.726
	1	v	olatility Br	eaks in LT	
	Date of co	mmencer	ment of De	rivative trading: 02-07-200	1
05/01/2000 21/11/2001		-0.023	1.026	1.003	11 359
22/11/2001 22/04/2004	0.041	0.031	0.956	0.987	3.098
23/04/2004 24/07/2006	1 172	0.283	0.532	0.815	6 352
			0.002	0.015	0.002

Period	ω	α	в	Total Persistence: (α+β)	Unconditional Volatility: ω/(1-α-β)
25/07/2006_02/02/2007	1.070	-0.072	0.602	0.531	2.278
03/02/2007 10/03/2008	0.540	0.205	0.742	0.947	10.227
11/03/2008_08/09/2009	6.429	0.237	0.380	0.617	16.790
09/09/2009_28/01/2011	2.361	0.120	-0.251	-0.131	2.087
29/01/2011_11/12/2013	0.071	0.049	0.935	0.985	4.583
12/12/2013 13/01/2016	1.073	-0.009	0.621	0.612	2.765
14/01/2016 05/07/2016	2.895	0.753	-0.085	0.668	8.710
06/07/2016 13/09/2019	1.767	0.047	-0.054	-0.007	1.755
		Vola	tility Break	s in MARUTI	· ·
	Date of co	mmencer	nent of Dei	rivative trading: 08-07-200	3
15/03/2001 06/02/2002	2.859	0.356	0.041	0.397	4.743
07/02/2002 05/12/2003	0.372	0.081	0.855	0.937	5.862
06/12/2003 21/01/2005	0.187	0.098	0.866	0.964	5.223
22/01/2005 05/08/2009	0.526	0.083	0.853	0.937	8.304
05/08/2009 31/05/2010	0.048	-0.065	1.045	0.980	2.422
01/06/2010 19/08/2011	0.118	-0.024	0.986	0.962	3.140
20/08/2011 21/06/2013	2.590	0.232	0.087	0.319	3.804
22/06/2013 17/12/2014	2.754	0.147	-0.080	0.067	2.951
18/12/2014 06/12/2016	0.547	0.195	0.602	0.798	2.703
07/12/2016 29/03/2019	0.025	0.031	0.955	0.986	1.830
	÷	Volat	ility Breaks	in RELIANCE	
	Date of co	mmencei	nent of De	rivative trading: 29-11-200	1
05/01/2000_28/02/2001	0.322	0.128	0.808	0.936	4.993
28/02/2001_21/12/2001	1.941	0.386	0.471	0.857	13.560
22/12/2001_02/08/2004	1.299	0.288	0.399	0.687	4.154
03/08/2004_24/07/2006	1.832	0.870	0.072	0.942	31.407
25/07/2006_24/12/2009	0.101	0.096	0.898	0.994	17.804
25/12/2009_22/08/2011	1.164	-0.014	0.510	0.496	2.309
23/08/2011_02/04/2012	1.980	-0.100	0.733	0.633	5.401
03/04/2012_29/03/2019	0.372	0.067	0.774	0.840	2.332
		Vo	latility Brea	aks in SBIN	
	Date of co	mmence	ment of Dei	rivative trading: 02-07-200	1
05/01/2000_15/03/2001	0.263	0.071	0.893	0.964	7.290
16/03/2001_25/09/2001	0.427	0.266	0.707	0.973	15.935
26/09/2001_27/04/2004	0.073	0.082	0.900	0.981	3.916
28/04/2004_26/07/2005	0.149	0.047	0.919	0.966	4.339
27/07/2005_15/05/2006	0.767	0.074	0.639	0.713	2.671
16/05/2006_08/10/2007	0.305	0.015	0.919	0.935	4.669
09/10/2007_31/07/2009	0.569	0.079	0.875	0.954	12.340
01/08/2009_01/08/2011	0.271	0.060	0.861	0.921	3.418
02/08/2011_24/10/2017	0.215	0.071	0.874	0.946	3.953
25/10/2017_08/06/2018	0.484	-0.111	0.976	0.865	3.582
09/06/2018_29/03/2019	0.179	0.081	0.869	0.950	3.598
		Vo	platility Bre	aks in TCS	
	Date of co	mmencer	nent of Der	rivative trading: 23-08-200	4
27/04/2004_26/08/2004	0.149	0.047	0.919	0.966	4.339
2//08/2004_15/04/2005	0.634	0.012	0.608	0.619	1.666
16/04/2005_25/07/2006	0.463	U.169	0.709	0.878	3./96
26/07/2006_06/07/2007	1.290	0.342	0.159	0.501	2.588
0//0//200/_29/10/2008	0.333	0.116	0.856	0.971	11.669
30/10/2008_13/08/2009	11.142	0.241	-0.171	0.070	11.982
14/08/2009_05/08/2011	1.133	0.134	0.460	0.594	2.794
U6/08/2011_10/05/2012	0.168	0.041	0.921	0.961	4.366
11/05/2012_26/06/2013	0.021	-0.041	1.030	0.988	1.823
2//06/2013_20/10/2014	0.808	0.027	0.692	0.719	2.873

Period	ω	α	в	Total Persistence: (α+β)	Unconditional Volatility: ω/(1-α-β)
21/10/2014_29/12/2017	1.048	0.151	0.232	0.383	1.699
30/12/2017_29/03/2019	0.363	0.047	0.799	0.846	2.353
		Volat	ility Breaks	in TATAELXSI	
	Date of co	mmencer	nent of De	rivative trading: 26-02-2016	
05/01/2000_01/10/2001	0.369	0.029	0.953	0.982	20.001
02/10/2001_25/12/2003	3.042	-0.007	0.590	0.584	7.307
26/12/2003_29/03/2006	1.135	0.211	0.675	0.886	9.962
30/03/2006_29/11/2007	0.331	0.080	0.880	0.959	8.164
30/11/2007_11/12/2008	5.212	0.284	0.498	0.782	23.918
12/12/2008_30/08/2011	0.938	0.222	0.644	0.865	6.976
31/08/2011_23/05/2013	1.213	0.019	0.608	0.627	3.251
24/05/2013_16/07/2014	10.945	0.049	-0.712	-0.663	6.581
17/07/2014_01/11/2017	0.357	0.027	0.882	0.910	3.944
02/11/2017_29/03/2019	0.959	0.072	0.568	0.641	2.670
		Volati	ility Breaks	in TATASTEEL	
	Date of co	mmencer	ment of De	rivative trading: 02-07-2001	
05/01/2000_27/07/2000	3.121	0.193	0.559	0.752	12.592
28/07/2000_20/07/2001	1.703	0.276	0.534	0.810	8.986
21/07/2001_12/08/2002	1.010	0.084	0.750	0.834	6.066
13/08/2002_10/07/2003	1.885	0.270	-0.044	0.226	2.436
11/07/2003_28/06/2006	0.155	0.092	0.888	0.979	7.480
29/06/2006_26/01/2007	0.250	-0.002	0.923	0.921	3.177
27/01/2007_23/05/2013	0.089	0.079	0.912	0.990	8.987
24/05/2013_11/08/2015	0.262	0.061	0.887	0.948	4.996
12/08/2015_28/11/2016	0.233	0.007	0.955	0.962	6.183
29/11/2016_29/03/2019	5.508	-0.035	-0.640	-0.674	3.289
		Vo	latility Brea	aks in VEDL	
	Date of co	mmencer	nent of De	rivative trading: 29-12-2006	
05/01/2000_24/12/2001	4.740	0.234	0.385	0.618	12.419
25/12/2001_07/04/2003	0.624	0.096	0.839	0.935	9.618
08/04/2003_08/03/2007	2.851	0.203	0.597	0.800	14.255
09/03/2007_26/03/2008	0.698	0.468	0.591	1.059	-11.844
27/03/2008_08/08/2016	0.206	0.076	0.904	0.980	10.189
09/08/2016_29/05/2017	5.367	-0.044	0.112	0.068	5.756
30/05/2017_29/03/2019	2.337	-0.018	0.582	0.565	5.369
		Volat	tility Break	s in YESBANK	
	Date of co	mmencer	nent of De	rivative trading: 06-09-2007	
14/07/2005_03/02/2008	2.310	0.158	0.510	0.668	6.951
04/02/2008_16/03/2012	0.244	0.112	0.870	0.982	13.216
17/03/2012_03/03/2016	0.196	0.068	0.900	0.968	6.100
04/03/2016_24/08/2018	0.676	0.083	0.689	0.772	2.969
25/08/2018_29/03/2019	9.789	-0.024	0.577	0.553	21.892
		Vo	latility Bre	aks in ZEEL	
	Date of co	mmencei	ment of De	rivative trading: 07-10-2010	
05/01/2000_12/10/2001	3.815	0.137	0.740	0.877	31.114
13/10/2001_08/01/2004	2.143	0.105	0.698	0.803	10.882
09/01/2004_02/10/2008	4.177	0.163	0.307	0.470	7.887
03/10/2008_28/10/2009	0.719	0.114	0.851	0.966	20.982
29/10/2009_09/06/2010	3.145	0.321	0.032	0.354	4.865
10/06/2010_29/11/2015	0.538	0.053	0.802	0.855	3.710
30/11/2015_05/10/2018	0.562	0.050	0.649	0.699	1.868
06/10/2018_29/03/2019	17.897	0.138	-0.556	-0.417	12.629