Link between Spot and Subsequent Closing Prices with Subsequent of Stock and Index in India

Vigneshwaran. T and Srinivasan. R

Abstract--- This research examines whether a relationship exists in India between spot and subsequent closing prices for NIFTY 50 stock and index subsequent on NSE. To achieve this objective, Nifty 50 Index & 25 select stocks are listed on the NIFTY 50 Index traded on NSE India for a period from April 2005 to December 2015, taking into account the inclusions and exclusions of the constituents of the Nifty 50 Index during the study period. The sample used in this study includes for Nifty Index & 25 select stocks traded on NSE (www.nseindia.com) daily subsequent close prices and spot closing prices. Since most trading activity occurs in near-month contracts, only near-month contracts are analyzed using econometric tools descriptive statistics, unit root check, granger causality, and correction of vector errors. The analysis reveals that spot-to-subsequent markets have a bidirectional relationship. The study also provides evidence of a long-run balance link between the index of spot market prices and their subsequent prices. This means that either of these two historic prices would help predict the other.

Keywords--- price closing subsequent, price closing spot, causality of grangers, co-integration, correction of vector errors.

I. INTRODUCTION

Financial derivatives have been playing a central role in the global financial markets. The derivatives market size has surpassed growth in other regions. Since its inception in 2000, the advent and growth of India's derivatives market has been phenomenal. Indian markets and Indian derivative products are listed in the Global rankings within the specified span of time. Generally, market participants are commonly utilizing derivative products on volatility indexes as tools for risk management and hedging toward market volatility. (SEBI, Yearly Report). Trading in derivatives has evolved more and more on NSE since its inception. In terms of price discovery, portfolio diversification and risk management, the derivatives industry assists in various functions on the financial market.

Vigneshwaran. T, Research Scholar, Bharathiar University, Coimbatore.

Srinivasan. R, Associate Professor, P.G. & Research, Department of Corporate Secretaryship, Bharathidasan Government College for Women (Autonomous), Puducherry – 605003.

Trading in derivatives has therefore brought about a major change in the spot market. Spot market fluctuations were mainly due to the practices of trading, hedging and arbitration on the subsequent market. Therefore, in the short run it is important to investigate the cause and effect connection between the spot and the subsequent as well as in the long run. In addition, which market first reacts is the reaction spot or subsequent market in India's dynamic financial market.

II. LITERATURE REVIEW

Dr. Jain Mathew, Dr. K. Srinivasan, Miss. Aditi Davidson (2012) this paper analyzed the effect on the underlying spot market volatility due to the introduction of a subsequent market in India for the period 1 January 1995 to 31 December 2011. The study concluded that index subsequent have a very significant role to play in mitigating market volatility and have contributed to improved market efficiency and spill over on the subsequent market leading to spot market, making the spot market unstable.

SahuDhananjay, (2012), this paper analyzed the effects of equity derivatives trading on spot market volatility, in particular the effect of the implementation of equity derivatives on spot market volatility in the Indian stock market by using the daily returns of 73 companies from April 01, 1998 to March 31, 2008, except holidays when no transactions occurred. Applied to the study of market volatility was the GARCH (1, 1) model capturing heteroscedasticity in returns. All the companies under review, however, showed asymmetric response and accordingly the GJR GARCH model that captured the asymmetric response was applied using CNX Nifty index return as the independent variable to reduce the effect of market-wide factors on equity returns. The results indicated that the dummy variable coefficient was large and negative and therefore it suggested that the advent of equity derivatives trading decreased the volatility of the spot market.

KapilChoudhary, Sushil Bajaj, Intraday (2012), this study examined whether spot and subsequent markets play an important role in assimilating the Indian stock market's knowledge and price discovery. The research applied the co-integration of Johansen and the residual-based approach of Engle and Granger to assess the long-term equilibrium between the two markets. In addition, the Granger causality test and VECM (Vector Error Correction Model) were used to determine causality direction and the leading market. The study results showed that there is a bi-directional flow of information or feedback between the spot and subsequent markets.

Mall Manmohan (2011), this paper based from 2000 to 2011 on analyzing India's stock index subsequent market. It also empirically checked the efficiency of the Indian spot and index subsequent market, analyzed the time-varying properties of Indian spot and index subsequent market volatility using the cointegration test of GARCH, ADF, PP, VECM and Johansen for the period from June 2000 to May 2011 after collecting the closing prices of the near-month subsequent contract and the daily closing prices of S&P CNX NIFTY. The conclusions drawn were that the capital market efficiency check provides evidence of the Indian spot market's poor form inefficiency and that the subsequent market forms the spot market relatively productive. The relative success of the subsequent market was the product of index subsequent being implemented. High volatility persistence in the subsequent market has resulted in unidirectional volatility flowing from index subsequent to spot market. It also concluded that there is long-term causality from index subsequent to spot market prices. ShiqingXie, Jiajun Huang (2013), this paper presented an empirical analysis using the Vector Error Correction Model (VECM) on the price discovery function of index subsequent in China for the study's CSI 300 index period from April 2010 to April 2012. The conclusions drawn were that there is a long-term solid cointegration relationship between the CSI 300 index and its index subsequent; where prices deviate from the long-term equilibrium, the stock index reverses slightly, while the index subsequent reversal is much stronger; the daily lead-lag relationship between the prices of the CSI 300 index and its index subsequent market, but not vice versa because of the spot-market's limited short-term adjustment capacity.

GoyalNiti (2012), this paper analyzed the volatility estimate prevailing on the Indian stock market, whether volatility is static or has changed over time, the effect on price discovery of derivatives trading, has the effect of derivatives trading on indices volatility on the stock market and that of individual shares been examined and has the introduction of derivatives been successful in reducing stock market volatility or has some other macroeconomic factors led to changes in volatility? The study concluded that there was volatility in the Indian stock market, adjustments in stock market volatility after derivatives were introduced, derivatives trading were not helpful in finding equity stock prices, trading derivatives had an effect on stock market volatility, and the introduction of derivatives was ineffective in reducing volatility.

III. RESEARCH METHODOLOGY

The research is performed for Nifty 50 Index & 25 select stocks on NIFTY 50 Index traded on NSE India for a period from April 2005 to December 2015, taking into account inclusions and exclusions from the constituents of the Nifty 50 Index during the study period, using various tools to achieve the objective. The research period is kept standardized from 1st April 2005 to 31 December 2015 in order to assist in comparative analysis. The sample used in this study includes Nifty Index & 25 select stocks traded on NSE's regular subsequent close prices and spot closing prices. Since most of the trading activity happens in contracts close to the month, only contracts close to the month are reviewed. All values are converted to natural logarithms, calculated as $R_t = LN(P_t / P_{t-1})$ where P_t and P_{t-1} are natural logarithms on day t and t-1 respectively to prevent non-stationary in order to obtain accurate results for the incorporated test.

| INDEX | Nifty 50 | | | | | | |
|-------|--------------------------|--------------------|-------------|--|--|--|--|
| | Company Name | Industry | Symbol | | | | |
| | ACC Ltd. | CEMENT & CEMENT | ACC | | | | |
| | | PRODUCT | | | | | |
| | Ambuja Cement Ltd. | CEMENT & CEMENT | AMBUJACEM | | | | |
| | | PRODUCT | | | | | |
| | Bank of Baroda | FINANCIAL SERVICES | BANK BARODA | | | | |
| | Bharat Heavy Electricals | INDUSTRIAL | BHEL | | | | |
| | Ltd. | MANUFACTURING | | | | | |

| | Bharat Petroleum | ENERGY | BPCL | | | |
|----------|--|--------------------|------------|--|--|--|
| | Corporation Ltd. | | | | | |
| | Cipla Ltd. | PHARMA | CIPLA | | | |
| | Gail (India) Ltd. | ENERGY | GAIL | | | |
| | HCL Technologies Ltd. | IT | HCLTECH | | | |
| | Housing Development | FINANCIAL SERVICES | HDFC | | | |
| | Finance Corporation Ltd. | | | | | |
| | HDFC Bank Ltd. | FINANCIAL SERVICES | HDFC BANK | | | |
| | Hero MotoCorp Ltd. | AUTOMOBILE | HEROMOTOCO | | | |
| | Hindalco Industries Ltd. | METALS | HINDALCO | | | |
| | Hindustan Unilever Ltd. | CONSUMER GOODS | HINDUNILVR | | | |
| | ICICI Bank Ltd. | FINANCIAL SERVICES | ICICIBANK | | | |
| | Infosys Ltd. | IT | INFY | | | |
| | ITC Ltd.CONSUMER GOODSITCMahindra & Mahindra Ltd.AUTOMOBILEM&MMaruti Suzuki India Ltd.AUTOMOBILEMARUTIOil & Natural GasENERGYONGCCorporation Ltd.ENERGYRELIANCEReliance Industries Ltd.ENERGYRELIANCEState Bank of IndiaFINANCIAL SERVICESSBINTata Motors Ltd.AUTOMOBILETATAMOTORS | | | | | |
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| | Tata Power Co. Ltd | ENERGY | TATAPOWER | | | |
| | Tata Steel Ltd. | METALS | TATASTEEL | | | |
| | Tata Consultancy Services | IT | TCS | | | |
| | Ltd. | | | | | |
| DATA | Futures Close Prices | | | | | |
| VARIABLE | Spot Closing Prices | | | | | |
| PERIOD | April 2005 – December 2015 | | | | | |
| TOOLS | Descriptive Statistics | | | | | |
| | • Unit Root Test | | | | | |
| | Ganger Causality | | | | | |
| | Co-integration | | | | | |
| | Vector error correction | | | | | |
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Objective

To investigate the cause-effect relationship in India between Stock Index Subsequent& Stock Subsequent and Spot Index & Spot Subsequent.

Hypothesis

H0- The relationship between Spot Close Price and Subsequent Close Price is not important

H1- Subsequent Close Price and Spot Close Price are not significantly related

IV. STATISTICS DESCRIPTIVE

To investigate the cause and effect relationship between subsequent close price and spot close price, we measure the NIFTY Index's daily log returns and the selected 25 stocks based on its subsequent daily close price and spot close price between 1st April 2005 and 31st December 2015. To know the pattern of distribution and also the output of stocks, we analyze the descriptive analysis of the spot close price. The descriptive spot close price statistics are summarized in the table below 1.9 in terms of mean, standard deviation, Skewness, Kurtosis and JarqueBera for the Nifty 50 Index and select 25 stocks for the period from April 1st, 2005 to December 31st, 2015.

| | Mean | Standard | Skewness | Kurtosis | Jarque-Bera | Probability | Observance |
|------------|-----------|-----------|----------|----------|-------------|-------------|------------|
| | | deviation | | | | | |
| ACC | 0.000495 | 0.021014 | -0.35346 | 7.948804 | 2772.894 | 0 | 2669 |
| AMBUJACEM | -0.00028 | 0.045406 | -33.5038 | 1498.646 | 2.49E+08 | 0 | 2669 |
| BANKBARODA | -7.70E-05 | 0.040686 | -23.5047 | 938.7657 | 97406699 | 0 | 2669 |
| BHEL | -0.00058 | 0.04264 | -22.6509 | 835.1935 | 77071457 | 0 | 2669 |
| BPCL | 0.000327 | 0.027625 | -6.12171 | 166.6597 | 2988596 | 0 | 2669 |
| CIPLA | 0.000349 | 0.026196 | -17.0018 | 604.9623 | 40334996 | 0 | 2669 |
| GAIL | 0.000195 | 0.023552 | -2.62723 | 54.5215 | 297598.5 | 0 | 2669 |
| HCLTECH | 0.000291 | 0.032185 | -8.14585 | 185.2833 | 3716286 | 0 | 2669 |
| HDFC | 0.000207 | 0.038727 | -25.3167 | 1036.782 | 1.19E+08 | 0 | 2669 |
| HDFC BANK | 0.000252 | 0.03718 | -30.1994 | 1307.906 | 1.89E+08 | 0 | 2669 |
| HEROMOTOCO | 0.000605 | 0.020006 | 0.498345 | 9.347388 | 4580.662 | 0 | 2669 |
| HINDALCO | -0.00104 | 0.053104 | -29.6867 | 1274.177 | 1.80E+08 | 0 | 2669 |
| HINDUNILVR | 0.000689 | 0.018713 | 0.437624 | 7.204359 | 2046.371 | 0 | 2669 |
| ICICI BANK | -0.00016 | 0.041202 | -22.0019 | 856.0108 | 80951178 | 0 | 2669 |
| INFY | -0.00027 | 0.031061 | -13.123 | 296.5476 | 9637733 | 0 | 2669 |
| ITC | -0.00053 | 0.0558 | -39.7117 | 1840.265 | 3.75E+08 | 0 | 2669 |
| MAHINDRA | 0.000344 | 0.030417 | -8.20243 | 190.9113 | 3947872 | 0 | 2669 |
| MARUTI | 0.000898 | 0.021448 | -0.07396 | 5.98711 | 992.4896 | 0 | 2669 |
| ONGC | -0.00047 | 0.036193 | -24.2226 | 954.9631 | 1.01E+08 | 0 | 2669 |
| RELIANCE | 0.00023 | 0.026607 | -8.11114 | 216.4365 | 5083923 | 0 | 2669 |
| SBIN | -0.0004 | 0.050065 | -35.539 | 1621.337 | 2.91E+08 | 0 | 2669 |
| TATAMOTORS | -3.50E-05 | 0.042242 | -22.5655 | 883.9484 | 86337445 | 0 | 2669 |

Table 4.1: Statistics descriptive of Log Spot Close Price (LSCP)

| TATAPOWER | -0.0062 | 0.051322 | -34.8822 | 1580.987 | 2.77E+08 | 0 | 2669 |
|-----------|----------|----------|----------|----------|----------|---|------|
| TATASTEEL | -0.00018 | 0.029114 | -0.27824 | 6.386945 | 1307.207 | 0 | 2669 |
| TCS | 0.000217 | 0.028535 | -11.2354 | 282.1195 | 8700533 | 0 | 2669 |
| NIFTY 50 | 0.000522 | 0.01536 | -0.04439 | 12.23048 | 9447.621 | 0 | 2669 |

Source: Computed Value

Table 4.1 displays the spot-close return sequence descriptive statistics for selecting 25 stocks and the NIFTY50 index. It shows positive mean returns for ACC, BPCL, CIPLA, GAIL, HDCLTECH, HDFCBANK, HEROMOTOCORP, HINDULVR, MAHINDRA, MARUTI, RELIANCE, TCS & NIFTY50 Index showing superior spot close prices while negative mean returns for MBUJJACEM, BANKBARODA, BHEL, HINDALCO, ICICIBANK, INFOSYS, ITC, ONGC, SBI, TATAMOTORS, TATAPOWER & TATASTEEL. The standard deviation for ITC stock is high with 0.0558 and lowest with 0.018713 for HINDULVR followed by NIFTY50 as 0.01536. With the exception of HERO & HINDULVR stock, the skewedness for all stocks is found to be negatively skewed. The kurtosis value exceeds 3, showing that the unconditional return distributions are not normal. JB test confirms that normality is rejected at a meaning level of nearly 1 percent at p-value.

| | Mean | Standard deviation | Skewness | Kurtosis | Jarque-Bera | Probability | Observance |
|------------|----------|--------------------|----------|----------|-------------|-------------|------------|
| ACC | 0.000495 | 0.021327 | -0.37702 | 8.171455 | 3037.38 | 0 | 2669 |
| AMBUJACEM | -0.00027 | 0.045586 | -33.6615 | 1509.123 | 2.53E+08 | 0 | 2669 |
| BANKBARODA | -0.00013 | 0.040924 | -23.1398 | 920.5788 | 93870142 | 0 | 2669 |
| BHEL | -0.00058 | 0.042624 | -22.3122 | 819.2022 | 74306883 | 0 | 2669 |
| BPCL | 0.000332 | 0.027364 | -5.85755 | 157.921 | 2684320 | 0 | 2669 |
| CIPLA | 0.000342 | 0.026352 | -17.3538 | 622.7336 | 42845723 | 0 | 2669 |
| GAIL | 0.000199 | 0.023739 | -2.62723 | 54.03742 | 292754.5 | 0 | 2669 |
| HCLTECH | 0.000302 | 0.032135 | -7.39226 | 165.5796 | 2963780 | 0 | 2669 |
| HDFC | 0.00021 | 0.038564 | -25.8802 | 1068.587 | 1.27E+08 | 0 | 2669 |
| HDFC BANK | 0.000255 | 0.037103 | -30.4198 | 1321.784 | 1.94E+08 | 0 | 2669 |
| HEROMOTOCO | 0.0006 | 0.019482 | 0.43559 | 8.927979 | 3992.366 | 0 | 2669 |
| HINDALCO | -0.00104 | 0.053214 | -29.6213 | 1271.6 | 1.79E+08 | 0 | 2669 |
| HINDUNILVR | 0.0007 | 0.018351 | 0.373241 | 7.438357 | 2252.663 | 0 | 2669 |
| ICICI BANK | -0.00017 | 0.041191 | -22.1308 | 863.7654 | 82614012 | 0 | 2669 |
| INFY | -0.00026 | 0.030506 | -13.4926 | 308.3661 | 10450990 | 0 | 2669 |
| ITC | -0.00053 | 0.056071 | -40.0418 | 1862.726 | 3.85E+08 | 0 | 2669 |
| MAHINDRA | 0.000343 | 0.03037 | -8.12283 | 201.7414 | 4426443 | 0 | 2669 |
| MARUTI | 0.000887 | 0.021347 | -0.12623 | 6.341993 | 1249.164 | 0 | 2669 |
| ONGC | -0.00049 | 0.036324 | -23.5156 | 920.9992 | 93963760 | 0 | 2669 |

Table 4.2: Statistics Descriptive of Log Subsequent Close Price (LNFCL)

| RELIANCE | 0.000217 | 0.02658 | -8.12283 | 216.8377 | 5114523 | 0 | 2669 |
|------------|-----------|----------|----------|----------|----------|---|------|
| SBIN | -0.00041 | 0.050314 | -35.114 | 1596.692 | 2.83E+08 | 0 | 2669 |
| TATAMOTORS | -3.56E-05 | 0.042239 | -22.8511 | 898.3168 | 89376019 | 0 | 2669 |
| TATAPOWER | -0.0063 | 0.051452 | -34.5541 | 1562.17 | 2.71E+08 | 0 | 2669 |
| TATASTEEL | -0.00017 | 0.029492 | -0.28732 | 6.2517 | 212.589 | 0 | 2669 |
| TCS | 0.000195 | 0.028387 | -11.5283 | 291.1232 | 9291074 | 0 | 2669 |
| NIFTY 50 | 0.000505 | 0.016255 | -0.13364 | 11.26912 | 7612.185 | 0 | 2669 |

Source: Computed Value

Table 4.2 provides the following relevant observations

The mean returns of the subsequent close prices of the stocks namely ACC, BPCL, CIPLA, GAIL, HECLTECH, HDFC, HDFCBANK, HEROMOTOCORP, HINDULVR, MAHINDRA, MARUTI, RELIANCE, TCS & NIFTY INDEX are optimistic, suggesting that the price series had increased and that of AMBUJACEM, BANKBARODA, BHEL, HINDALCO, ICICIBANK, INFOSYS, ITC, ONGC, SBI, TATAMOTORS, TATAPOWER & TATASTEEL are negative, implying that the price series had dropped from April 2005 to December 2015. The volatile nature of stocks is evident from the statistics on the standard deviation of the close price returns of the daily subsequent. The least volatile stock is HINDULVR with 0.018351 & 0.016255 NIFTY50 Index standard deviation. At 0.056071 suggesting the highest volatile stock in terms of subsequent close values, the highest standard deviation is observed in the ITC. Negative skewed implies that the return distribution of stocks except for HEROMOTOCORP & HINDULVR having positive skewedness, which means there are higher chances of lower returns. The Kurtosis value exceeds 3, showing that the unconditional return distributions are not natural. JB check indicates that normality is rejected at a significance point of approximately 1 percent at p-value.

V. UNIT ROOT TEST

5.1. Important Deckey Fuller test

This study uses the standard Augmented Dickey-Fuller (ADF) test to determine whether the presumed time series is I, a prerequisite for further testing. Next, check the unit roots when interception is present in regression, then when interception and trend are present, and finally without interception and trend. If the null hypothesis about the unit root cannot be rejected, run the ADF on the initial time series differences. In this step, to infer that the original time series is I, we should reject the null hypothesis about the unit root. The data used are closing prices for the daily subsequent and closing prices for the period 1st April 2005 to 31st December 2015. All the regular values are translated to natural logarithms, measured as $R_t = LN (P_t / P_{t-1})$, where P_t and P_{t-1} are natural logarithms on day t and t-1. The variables for the analysis are found to be stationary at rates after translating the series to natural logarithms and therefore we reject the null concluding that the series has a root unit. Therefore, since the null hypothesis is rejected that the data is non-stationary or has a unit root as shown in table 4, the sequence is stationary.

H0: Has a unit root (i.e. non-stationary data)

H1: Has no unit root (i.e. stationary data)

Table 5.1.1: Test Result of ADF for Spot Close Price & Subsequent Close Price

| Intercept Trend None Intercept Trend None ACC -49.4427* -49.45073* -49.42802* -48.89827* -48.90490* -48.88 AMBUJACEM -53.1746 -53.18925* -53.18269* -53.03973* -53.05376* -53.04 BANKBARODA -50.92400* -50.94875* -50.93301* -50.78092* -50.80379* -50.799 BHEL -49.80415* -49.82187* -49.80461* -49.69829* -49.71556* -49.699 BPCL -51.00927* -51.00404* -51.01144* -51.10742* -51.10160* -51.110 CIPLA -50.54330* -50.53977* -50.54443* -50.33647* -50.332001* -53.332 GAIL -53.82910* -53.82450* 53.83524* -53.32427* -53.32001* -53.332 HCLTECH -51.6471* -51.25646* -51.66772* -51.64434* -51.63831* -51.655 HDFC -51.36043* -51.35435* -51.36765* -51.19196* -51.18599* -51.195 <t< th=""><th colspan="4">Spot Close Price</th></t<> | Spot Close Price | | | |
|---|------------------|--|--|--|
| ACC - 49.4427* - 49.45073* - 49.42802* - 48.89827* - 48.90490* - 48.88 AMBUJACEM - 53.1746 - 53.18925* - 53.18269* - 53.03973* - 53.05376* - 53.054 BANKBARODA -50.92400* -50.94875* - 50.93301* -50.78092* - 50.80379* - 50.799 BHEL - 49.80415* -49.82187* - 49.80461* -49.69829* - 49.71556* - 49.699 BPCL - 51.00927* -51.00404* - 51.01144* -51.10742* - 51.10160* - 51.101 CIPLA - 50.54330* -50.53977* - 50.54443* -50.33647* - 50.33316* - 50.33 GAIL - 51.26471* -51.25646* -51.26982* -50.66872* -50.66025* -50.674 HDFC - 51.36043* -51.35435* -51.3675* -51.19196* -51.18539* -51.199 HEROMOTOCO -50.38299* -50.37687* -50.34587* -32.88722* -32.88477* -32.820 HINDALCO -51.39215* -51.40389* -51.38235* -51.13204* -51.14356* -51.122 HINDUNILVR -51.67432* <th< th=""><th></th></th<> | | | | |
| AMBUJACEM - 53.1746 - 53.18925* - 53.18269* - 53.03973* - 53.05376* - 53.044 BANKBARODA -50.92400* -50.94875* - 50.93301* -50.78092* - 50.80379* - 50.799 BHEL - 49.80415* -49.82187* - 49.80461* -49.69829* - 49.71556* - 49.699 BPCL - 51.00927* -51.00404* - 51.01144* -51.10742* - 51.10160* - 51.110 CIPLA - 50.54330* -50.53977* - 50.54443* -50.33647* - 50.33316* - 50.333 GAIL - 53.82910* -53.82450* 53.83524* -53.32427* -53.32001* -53.330 HCLTECH -51.26471* -51.25646* -51.26982* -50.66872* -50.66025* -50.674 HDFC -51.36043* -51.35435* -51.36772* -51.64434* -51.63831* -51.695 HDFC BANK -51.36043* -51.35435* -51.36765* -51.19196* -51.18559* -51.199 HEROMOTOCO -50.38299* -50.37687* -50.34587* -32.88722* -32.88477* -32.820 HINDALCO -51.57432* | | | | |
| BANKBARODA -50.92400* -50.94875* - 50.93301* -50.78092* - 50.80379* - 50.790 BHEL -49.80415* -49.82187* -49.80461* -49.69829* -49.71556* -49.698 BPCL - 51.00927* -51.00404* - 51.01144* -51.10742* - 51.10160* - 51.116 CIPLA - 50.54330* -50.53977* - 50.54443* -50.33647* - 50.33316* - 50.333 GAIL - 53.82910* -53.82450* 53.83524* -53.32427* -53.32001* -53.330 HCLTECH -51.26471* -51.25646* -51.26982* -50.66072* -50.66025* -50.674 HDFC -51.36043* -51.35435* -51.36765* -51.19196* -51.18559* -51.199 HEROMOTOCO -50.38299* -50.37687* -50.34587* -32.88722* -32.88477* -32.820 HINDALCO -51.67432* -51.66463* -51.70824* -51.69850* -51.6443 INPY -52.20194* -52.2019* -52.20778* -52.20384* -51.69850* -51. | 98* | | | |
| BHEL-49.80415*-49.82187*-49.80461*-49.69829*-49.71556*-49.698BPCL-51.00927*-51.00404*-51.01144*-51.10742*-51.10160*-51.114CIPLA-50.54330*-50.53977*-50.54443*-50.33647*-50.33316*-50.333GAIL-53.82910*-53.82450*53.83524*-53.32427*-53.32001*-53.330HCLTECH-51.26471*-51.25646*-51.26982*-50.66872*-50.66025*-50.674HDFC-51.65956*-51.65341*-51.66772*-51.64434*-51.63831*-51.655HDFC BANK-51.36043*-51.35435*-51.36765*-51.19196*-51.18559*-51.199HEROMOTOCO-50.38299*-50.37687*-50.34587*-32.88722*-32.88477*-32.820HINDALCO-51.39215*-51.66463*-51.60887*-51.13204*-51.14356*-51.122HINDUNILVR-51.67432*-51.66463*-51.60887*-51.70824*-51.69850*-51.6474ICICI BANK-49.88977*-49.91406*-49.89833*-49.44036*-49.46400*-49.4449INFY-52.20194*-52.20519*-52.20778*-52.20384*-52.20689*-52.2059ITC-51.51800*-51.53465*-51.52308*-51.27931*-51.29644*-51.284MAHINDRA-48.91175*-48.90294*-48.91500*-48.71005*-48.70123*-48.713MARUTI-49.36379*-49.35789*-49.29166*-49.05111*-49.04524*-48.975 <td>769*</td> | 769* | | | |
| BPCL- 51.00927*-51.00404*- 51.01144*-51.10742*- 51.10160*- 51.1160*CIPLA- 50.54330*-50.53977*- 50.54443*-50.33647*- 50.33316*- 50.337GAIL- 53.82910*-53.82450*53.83524*-53.32427*-53.32001*- 53.330HCLTECH- 51.26471*-51.25646*-51.26982*-50.66872*-50.66025*- 50.674HDFC- 51.65956*- 51.65341*- 51.36765*-51.19196*-51.18559*- 51.657HDFC BANK- 51.36043*-51.35435*-51.36765*-51.19196*-51.18559*-51.199HEROMOTOCO-50.38299*-50.37687*-50.34587*-32.88722*-32.88477*-32.820HINDALCO-51.39215*-51.40389*-51.38235*-51.13204*-51.14356*-51.122HINDUNILVR-51.67432*-51.66463*-51.60887*-51.70824*-51.69850*-51.6474INFY-52.20194*-52.20519*-52.20778*-52.20384*-52.20689*-52.209ITC-51.51800*-51.53465*-51.52308*-51.27931*-51.29644*-51.284MAHINDRA-48.91175*-48.90294*-48.91500*-48.71005*-48.70123*-48.713MARUTI-49.36379*-49.35789*-49.29166*-49.05111*-49.04524*-48.979 |)29* | | | |
| CIPLA- 50.54330*-50.53977*- 50.54443*-50.33647*- 50.33316*- 50.33GAIL- 53.82910*- 53.82450*53.83524*- 53.32427*- 53.32001*- 53.330HCLTECH- 51.26471*- 51.25646*- 51.26982*- 50.66872*- 50.66025*- 50.674HDFC- 51.65956*- 51.65341*- 51.66772*- 51.64434*- 51.63831*- 51.657HDFC BANK- 51.36043*- 51.35435*- 51.36765*- 51.19196*- 51.18559*- 51.199HEROMOTOCO- 50.38299*- 50.37687*- 50.34587*- 32.88722*- 32.88477*- 32.820HINDALCO- 51.39215*- 51.66463*- 51.60887*- 51.70824*- 51.69850*- 51.64434*INFY- 52.20194*- 52.20519*- 52.20778*- 52.20384*- 52.20689*- 52.209ITC- 51.51800*- 51.53465*- 51.52308*- 51.27931*- 51.29644*- 51.284MAHINDRA- 48.91175*- 48.90294*- 48.91500*- 48.7105*- 48.70123*- 48.713MARUTI- 49.36379*- 49.35789*- 49.29166*- 49.05111*- 49.04524*- 48.979 | 397* | | | |
| GAIL- 53.82910*-53.82450*53.83524*-53.32427*-53.32001*-53.330HCLTECH-51.26471*-51.25646*-51.26982*-50.66872*-50.66025*-50.674HDFC- 51.65956*- 51.65341*- 51.66772*-51.64434*- 51.63831*- 51.653HDFC BANK-51.36043*-51.35435*-51.36765*-51.19196*-51.18559*-51.199HEROMOTOCO-50.38299*-50.37687*-50.34587*-32.88722*-32.88477*-32.820HINDALCO-51.39215*-51.66463*-51.60887*-51.70824*-51.69850*-51.64434*INDUNILVR-51.67432*-51.66463*-51.60887*-51.70824*-51.69850*-51.64434*INFY-52.20194*-52.20519*-52.20778*-52.20384*-52.20689*-52.209ITC-51.51800*-51.53465*-51.52308*-51.27931*-51.29644*-51.284MAHINDRA-48.91175*-48.90294*-48.91500*-48.71005*-48.70123*-48.713MARUTI-49.36379*-49.35789*-49.29166*-49.05111*-49.04524*-48.979 |)13* | | | |
| HCLTECH-51.26471*-51.25646*-51.26982*-50.66872*-50.66025*-50.6744HDFC-51.65956*-51.65341*-51.66772*-51.64434*-51.63831*-51.653HDFC BANK-51.36043*-51.35435*-51.36765*-51.19196*-51.18559*-51.1999HEROMOTOCO-50.38299*-50.37687*-50.34587*-32.88722*-32.88477*-32.820HINDALCO-51.39215*-51.40389*-51.38235*-51.13204*-51.14356*-51.122HINDUNILVR-51.67432*-51.66463*-51.60887*-51.70824*-51.69850*-51.6449INFY-52.20194*-52.20519*-52.20778*-52.20384*-52.20689*-52.2099ITC-51.51800*-51.53465*-51.52308*-51.27931*-51.29644*-51.284MAHINDRA-48.91175*-48.90294*-48.91500*-48.71005*-48.70123*-48.979MARUTI-49.36379*-49.35789*-49.29166*-49.05111*-49.04524*-48.979 | 749* | | | |
| HDFC- 51.65956*- 51.65341*- 51.66772*- 51.64434*- 51.63831*- 51.653HDFC BANK- 51.36043*- 51.35435*- 51.36765*- 51.19196*- 51.18559*- 51.199HEROMOTOCO- 50.38299*- 50.37687*- 50.34587*- 32.88722*- 32.88477*- 32.820HINDALCO- 51.39215*- 51.40389*- 51.38235*- 51.13204*- 51.14356*- 51.122HINDUNILVR- 51.67432*- 51.66463*- 51.60887*- 51.70824*- 51.69850*- 51.6447ICICI BANK- 49.88977*- 49.91406*- 49.89833*- 49.44036*- 49.46400*- 49.449INFY- 52.20194*- 52.20519*- 52.20778*- 52.20384*- 52.20689*- 52.209ITC- 51.51800*- 51.53465*- 51.52308*- 51.27931*- 51.29644*- 51.284MAHINDRA- 48.91175*- 48.90294*- 48.91500*- 48.71005*- 48.70123*- 48.979MARUTI- 49.36379*- 49.35789*- 49.29166*- 49.05111*- 49.04524*- 48.979 | 54* | | | |
| HDFC BANK-51.36043*-51.35435*-51.36765*-51.19196*-51.18559*-51.1996HEROMOTOCO-50.38299*-50.37687*-50.34587*-32.88722*-32.88477*-32.820HINDALCO-51.39215*-51.40389*-51.38235*-51.13204*-51.14356*-51.122HINDUNILVR-51.67432*-51.66463*-51.60887*-51.70824*-51.69850*-51.647ICICI BANK-49.88977*-49.91406*-49.89833*-49.44036*-49.46400*-49.449INFY-52.20194*-52.20519*-52.20778*-52.20384*-52.20689*-52.209ITC-51.51800*-51.53465*-51.52308*-51.27931*-51.29644*-51.284MAHINDRA-49.36379*-49.35789*-49.29166*-49.05111*-49.04524*-48.979 | 79* | | | |
| HEROMOTOCO-50.38299*-50.37687*-50.34587*-32.88722*-32.88477*-32.820HINDALCO-51.39215*-51.40389*-51.38235*-51.13204*-51.14356*-51.122HINDUNILVR-51.67432*-51.66463*-51.60887*-51.70824*-51.69850*-51.647ICICI BANK-49.88977*-49.91406*-49.89833*-49.44036*-49.46400*-49.449INFY-52.20194*-52.20519*-52.20778*-52.20384*-52.20689*-52.209ITC-51.51800*-51.53465*-51.52308*-51.27931*-51.29644*-51.284MAHINDRA-48.91175*-48.90294*-48.91500*-48.71005*-48.70123*-48.713MARUTI-49.36379*-49.35789*-49.29166*-49.05111*-49.04524*-48.979 | 264* | | | |
| HINDALCO-51.39215*-51.40389*-51.38235*-51.13204*-51.14356*-51.122HINDUNILVR-51.67432*-51.66463*-51.60887*-51.70824*-51.69850*-51.647ICICI BANK-49.88977*-49.91406*-49.89833*-49.44036*-49.46400*-49.494INFY-52.20194*-52.20519*-52.20778*-52.20384*-52.20689*-52.209ITC-51.51800*-51.53465*-51.52308*-51.27931*-51.29644*-51.284MAHINDRA-48.91175*-48.90294*-48.91500*-48.71005*-48.70123*-48.713MARUTI-49.36379*-49.35789*-49.29166*-49.05111*-49.04524*-48.979 | 12* | | | |
| HINDUNILVR-51.67432*-51.66463*-51.60887*-51.70824*-51.69850*-51.647ICICI BANK-49.88977*-49.91406*-49.89833*-49.44036*-49.46400*-49.49.49INFY-52.20194*-52.20519*-52.20778*-52.20384*-52.20689*-52.209ITC-51.51800*-51.53465*-51.52308*-51.27931*-51.29644*-51.284MAHINDRA-48.91175*-48.90294*-48.91500*-48.71005*-48.70123*-48.713MARUTI-49.36379*-49.35789*-49.29166*-49.05111*-49.04524*-48.979 | 99* | | | |
| ICICI BANK -49.88977* -49.91406* -49.89833* -49.44036* -49.46400* -49.4494 INFY -52.20194* -52.20519* -52.20778* -52.20384* -52.20689* -52.209 ITC -51.51800* -51.53465* -51.52308* -51.27931* -51.29644* -51.284 MAHINDRA -48.91175* -48.90294* -48.91500* -48.71005* -48.70123* -48.713 MARUTI -49.36379* -49.35789* -49.29166* -49.05111* -49.04524* -48.979 | 02* | | | |
| INFY -52.20194* -52.20519* -52.20778* -52.20384* -52.20689* -52.209 ITC -51.51800* -51.53465* -51.52308* -51.27931* -51.29644* -51.284 MAHINDRA -48.91175* -48.90294* -48.91500* -48.71005* -48.70123* -48.713 MARUTI -49.36379* -49.35789* -49.29166* -49.05111* -49.04524* -48.979 | 09* | | | |
| ITC -51.51800* -51.53465* -51.52308* -51.27931* -51.29644* -51.284 MAHINDRA -48.91175* -48.90294* -48.91500* -48.71005* -48.70123* -48.713 MARUTI -49.36379* -49.35789* -49.29166* -49.05111* -49.04524* -48.979 | 11* | | | |
| MAHINDRA -48.91175* -48.90294* -48.91500* -48.71005* -48.70123* -48.713 MARUTI -49.36379* -49.35789* -49.29166* -49.05111* -49.04524* -48.979 | 57* | | | |
| MARUTI -49.36379* -49.35789* -49.29166* -49.05111* -49.04524* -48.979 | 19* | | | |
| | 27* | | | |
| ONGC -51.32418* -51.31807* -51.32469* -50.72645* -50.72007* -50.727 | 82* | | | |
| | 65* | | | |
| RELIANCE -50.59414* -50.59995* -50.60032* -50.09545* -50.10025* -50.101 | 28* | | | |
| SBIN -51.17824* -51.22239* -51.18442* -50.73245* -50.77628* -50.738 | 95* | | | |
| TATAMOTORS -45.56743* -48.55966* -48.57650* -48.10293* -48.09518* -48.111 | 96* | | | |
| TATAPOWER -52.42049* -52.43148* -52.42237* -52.29445* -52.30516* -52.296 | 56* | | | |
| TATASTEEL -49.56544* -49.56415* -49.57308* -48.06210* -48.06097* -48.069 | 54* | | | |
| TCS -51.33602* -51.34425* -51.34323* -51.29086* -51.29920* -51.298 | 07* | | | |
| NIFTY 50 -50.78146* -50.78380* -50.74270* -48.46953* -48.47072* -48.425 | 91* | | | |

Source: Computed Value. Note: * denotes rejection of null hypothesis at 5% level of significance

5.2. Granger Test Causality

A direct "Granger-causality" test proposed by C is the method for evaluating statistical causality between subsequent close prices and spot close prices. J. Granger was used in 1969. Granger causality in the usual sense may have more to do with precedent, or inference, than with causation.

H₀: Spot Close Price is not the cause of Subsequent Close Price

 H_{01} : Subsequent Close Price is not the cause of Spot Close Price

Table 5.2.1: Results of Granger Causality Test

| COMPANY | LAGS | LNSCL ->LNFCL | LNFCL ->LNSCL |
|------------|------|-------------------|-------------------|
| ACC | 7 | 1.91156(0.0638) | 2.14629(0.0361)* |
| AMBUJACEM | 7 | 0.71746(0.6573) | 0.51418(0.8245) |
| BANKBARODA | 3 | 0.04576(0.9870) | 2.27094(0.0784) |
| BHEL | 8 | 0.39637(0.9231) | 2.25243 (0.0214)* |
| BPCL | 7 | 2.02020(0.0491)* | 2.89409(0.0052)* |
| CIPLA | 8 | 0.61831(0.7632) | 2.04400 (0.0380)* |
| GAIL | 8 | 1.34921(0.2143) | 2.02504(0.0400)* |
| HCLTECH | 7 | 3.03596(0.0035)* | 0.94464(0.4705) |
| HDFC | 7 | 0.96561(0.4546) | 0.86443(0.5339) |
| HDFCBANK | 6 | 0.50583(0.8044) | 0.93952(0.4653) |
| HEROMOTOCO | 7 | 1.14032(0.3346) | 2.08086(0.0424)* |
| HINDALCO | 7 | 1.45562(0.1786) | 1.13607(0.3372) |
| HINDUNILVR | 8 | 2.04364(0.0381)* | 2.90439(0.0032)* |
| ICICIBANK | 7 | 0.63679(0.7258) | 0.32267(0.9441) |
| INFY | 7 | 0.92503(0.4857) | 0.98560(0.4397) |
| ITC | 5 | 0.66073(0.6533) | 0.28123(0.9236) |
| MAHINDRA | 8 | 1.43996(0.1746) | 1.16155(0.3187) |
| MARUTI | 8 | 2.10917 (0.0318)* | 1.73602(0.0853) |
| ONGC | 8 | 0.52971(0.8349) | 0.41292(0.9138) |
| RELIANCE | 8 | 0.54803(0.8208) | 0.46205(0.8833) |
| SBIN | 7 | 1.37511(0.2113) | 0.74493(0.6338) |
| TATAMOTORS | 7 | 0.48140(0.8487) | 1.01644(0.4173) |
| TATAPOWER | 7 | 0.33248(0.9394) | 0.71093(0.6628) |
| TATASTEEL | 8 | 1.73528(0.0855) | 1.21700(0.2846) |
| TCS | 8 | 1.38038(0.1999) | 0.59523(0.7826) |
| NIFTY50 | 7 | 1.59808(0.1312) | 1.05738(0.3886) |

Source: Computed Value. Note: * denotes rejection of hypothesis at 5% level of significance

Table 5.2.1 reflects the results of the Granger Causality test in which it is shown that a bi-directional causality exists for BPCL & HINDULVR stocks from spot to subsequent close price returns. There is unidirectional causality for ACC, BHEL, CIPLA, GAIL & HEROMOTOCORP from subsequent to spot, whereas in HCLTECH & MARUTI it is observed from spot to subsequent. No causality has been found between spot and subsequent for AMBUJACEM, HDFC, HDFCBANK, HINDALCO, ICICIBANK,

INFOSYS, ITC, MAHINDRA, ONGC, RELIANCE, SBI, TATAMOTORS, TATASTEEL, TCS & NIFTY50

indicates that spot causes no subsequent, so the subsequent does not affect the spot for the study period.

5.3. Co-integration

H₀: Subsequent Close Price & Spot Close Price

H₁: It is not co-integrated: Subsequent Close Price & Spot Close Price is co-integrated.

Table 5.3.1: Results of Johansen Co-integration

| STOCK | NO.OF CE(S) | EIGENVALUE | TRACE | PROBABILITY |
|------------|-------------|------------|-----------|-------------|
| | | | STATISTIC | |
| ACC | NONE | 0.049589 | 138.5750 | 0.0001* |
| | AT MOST 1 | 0.001510 | 3.997541 | 0.0456* |
| AMBUJACEM | NONE | 0.056096 | 173.3701 | 0.0001* |
| | AT MOST 1 | 0.007760 | 20.61387 | 0.0000* |
| BANKBARODA | NONE | 0.048216 | 134.0132 | 0.0001* |
| | AT MOST 1 | 0.001229 | 3.255016 | 0.0712 |
| BHEL | NONE | 0.060916 | 167.7986 | 0.0001* |
| | AT MOST 1 | 0.000566 | 1.497355 | 0.2211 |
| BPCL | NONE | 0.075849 | 210.4941 | 0.0001* |
| | AT MOST 1 | 0.000672 | 1.778567 | 0.1823 |
| CIPLA | NONE | 0.063727 | 176.1511 | 0.0001* |
| | AT MOST 1 | 0.000724 | 1.916928 | 0.1662 |
| GAIL | NONE | 0.045385 | 130.3821 | 0.0001* |
| | AT MOST 1 | 0.002824 | 7.483884 | 0.0062* |
| HCLTECH | NONE | 0.066850 | 185.9320 | 0.0001* |
| | AT MOST 1 | 0.001079 | 2.857813 | 0.0909 |
| HDFC | NONE | 0.068227 | 191.8367 | 0.0001* |
| | AT MOST 1 | 0.001833 | 4.853971 | 0.0276* |
| HDFCBANK | NONE | 0.046008 | 129.8842 | 0.0001* |
| | AT MOST 1 | 0.001985 | 5.257821 | 0.0218* |
| HEROMOTOCO | NONE | 0.061570 | 168.8129 | 0.0001* |
| | AT MOST 1 | 0.000252 | 0.666425 | 0.4143 |
| HINDALCO | NONE | 0.051596 | 167.8520 | 0.0001* |
| | AT MOST 1 | 0.010407 | 27.68205 | 0.0000* |
| HINDULVR | NONE | 0.068383 | 187.4304 | 0.0001* |
| | AT MOST 1 | 1.59E-06 | 0.004204 | 0.9470 |
| ICICIBANK | NONE | 0.047472 | 135.1798 | 0.0001* |
| | AT MOST 1 | 0.002450 | 6.489613 | 0.0108* |

| INFOSYS | NONE | 0.047302 | 134.1244 | 0.0001* |
|------------|-----------|----------|----------|---------|
| | AT MOST 1 | 0.002229 | 5.905144 | 0.0151* |
| ITC | NONE | 0.077393 | 231.1271 | 0.0001* |
| | AT MOST 1 | 0.006774 | 17.98619 | 0.0000* |
| M&M | NONE | 0.044750 | 124.7318 | 0.0001* |
| | AT MOST 1 | 0.001356 | 3.591730 | 0.0581 |
| MARUTI | NONE | 0.066764 | 187.2785 | 0.0001* |
| | AT MOST 1 | 0.001679 | 4.447678 | 0.0349* |
| ONGC | NONE | 0.050252 | 138.6104 | 0.0001* |
| | AT MOST 1 | 0.000826 | 2.187188 | 0.1392 |
| RELIANCE | NONE | 0.057926 | 161.5341 | 0.0001* |
| | AT MOST 1 | 0.001376 | 3.643175 | 0.0563 |
| SBIN | NONE | 0.048992 | 136.2750 | 0.0001* |
| | AT MOST 1 | 0.001269 | 3.359951 | 0.0668 |
| TATAMOTORS | NONE | 0.040291 | 110.8295 | 0.0001* |
| | AT MOST 1 | 0.000760 | 2.011177 | 0.1561 |
| TATAPOWER | NONE | 0.049731 | 136.4213 | 0.0001* |
| | AT MOST 1 | 0.000547 | 1.447487 | 0.2289 |
| TATASTEEL | NONE | 0.052245 | 146.2656 | 0.0001* |
| | AT MOST 1 | 0.001618 | 4.284046 | 0.0385* |
| TCS | NONE | 0.065401 | 179.2513 | 0.0001* |
| | AT MOST 1 | 0.000107 | 0.282574 | 0.5950 |
| NIFTY50 | NONE | 0.049519 | 135.6209 | 0.0001* |
| | AT MOST 1 | 0.000584 | 1.543138 | 0.2142 |

Source: Computed Value. Note: * denotes rejection of hypothesis at 5% level of significance

To examine the long-term relationship, the Johansen Co-integration test is used. Johansen Cointegration is well known to be very sensitive to choosing the length of the lag. In order to find a suitable lag structure, a VAR model is first fitted to the time series data. The number of lags required in the co-integration test is selected using the AIC, SC, LR. The co-integration test indicates that at the meaning level of 5 percent there is one co-integrating vector. This indicates the long-term co-integration of the subsequent close price and spot close price. The trace test shows the presence at 5 percent level of significance of two co-integrating equations. This result is confirmed by the Maximum Eigen Value test. Thus the study's two variables have a long-term relationship of balance between them. But there may be short-term deviations from this equilibrium, and we have to check whether or not such equilibrium converges to long-term equilibrium. VECM can therefore be used to generate the dynamics of the short run.

| | C(1) LNF CL (- | C(2) D | C(3) D LN | C(4) D | C(5) D | |
|------------|----------------|----------|-----------|-----------|-----------|----------|
| | 1) | LNF CL(- | F CL(-2) | LNSCL(-1) | LNSCL(-2) | C(6) C |
| | | 1) | | | | |
| ACC | -1.06405 | 0.054286 | -0.06243 | -0.68117 | -0.25671 | 2.85E-05 |
| | -4.56783 | 0.302587 | -0.56562 | -3.77518 | -2.30433 | 0.059816 |
| | 0* | 0.7622 | 0.5717 | 0.0002* | 0.0213* | 0.9523 |
| AMBUJACEM | 1.085462 | -1.66052 | -0.70006 | 0.983805 | 0.3632 | 5.41E-05 |
| | 2.243769 | -4.4762 | -3.19113 | 2.611804 | 1.636261 | 0.05291 |
| | 0.0249* | 0* | 0.0014* | 0.0091* | 0.1019 | 0.9578 |
| | 0.088536 | -0.62739 | -0.22467 | -0.0288 | -0.1111 | 1.60E-05 |
| BANKBARODA | 0.211452 | -1.93522 | -1.16031 | -0.08817 | -0.57061 | 0.017421 |
| | 0.8326 | 0.0531* | 0.246 | 0.9297 | 0.5683 | 0.9861 |
| BHEL | -0.08157 | -0.26729 | -0.09651 | -0.3637 | -0.22224 | 6.03E-06 |
| | -0.17709 | -0.75674 | -0.46371 | -1.03273 | -1.07244 | 0.006285 |
| | 0.8595 | 0.4493 | 0.6429 | 0.3018 | 0.2836 | 0.995 |
| BPCL | -0.95534 | 0.019178 | 0.098869 | -0.6756 | -0.41478 | 3.49E-05 |
| | -3.2871 | 0.086275 | 0.740379 | -3.04981 | -3.12483 | 0.056866 |
| | 0.001* | 0.9313 | 0.4591 | 0.0023* | 0.0018* | 0.9547 |
| CIPLA | -0.97781 | 0.114521 | 0.058378 | -0.75754 | -0.39906 | 3.18E-05 |
| | -2.59069 | 0.39971 | 0.350912 | -2.63438 | -2.39143 | 0.053591 |
| | 0.0096* | 0.6894 | 0.7257 | 0.0085* | 0.0169* | 0.9573 |
| GAIL | -0.12142 | -0.75202 | -0.36928 | 0.06773 | 0.009082 | 3.23E-05 |
| | -0.55397 | -4.47104 | -3.57275 | 0.39547 | 0.086548 | 0.060111 |
| | 0.5796 | 0* | 0.0004* | 0.6925 | 0.931 | 0.9521 |
| HCLTECH | -1.21413 | -0.08533 | -0.09829 | -0.54003 | -0.21472 | 9.45E-06 |
| | -3.99119 | -0.36871 | -0.72456 | -2.34542 | -1.58496 | 0.012961 |
| | 0.0001* | 0.7124 | 0.4688 | 0.0191* | 0.1131 | 0.9897 |
| HDFC | -0.14234 | -0.67221 | -0.19102 | 0.019917 | -0.14216 | 3.53E-05 |
| | -0.2539 | -1.58509 | -0.76597 | 0.046843 | -0.56959 | 0.04043 |
| | 0.7996 | 0.1131 | 0.4438 | 0.9626 | 0.569 | 0.9678 |
| HDFCBANK | 0.079437 | -0.88717 | -0.39177 | 0.23329 | 0.055773 | 1.05E-05 |
| | 0.156036 | -2.30488 | -1.76737 | 0.605254 | 0.251452 | 0.012586 |
| | 0.876 | 0.0213* | 0.0773 | 0.5451 | 0.8015 | 0.99 |
| HEROMOTOCO | 0.572622 | -1.08862 | -0.53412 | 0.500009 | 0.22756 | 2.44E-05 |

Table 5.3.2: Results of Vector Error Correction

| | 4.044371 | -9.80837 | -7.45515 | 4.362621 | 3.15004 | 0.054954 |
|------------|----------|----------|----------|----------|----------|-----------|
| | 0.0001* | 0* | 0* | 0* | 0.0017* | 0.9562 |
| HINDALCO | -1.79636 | 0.441034 | 0.080947 | -1.10248 | -0.40079 | 4.13E-06 |
| | -2.35142 | 0.770268 | 0.24732 | -1.92408 | -1.22237 | 0.00345 |
| | 0.0188* | 0.4412 | 0.8047 | 0.0545 | 0.2217 | 0.9972 |
| HINDULVR | -0.20736 | -0.64863 | -0.3319 | -0.01589 | 0.003136 | 2.60E-05 |
| | -0.98263 | -3.99703 | -3.39474 | -0.09854 | 0.03244 | 0.062974 |
| | 0.3259 | 0.0001* | 0.0007* | 0.9215 | 0.9741 | 0.9498 |
| ICICIBANK | -0.78251 | -0.25594 | -0.06704 | -0.3727 | -0.25743 | 1.28E-05 |
| | -1.79936 | -0.76615 | -0.33482 | -1.1199 | -1.28831 | 0.013797 |
| | 0.0721 | 0.4437 | 0.7378 | 0.2629 | 0.1977 | 0.989 |
| INFOSYS | 1.856668 | -2.37324 | -1.17131 | 1.709242 | 0.851672 | -3.05E-06 |
| | 3.656638 | -6.10658 | -5.11143 | 4.40844 | 3.741565 | -0.00441 |
| | 0.0003* | 0* | 0* | 0* | 0.0002* | 0.9965 |
| ITC | -2.02594 | 0.598906 | 0.355924 | -1.24985 | -0.68267 | 1.80E-05 |
| | -2.83536 | 1.088138 | 1.082868 | -2.26606 | -2.06944 | 0.014222 |
| | 0.0046* | 0.2766 | 0.279 | 0.0235* | 0.0386* | 0.9887 |
| M&M | -1.65834 | 0.429311 | 0.081386 | -1.03886 | -0.36912 | 2.10E-05 |
| | -4.68824 | 1.608924 | 0.541775 | -3.91799 | -2.46924 | 0.030938 |
| | 0* | 0.1078 | 0.588 | 0.0001* | 0.0136* | 0.9753 |
| MARUTI | -0.79969 | -0.20779 | -0.14531 | -0.41169 | -0.15085 | 3.96E-05 |
| | -3.0809 | -1.05885 | -1.27565 | -2.10142 | -1.32667 | 0.082496 |
| | 0.0021* | 0.2898 | 0.2022 | 0.0357* | 0.1847 | 0.9343 |
| ONGC | -0.25008 | -0.63382 | -0.31338 | -0.01734 | -0.00505 | 1.58E-06 |
| | -0.64674 | -2.12992 | -1.75563 | -0.0582 | -0.02821 | 0.00192 |
| | 0.5179 | 0.0333* | 0.0793 | 0.9536 | 0.9775 | 0.9985 |
| RELIANCE | 2.544755 | -2.6471 | -1.1586 | 2.046858 | 0.880003 | 7.01E-06 |
| | 5.162687 | -6.92898 | -5.06112 | 5.303133 | 3.819922 | 0.011613 |
| | 0* | 0* | 0* | 0* | 0.0001* | 0.9907 |
| SBIN | -1.62161 | -0.00574 | -0.09994 | -0.63917 | -0.22379 | 3.61E-05 |
| | -2.11537 | -0.0098 | -0.29337 | -1.09115 | -0.65531 | 0.031846 |
| | 0.0345* | 0.9922 | 0.7693 | 0.2753 | 0.5123 | 0.9746 |
| TATAMOTORS | -0.41321 | -0.36246 | -0.05256 | -0.23633 | -0.24953 | 3.28E-05 |
| | -0.9866 | -1.1271 | -0.27957 | -0.7367 | -1.33048 | 0.034357 |
| | 0.3239 | 0.2598 | 0.7798 | 0.4614 | 0.1835 | 0.9726 |
| TATAPOWER | -0.92318 | -0.05562 | 0.070614 | -0.61497 | -0.39959 | 2.45E-05 |

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| | -1.78633 | -0.13989 | 0.300413 | -1.54579 | -1.69803 | 0.021054 |
|-----------|----------|----------|----------|----------|----------|-----------|
| | 0.0742 | 0.8888 | 0.7639 | 0.1223 | 0.0896 | 0.9832 |
| TATASTEEL | -0.50291 | -0.53312 | -0.21351 | -0.10864 | -0.11351 | 4.11E-05 |
| | -1.47815 | -2.03046 | -1.35304 | -0.41402 | -0.7155 | 0.062706 |
| | 0.1395 | 0.0424* | 0.1762 | 0.6789 | 0.4744 | 0.95 |
| TCS | -0.7699 | -0.27832 | -0.30863 | -0.35159 | -0.00987 | -1.21E-05 |
| | -1.72471 | -0.82219 | -1.57065 | -1.03918 | -0.05027 | -0.01877 |
| | 0.0847 | 0.411 | 0.1164 | 0.2988 | 0.9599 | 0.985 |
| NIFTY50 | -2.21808 | 0.921352 | 0.616827 | -1.57277 | -0.95665 | -2.05E-06 |
| | -5.89006 | 3.204017 | 3.747055 | -5.39175 | -5.65961 | -0.00563 |
| | 0* | 0.0014* | 0.0002* | 0* | 0* | 0.9955 |

Source: Computed Value

It is noted that the results of the error correction co-integrating term C(1) indicate the long-term relationship in the short-term dynamics, and C(2) to C(6) indicates the short-term relationship between the variables. It is reflected that for the majority of stocks, there is a short-term relationship between the variables. In all these cases where the co-integrating term is negative and small, it is shown that the errors go back to the equilibrium and the error is corrected while the positive and significant co-integrating term implies that the errors are bursting.

VI. CONCLUSION

Higher liquidity, lower transaction costs, lower margins, easier leverage positions, rapid execution and greater flexibility for short positions are mainly due to the advantages offered by the subsequent market. These advantages attract larger knowledgeable traders and make the subsequent markets respond first when marketwide or stock-specific information comes in. Subsequent prices therefore lead the prices of the spot market. On the other hand, low-cost contingent strategies and high degree of leverage gains in the subsequent market draw larger speculative traders to a more regulated subsequent market segment from a spot market. It effectively decreases the spot market's knowledge asymmetries by reducing the amount of noise trading and helps in price discovery, enhancing the overall scope of the market, improving market efficiency and increasing market liquidity. When market-wide information or major stock-specific information arrives, this makes the spot market react first. In addition, there is a two-way relationship between subsequent and spot markets through the process of price discovery. This may be due primarily to subsequent markets that attract larger informed traders to enjoy the benefits of higher liquidity, lower transaction costs, lower margins, and greater flexibility for short positions. Therefore, these benefits make subsequent markets lead the spot markets around macroeconomic or major stock-specific releases of information. The spot markets will therefore lead the subsequent market under the circumstances that these subsequent markets draw greater speculative traders from the spot market and reduce spot market knowledge asymmetries by reducing the amount of noise trading and helping to discover prices, improve the overall market size, improve market efficiency and increase market liquidity. It helps the spot market react quickly when market-wide or stock-specific information comes in.

Johansen's technique of co-integration followed by the Vector Error Correction Model (VECM) was used to investigate the long-term relationship between stock subsequent and stock index subsequent. The empirical analysis was carried out from April 2005 to December 2015 for the daily data series. The analysis reveals the spot-to-subsequent demand bidirectional relationship. The research also provides evidence of a longrun equilibrium relationship between the index of spot market prices and their subsequent prices. It means that either of these two historical prices would help predict the other, which is the proof that these two markets disapprove of the theory of market efficiency.

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