

THE ROLE OF STRUCTURAL BREAK AND VOLATILE INNOVATIONS ON COINTEGRATION TESTS: TSUNAMI AND GLOBAL ECONOMICS CRISIS

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This paper examines the impact of structural break and volatile innovations on contegration tests. It is suggested that eartquake-tsunami effects and Standard and Poor's downgrades US credit rating from AAA to AA⁻ might be considered as a structural break or breaks on stock market indexes of the some countries. The evidences of structural break/breaks are investigated on Turkey, Japan, United Kingdom, United States of America and China economy for period 2011. By dealing with these two challenges in this study, we find evidence to support the failure of Engle-Granger (E-G), Gregory-Hansen (G-H) and conventional Lagrange-Multiplier (LM) cointegration tests to identify stuructural break. This failure may arise from their not taking into account potential structural breaks and volatile innovations. We conclude that if heteroscedasticity is more prominent than structural break, Westerlund-LM test should be performed.

Keywords: Structural break, Heteroscedasticity, Stock market, Cointegration, ARCH effect.

I. Introduction

Economic and finance theory usually indicates the existence of long-run equilibrium relationships among nonstationary time series variables. Most economic time series behaves nonstationary in clearly for two distinct reasons: heteroscedasticity (i.e. inconstant variance) and level shift (i.e. structural break). Even if some transformations and differentiations may fix the instable variance and imbalance mean, to stabilize the series could be difficult due to persistent effects of unexpected shocks to economic activity. Cointegration analysis in the presence of structural break/breaks is basically relevant, when existence of long-run relationships among economic variables has still engaged the attention of full spread of researchers from economics, econometrics, statistics and probability.

Gregory and Hansen (1996) developed residual-based tests for cointegration that explicitly allow one structural break in the cointegrating vector. Then, Leybourne and Newbold (2003) and Cook (2004) combined that misspecification of the structural breaks may be more challenging, in hypothesis testing, than ignorance of them. Perron (2006) revealed that two types of structural change have occurred in relation to cointegration testing. Recently, Westerlund and Edgerton (2007) have proposed cointegration tests for the case of no break, a break in the intercept as well as a break in the intercept and cointegrated vector. These tests are derived from Lagrange Multiplier (LM)-based unit root tests of Schmidt and Phillips (1992). LM test takes into account both structural breaks and volatility of the series. The organization of this paper proceeds as follows: In Section II we review the unit root tests: Augmented-

Dickey Fuller (ADF) and Zivot-Andrews which takes into account structural break/breaks. Section III outlines Engle Granger (EG), Gregory Hansen (GH), Engle's ARCH test, LM Test of Westerlund-Edgerton (LM-Test-WE) cointegration tests. In Section IV we present application of these tests on stock market price index. This empirical part of the study includes macro economical impacts of earthquake and tsunami disasters in Japan and downgrades USA credit rating in 2011. We suggest that the important international issue of whether economic growth paths of different countries converge in some sense over time. These extraordinary events considered as structural break as exogenous variables and known time period. In Section V we give some brief remarks.

Here, we address a proposition in detail with the motivation that, if one can identify the presence of structural break correctly and heteroscedasticity, one can in principle aware of the importance the selection of the proper cointegration test. Hereby, we may suggest a route map for practitioners when they use cointegration test.

II. Unit Root Analyses

Augmented Dickey Fuller

Unit roots of a series is a precondition to the existence of cointegration relationship, originally, the Augmented Dickey-Fuller (1979) test was extensively used to test for stationarity. The ADF test which analyzes whether any series included has unit root or not, gives information about whether the series are cointegrated or not, since a similar operation is applied on the error term in cointegration analysis. As the error terms obtained from the linear correlation between the series, under cointegration investigation, can be modeled with their lagged values, Dickey – Fuller test can be applied on this data. ADF test primarily concerned with the estimate of α . In the following equation, we test the null hypothesis of $\rho = 0$ against the alternative hypothesis of $\rho < 0$:

$$\Delta y_t = u + \beta t + \rho y_{t-1} + \sum_{i=1}^k c_i \Delta y_{t-1} + \varepsilon_t$$
⁽¹⁾

where y_t is the time series being tested, Δ denotes the first difference and k is the number of lags which are added to the model to ensure that the residuals, ε_t are white noise. When $\rho < 0$, this model has a unit root and becomes a random walk process.

However, Perron (1989) demonstrated that failure to allow for an existing break leads to a bias that decreased the ability to reject a false unit root null hypothesis. A problem common with the conventional unit root tests —such as the ADF tests, is that they do not allow for the possibility of a structural break. Perron argues that most macroeconomic series are not characterized by a unit root but rather that persistence arises only from large and infrequent shocks, and that the economy returns to deterministic trend after small and frequent shocks Glynn et al.(2007). Perron (1989) test includes the time of break into the model externally. Zivot and Andrews (1992), on the other hand, developed a test that includes the time of the break internally.

III. Cointegration Tests

Engle-Granger

The residual-based cointegration test suggested by Engle and Granger have been used commonly in the literature. The basic approach in Engle – Granger method is the error terms of a linear combination between two non-stationary time series having the property of stationarity.

$$y_t = \mu_1 + \alpha x_t + \mathcal{E}_t \tag{2}$$

A general model that can be built between two series can be presented as in equation (2). In this model the dependent variable y_t , the independent variable x_t , and the error term ε_t , which is random, is presented.

In order to variables in the model to be cointegrated, it is both assumed that the difference of both variables are obtained (I(1) distributed) and at the same time the error term is non-differenced (I(0) distributed). In other words, the error term is $\mathcal{E}_t \sim IN(0, \sigma^2)$.

In order to determine the existence of the cointegraiton between the series Engle – Granger proposed a procedure comprising of two steps. According to this procedure, first a linear equation (ordinary least squares, OLS) is built and the parameter estimations are obtained by using the least square method. As the second step the unit root test is applied on the error terms obtained from the model. In order to determine whether the error terms are stationary or not, the Dickey-Fuller test is widely used.

Gregory - Hansen

The study results of Gregory et al. (1996) indicate that the power of the Engle–Granger (1987) test of the null of no cointegration is dramatically reduced if a break in the cointegrating relationship. In order to overcome this drawback, Gregory and Hansen (1996) extend the Engle–Granger test to allow for single structural break in either the intercept or the intercept and trend of the cointegrating relationship at an unknown time. The breaks are tried to be determined by adding dummy variables to the Engle - Granger method.

Gregory – Hansen (1996) test investigates the determination of structural breaks in long term relation under three different models. these models being specified and denoted as follows:

Model C : Level shift
$$y_t = \mu_1 + \mu_2 \varphi_\tau + \alpha x_t + \varepsilon_t$$
 (3)

Model C/T: Level shift with trend $y_t = \mu_1 + \mu_2 \varphi_\tau + \beta t + \alpha x_t + \varepsilon_t$ (4)

Model C/S : Regime shift
$$y_t = \mu_1 + \mu_2 \varphi_\tau + \alpha_1 x_t + \alpha_2 \varphi_\tau x_t + \varepsilon_t$$
 (5)

The dummy variable φ_{τ} which is included in the model for the determination of the structural break can be defined as below.

$$\varphi_{\tau} = \begin{cases} 1, & if \ t > \tau \\ 0, & otherwise \end{cases}$$

with τ is a coefficient which shows the break point occurs in the sample and takes the value of 0 or 1. Non-stationarity of the obtained residuals is controlled by the ADF test. Finding the test statistics to the minimum value of the ADF statistics in the sequence, we choosed the value that constitutes the powerful evidence against the null hypothesis of no cointegration.

Lagrange Multiplier Cointegration Test

Westerlund and Edgerton (2007) have proposed cointegration tests that are derived from the Lagrange Multiplier (LM)-based unit root tests of Schmidt and Phillips (1992). They consider the cases of no break, a break in the intercept as well as a break in the intercept and cointegrating vector of the cointegrating

relationship (Tam, 2012). This test take into account heteroskedastic and serially correlated errors, deterministic time trends and a single break in both the intercept and slope.

Bivariate system of integrated series (y_t, x_t)

$$y_t = \mu_1 + \beta t + \mu_2 \varphi_\tau + \alpha_1 x_t + \alpha_2 \varphi_\tau x_t + \varepsilon_t \qquad t=1,..,T$$

$$x_t = x_{t-1} + w_t \qquad (6)$$

$$\mathcal{E}_{t} = \rho \mathcal{E}_{t-1} + v_{t}$$

$$\varphi_{\tau} = \begin{cases} 1, & \text{if } t > \tau \\ 0, & \text{otherwise} \end{cases}$$

where *t* represents the time index, μ_1 is the intercept, β , μ_2 , β_1 , β_2 and α are coefficients, with ε_t , w_t and v_t are error process. Test use a t-statistics for testing the null of $\phi = 0$ against the alternative of $\phi < 0$ in the following regression:

$$\Delta \widetilde{S}_{t} = \gamma + \phi \widetilde{S}_{t-1} + \beta t$$

$$\widetilde{S}_{t} = y_{t} - \widetilde{\mu}_{e1} - \widetilde{\beta} t - \widetilde{\mu}_{2} \varphi_{t} - x_{t} \widetilde{\alpha}_{1} - \varphi_{t} x_{t} \widetilde{\alpha}_{2} \qquad t=2, 3, 4, ..., T \qquad (7)$$

with $\tilde{S}_1 = 0$, $\tilde{\mu}_{e_1} \equiv \tilde{\mu}_1 + e_0$, $\phi = \alpha - 1$ and \tilde{S}_t being the regression errors. The parameter estimates can be obtained by an ordinary least squares regression of the equation.

Engle's ARCH Test

Financial time series often display some well-known property. First, large changes tend to be followed by large changes and small changes tend to be followed by small changes. Secondly financial time series often exhibit leptokurtosis and volatility clastering, which means that the distribution of their returns is fat-tailed.

The standard warning is that in the presence of heteroskedasticity, the regression coefficients for an ordinary least squares regression are still unbiased, but the standard errors and confidence intervals estimated by conventional procedures will be too narrow, giving a false sense of precision. ARCH and GARCH models treat heteroskedasticity as a variance to be modelled (Engle,1982).

Various tests have been developed in order to determine the heteroscedasticity in time series model. These test are; Goldfeld – Quandt Test (GQ), Breusch – Pagan – Godfrey Test (BPG) ARCH (LM) and GARCH, Mcleod and Li's Q. The most widely used, among these tests, Robert Engle' ARCH-LM. The test's null hypothesis is there is no ARCH up to order q in the residuals. We use ARCH LM test in order to understand whether the standardized residuals exhibit additional ARCH.

The general ARCH model is;

$$\varepsilon_{t} = v_{t} \sqrt{h_{t}}$$

$$h_{t} = \alpha_{0} + \alpha_{1} \varepsilon_{t-1}^{2}$$
(8)

 $v_t \sim \text{WN}(0, \sigma^2 = 1) v_t$ and \mathcal{E}_{t-1} independent each other. In addition α_0, α_1 are constant $\alpha_0 > 0, 0 < \alpha_1 < 1$

Thus ARCH process have zero mean, uncorrelated, unconditional variance is constant (if exist) but conditional variance varying depends on time. If the coefficient of variance equation are not significant, there is no ARCH effect in the residual series. Otherwise it can be said that, series have ARCH component.

IV. Empirical Study

Earthquake and tsunami hit Japan in March 11th, 2011, both were having a large negative economic impact on Japan. Than, not the exactly same date, but in the same year, USA credit rating was decreased in August 5th, 2011. As a consequence of these events, global crisis had spreaded all around the world. We consider that these large disasters' might reveal a structural break or breaks on Japan economy and also world economy. For this reason, some stock markets which were thought in related with Japan and USA economy had examined for detecting structural break. Stock market indexes were represented on a daily basis ranging from January 2011 and December 2011. The sample was chosen reflect the periods on the basis of common working hours for all markets of countries, so that the data was gathered by paying attention to the local time difference between countries. The sample was performed with 223 observations in Eviews Version 7, R, Gauss 15, MATLAB.

As you see in Figure 1, the impacts of Earthquake & Tsumai disasters and global crisis are shown up in the series.

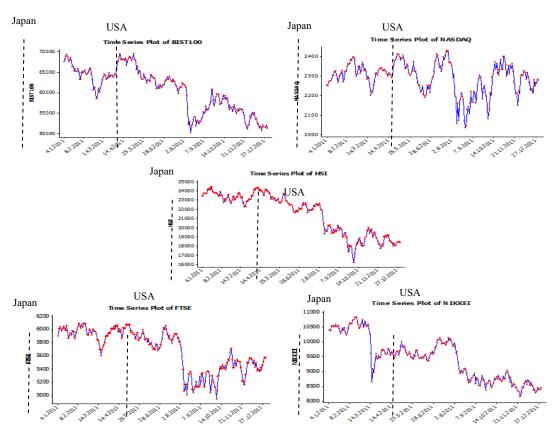


Figure 1. Time series graphs of BIST, NASDAQ, HSI, FTSE and NIKKEI in 2011

Examining the stock price indexes of five countries, we can observe that all the series have sharply decreasing in their level, occurring in March 11th and August 5th, 2011, but the decreasing degrees are not similar to each other. First sharp rush in NIKKEI on 43th data can be explained by the impact of earthquake and tsunami, infact same sharply impacts of them are more difficult to see on the other stock market price index series. However, the second sharply decreasing on 135th data has been shown up on the all series. All series seem to be affected by global crisis.

Testing for unit root in stock market series under structural break

Before performing cointegration test, the data should be controlled by unit root tests. ADF is the mostly used unit root test in literature. Unfortunately it does not take into account any structural break. The results of ADF test in Table 1 show that for the natural logarithm of each stock market index turns out to be stationary in first differences therefore is integrated of order 1, I(1).

Residual plots have examined for checking assumptions of white noise and normality of each series. It is realized that normality assumption has been violated for all the series. A structural break may cause normality problem even if ADF conclude absence of a unit root, so ADF conclusion can be biased. Therefore, Zivot-Andrews unit root test in the presence of a structural break should be applied to the series.

Variables	ADF	p-value
$ abla^1$ lnBIST	-14.827	0.000
∇^1 lnFTSE	-14.428	0.000
$ abla^1$ lnNASDAQ	-16.467	0.000
$ abla^1$ lnNIKKEI	-14.677	0.000
$ abla^1$ lnHSI	-14.728	0.000

Table 1. ADF Unit Root test results

While allowing for structural break/breaks in the series Zivot-Andrews' unit root test with unknown timing of the structural break is performed. The results of the Zivot-Andrews test in Table 2 suggest two clear breaks in two distinct time points. Thus, one can conclude that all series are I(0) stationary with at least one clear break. According to Zivot-Andrews test results, $\nabla^1 lnBIST$ and $\nabla^1 lnNIKKEI$ are stationary with one structural break on 41th and 47th data, respectively. Besides, $\nabla^1 lnFTSE$, $\nabla^1 lnNASDAQ$ and $\nabla^1 lnHSI$ are stationary with one structural break on 134th, 135th and 136th data, respectively. The structural break in BIST data is less likely linked to eartquake and tsunami, whereas there is logical connection on 47th data. Because NIKKEI stock market had closed as soon as earthquake had occurred.

Table 2. Zivot-Andrews Unit Root test results

Variables	Zivot -Andrews	p-value	Potential Break	Break date
∇^1 lnBIST	-3.688	0.010	41	08.03.2011
∇^1 lnFTSE	-6.345	0.000	134	03.08.2011
∇^1 lnNASDAQ	-5.190	0.002	135	04.08.2011
$ abla^1$ lnNIKKEI	-3.931	0.031	47	16.03.2011
∇^1 lnHSI	-4.713	0.000	136	05.08.2011

Interesting point is that HSI has been affecting by global crisis not Japan's tsunami. 134th, 135th and 136th are logical points for global crisis of USA core, because the rule of "purchase expectations, realities sold" is valid in stock market. "The market's already been shaken out," said Harvey Neiman, a portfolio manager of the Neiman Large Cap Value Fund. "It knew it was coming." (*http://www.thejournal.ie*).

According to these discussions, downgrade of US credit was not surprise, so these points may show the prior flag of expectations.

Diagnosting Checking of Cointegration Under Structural Break

To implement the Engle-Granger method on stock markets data set, we begin by regressing one series on each other and then assess the model fit. If variables cointegrate, the resulting OLS regression yields a "super-consistent" estimator of the cointegrating parameters. By this we mean that there is a very strong relationship between the estimated parameters. For example, taking the ∇^1 lnBIST series as the dependent variable and the other ∇^1 lnFTSE series as the independent variable, we yield the following regression equation:

$$\ln BIST = -0.42 + 1.32 \ln FTSE$$
 (9)

There is no unit root on the residuals of (9) model –ADF test statistics=-2.126 p-value=0.03. The p-values of the independent variables are very small; this means that these regression coefficients are statistically significant at the 0.05 level of significance.

When lnFTSE and lnBIST series are taken to be the dependent and independent variables, respectively, the following regression equations are obtained: To determine if the variables actually cointegrate, we test whether the residuals from the regression relationship(s) are stationary.

$$\ln FTSE = 2.80 + 0.53 \ln BIST$$
 (10)

There is no unit root on the residuals of (10) model –ADF test statistics=-2.64 p-value=0.09. As a result of these models in Table 3, FTSE and BIST are cointegrated each other, so do FTSE and NASDAQ series. Test represents the NASDAQ series have cointegrated with all other series. Generally the test was able to capture a total of six cointegration relationship. Symmetry is expected in the cointegration system, but in this example it does not exist. Hence, we conclude that EG cointegration test is not suitable for the data set.

One of the most commonly used cointegration tests in presence of a structural break is Gregory – Hansen (GH). This test is similar to the Engle – Granger cointegration method. The breaks are tried to be determined by adding dummy variables to the model apart from the Engle - Granger method. Gregory – Hansen test investigates the determination of structural breaks in long term relation under three different models: Level Shift (C), Level Shift with trend (C/T) and Regime Shift (C/S). α and τ is including to the model for determination of the structural break.

The results of GH test are shown in Table 4. According to GH results, the test captures the cointegration at different models. For example NASDAQ is cointegrated just in level shift model, but the break point shows up on 174th point instead of 135th point. It means even if there exist little bit shifting on the NASDAQ and BIST modeling residuals, there is another dominant point and that is the 174th point. BIST index does not have any cointegration relationship with other index series. However, in level shift with trend model, break has been detected in 46th observation in cointegration with NIKKEI series. But this break is not statistically significant. All the same time in all three models, FTSE and NASDAQ series are cointegrated along with structural break in 132th observation. Similarly, FTSE and NIKKEI are cointegrated along with structural break in 163th observations. For all models NASDAQ and FTSE are cointegrated along with structural breaks at 132th observation. Similarly same series have relation with HSI index.

NDEPENDEN	T VARIA	ABLES													
	LnBIST		LnFTS	LnFTSE			LnNASDAQ			LnNIKKEI			LnHSI		
	βo	β1	ADF	β₀	βı	ADF	β₀	β1	ADF	β。	β1	ADF	βo	β1	ADF
nBIST -value	_			-0.42 <i>0.00</i>	1.32 0.00	-2.13 <i>0.03</i>	0.78 <i>0.49</i>	1.32 0.00	-1.87 <i>0.05</i>	3.04 0.00	0.87 0.00	-2.50 0.12	3.93 0.00	0.71 0.00	-2.279 0.18
nFTSE -value	2.80 0.00	0.53 0.00	-2.64 0.09				0.64 0.25	1.20 0.00	-2.18 0.03	3.20 0.00	0.60 0.00	-1.68 0.44	4.04 0.00	0.46 0.00	-2.690 0.08
nNASDAQ -value	5.51 0.00	0.20 <i>0.00</i>	-3.89 0.00	3.78 0. <i>00</i>	0.46 <i>0.00</i>	-3.83 <i>0.00</i>		_	_	5.77 0.00	0.22 0.00	-3.75 0.00	6.26 <i>0.00</i>	0.15 <i>0.00</i>	-3.710 0.00
nNIKKEI -value	1.67 0.00	0.68 0.00	-2.27 0.19	-0.86 0.06	1.16 0.00	-1.39 0.15	0.67 0.51	1.10 0.00	-1.26 0.19	-		-	2.75 0.00	0.64 0.00	-2.466 0.13
nHSI -value	-1.39 0.00	1.03 0.00	-2.17 0.22	-4.47 0.00	1.67 0.00	-1.98 0.30	-0.90 0.53	1.40 0.00	-1.13 0.23	-0.98 .02	1.20 0.00	-2.21 0.20	_		

Table 3. Cointegration test results of all stock market price indexes.

INDEPEN	DENT VARIA	BLES										
	InBIST			InFTSE InNAS		InNASD	nNASDAQ		InNIKKEI			
Critical Values	MODEL	break point	GH-ADF statistics	break point	GH-ADF statistics	break point	GH-ADF statistics	break point	GH-ADF statistics	break point	GH-ADF statistics	
-4.69	С			183	-3.625	174	-3.940	43	-3.402	156	-3.699	
-5.03	C/T			50	-3.435	45	-4.086	46	-3.694	156	-3.567	
-5.23	C/S			157	-3.796	174 -3.945 95 -3.601 149 - 132 -6.640 132 -4.489 163 -	-4.029					
-4.69	С	132	-4.144			132	-6.640	132	-4.489	163	-5.783	
-5.03	C/T	132	-4			132	-6.778	105	-5.008	160	-5.894	
-5.23	C/S	162	-4.666			132	-6.873	133	-4.844	163	-5.802	
-4.69	С	174	-4.886	123	-6.376			172	-5.185	156	-5.635	
-5.03	C/T	174	-4.845	123	-6.620			97	-6.176	156 156	-6.154	
-5.23	C/S	177	-4.895	123	-6.384			172	Ak nt GH-ADF statistics break point -3.402 156 -3.694 156 -3.601 149 -4.489 163 -5.008 160 -4.844 163 -5.185 156 -6.176 156 -5.319 156 -4.84 46 -3.626 -4.269	-5.698		
-4.69	С	46	-4.036	177	-4.104	172	-4.362			46	-3.828	
-5.03	C/T	134	-3.923	105	-5.327	97	-5.848			46	-3.850	
-5.23	C/S	46	-4.188	55	-4.061	160	-4.396			43	-4.033	
-4.69	С	154	-4.357	160	-6.026	156	-5.224	138	-3.626			
-5.03	C/T	138	-4.556	160	-6.192	156	-5.840	138	-4.269			
-5.23	C/S	162	-5.439	160	-5.999	156	-5.140	104	-3.717			
	Critical Values -4.69 -5.03 -5.23 -4.69 -5.03 -5.23 -4.69 -5.03 -5.23 -4.69 -5.03 -5.23 -4.69 -5.03 -5.23 -4.69 -5.03 -5.23 -4.69 -5.03 -5.23	InBIST Critical Values MODEL -4.69 C -5.03 C/T -5.23 C/S -4.69 C -5.03 C/T	Critical Values MODEL break point -4.69 C - -5.03 C/T - -5.23 C/S - -4.69 C 132 -5.23 C/T 132 -5.03 C/T 132 -5.03 C/T 132 -5.23 C/S 162 -4.69 C 174 -5.03 C/T 174 -5.23 C/S 177 -4.69 C 46 -5.03 C/T 134 -5.23 C/S 46 -5.03 C/S 46 -5.03 C/S 46 -4.69 C 154 -5.03 C/T 138	InBIST Critical Values MODEL break point GH-ADF statistics -4.69 C - -5.03 C/T - -5.23 C/S - -4.69 C 132 -4.69 C 132 -4.69 C 132 -4.69 C 132 -5.03 C/T 132 -5.03 C/T 132 -5.23 C/S 162 -4.69 C 174 -5.03 C/T 174 -5.23 C/S 177 -4.886 -5.03 C/T -5.23 C/S 177 -4.69 C 46 -5.03 C/T 134 -5.23 C/S 46 -5.23 C/S 46 -4.69 C 154 -4.69 C 154 -4.69 C 154 -4.69 </td <td>InBIST InFTSE Critical Values MODEL break point GH-ADF statistics break point -4.69 C 183 -5.03 C/T 50 -5.23 C/S 157 -4.69 C 132 -4.144 -5.03 C/T 132 -4 -5.03 C/T 132 -4 -5.03 C/T 132 -4 -5.03 C/T 132 -4 -5.03 C/T 174 -4.886 123 -5.03 C/T 174 -4.845 123 -5.03 C/T 174 -4.895 123 -5.03 C/T 134 -3.923 105 -5.23 C/S 46 -4.188 55 -4.69 C 154 -4.357 160 -5.23 C/S 46 -4.357 160 -5.03 C/T 138 -4.556 160 </td> <td>InBIST InFTSE Critical Values MODEL break point GH-ADF statistics break point GH-ADF statistics -4.69 C 183 -3.625 -5.03 C/T 50 -3.435 -5.23 C/S 157 -3.796 -4.69 C 132 -4.144 - -5.03 C/T 132 -4 - - -5.03 C/T 174 -4.866 123 -6.376 -5.03 C/T 174 -4.845 123 -6.620 -5.23 C/S 177 -4.895 123 -6.384 -4.69 C 46 -4.036 177 -4.104 -5.03 C/T 134 -3.923 105 -5.327 -5.23 C/S</td> <td>InBIST InFTSE InNASD Critical Values MODEL break point GH-ADF statistics break point GH-ADF statistics break point GH-ADF statistics break point -4.69 C 183 -3.625 174 -5.03 C/T 50 -3.435 45 -5.23 C/S 157 -3.796 174 -4.69 C 132 -4.144 132 -5.03 C/T 132 -4 132 -5.03 C/T 174 -4.886 123 -6.376 -5.03 C/T 174 -4.895 123 -6.384 -4.69 C 46 -4.036 177 -4.104 172 -5.03 C/T 134 -3.923 105 -5.327 97 <td>InBIST InFTSE InNASDAQ Critical Values MODEL break point GH-ADF statistics break point GH-ADF statistics break point GH-ADF statistics -4.69 C 183 -3.625 174 -3.940 -5.03 C/T 183 -3.625 174 -3.940 -5.03 C/T 157 -3.796 174 -3.945 -4.69 C 132 -4.144 132 -6.640 -5.03 C/T 132 -4 132 -6.778 -5.23 C/S 162 -4.666 132 -6.873 -4.69 C 174 -4.886 123 -6.620 124 -5.03 C/T 174 -4.895 123 -6.620 124 -5.03 C/T 174 -4.895 123 -6.384 124 -4.69 C 46 -4.036 177 -4.104 172 -4.362 -5.03 C/T</td><td>$\begin{array}{ c c c c c c c c c c c c c c c c c c c$</td><td>$\begin{array}{ c c c c c c c c c c c c c c c c c c c$</td><td>$\begin{array}{ c c c c c c c c c c c c c c c c c c c$</td></td>	InBIST InFTSE Critical Values MODEL break point GH-ADF statistics break point -4.69 C 183 -5.03 C/T 50 -5.23 C/S 157 -4.69 C 132 -4.144 -5.03 C/T 132 -4 -5.03 C/T 132 -4 -5.03 C/T 132 -4 -5.03 C/T 132 -4 -5.03 C/T 174 -4.886 123 -5.03 C/T 174 -4.845 123 -5.03 C/T 174 -4.895 123 -5.03 C/T 134 -3.923 105 -5.23 C/S 46 -4.188 55 -4.69 C 154 -4.357 160 -5.23 C/S 46 -4.357 160 -5.03 C/T 138 -4.556 160	InBIST InFTSE Critical Values MODEL break point GH-ADF statistics break point GH-ADF statistics -4.69 C 183 -3.625 -5.03 C/T 50 -3.435 -5.23 C/S 157 -3.796 -4.69 C 132 -4.144 - 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.Table 4. Gregory-Hansen cointegration results

Investigation of relation between NASDAQ and NIKKEI series, for C and C/S models break points indicated 172th observation and for C/T model, 97th observation is the break point. When GH test examination results for the NIKKEI stock market, it can be said that for C/T model, cointegration relation with FTSE and NASDAQ series along with break points at 105th and 97th observations respectively. 46th observation is reflected tsunami disaster date, test could find out this break point. Although the break point has determined correctly by test, this is not affected by cointegration relationship between series. When the considering China's HSI index, it can be expressed that series cointegrated with BIST along with break point 162th observation for C/S model. Moreover, HSI is cointegrated with FTSE and NASDAQ series along with the break point 160th and 156th observation respectively for all models.

In the overall, GH test is able to capture more cointegrated relationship than the EG test. The nature of mutually relations in the series caused the variation of the break point location is the reason of the good performance of GH test. While FTSE and NASDAQ series are affected by break point 132th observation, NIKKEI series is affected from 46th observation. For HSI index, break points are significant which is located around 160th observations.

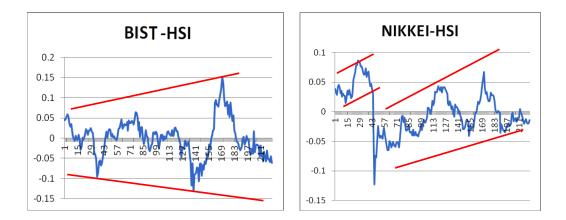
Engle's ARCH Test Result

We check out the cointegration relationship, this implies that residuals are stationary after modeling cointegration system. Even if residuals are stationary, we believe in some dominant points may cause the wrong decision (i.e. actually there is no cointegrated relationship, but the test could conclude that there exist cointegration, we can call it Type I and Type II error). Having a structural break may mask on the data, so we must suspect about the correct decision on GH cointegration test. That's why we check them out for heteroscedasticity of residuals which produced by regression models.

This section will examine volatility in series with Robert Engle ARCH test. Hypotheses are:

- H_o: Residuals do not have volatility. There is a not ARCH effect
- H_a: Residuals have volatility. There is an ARCH effect

Previous analyses show that series have structural breaks and this component affects cointegration test's performance. Besides the break, the series were investigated whether they have heteroscedasticity.



Some residual plots of series are shown as follows:

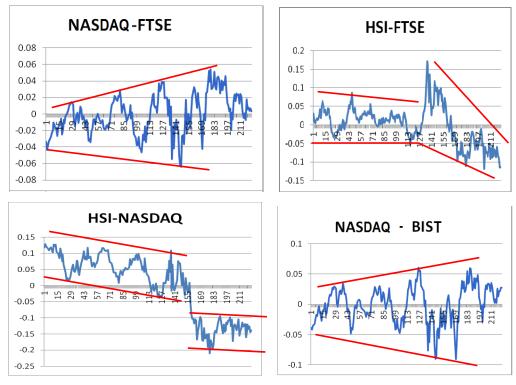


Figure 2. Time series graph of residuals produced by regression models

Table 5.	Engle's	ARCH	test result
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		InBIST	InFTSE	lnNASDAQ	InNIKKEI	lnHSI
	Hypothesis		NO ARCH EFFECT	NO ARCH EFFECT	ARCH EFFECT	ARCH EFFECT
InBIST	Test Stat.		0.061	0.733	24.55	15.565
	<i>p</i> -value		0.805	0.391	0.000	0.000
	Hypothesis	ARCH EFFECT		ARCH EFFECT	ARCH EFFECT	ARCH EFFECT
InFTSE	Test Stat.	8.403		47.88	19.33	17.22
	<i>p</i> -value	0.003		0.000	0.000	0.000
	Hypothesis	ARCH EFFECT	ARCH EFFECT		ARCH EFFECT	ARCH EFFECT
InNASDAQ	Test Stat.	10.49	17.77]	8.403	14.702
	<i>p</i> -value	0.001	0.000]	0.003	0.000
	Hypothesis	ARCH EFFECT	ARCH EFFECT	ARCH EFFECT		ARCH EFFECT
InNIKKEI	Test Stat.	44.17	42.02	71.94]	63.47
	<i>p</i> -value	0.000	0.000	0.000]	0.000
	Hypothesis	ARCH EFFECT	ARCH EFFECT	ARCH EFFECT	ARCH EFFECT	
lnHSI	Test Stat.	21.73	30.98	50.77	66.78	
	p-value	0.000	0.000	0.000	0.000	

Critical value: 3.841

					-		connegiu		our						
	LnBIST	Г		LnFTS	Е		LnNAS	SDAQ		LnNIK	KEI		LnHSI		
	break	break LMCOINT		break	LMCOINT		break	LMCOINT		break	LMCOI	NT	break		
MODEL	point	statistics	5	point	statistic	statistics		statistics		point	statistic	statistics		LMCOINT statistics	
		t	¢		t	ф		t	¢		t	¢		t	ф
C/T				136	-1.614	-7.146	161	-1.932	-10.228	136	-2.526	-15.032	155	-2.308	-15.069
C/S	1			136	-1.681	-7.670	161	-1.956	-10.359	134	-2.148	-14.459	136	-2.395	-12.788
C/T	171	-2.249	-9.285				135	-2.283	-19.680	135	-1.809	-12.465	134	-2.211	-16.703
C/S	135	-1.836	-8.956	-			135	-2.283	-19.681	134	-1.883	-12.656	135	-2.287	-18.051
C/T	171	-3.367	-24.207	150	-4.017	-46.850				144	-3.279	-26.842	144	-3.260	-23.898
C/S	92	-3.088	-25.656	150	-3.970	-45.640	-			144	-3.298	-25.914	144	-3.080	-22.792
C/T	45	-2.438	-12.912	45	-2.140	-12.425	45	-2.140	-12.425		1		45	-2.593	-15.939
C/S	45	-2.454	-13.154	45	-2.038	-11.118	45	-2.038	-11.118	1			45	-2.589	-15.968
C/T	178	-2.072	-10.300	137	-3.044	-22.244	137	-2.929	-19.797	45	-2.558	-12.447			_
C/S	178	-2.074	-10.377	137	-3.059	-22.424	137	-2.952	-20.106	108	-2.615	-12.662			
		1									1	1			

Table 6. LM Cointegration test result

b10 critical values for t and **\phi** -2.75 and -15.00, respectively

This test searches residuals which are obtain regression equation of stock exchange indexes. Having the volatile residuals means that series are under ARCH effects. As a result of analyses, all residuals have an ARCH effect except residuals of BIST-FTSE and BIST-NASDAQ models. Being under the ARCH effects has also affected unit root and cointegration tests performance according to the literature (Guidi, 2012).

Lagrange Multiplier Cointegration Test

LM test suggested by Westerlund J. ve Edgerton D.L. in 2006. This test take into account both structural breaks and volatility of the series. Former analyses demonstrated that the stock market index have all of two components. Therefore LM cointegration test has applied and results are shown in the Table 6.

According to the result BIST and NIKKEI series are cointegrated for level shift model with trend (C/T) along with break point 136th observation. Similarly BIST has a relationship with HSI index. When results are examined, FTSE series were cointegrated with NASDAQ along with break point 135th observation. In addition to FTSE series are cointegrated NIKKEI and HSI index. NASDAQ series has cointegrated relationship all index along with the different break points such as 171th observation for BIST in C/T model and 92th observation in regime shift model (C/S). In addition, it is cointegrated with FTSE for C/T and C/S model in same break point. The NIKKEI and the HSI in the series were cointegrated along with the break point 144th.

HSI series are cointegrated FTSE and NASDAQ series along with the breakpoint 137 for C/T and C/S models. When the results are analyzed, Nikkei series has a structural break point in 45th observation for all models. But this break is not statistically significant.

V. Conclusion

In this paper, we have investigated the impact of structural break and volatile innovations on contegration tests via Turkey, Japan, United Kingdom, United States of America and China stock market indexes for period 2011.By applying the unit root test we find that all stock prices are nonstationary. According to Zivot Andrews test series two clear breaks in two distinct time points for all stock markets. Result of unit root tests all series are I(0) stationary with at least one clear break. Overview the cointegration test results, conventional EG test failed to catch cointegrated relations because of structural break. The results of GH test indicates that there are many long-run relationship between the stock markets with break.captures the cointegration at different models.GH and LM test have some outcome under some condition. The nature of mutually relations in the series caused the variation of the break point location is the reason of the good performance of GH test. While FTSE and NASDAQ series are affected by break point 132th observation, NIKKEI series is affected from 46th observation. EG and GH cointegration analyses show that series have structural breaks and this component affects cointegration test's performance. Besides the break, the series were investigated whether they have heteroscedasticity. We have performed Engle's ARCH test to obtain heteroscedasticity of error term. Test results show that most of the residuals series are under ARCH effects. Then we investigate the long-run relationship between the series by Westerlun&Edgerton LM test. This test takes into account heteroskedastic and serially correlated errors. when we compare GH and LM test results. LM test more accurate than GH test with regard to estimate of break date. Furthermore LM test determined more cointegration relationship according to GH test. Consequently if the heteroscedasticity is more prominent than structural breaks, LM test result should be preferred.

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