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Abstract

The interdependence of international asset pricing is one of the key topics of modern finance. In this research domain, a specific category is composed of studies focusing on the interrelations between relatively thin or incompletely developed stock markets and the leading capital markets of the world. In this paper we study the Granger causality between a representative set of global asset returns and a thin Nordic financial market within a recursive framework. The database is initially subjected to factor analysis, where three global factors are identified: a European, an American and an Asian. Next, Granger causality between these factors and the Finnish stock market is studied. The results reveal that - in most cases - the global factors are relevant and have an incremental impact on the Finnish stock return. The timeliness hypothesis is accepted for the American factor but not for the Asian and European. The global factors show interesting interrelations in the rolling Granger framework.

Keywords: Rolling Granger causality, global asset returns.

1. Introduction

The interdependence of international asset pricing is one of the key topics of modern finance (cf., e.g., Masih and Masih, 1997, Blackman et al., 1994, Malkamäki et al., 1991). Empirical evidence indicates an increasing comovement of international capital markets (cf., e.g., Malliaris and Urrutia, 1992, Bos et al., 1995). In this research domain, a specific category is composed of studies focusing on the interrelationships between relatively thin or incompletely developed stock markets and the leading capital markets of the world. For example, Masih and Masih [1997] examined the patterns of dynamic linkages among national stock prices of four newly industrializing Asian countries - Taiwan, South Korea, Singapore, and Hong Kong - in models incorporating the established markets of Japan, USA, UK, and Germany. Another example is the increased attention paid to the interrelationships between the Scandinavian financial markets and the leading economies of the world (cf. e.g., Martikainen et al. [1994] and Mathur and Subrahmanyam [1991, Malkamäki [1993])]. For example, Malkamäki et al. [1993] investigated the lead-lag structures and causality patterns of the Scandinavian stock markets relative to world wide returns. They found that the Swedish market was the leader of the four, while the other three appeared to have no significant influence on the other markets. The presence of a US influence on Finnish stock returns was documented in Östermark and Aoki [1995] and of a Japanese impact in Östermark [1998]. Multivariate causality in consecutive subperiods was examined using the same database as in Östermark [1997]. It should be noted that, as evidenced by Martikainen et al. [1991], the relationship between the Finnish and US stock markets seems to be weaker than between the Finnish and the Swedish stock markets.

The present study is a continuation of the empirical research on the relationship between the Finnish and global asset returns. We will focus on the Granger causality between the Finnish market return and some factors representing global asset returns. The research issue is twofold. Firstly, three core hypotheses describing the features of the global return generating mechanism are developed and tested, using the well-known Granger framework. Secondly, the dynamics of Granger causality is analyzed within the Rolling Granger framework introduced by Smith et al. [1993]. This framework has been extensively applied by Östermark and Aaltonen [1998] and Aaltonen and Östermark [1997].

The study is organized as follows. In the next chapter, the methodology of the paper is presented: the Granger causality framework, a set of relevant unit root tests and the diagnostic tests applied to check the statistical properties of the residuals are discussed. In chapter 3 we present the global returns database and in chapter 4 the test setting. The empirical tests and the statistical properties of the residuals are examined in chapter 5. Chapter 6 concludes.

2. Methodology

2.1. Univariate Granger-causality

A time series $\langle x_t \rangle$ causes another time series $\langle y_t \rangle$ in the Granger sense if present y can be predicted better by using past values of x than by not doing so, considering also other relevant information, including past values of y. Formally, y is caused by x, if (cf., e. g. Pehkonen, 1991)

$$\sigma^{2}(y_{t}|y) > \sigma^{2}(y_{t}|y,x)$$
(2.1)
where $y = \{y_{t-1}, y_{t-2}, ..., y_{t-r}\}$
 $x = \{x_{t-1}, x_{t-2}, ..., x_{t-s}\}$

 $\sigma^2(y_t|y)$ and $\sigma^2(y_t|y,x)$ represent the minimum predictive error variance of y_t obtained by regressing y_t respectively on y and (y, x).

In mathematical terms, x is said to cause y, provided some β_j is nonzero in the full regression equation (2.2a):

$$y_t = \delta_0 + \sum_{i=1}^r \alpha_i y_{t-i} + \sum_{j=1}^s \beta_j x_{t-j} + \varepsilon_t.$$
(2.2a)

The relevance of x is indicated when comparing the error in (2.2a) to that of the reduced equation

$$y_t = \delta_0 + \sum_{i=1}^r \alpha_i y_{t-i} + \varepsilon$$
(2.2b)

The error terms are compared formally in the following *F*-statistic:

$$F = \frac{(SSE_r - SSE_f)/s}{SSE_f/(T - r - s - 1)}$$
(2.3)

where SSE_r, SSE_f = residual sum of squares of the reduced (2.2b) and full (2.2a) models respectively T = total number of observations r = number of lags for the y-variable s = number of lags for the x-variable

F has an asymptotic F-distribution with s and T-r-s-1 degrees of freedom.

In this study, a rolling Granger causality test is employed. With this method, possible changes

(structural breaks) in the causality pattern can be detected. Initially, Granger causality is estimated for the first 100 observations in the database, i.e., with a window length of 100. Then, the first observation is dropped from the sample and a new one is added to the end, after which the relationship is reestimated. The procedure is repeated throughout the sample. A similar procedure was applied in Smith et al. [1993] to test for the causality between four major equity markets.

2.1.1. Determining the Optimal Lag Structure

Studies by Guilkey and Salemi [1982], Geweke [1984], and Kang [1985] have indicated that Granger causality tests are sensitive - often critically so - to the choice of lag length. (cf. Thornton and Batten [1985], Jones [1989], and Kang [1989]).

The optimal lag-length is here defined by a two-step procedure based on minimizing Akaike's [1969] final prediction error (FPE) criterion, as suggested by Hsiao [1981]. We begin by determining the optimal lag of *Y*. We estimate the auto-regression equation (2.2b) with r = 1,...,5 and compute the sum of squared residuals for each regression. The optimal lag is selected by minimizing FPE, defined as

FPE (r) =
$$\frac{T + r + 1}{T - r - 1} * \frac{SSR_r}{T}$$
 (2.4)

Let r^* denote the lag that minimizes FPE(r). The optimal lag for the exogenous variable x is then determined by running the bivariate regressions (2.2a) with the lag for y fixed at r^* and lags s = 1,...,5 for x, and calculating the FPEs for each lag as follows:

FPE
$$(r^*, s) = \frac{T + r^* + s + 1}{T - r^* - s - 1}$$
 (2.5)

The optimal lag structure (r^*, s^*) is determined separately for each iteration in the rolling Granger regressions.

2.1.2. Testing for Stationarity and Co-integration

Since the return series are defined as first differences of the natural logarithms, we first test for stationarity of the logarithmic index series using the Augmented Dickey-Fuller (ADF) -statistic. The presence of a unit root in a time series indicates non-stationarity (cf. Engle and Granger [1987]). The test statistic is based on the following ADF-regression:

$$\Delta y_t = \delta + \alpha y_{t-1} + \sum_{i=1}^p \beta_i \Delta y_{t-i} + \varepsilon_t, \qquad (2.6)$$

where Δ is the difference operator, the residual term ε_t is assumed to be Gaussian white noise and the number of lagged terms *p* is chosen such that the errors are uncorrelated (cf. Dickey and Fuller, 1979, SHAZAM, 1993. The critical values for the *t*-statistic are tabulated in Fuller, 1976). The null hypothesis H₀ is that $\alpha = 0$, i.e., there exists a unit root and the time series is nonstationary. If the logarithmic index is non-stationary, but the first difference of the series, i.e. returns are stationary, the prices are said to be integrated of order one, denoted *I*(1). We test for stationarity of the total data series (752 observations) using three different tests: A standard augmented Dickey-Fuller (ADF)-test, the Phillips-Perron [1988] test, and a refined DF-GLS^T test due to Elliott et al. [1992].

Phillips and Perron [1988] suggested some non-parametric corrections to the conventional test statistics which eliminate the nuisance parameter dependencies asymptotically. (For the exact formula, see Phillips and Perron, 1988.) The Dickey-Fuller critical values can still be directly used.

The third stationarity test is a modified Dickey-Fuller test called DF-GLS test due to Elliott et al. [1992]. The DF-GLS^{τ} test that allows for a linear time trend applies the ADF-regression (2.6)

$$\Delta y_t^{\tau} = a_0 \ y_{t-1}^{\tau} + \sum_{j=1}^p a_j \Delta y_{t-j}^{\tau} + u_t$$
(2.7)

where y_t^{τ} , is the locally detrended data process given by

$$y_t^{\tau} = y_t - z_t \,\widetilde{\beta} \,. \tag{2.8}$$

In (2.8), $z_t = (1,t)$ and $\tilde{\beta}$ is the regression coefficient of \tilde{y} on \tilde{z} , for which

$$\widetilde{y}_t = (y_1, (l - \overline{\alpha}L) y_2, ..., (l - \overline{\alpha}L) y_T)'$$
 and
 $\widetilde{z}_t = (z_1, (l - \overline{\alpha}L) z_2, ..., (l - \overline{\alpha}L) z_T)'.$

Elliott et al. [1992] analyzed a sequence of Neyman-Pearson tests of the unit root null hypothesis against the local alternative $\overline{\alpha} = 1 + \overline{c} / T$. The DF-GLS^{τ} test statistic is given by the conventional *t*-statistic, with the null ($a_0 = 0$) tested against the alternative ($a_0 < 0$). Elliott et al. [1992] recommend that the parameter \overline{c} be set equal to -13.5. A detailed description of the DF-GLS test is presented in Cheung and Lau [1995].

2.1.3. Diagnostic Testing of the Residuals

We test for the statistical properties of the residuals in each full regression model (2.2a) using five tests (cf. Hendry and Doornik [1996]).

1. Error Autocorrelation Test Test equation: $\varepsilon_t = \beta_0 + \sum_{i=1}^{k-1} \beta_i x_i + \sum_{i=1}^p \alpha_i \varepsilon_{t-i} + e_t$ H₀: $\alpha_1 = ... = \alpha_p = 0$ vs. H₁: At least one $\alpha_i = 0, i = 1,...,p$. $F_c = \frac{R^2 / p}{(1 - R^2) / (n - k - p)} - F_{p,n-k-p}$ Hypotheses: Statistic: 2. ARCH-test Test equation: $\varepsilon_t^2 = \alpha_0 + \sum_{i=1}^p \alpha_i \varepsilon_{t-i}^2 + e_t$ *H*₀: $\alpha_1 = ... = \alpha_p = 0$ vs. *H*₁: At least one $\alpha_i = 0, i = 1,...,p$. $F_c = \frac{R^2 / p}{(1 - R^2) / (n - k - p - 1)} - F_{p,n-k-p-1}$ Hypotheses: Statistic: 3. Normality test $\chi^2 = n * \left(\frac{S^2}{6} + \frac{(K - 3)^2}{24} \right) - \chi_2^2$

where n = sample size, S = skewness, K = kurtosis

4. Heteroskedasticity

Statistic:

Test equation: $u_t = \beta_0 + \sum_{i=1}^{k-1} \beta_i x_i + \sum_{i=1}^p \alpha_i u_{t-i} + \varepsilon_t$ *H*₀: $\alpha_1 = ... = \alpha_p = 0$ vs. *H*₁: At least one $\alpha_i = 0, i = 1,...,p$. $F_c = \frac{R^2 / p}{(1 - R^2) / (n - k - p)} - F_{p,n-k-p}$ Hypotheses: Statistic:

5. The RESET-test

The RESET test (Regression Specification test) due to Ramsey [1969] tests the null of correct model specification against the alternative that powers of the endogenous variable (actually its predictions) have been omitted from the model.

3. Data

Our database consists of daily market indices of 15 stock exchanges all around the world between February 1, 1994 and December 31, 1996 (752 observations). The statistical properties of the return series are presented in tables (3.1) - (3.2).

Most of the return series are non-normal, leptocurtic, autocorrelated, and heteroskedastic. At the 1% level normality is accepted for Paris and Stockholm. London, Paris, Sidney, Tokyo and Oslo exhibit no autocorrelation in their market returns. Furthermore, Paris and Sidney are homoskedastic.

The three unit root tests support the nonstationarity hypothesis for most exchanges. When including a trend, nonstationarity is rejected for more than half of the exchanges by the Dickey-Fuller or Phillips-Perron test in at least one of the null hypotheses. The evidence is contradictory because none of the tests indicates stationarity simultaneously for any exhange index. In fact, the DF-GLS^{τ} test indicates nonstationarity for the index of all exchanges at the 10% level. We therefore interpret the results as an indication of unit roots in all index series. The return series are unequivocally *I*(0).

Table 3.1: Descriptive statistics for the return series										
Index	Country	Mean	Standard Deviation	Excess Kurtosis	Skewness	$K\text{-}S~Z^{1)}$	Kiefer-Salmon ²⁾			
Helsinki FOX	FIN	0.0414	1.2787	3.0709	-0.3893	1.366**	314.489***			
Standard & Poors 500	USA	0.0578	0.6275	2.1094	-0.4457	1.908***	164.318***			
Frankfurt Dax	GER	0.0375	0.8865	1.9170	-0.5354	1.675***				
Hong Kong Hang Seng	HKG	0.0171	1.3230	3.2695	-0.3723	2.332***	352.319***			
London Financial Times 100	GBR	0.0223	0.6876	0.4758	-0.2934	1.688***				
Mexico IPC Stock	MEX	0.0249	1.8055	3.7909	0.1894	2.228***				
Paris CAC 40	FRA	-0.0009	0.9944	0.3630	-0.0128	1.284*	4.149			
Singapore Straits	SIN	-0.0077	0.8845	3.3404	-0.2349	1.664***	356.54***			
Sydney Australian All Ordinary	AUS	0.0054	0.7598	2.4380	-0.0065	1.491**	186.244***			
Tokyo Nikkei	JPN	-0.0071	1.1049	3.1011	0.0983	1.551***	302.538***			
Toronto Stock Exchange 300	CAN	0.0340	0.5724	2.7965	-0.6175	2.103***	292.835***			
Zürich Swiss	SWI	0.0308	0.6983	2.5180	-0.3131	1.948***	210.957***			
Stockholm General Index	SWE	0.0551	0.8429	0.3512	-0.1264	0.544	5.866*			
Oslo General Index	NOR	0.0466	0.7584	3.7367	-0.0623	1.542***	437.999***			
Copenhagen General Index	DEN	0.0197	0.7190	1.2979	-0.2966	1.502**	63.812***			
Critical values: 10%						1.2238	4.605			
5%					i	1.3581	5.991			
1%						1.5174	9.210			
		LB6 ³⁾	LB12 ³⁾	QLB6 ⁴⁾	QLB12 ⁴⁾	LM ⁵⁾				
Helsinki FOX	FIN	13.846**	33.988***	104.114***	195.247***	47.258***				
Standard & Poors 500	USA	13.410**	15.830	30.310***	35.272***	0.674				
Frankfurt Dax	GER	12.725**	21.098**	25.492***	32.710***	8.934***				
Hong Kong Hang Seng	HKG	10.786*	14.415	37.183***	62.609***	4.636**				
London Financial Times 100	GBR	8.635	15.002	26.885***	56.014***	0.309				
Mexico IPC Stock	MEX	14.862**	22.611**	100.299***	133.869***	51.591***				
Paris CAC 40	FRA	2.844	10.624	8.800	17.269	0.604				
Singapore Straits	SIN	17.153***	18.620*	55.544***	72.723***	21.834***				
Sydney Australian All Ordinary	AUS	6.154	13.565	3.801	4.703	0.175				
Tokyo Nikkei	JPN	5.961	14.996	20.704***	34.533***	9.988***				
Toronto Stock Exchange 300	CAN	30.561***	36.321***	34.562***	37.999***	5.747**				
Zürich Swiss	SWI	14.858**	30.097***	59.791***	69.304***	3.767*				
Stockholm General Index	SWE	14.107**	28.377***	15.857**	25.905**	4.667**				
Oslo General Index	NOR	9.778	17.167	58.034***	61.784***	44.472***				
Copenhagen General Index	DEN	14.107**	28.377***	20.606***	25.532**	0.702				
Critical values: 10%		10.645	18.549	10.645	18.549	2.706				
5%		12.592	21.026	12.592	21.026	3.841				
1%		16.812	26.217	16.812	26.217	6.635				

Table 3.1. Descriptive statistics for the return series

1) One-sample Kolmogorov-Smirnov test for normality (cf. Neave [1981], pp. 26-27)

2) Kiefer-Salmon test for normality: KS = $(T/6)^*$ sk² + $(T/24)^*$ ku² χ^2_2 , T = number of observations (752) 3) Ljung-Box-Pierce test for autocorrelations: LB[J] = T(T + 2) $\sum_{j=1}^{J} \frac{1}{T - j} r_j^2 \sim \chi^2_J$

4) Ljung-Box-Pierce test for autocorrelations on squared series

5) Lagrange Multiplier -test for ARCH(1) ~ χ^{2}_{1}

Ι.	(1) Undifferenced variables (Logarithmic indexes)												
		Withou	ut trend				With	trend			DF-GLS ^T		
	<i>H</i> ₀ : 0	$\alpha = 0$	<i>H</i> ₀ : δ =	$= \alpha = 0$	<i>H</i> ₀ : 0	$\alpha = 0$	Н₀: б=с	$\lambda = 0$	<i>H</i> ₀ : α =	$=\lambda=0$			
	`	value at	`	value at		value at	`	value at	`	value at	(critical va	alue at $\alpha = 1$	10%: 1.96)
	$\alpha = 10\%$	6: -2.57)	$\alpha = 10\%$	6: 3.78)	$\alpha = 10\%$	6: -3.13)	$\alpha = 10\%$	6: 4.03)	$\alpha = 10\%$	6: 5.34)			
	Dickey-	Phillips-	Dickey-	Phillips-	Dickey-	Phillips-	Dickey-	Phillips-	Dickey-	Phillips-	<i>p</i> = 1	<i>p</i> = 2	<i>p</i> = 3
	Fuller	Perron	Fuller	Perron	Fuller	Perron	Fuller	Perron	Fuller	Perron			
FIN	-1.2518	-0.9454	1.2555	0.8130	-2.4579	-2.3777	2.4275	2.3633	3.1668	3.1964	-1.2964	-1.3419	-1.3327
USA	0.8446	0.7515	4.6961	3.8615	-3.0588	-3.3670	6.9022	6.9932	5.9519	6.9762	-0.0263	-0.0725	0.0081
GER	0.0887	0.2975	0.8665	0.9297	-1.9579	-2.0437	2.5055	2.8074	2.8900	3.3803	-0.5601	-0.5836	-0.5995
HKG	-0.1374	-0.8725	0.2693	0.4343	-2.5525	-3.1711	3.9602	5.1632	5.6766	7.6837	0.0828	-0.0019	-0.0087
GBR	0.1054	0.0712	0.6241	0.4504	-3.8744	-3.9315	6.4560	6.8594	9.0509	9.8074	0.1951	0.1247	0.1762
MEX	-1.1766	-1.0611	0.7483	0.6263	-2.3081	-2.2124	2.1858	2.1009	3.2222	3.0877	-0.3961	-0.3822	-0.3402
FRA	-2.0996	-2.0724	2.2045	2.1477	-2.3609	-2.3322	3.1619	3.1615	4.7426	4.7419	-0.2853	-0.3159	-0.3104
SIN	-1.9559	-2.7040	1.9131	3.6812	-1.9539	-2.6747	1.2740	2.4567	1.9106	3.6609	-1.5122	-1.4763	-1.4050
AUS	-0.4661	-1.1700	0.3065	0.7044	-2.4291	-3.1079	3.0712	4.5186	4.4068	6.7551	-0.0177	0.0005	-0.0279
JPN	-1.4293	-1.6145	1.0246	1.3179	-1.5153	-1.7293	0.7762	1.0333	1.1612	1.5354	-0.7633	-0.8163	-0.7872
CAN	0.6617	0.9639	1.1340	1.5762	-2.5405	-2.4518	4.3464	4.8092	5.5927	6.0849	-0.1746	-0.1658	-0.2020
SWI	0.7582	0.4279	1.6217	0.810	-3.0400	-3.3133	6.0036	6.1522	7.6461	8.4356	0.4059	0.3741	0.4271
SWE	1.0085	0.9515	2.3312	1.9760	-2.5237	-2.7802	4.6660	5.0199	5.1551	6.0486	-0.2785	-0.3274	-0.3114
NOR	0.4888	0.8292	1.1251	1.6429	-2.8627	-2.5582	4.5030	4.5543	5.7347	5.5208	-0.1650	-0.2056	-0.2232
DEN	0.0498	0.2277	0.2035	0.2787	-2.2893	-2.1321	4.6346	4.6682	6.7462	6.7427	0.0135	0.0223	0.0225
		**== * *	0.2000								•		
						ferenced va					1		
			at trend				ariables (R					DF-GLS ^t	
	H ₀ : c	Without $\alpha = 0$	it trend	$= \alpha = 0$	(2) Diff <i>H</i> ₀ : c	ferenced vanishing $\alpha = 0$	ariables (R With H ₀ : δ=e	eturn serie trend $\chi = \lambda = 0$	s) <i>H</i> ₀ : α =	$=\lambda=0$	(oritical v	$DF-GLS^{\tau}$ $H_0: \alpha_0 = 0$	·
	H ₀ : o	Without $\alpha = 0$ value at	ut trend H ₀ : δ = (critical	$= \alpha = 0$ value at	(2) Diff H ₀ : (critical	ferenced van $\alpha = 0$ value at	ariables (R With H ₀ : δ=α (critical	eturn serie trend $\chi = \lambda = 0$ value at	s) H ₀ : α = (critical	$= \lambda = 0$ value at	(critical va	$DF-GLS^{\tau}$ $H_0: \alpha_0 = 0$	
	H_0 : α (critical $\alpha = 10\%$	Without $\alpha = 0$ value at $(6: -2.57)$	it trend $H_0: \delta =$ (critical $\alpha = 109$	$= \alpha = 0$ value at %: 3.78)	(2) Diff H_0 : α (critical $\alpha = 10\%$	ferenced van $\alpha = 0$ value at 6: -3.13)	ariables (R With $H_0: \delta = 0$ (critical $\alpha = 109$	eturn serie trend $\alpha = \lambda = 0$ value at 6: 4.03	s) $H_0: \alpha =$ (critical $\alpha = 10\%$	$= \lambda = 0$ value at (5.34)	Ì	$DF-GLS^{\tau}$ $H_0: \alpha_0 = 0$ alue at $\alpha = 1$	10%: 1.96)
	H_0 : α (critical $\alpha = 10\%$) Dickey-	Without $\alpha = 0$ value at $6: -2.57$) Phillips-	It trend $H_0: \delta =$ (critical $\alpha = 10\%$ Dickey-	$= \alpha = 0$ value at $\therefore 3.78$) Phillips-	(2) Diff H_0 : α (critical $\alpha = 10\%$ Dickey-	ferenced value at $\alpha = 0$ value at 6: -3.13) Phillips-	ariables (R With $H_0: \delta = 0$ (critical $\alpha = 10$? Dickey-	eturn serie trend $\chi = \lambda = 0$ value at 4.03 Phillips-	s) $H_0: \alpha =$ (critical $\alpha = 10\%$ Dickey-	$= \lambda = 0$ value at 6: 5.34) Phillips-	(critical va $p = 1$	$DF-GLS^{\tau}$ $H_0: \alpha_0 = 0$	·
	H_0 : c (critical $\alpha = 10\%$ Dickey- Fuller	Without $\alpha = 0$ value at 6: -2.57) Phillips- Perron	at trend $H_0: \delta =$ (critical $\alpha = 10\%$ Dickey- Fuller	= α = 0 value at %: 3.78) Phillips- Perron	(2) Diff H_0 : α (critical $\alpha = 10\%$ Dickey- Fuller	ferenced value at $\alpha = 0$ value at (6: -3.13) Phillips- Perron	ariables (R With $H_0: \delta = 0$ (critical $\alpha = 109$ Dickey- Fuller	eturn serie trend $\chi=\lambda=0$ value at \therefore 4.03) Phillips- Perron	s) $H_0: \alpha =$ (critical $\alpha = 10^9$ Dickey- Fuller	= λ = 0 value at %: 5.34) Phillips- Perron	<i>p</i> = 1	$DF-GLS^{T}$ $H_{0}: \alpha_{0} = 0$ alue at $\alpha = 1$ $p = 2$	10%: 1.96) p = 3
FIN	H_0 : α (critical $\alpha = 10\%$) Dickey- Fuller -4.0999	Without $x = 0$ value at 6: -2.57) Phillips- Perron -26.006	It trend $H_0: \delta =$ (critical $\alpha = 10^9$ Dickey- Fuller 8.4087	= α = 0 value at %: 3.78) Phillips- Perron 338.150	(2) Diff H_0 : c (critical $\alpha = 10\%$ Dickey- Fuller -4.1487	ferenced value at $\alpha = 0$ value at 6: -3.13) Phillips- Perron -26.012	ariables (R With $H_0: \delta = \epsilon$ (critical $\alpha = 10^9$ Dickey- Fuller 5.7413	eturn serie trend $\chi=\lambda=0$ value at $\langle : 4.03 \rangle$ Phillips- Perron 225.500	s) $H_0: \alpha =$ (critical $\alpha = 10^9$ Dickey- Fuller 8.6079	$= \lambda = 0$ value at (5.34) Phillips- Perron 338.260	<i>p</i> = 1 -10.723	DF-GLS ^{τ} H ₀ : $\alpha_0 = 0$ alue at $\alpha = 1$ p = 2 -8.175	10%: 1.96) p = 3 -6.692
FIN USA	H_0 : α (critical $\alpha = 10\%$ Dickey- Fuller -4.0999 -6.4186	Without $\alpha = 0$ value at 6: -2.57) Phillips- Perron -26.006 -25.110	at trend $H_0: \delta =$ (critical $\alpha = 10\%$ Dickey- Fuller 8.4087 20.6000	= α = 0 value at %: 3.78) Phillips- Perron 338.150 315.140	(2) Diff $H_0: \alpha$ (critical $\alpha = 10\%$ Dickey- Fuller -4.1487 -6.5710	ferenced value at $\alpha = 0$ value at δ : -3.13) Phillips- Perron -26.012 -25.208	ariables (R With $H_0: \delta = 0$ (critical $\alpha = 10\%$ Dickey- Fuller 5.7413 14.4280	eturn serie trend $\alpha = \lambda = 0$ value at \therefore 4.03) Phillips- Perron 225.500 211.670	s) $H_0: \alpha =$ (critical $\alpha = 10\%$ Dickey- Fuller 8.6079 21.6420	 = λ = 0 value at 6: 5.34) Phillips- Perron 338.260 317.490 	<i>p</i> = 1 -10.723 -10.680	DF-GLS ^T $H_0: \alpha_0 = 0$ alue at $\alpha = 1$ p = 2 -8.175 -8.811	10%: 1.96) p = 3 -6.692 -7.142
FIN USA GER	H_0 : c (critical $\alpha = 10\%$ Dickey- Fuller -4.0999 -6.4186 -5.7716	Withou x = 0 value at 6: -2.57) Phillips- Perron -26.006 -25.110 -29.704	It trend $H_0: \delta =$ (critical $\alpha = 10^9$ Dickey- Fuller 8.4087 20.6000 16.6880	$= \alpha = 0$ value at 6: 3.78) Phillips- Perron 338.150 315.140 441.110	(2) Diff $H_0: \alpha$ (critical $\alpha = 10\%$ Dickey- Fuller -4.1487 -6.5710 -5.8556	x = 0 value at 6: -3.13) Phillips- Perron -26.012 -25.208 -29.895	ariables (R With $H_0: \delta = 0$ (critical $\alpha = 10^9$ Dickey- Fuller 5.7413 14.4280 11.4870	eturn serie trend $\chi=\lambda=0$ value at (4.03) Phillips- Perron 225.500 211.670 297.810	s) $H_{0:} \alpha =$ (critical $\alpha = 109$ Dickey- Fuller 8.6079 21.6420 17.1980	$= \lambda = 0$ value at (5.34) Phillips- Perron 338.260 317.490 446.720	<i>p</i> = 1 -10.723 -10.680 -12.807	DF-GLS ^T $H_0: \alpha_0 = 0$ alue at $\alpha = 1$ p = 2 -8.175 -8.811 -9.411	10%: 1.96) <i>p</i> = 3 -6.692 -7.142 -7.270
FIN USA GER HKG	H_0 : (critical $\alpha = 10\%$ Dickey- Fuller -4.0999 -6.4186 -5.7716 -6.2907	Without $\alpha = 0$ value at $6: -2.57$) Phillips- Perron -26.006 -25.110 -29.704 -26.732	at trend $H_0: \delta =$ (critical $\alpha = 10\%$ Dickey- Fuller 8.4087 20.6000 16.6880 19.8430	$= \alpha = 0$ value at 6: 3.78) Phillips- Perron 338.150 315.140 441.110 357.330	(2) Diff $H_0: c$ (critical $\alpha = 10\%$ Dickey- Fuller -4.1487 -6.5710 -5.8556 -6.5270	Ferenced value at $\alpha = 0$ value at 6: -3.13) Phillips- Perron -26.012 -25.208 -29.895 -26.869	ariables (R With $H_0: \delta = 0$ (critical $\alpha = 10^{\circ}$ Dickey- Fuller 5.7413 14.4280 11.4870 14.3090	eturn serie trend $\chi = \lambda = 0$ value at 6: 4.03 Phillips- Perron 225.500 211.670 297.810 240.610	s) $H_0: \alpha =$ (critical $\alpha = 10^9$ Dickey- Fuller 8.6079 21.6420 17.1980 21.4070	$= \lambda = 0$ value at 6: 5.34) Phillips- Perron 338.260 317.490 446.720 360.920	<i>p</i> = 1 -10.723 -10.680 -12.807 -17.268	DF-GLS ^T $H_0: \alpha_0 = 0$ alue at $\alpha = 1$ p = 2 -8.175 -8.811 -9.411 -14.496	10%: 1.96) <i>p</i> = 3 -6.692 -7.142 -7.270 -13.700
FIN USA GER HKG GBR	H_0 : c (critical $\alpha = 10\%$ Dickey- Fuller -4.0999 -6.4186 -5.7716 -6.2907 -7.3386	Withou x = 0 value at 6: -2.57) Phillips- Perron -26.006 -25.110 -29.704 -26.732 -27.410	at trend $H_0: \delta =$ (critical $\alpha = 109$ Dickey- Fuller 8.4087 20.6000 16.6880 19.8430 26.9470	$= \alpha = 0$ value at (: 3.78) Phillips- Perron 338.150 315.140 441.110 357.330 375.610	(2) Diff $H_0: c$ (critical $\alpha = 10\%$ Dickey- Fuller -4.1487 -6.5710 -5.8556 -6.5270 -7.5510	α = 0 value at 6: -3.13) Phillips- Perron -26.012 -25.208 -29.895 -26.869 -27.595	ariables (R With $H_0: \delta = 0$ (critical $\alpha = 109$ Dickey- Fuller 5.7413 14.4280 11.4870 14.3090 19.0190	eturn serie trend $\chi=\lambda=0$ value at $\langle : 4.03 \rangle$ Phillips- Perron 225.500 211.670 297.810 240.610 253.740	s) $H_0: \alpha =$ (critical $\alpha = 109$ Dickey- Fuller 8.6079 21.6420 17.1980 21.4070 28.5090	$= \lambda = 0$ value at (5.34) Phillips- Perron 338.260 317.490 446.720 360.920 380.600	<i>p</i> = 1 -10.723 -10.680 -12.807 -17.268 -4.558	DF-GLS ^T $H_0: \alpha_0 = 0$ alue at $\alpha = 1$ p = 2 -8.175 -8.811 -9.411 -14.496 -3.476	10%: 1.96) p = 3 -6.692 -7.142 -7.270 -13.700 -2.635
FIN USA GER HKG GBR MEX	H_0 : c (critical $\alpha = 10\%$ Dickey- Fuller -4.0999 -6.4186 -5.7716 -6.2907 -7.3386 -5.3057	Without $\alpha = 0$ value at 6: -2.57) Phillips- Perron -26.006 -25.110 -29.704 -26.732 -27.410 -24.078	at trend $H_0: \delta =$ (critical $\alpha = 10^9$ Dickey- Fuller 8.4087 20.6000 16.6880 19.8430 26.9470 14.0950	$= \alpha = 0$ value at 6: 3.78) Phillips- Perron 338.150 315.140 441.110 357.330 375.610 289.880	(2) Diff $H_0: \alpha$ (critical $\alpha = 10\%$ Dickey- Fuller -4.1487 -6.5710 -5.8556 -6.5270 -7.5510 -5.3306	Ferenced value at $\alpha = 0$ value at 6: -3.13) Phillips- Perron -26.012 -25.208 -29.895 -26.869 -27.595 -24.068	ariables (R With $H_0: \delta = \epsilon$ (critical $\alpha = 10^{\circ}$ Dickey- Fuller 5.7413 14.4280 11.4870 14.3090 19.0190 9.5094	eturn serie trend $\alpha = \lambda = 0$ value at 6: 4.03) Phillips- Perron 225.500 211.670 297.810 240.610 253.740 193.080	s) $H_0: \alpha =$ (critical $\alpha = 10^9$ Dickey- Fuller 8.6079 21.6420 17.1980 21.4070 28.5090 14.2450	$= \lambda = 0$ value at 6: 5.34) Phillips- Perron 338.260 317.490 446.720 360.920 380.600 289.620	p = 1 -10.723 -10.680 -12.807 -17.268 -4.558 -18.186	DF-GLS ^T $H_0: \alpha_0 = 0$ alue at $\alpha = 1$ p = 2 -8.175 -8.811 -9.411 -14.496 -3.476 -15.708	10%: 1.96) p = 3 -6.692 -7.142 -7.270 -13.700 -2.635 -13.198
FIN USA GER HKG GBR MEX FRA	$H_0: c$ (critical $\alpha = 10\%$ Dickey- Fuller -4.0999 -6.4186 -5.7716 -6.2907 -7.3386 -5.3057 -5.8666	Withou x = 0 value at 6: -2.57) Phillips- Perron -26.006 -25.110 -29.704 -26.732 -27.410 -24.078 -28.345	at trend $H_0: \delta =$ (critical $\alpha = 109$ Dickey- Fuller 8.4087 20.6000 16.6880 19.8430 26.9470 14.0950 17.2330	$= \alpha = 0$ value at $\langle : 3.78 \rangle$ Phillips- Perron 338.150 315.140 441.110 357.330 375.610 289.880 401.670	(2) Diff $H_0: c$ (critical $\alpha = 10\%$ Dickey- Fuller -4.1487 -6.5710 -5.8556 -6.5270 -7.5510 -5.3306 -6.0989	x = 0 value at δ: -3.13) Phillips- Perron -26.012 -25.208 -29.895 -26.869 -27.595 -24.068 -28.604	ariables (R With H_0 : $\delta = 0$ (critical $\alpha = 109$ Dickey- Fuller 5.7413 14.4280 11.4870 14.3090 19.0190 9.5094 12.4210	eturn serie trend $\chi=\lambda=0$ value at $\langle:4.03\rangle$ Phillips- Perron 225.500 211.670 297.810 240.610 253.740 193.080 272.650	s) $H_0: \alpha =$ (critical $\alpha = 109$ Dickey- Fuller 8.6079 21.6420 17.1980 21.4070 28.5090 14.2450 18.6070	$=\lambda = 0$ value at $\langle 5.34 \rangle$ Phillips- Perron 338.260 317.490 446.720 360.920 380.600 289.620 408.980	p = 1 -10.723 -10.680 -12.807 -17.268 -4.558 -18.186 -6.497	DF-GLS ^T $H_0: \alpha_0 = 0$ alue at $\alpha = 1$ p = 2 -8.175 -8.811 -9.411 -14.496 -3.476 -15.708 -4.744	10%: 1.96) p = 3 -6.692 -7.142 -7.270 -13.700 -2.635 -13.198 -3.658
FIN USA GER HKG GBR MEX FRA SIN	H_0 : c (critical $\alpha = 10\%$ Dickey- Fuller -4.0999 -6.4186 -5.7716 -6.2907 -7.3386 -5.3057 -5.8666 -6.3385	Withor x = 0 value at 6: -2.57) Phillips- Perron -26.006 -25.110 -29.704 -26.732 -27.410 -24.078 -28.345 -23.683	at trend $H_0: \delta =$ (critical $\alpha = 10^9$ Dickey- Fuller 8.4087 20.6000 16.6880 19.8430 26.9470 14.0950 17.2330 20.1040	$= \alpha = 0$ value at 6: 3.78) Phillips- Perron 338.150 315.140 441.110 357.330 375.610 289.880 401.670 280.350	(2) Diff $H_0: c$ (critical $\alpha = 10\%$ Dickey- Fuller -4.1487 -6.5710 -5.8556 -6.5270 -7.5510 -5.3306 -6.0989 -6.3349	Ferenced value at $\alpha = 0$ value at 6: -3.13) Phillips- Perron -26.012 -25.208 -29.895 -26.869 -27.595 -24.068 -28.604 -23.669	ariables (R With H_0 : $\delta = 0$ (critical $\alpha = 10^{\circ}$ Dickey- Fuller 5.7413 14.4280 11.4870 14.3090 19.0190 9.5094 12.4210 13.3980	eturn serie trend $\chi=\lambda=0$ value at 6:4.03) Phillips- Perron 225.500 211.670 297.810 240.610 253.740 193.080 272.650 186.640	s) $H_0: \alpha =$ (critical $\alpha = 10^9$ Dickey- Fuller 8.6079 21.6420 17.1980 21.4070 28.5090 14.2450 18.6070 20.0820	$= \lambda = 0$ value at 6: 5.34) Phillips- Perron 338.260 317.490 446.720 360.920 380.600 289.620 408.980 279.960	p = 1 -10.723 -10.680 -12.807 -17.268 -4.558 -18.186 -6.497 -14.920	DF-GLS ^T $H_0: \alpha_0 = 0$ alue at $\alpha = 1$ p = 2 -8.175 -8.811 -9.411 -14.496 -3.476 -15.708 -4.744 -12.271	10%: 1.96) p = 3 -6.692 -7.142 -7.270 -13.700 -2.635 -13.198 -3.658 -10.338
FIN USA GER HKG GBR MEX FRA SIN AUS	H_0 : c (critical $\alpha = 10\%$ Dickey- Fuller -4.0999 -6.4186 -5.7716 -6.2907 -7.3386 -5.3057 -5.8666 -6.3385 -5.9895	Withou $\alpha = 0$ value at6: -2.57)Phillips-Perron-26.006-25.110-29.704-26.732-27.410-24.078-28.345-23.683-26.884	at trend $H_0: \delta =$ (critical $\alpha = 10^9$ Dickey- Fuller 8.4087 20.6000 16.6880 19.8430 26.9470 14.0950 17.2330 20.1040 18.0020	$= \alpha = 0$ value at 6: 3.78) Phillips- Perron 338.150 315.140 441.110 357.330 375.610 289.880 401.670 280.350 361.270	(2) Diff $H_0: o$ (critical $\alpha = 10\%$ Dickey- Fuller -4.1487 -6.5710 -5.8556 -6.5270 -7.5510 -5.3306 -6.0989 -6.3349 -6.1292	Ferenced value at $\alpha = 0$ value at 6: -3.13) Phillips- Perron -26.012 -25.208 -29.895 -26.869 -27.595 -24.068 -28.604 -23.669 -27.090	ariables (R With H_0 : $\delta = \epsilon$ (critical $\alpha = 10^{\circ}$ Dickey- Fuller 5.7413 14.4280 11.4870 14.3090 19.0190 9.5094 12.4210 13.3980 12.6120	eturn serie trend $\alpha = \lambda = 0$ value at 6: 4.03) Phillips- Perron 225.500 211.670 297.810 240.610 253.740 193.080 272.650 186.640 244.480	s) $H_0: \alpha =$ (critical $\alpha = 10^9$ Dickey- Fuller 8.6079 21.6420 17.1980 21.4070 28.5090 14.2450 18.6070 20.0820 18.8520	$= \lambda = 0$ value at 6: 5.34) Phillips- Perron 338.260 317.490 446.720 360.920 380.600 289.620 408.980 279.960 366.720	p = 1 -10.723 -10.680 -12.807 -17.268 -4.558 -18.186 -6.497 -14.920 -15.124	DF-GLS ^T $H_0: \alpha_0 = 0$ alue at $\alpha = 1$ p = 2 -8.175 -8.811 -9.411 -14.496 -3.476 -15.708 -4.744 -12.271 -11.277	10%: 1.96) p = 3 -6.692 -7.142 -7.270 -13.700 -2.635 -13.198 -3.658 -10.338 -9.455
FIN USA GER HKG GBR MEX FRA SIN AUS JPN	$H_0: c$ (critical $\alpha = 10\%$ Dickey- Fuller -4.0999 -6.4186 -5.7716 -6.2907 -7.3386 -5.3057 -5.8666 -6.3385 -5.9895 -4.8172	Withou x = 0 value at 6: -2.57) Phillips- Perron -26.006 -25.110 -29.704 -26.732 -27.410 -24.078 -28.345 -23.683 -26.884 -27.392	at trend $H_0: \delta =$ (critical $\alpha = 10^9$ Dickey- Fuller 8.4087 20.6000 16.6880 19.8430 26.9470 14.0950 17.2330 20.1040 18.0020 11.6280	$= \alpha = 0$ value at 6: 3.78) Phillips- Perron 338.150 315.140 441.110 357.330 375.610 289.880 401.670 280.350 361.270 375.140	(2) Diff $H_0: c$ (critical $\alpha = 10\%$ Dickey- Fuller -4.1487 -6.5710 -5.8556 -6.5270 -7.5510 -5.3306 -6.0989 -6.3349 -6.1292 -4.8059	Ferenced value at $\alpha = 0$ value at 6: -3.13) Phillips- Perron -26.012 -25.208 -29.895 -26.869 -27.595 -24.068 -28.604 -23.669 -27.090 -27.376	ariables (R With H_0 : $\delta = 0$ (critical $\alpha = 109$ Dickey- Fuller 5.7413 14.4280 11.4870 14.3090 19.0190 9.5094 12.4210 13.3980 12.6120 7.7456	eturn serie trend $\chi=\lambda=0$ value at $\langle :4.03\rangle$ Phillips- Perron 225.500 211.670 297.810 240.610 253.740 193.080 272.650 186.640 244.480 249.800	s) $H_0: \alpha =$ (critical $\alpha = 109$ Dickey- Fuller 8.6079 21.6420 17.1980 21.4070 28.5090 14.2450 18.6070 20.0820 18.8520 11.5930	$=\lambda = 0$ value at 6: 5.34) Phillips- Perron 338.260 317.490 446.720 360.920 380.600 289.620 408.980 279.960 366.720 374.690	p = 1 -10.723 -10.680 -12.807 -17.268 -4.558 -18.186 -6.497 -14.920 -15.124 -10.028	DF-GLS ^{τ} $H_0: \alpha_0 = 0$ alue at $\alpha = 1$ p = 2 -8.175 -8.811 -9.411 -14.496 -3.476 -15.708 -4.744 -12.271 -11.277 -7.626	p = 3 -6.692 -7.142 -7.270 -13.700 -2.635 -13.198 -3.658 -10.338 -9.455 -6.006
FIN USA GER HKG GBR MEX FRA SIN AUS JPN CAN	H_0 : c (critical $\alpha = 10\%$ Dickey- Fuller -4.0999 -6.4186 -5.7716 -6.2907 -7.3386 -5.3057 -5.8666 -6.3385 -5.9895 -4.8172 -6.1324	Withon x = 0 value at 6: -2.57) Phillips- Perron -26.006 -25.110 -29.704 -26.732 -27.410 -24.078 -28.345 -23.683 -23.683 -26.884 -27.392 -22.486	at trend $H_0: \delta =$ (critical $\alpha = 10^9$ Dickey- Fuller 8.4087 20.6000 16.6880 19.8430 26.9470 14.0950 17.2330 20.1040 18.0020 11.6280 18.8070	$= \alpha = 0$ value at 6: 3.78) Phillips- Perron 338.150 315.140 441.110 357.330 375.610 289.880 401.670 280.350 361.270 375.140 252.820	(2) Diff $H_0: c$ (critical $\alpha = 10\%$ Dickey- Fuller -4.1487 -6.5710 -5.8556 -6.5270 -7.5510 -5.3306 -6.0989 -6.3349 -6.1292 -4.8059 -6.4555	x = 0 value at 6: -3.13) Phillips- Perron -26.012 -25.208 -29.895 -26.869 -27.595 -24.068 -23.669 -27.090 -27.376 -22.484	ariables (R With H_0 : $\delta = \epsilon$ (critical $\alpha = 10^{\circ}$ Dickey- Fuller 5.7413 14.4280 11.4870 14.3090 19.0190 9.5094 12.4210 13.3980 12.6120 7.7456 13.8950	eturn serie trend $\alpha = \lambda = 0$ value at 6: 4.03) Phillips- Perron 225.500 211.670 297.810 240.610 253.740 193.080 272.650 186.640 244.480 249.800 168.480	s) $H_0: \alpha =$ (critical $\alpha = 10^{9}$ Dickey- Fuller 8.6079 21.6420 17.1980 21.4070 28.5090 14.2450 18.6070 20.0820 18.8520 11.5930 20.8390	$= \lambda = 0$ value at 6: 5.34) Phillips- Perron 338.260 317.490 446.720 360.920 380.600 289.620 408.980 279.960 366.720 374.690 252.720	p = 1 -10.723 -10.680 -12.807 -17.268 -4.558 -18.186 -6.497 -14.920 -15.124 -10.028 -14.592	DF-GLS ^{τ} $H_0: \alpha_0 = 0$ alue at $\alpha = 1$ p = 2 -8.175 -8.811 -9.411 -14.496 -3.476 -15.708 -4.744 -12.271 -11.277 -7.626 -11.286	p = 3 -6.692 -7.142 -7.270 -13.700 -2.635 -13.198 -3.658 -10.338 -9.455 -6.006 -9.733
FIN USA GER HKG GBR MEX FRA SIN AUS JPN CAN SWI	H_0 : c (critical $\alpha = 10\%$ Dickey- Fuller -4.0999 -6.4186 -5.7716 -6.2907 -7.3386 -5.3057 -5.8666 -6.3385 -5.9895 -4.8172 -6.1324 -5.6761	Without $\alpha = 0$ value at $6: -2.57$) Phillips- Perron -26.006 -25.110 -29.704 -26.732 -27.410 -24.078 -23.683 -26.884 -27.392 -22.486 -25.967	at trend $H_0: \delta =$ (critical $\alpha = 10\%$ Dickey- Fuller 8.4087 20.6000 16.6880 19.8430 26.9470 14.0950 17.2330 20.1040 18.0020 11.6280 18.8070 16.1410	$= \alpha = 0$ value at 6: 3.78) Phillips- Perron 338.150 315.140 441.110 357.330 375.610 289.880 401.670 280.350 361.270 375.140 252.820 337.080	(2) Diff $H_0: c$ (critical $\alpha = 10\%$ Dickey- Fuller -4.1487 -6.5710 -5.8556 -6.5270 -7.5510 -5.3306 -6.0989 -6.1292 -4.8059 -6.4555 -6.0550	Ferenced value at $\alpha = 0$ value at 6: -3.13) Phillips- Perron -26.012 -25.208 -29.895 -26.869 -27.595 -24.068 -28.604 -23.669 -27.090 -27.376 -22.484 -26.276	ariables (R With H_0 : $\delta = 0$ (critical $\alpha = 109$ Dickey- Fuller 5.7413 14.4280 11.4870 14.3090 19.0190 9.5094 12.4210 13.3980 12.6120 7.7456 13.8950 12.2750	eturn serie trend $\chi=\lambda=0$ value at 6:4.03) Phillips- Perron 225.500 211.670 297.810 240.610 253.740 193.080 272.650 186.640 244.480 249.800 168.480 229.980	s) $H_0: \alpha =$ (critical $\alpha = 10^{9}$ Dickey- Fuller 8.6079 21.6420 17.1980 21.4070 28.5090 14.2450 18.6070 20.0820 18.8520 11.5930 20.8390 18.3800	$= \lambda = 0$ value at 6: 5.34) Phillips- Perron 338.260 317.490 446.720 360.920 380.600 289.620 408.980 279.960 366.720 366.720 374.690 252.720 344.970	p = 1 -10.723 -10.680 -12.807 -17.268 -4.558 -18.186 -6.497 -14.920 -15.124 -10.028 -14.592 -7.177	DF-GLS ^T $H_0: \alpha_0 = 0$ alue at $\alpha = 1$ p = 2 -8.175 -8.811 -9.411 -14.496 -3.476 -15.708 -4.744 -12.271 -11.277 -7.626 -5.496	p = 3 -6.692 -7.142 -7.270 -13.700 -2.635 -13.198 -3.658 -10.338 -9.455 -6.006 -9.733 -3.940
FIN USA GER HKG GBR MEX FRA SIN AUS JPN CAN SWI SWE	H_0 : c (critical $\alpha = 10\%$ Dickey- Fuller -4.0999 -6.4186 -5.7716 -6.2907 -7.3386 -5.3057 -5.8666 -6.3385 -5.9895 -4.8172 -6.1324 -5.5102	Withon x = 0 value at 6: -2.57) Phillips- Perron -26.006 -25.110 -29.704 -26.732 -27.410 -24.078 -23.683 -23.683 -26.884 -27.392 -22.486 -25.967 -24.917	at trend $H_0: \delta =$ (critical $\alpha = 10^9$ Dickey- Fuller 8.4087 20.6000 16.6880 19.8430 26.9470 14.0950 17.2330 20.1040 18.0020 11.6280 18.8070 16.1410 15.2010	$= \alpha = 0$ value at 6: 3.78) Phillips- Perron 338.150 315.140 441.110 357.330 375.610 289.880 401.670 280.350 361.270 375.140 252.820 337.080 310.390	(2) Diff $H_0: c$ (critical $\alpha = 10\%$ Dickey- Fuller -4.1487 -6.5710 -5.8556 -6.5270 -7.5510 -5.3306 -6.0989 -6.3349 -6.1292 -4.8059 -6.4555 -6.0550 -5.8728	Ferenced value at $\alpha = 0$ value at 6: -3.13) Phillips- Perron -26.012 -25.208 -29.895 -26.869 -27.595 -24.068 -23.669 -27.090 -27.376 -22.484 -26.276 -25.049	ariables (R With H_0 : $\delta = 0$ (critical $\alpha = 10^{\circ}$ Dickey- Fuller 5.7413 14.4280 11.4870 14.3090 19.0190 9.5094 12.4210 13.3980 12.6120 7.7456 13.8950 12.2750 11.5100	eturn serie trend $\chi=\lambda=0$ value at 6:4.03) Phillips- Perron 225.500 211.670 297.810 240.610 253.740 193.080 272.650 186.640 244.480 249.800 168.480 229.980 209.040	s) $H_0: \alpha =$ (critical $\alpha = 10^9$ Dickey- Fuller 8.6079 21.6420 17.1980 21.4070 28.5090 14.2450 18.6070 20.0820 18.8520 11.5930 20.8390 18.3800 17.2450	$= \lambda = 0$ value at 6: 5.34) Phillips- Perron 338.260 317.490 446.720 360.920 380.600 289.620 408.980 279.960 366.720 374.690 252.720 344.970 313.560	p = 1 -10.723 -10.680 -12.807 -17.268 -4.558 -18.186 -6.497 -14.920 -15.124 -10.028 -14.592 -7.177 -6.074	DF-GLS ^T $H_0: \alpha_0 = 0$ alue at $\alpha = 1$ p = 2 -8.175 -8.811 -9.411 -14.496 -3.476 -15.708 -4.744 -12.271 -11.277 -7.626 -11.286 -5.496 -4.648	p = 3 -6.692 -7.142 -7.270 -13.700 -2.635 -13.198 -3.658 -10.338 -9.455 -6.006 -9.733 -3.940 -3.603
FIN USA GER HKG GBR MEX FRA SIN AUS JPN CAN SWI	H_0 : c (critical $\alpha = 10\%$ Dickey- Fuller -4.0999 -6.4186 -5.7716 -6.2907 -7.3386 -5.3057 -5.8666 -6.3385 -5.9895 -4.8172 -6.1324 -5.6761	Without $\alpha = 0$ value at $6: -2.57$) Phillips- Perron -26.006 -25.110 -29.704 -26.732 -27.410 -24.078 -23.683 -26.884 -27.392 -22.486 -25.967	at trend $H_0: \delta =$ (critical $\alpha = 10\%$ Dickey- Fuller 8.4087 20.6000 16.6880 19.8430 26.9470 14.0950 17.2330 20.1040 18.0020 11.6280 18.8070 16.1410	$= \alpha = 0$ value at 6: 3.78) Phillips- Perron 338.150 315.140 441.110 357.330 375.610 289.880 401.670 280.350 361.270 375.140 252.820 337.080	(2) Diff $H_0: c$ (critical $\alpha = 10\%$ Dickey- Fuller -4.1487 -6.5710 -5.8556 -6.5270 -7.5510 -5.3306 -6.0989 -6.1292 -4.8059 -6.4555 -6.0550	Ferenced value at $\alpha = 0$ value at 6: -3.13) Phillips- Perron -26.012 -25.208 -29.895 -26.869 -27.595 -24.068 -28.604 -23.669 -27.090 -27.376 -22.484 -26.276	ariables (R With H_0 : $\delta = 0$ (critical $\alpha = 109$ Dickey- Fuller 5.7413 14.4280 11.4870 14.3090 19.0190 9.5094 12.4210 13.3980 12.6120 7.7456 13.8950 12.2750	eturn serie trend $\chi=\lambda=0$ value at 6:4.03) Phillips- Perron 225.500 211.670 297.810 240.610 253.740 193.080 272.650 186.640 244.480 249.800 168.480 229.980	s) $H_0: \alpha =$ (critical $\alpha = 10^{9}$ Dickey- Fuller 8.6079 21.6420 17.1980 21.4070 28.5090 14.2450 18.6070 20.0820 18.8520 11.5930 20.8390 18.3800	$= \lambda = 0$ value at 6: 5.34) Phillips- Perron 338.260 317.490 446.720 360.920 380.600 289.620 408.980 279.960 366.720 366.720 374.690 252.720 344.970	p = 1 -10.723 -10.680 -12.807 -17.268 -4.558 -18.186 -6.497 -14.920 -15.124 -10.028 -14.592 -7.177	DF-GLS ^T $H_0: \alpha_0 = 0$ alue at $\alpha = 1$ p = 2 -8.175 -8.811 -9.411 -14.496 -3.476 -15.708 -4.744 -12.271 -11.277 -7.626 -5.496	p = 3 -6.692 -7.142 -7.270 -13.700 -2.635 -13.198 -3.658 -10.338 -9.455 -6.006 -9.733 -3.940

Table 3.2: Unit root tests for the global returns database.

Having demonstrated the presence of unit roots in the indexes, we conducted pair-wise cointegration tests between the Finnish stock exchange and the major world markets. No pair-wise co-integration seems to be present in the data (cf. table (3.3)). The result implies that there is no need for an error correction term in the Granger regressions. Theoretically, co-integration is stronger than causality, i.e., the latter can prevail with or without the former. However, cointegration would automatically indicate the presence of causality.

	No trend	in cointegrating re	gression	Trend in cointegrating regression			
	Dickey-Fuller	Phillps-	-Perron	Dickey-Fuller	Phillps-Perron		
	t	Ζ	t	t	Z	t	
10%	-3.04	-17.1	-3.04	-3.5	-23.4	-3.5	
USA	-2.4880	-10.9210	-2.2396	-2.4849	-11.3080	-2.2887	
GER	-2.7075	-9.3254	-2.1390	-3.0952	-11.6320	-2.3743	
HKG	-2.5889	-10.8950	-2.4483	-2.6560	-12.1800	-2.4085	
GBR	-2.5484	-10.8970	-2.3094	-2.5295	-11.1360	-2.2784	
MEX	-2.0232	-7.6084	-1.8264	-2.9024	-12.2950	-2.3983	
FRA	-1.7335	-4.6294	-1.3940	-2.5926	-11.7860	-2.3858	
SIN	-1.3836	-4.0248	-1.1066	-2.6028	-13.4140	-2.5471	
AUS	-2.3364	-8.3548	-2.0830	-2.5313	-11.3020	-2.3021	
JPN	-1.3471	-3.2837	-0.9703	-2.4889	-12.4780	-2.4533	
CAN	-2.3210	-10.5130	-2.3107	-2.6648	-12.2770	-2.4419	
SWI	-2.4560	-9.6494	-2.1299	-2.5271	-11.5780	-2.3244	
SWE	-2.4642	-10.4370	-2.2900	-2.4561	-10.1570	-2.3170	
NOR	-2.4719	-9.9744	-2.2335	-2.4843	-10.1650	-2.2393	
DEN	-2.3533	-8.1466	-2.0983	-2.5497	-11.8060	-2.3821	

Table 3.3: Co-integration tests between Finland and the other 14 market indices

Next, the global database was subjected to principal components factor analysis. The VARIMAX-rotated factor loadings matrix computed on the return database is presented in table 3.4. The corresponding price series along with the Finnish stock market return are presented in figures (3.1) - (3.4). The factors clearly reflect three continental areas: Asia, Europe, and America. Daily stock returns are effectively governed by the trading activity within the continental time zones.

	1 Ioaungs ma		100ul uuluou	.
Country		FACTOR 1	FACTOR 2	FACTOR 3
SWI	Zürich	0.772	0.176	0.091
FRA	Paris	0.759	-0.015	0.201
SWE	Stockholm	0.755	0.162	0.149
GBR	London	0.739	0.031	0.301
GER	Frankfurt	0.719	0.357	0.012
NOR	Oslo	0.662	0.284	0.042
DEN	Copenhagen	0.647	0.295	-0.044
HKG	Hongkong	0.198	0.781	0.057
SIN	Singapore	0.073	0.744	0.091
AUS	Sydney	0.280	0.660	0.088
JPN	Tokyo	0.113	0.523	0.010
USA	S&P 500	0.176	-0.029	0.832
CAN	Toronto	0.237	0.131	0.781
MEX	Mexico	-0.006	0.088	0.634

 Table 3.4:
 Rotated factor loadings matrix for the global database

4. Test Setting

In this paper we study the impact of three geographical factors - the European, Asian, and American - on the Finnish market returns in a rolling framework. We expect a thin stock market to be continuously and significantly caused by the global market movements. The time zone differences between the three continents create a continuous information flow, where the stock markets of one continent open approximately the same time as the markets of another continent close. At each time point we have one active continent and two continents closed.

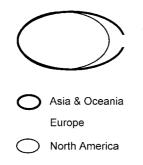
The Finnish market operates within the time zone of the European markets. On the daily level, the most recent European and American information available for the Finnish investors is that of the previous day. The major Asian markets, again close before or slightly after the Finnish market opens. Thus, even on the daily level the Asian market information is available to the Finnish investors most of the trading hours.

Due to the rotation of the earth and the geographical position of the continents, it seems natural to assume that the daily stock market operations within each continent has its specific impact on the global return generating forces. An interesting question is whether the activity within a continent is fully absorbed into the continent in the succeeding time zone. In particular, the following three aspects of information are addressed (cf. figures (4.1) and (4.2)): relevance, incremental importance and timeliness.

Figure 4.1: Relevance of the three continents w.r.t. the Finnish return generating mechanism.



Figure 4.2: Incremental importance and timeliness of the three continents w.r.t. the Finnish return generating mechanism.



We summarize the assumptions in the following research hypotheses:

- A. Relevance. All three geographical areas contain relevant information. This assumption is tested by bivariate regression models separately for each continent. A significant causality is expected in each test.
- B. Incrementality. New information is added to the total information set in each continent. This assumption is tested separately for each continent by controlling for the previous information set. A significant causality is expected in each test.
- C. Timeliness. The old information set is completely absorbed by the current one. This assumption is tested for each continent by controlling for the next information set. We expect no significant causality for the old information sets after controlling for the more recent one.

The hypotheses are operationalized in table 4.1.

Table 4.1:	Testing for relevance, incremental importance and timeliness of geographical
	areas w.r.t. the Finnish return generating mechanism

Hypothesis testing	Expected causality	Explanation
A Relevance		
E -> F	+	Europe is relevant in causing Finnish returns
Am -> F	+	America is relevant in causing Finnish returns
As -> F	+	Asia is relevant in causing Finnish returns
B Incremental information		
Am E -> F	+	America has incremental information after controlling for the European factor
As Am -> F	+	Asia has incremental information after controlling for the American factor
E As -> F	+	Europe has incremental information after controlling for the Asian factor
C Timeliness		
Am As -> F	~	The American information has been absorbed into the more recent Asian factor
E Am -> F	~	The European information has been absorbed into the more recent American factor
$As E \rightarrow F$	~	The Asian information has been absorbed into the more recent European factor

The following symbols are used in the table:

F = the Finnish market return

E = the European factor extracted from the global data base

As = the Asian-Oceanian factor extracted from the global data base

Am = the American factor extracted from the global data base

X -> Y

+ = significant causality

 \sim = nonexistent causality

5. Empirical Tests

The hypotheses A-C are first tested by running the Granger regressions (2.2a) and (2.2b) - augmented by control variables as implied by table 4.1. - once throughout the whole test period. The dynamics of the observed causalities is then further examined within the rolling framework of Smith et al [1993] using the window length of 100 observations.

5.1. The Relevance Hypothesis

The optimal lag length is determined by the FPE-criterion. As the modern technology allows rapid flow of information, we may expect a low-order optimal lag. Thus, a search interval from

one to five days for both the endogenous and exogenous variables was applied. For both the European and Asian factors, the optimal lag is one, whereas for the American one, two lags was found to be optimal. The global market information seems to be reflected in the Finnish market within at most two days. In consequence, lags between one to two days are applied in the subsequent tests. The results for the relevance tests are presented in table 5.1.

Hypothesis	Test variable	Test equation / [restriction]	F-value	Significance
E -> F	Europe, previous day	$F_t = \alpha F_{t-1} [+\beta E_{t-1}] + \varepsilon_t$	0.1231	0.7258
E -> F	Europe, current day	$F_t = \alpha F_{t-1} [+\beta E_t] + \varepsilon_t$	213.3251	0.0000***
Am -> F	America, previous day	$F_t = \alpha F_{t-1} [+\beta A m_{t-1}] + \varepsilon_t$	87.7082	0.0000***
Am -> F	America, lag 2	$F_t = \alpha F_{t-1} [+\beta Am_{t-2}] + \varepsilon_t$	7.5972	0.0060***
A s -> F	Asia, current day	$F_t = \alpha F_{t-1} [+\beta A_{s_t}] + \varepsilon_t$	27.0181	0.0000***
As -> F	Asia, previous day	$F_t = \alpha F_{t-1} [+\beta As_{t-1}] + \varepsilon_t$	8.8440	0.0000***

Table 5.1:	Relevance tests
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 E_t = The European factor return

 Am_t = The American factor return

 As_t = The Asian factor return

The relevance hypothesis for Europe is, somewhat surprisingly, rejected when using the previous day's observations (t-1). On the other hand, for contemponareous causality (t), a highly significant test statistic is achieved. The Finnish market clearly and immediately reflects general changes in the European market. Note, however, that the European exchanges close one to two hours after the Finnish market. Thus, the daily closing prices include some two hours' information that impossibly can be reflected in the Finnish market. On the other hand, the contemporaneous information emanating from the six hours during which all European markets are jointly operating obviously is relevant.

For the American and Asian markets, significant causality is detected for lags one and two. The evidence is relevant when testing the incrementality and timeliness. For the American factor, the previous day constitutes the latest possible information set with relevance for the Finnish returns.

5.2. The Incremental Information Hypothesis

In testing for the incremental information content of the three continents, the relevant causalities were re-estimated using the previous information set as a control variable. If a significant causality from a continent is maintained even after controlling for the previous information set (cf. figure 4.2), the continent adds some intrinsic value to the global set of information. In the opposite case, the continent may act as a filter, through which the information created otherwise is forwarded to the Finnish market. The test equations and results for the incrementality tests are presented in table 5.2.

1 auto 5.2.	10515 101		Jillation			
Hypothesis	Test variable	Control variable	Test equation / [1	restriction]	F-value	Significance
E As -> F	Europe, current day	Asia, current day	$F_t = \alpha F_{t-1} [+\beta]$	$[E_t] + \gamma As_t + \varepsilon_t$	224.2863	0.0000***
E As -> F	Europe, current day	Asia, previous day	$F_t = \alpha F_{t-1} [+\beta]$	$[E_t] + \gamma As_{t-1} +$	206.7510	0.0000***
Am E -> F	America, previous day	Europe, previous day	$F_t = \alpha F_{t-1} \left[+ \beta_1 \right]$	$[Am_{t-1}] + \gamma E_t$	88.3769	0.0000***
As Am -> F	Asia, previous day	America, lag 2	$F_t = \alpha F_{t-1} [+\beta_1]$	$As_{t-1} + \gamma Am$	4.2221	0.0402**
	The Finnish stock					

 Table 5.2:
 Tests for incremental information

 Am_t = The American factor return

 As_t = The Asian factor return

For the European factor, highly significant incremental information value was observed even after controlling for the latest Asian observation. As the previous day Asian factor was significant in the relevance test with respect to the Finnish returns, we replicated the test by using the significant previous day information from the Asian market as the control variable. Practically no impact on the European current day information was observed. The high significance of the European factor may partly be caused by the one to two hours lag between the Finnish and the other European markets.

For the American factor, highly significant incremental causality was observed even after controlling for the latest European information set. As expected, the huge American markets clearly add value to the total set of information reflected in the Finnish market. For the Asian factor, again, the significance observed for the previous day's information in testing for the relevance is weaker when controlling for the previous American information (lag 2). The impact of the American markets seems to be so strong that it is reflected to the Finnish markets both directly and indirectly through the Asian market. The relevance observed for the previous day's Asian information might in fact not be a pure Asian impact, but instead a reflection of the American influence filtered through the American markets.

5.3. The Timeliness Hypothesis

The timeliness hypothesis was tested using the next (newer) information set as a control variable. According to the timeliness hypothesis, the information from one continent should be completely absorbed by that from the next. Thus, no significant causality should exist after controlling for the more recent information set. The test results are presented in table 5.3.

The timeliness hypothesis is accepted for the American factor (E|Am / F). For the Asian and European factors, the timeliness hypothesis is rejected.

Hypothesis	Test variable	Control variable	Test equation / [restriction]	F-value	Significance
E Am -> F	Europe, previous day	America, previous day	$F_t = \alpha F_{t-1} \left[+ \beta E_{t-1} \right] + \gamma Am_t.$	0.8265	0.3636
Am∣As -> F	America, previous day	Asia, current day	$F_t = \alpha F_{t-1} \left[+ \beta A m_{t-1} \right] + \gamma A s_t$	61.7365	0.0000***
	America, lag 2	Asia, previous day	$F_t = \alpha F_{t-1} \left[+\beta Am_{t-2} \right] + \gamma As_t$	3.1408	0.0768*
As E -> F	Asia, previous day	Europe, previous day	$F_t = \alpha F_{t-1} \left[+ \beta As_{t-1} \right] + \gamma E_t -$	8.7097	0.0033***
$F_t =$	The Finnish stock 1	narket return			
	The European facto				
Am_t	= The American fac	ctor return			
$As_t =$	The Asian factor r	eturn			

Table 5.3: Tests for timeliness of the information

5.4. The Dynamics of the Causality Pattern

The rolling regressions for the hypotheses A-C are depicted in figures 5.1 - 5.3. In all figures, the time axis shows the ending point of the rolling regressions. For example, the *F*-value for February 1996 corresponds to the period of 100 market days (approximately 4 months) ending at February 1996.

The relative strength of the three hypotheses varies significantly over time. The relevance, incrementality and timeliness of the European information set is significant almost throughout the study period. The relevance, incrementality and timeliness of the Asian factor is notable in the beginning and in the middle of the study period. For the American and European factors, these features are notable most of the study period.

The dynamics of the relevance hypothesis is shown in figures 5.1a-f (cf. table 5.1). There is an interesting common pattern in the behaviour of the global return factors: in all three cases, a clear shift from old to new information is observed. However, also old (yesterday) news have a significant impact on the Finnish returns, especially in the beginning and towards the end of the study period. For the American factor, the shift from old to new information usage occurs somewhat later than for the European. The common shift towards more efficient information processing holds also for the Asian factor.

In figures 5.2a-f we show the dynamics of the incremental information hypotheses (cf. table 5.2). As expected, the significant contemporaneous causality from the European factor prevails over the whole test period. For the American factor, the incremental information pattern closely resembles the relevance pattern, i.e., when the latest American news has significant relevance in causing the Finnish returns, it also has significant incremental value over the older European impact. It is interesting to note that - for a large part of the test period - the European information for the previous day is strong enough to almost annihilate the incremental impact of the latest American information.

Figures 5.3a-f show the evolution of the timeliness hypotheses (cf. table 5.3). For all factors, the timeliness profiles resemble the relevance profiles closely. The timeliness hypothesis is accepted over a larger consecutive time span only for the European factor (figure 5.3f). Yet, in total the evidence suggests that the timeliness hypothesis is mostly rejected for all three factors. Since the intrinsic informational content of each factor persists even after controlling for the newer information, the stock pricing information is not completely absorbed between continents of different time zones. The trading activity within a continental time zone hence contains informational components that are orthogonal to those of the succeeding continent.

5.5. Testing the Statistical Quality of the Regression Models

In order to verify the reliability of the regression models, the key statistical tests presented previously were conducted on the residuals of each regression. The test results are summarized in table (5.4). The table shows that normality of the residuals is rejected in approximately 30-40% of the regressions. The other statistical tests show a considerably lower frequency of rejection. The RESET test, measuring the validity of specification indicates acceptable rejection frequencies at both the 5% and 10% level of significance. The error autocorrelation test indicates moderate rejection frequencies. In no more than seven out of eighteen regression specifications do the rejection frequencies exceed the nominal levels clearly. The ARCH and heteroskedasticity tests exhibit a somewhat higher rejection frequency. At least one of the above statistical tests is rejected frequently mostly because of violations of normality. In summary, rejection of normality appears to be the critical factor among the statistical tests. The statistical quality of the regressions might be somewhat improved by allowing for some type of ARCH-effects the overall evidence of Granger causality would change only marginally. The verification of this conjecture is left for future research, however.

											A. A			
		Erro	r AC	ARCH		Normality		Heterosk		RESET		At le	At least one	
		5%	10%	5%	10%	5%	10%	5%	10%	5%	10%	5%	10%	
Relevance														
Europe	E _{t-1}	1.84%	4.45%	14.11%	18.71%	34.97%	38.80%	14.26%	20.86%	7.06%	10.74%	40.34%	51.99%	
	Et	16.72%	26.84%	0.31%	1.69%	36.04%	42.48%	0.31%	1.84%	0.00%	0.61%	45.71%	54.91%	
America	Am _{t-2}	7.22%	11.67%	6.91%	15.05%	41.47%	44.24%	8.91%	15.98%	3.84%	6.91%	48.39%	55.30%	
	Am _{t-1}	9.51%	15.18%	15.18%	17.33%	41.10%	43.40%	17.33%	17.02%	4.75%	9.36%	51.23%	61.20%	
Asia	As _{t-1}	3.07%	5.98%	12.88%	16.41%	42.02%	45.55%	6.75%	12.73%	5.83%	8.74%	46.93%	51.84%	
	As _t	2.76%	5.06%	17.33%	21.78%	38.34%	40.80%	21.93%	24.85%	3.83%	9.20%	48.16%	54.91%	
Incrementa	ılity													
Europe	Et-1 Ast-1	10.43%	16.10%	0.46%	2.45%	34.36%	44.48%	0.61%	2.61%	5.67%	13.04%	39.26%	55.21%	
	$E_t As_t$	7.67%	14.26%	0.15%	2.15%	38.19%	43.56%	0.15%	0.31%	0.00%	0.92%	39.26%	45.40%	
America	Am _{t-2} E _{t-2}	4.14%	9.20%	6.60%	14.72%	38.04%	42.33%	11.81%	18.25%	6.13%	12.88%	48.93%	60.74%	
	$Am_{t-1} E_{t-1}$	5.21%	8.59%	15.49%	18.25%	34.20%	39.57%	17.02%	16.41%	5.98%	7.98%	43.71%	52.61%	
Asia	Ast-1 Amt-2	5.38%	10.60%	5.84%	13.21%	40.71%	43.93%	10.29%	10.14%	4.45%	9.37%	49.46%	57.30%	
	As _t Am _{t-1}	5.21%	10.74%	16.26%	17.94%	40.03%	42.94%	17.02%	17.18%	4.29%	5.98%	49.23%	57.21%	
Timeliness														
Europe	$E_{t-1} Am_{t-1} $	5.21%	8.59%	15.49%	18.25%	34.20%	39.57%	17.02%	16.41%	5.98%	7.98%	43.71%	52.61%	
	E _t Am _t	11.50%	21.78%	0.77%	3.68%	39.11%	44.79%	5.06%	9.36%	3.83%	6.75%	51.84%	65.18%	
America	Am _{t-2} As _{t-1}	5.38%	10.60%	5.84%	13.21%	40.71%	43.93%	10.29%	10.14%	4.45%	9.37%	49.46%	57.30%	
	Am _{t-1} As _t	5.21%	10.74%	16.26%	17.94%	40.03%	42.94%	17.02%	17.18%	4.29%	5.98%	49.23%	57.21%	
Asia	As _{t-1} E _{t-1}	4.75%	9.20%	13.80%	18.25%	35.12%	39.72%	4.14%	6.75%	9.20%	13.80%	39.72%	50.15%	
	A _{st} E _t	10.43%	16.10%	0.46%	2.45%	34.36%	44.48%	0.61%	2.61%	5.67%	13.04%	39.26%	55.21%	

Table 5.4.: Percentages of iterations with unacceptable residual statistics.

6. Conclusion

In the paper we have formulated and tested three hypotheses concerning the impact of global stock return factors on the Finnish return generating mechanism. The study is carried out using the well-known Granger causality framework. The relevance and incrementality hypotheses are corroborated for all continents (America, Europe, Asia). The timeliness hypothesis is accepted for America but not for Europe and Asia, obviously due to the strong information value in the American return factor. The dynamics of the causality patterns - studied within the rolling Granger framework - exhibits features particular to each continent included in the study. This is a reflection of the time-variability of the importance of the geographical continents as driving forces of the global financial markets.

Our study has several implications for future research. Firstly, the study could be extended to cover other Scandinavian markets as well. Furthermore, a comparison between the rather small Nordic stock markets to larger markets such as, e.g., the New York, London or Frankfurt stock markets could reveal interesting relationships. Finally, the three hypotheses tested in this study might be further challenged by extending the set of control information, by for example domestic interest rates, production volume and other macroeconomic variables.

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Figure 3.1: Daily returns on the Finnish general index FOX

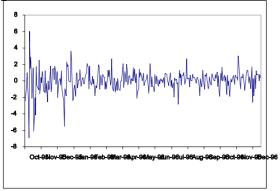


Figure 3.3: Factor 2 - Asia

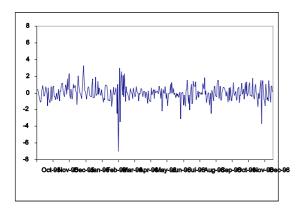


Figure 3.2: Factor 1 - Europe

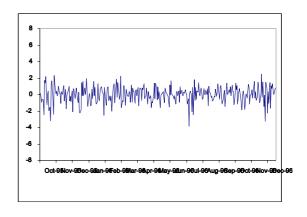
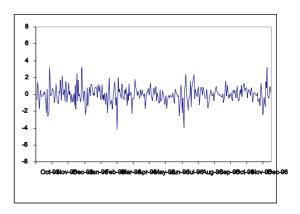


Figure 3.4: Factor 3 - America



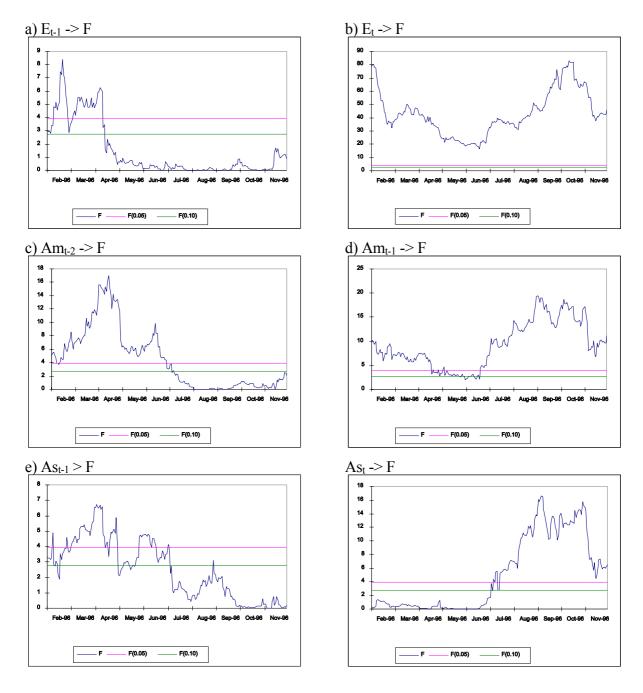


Figure 5.1: Dynamics of the relevance hypothesis measured by rolling Granger causality with window length 100

Figure 5.2: Dynamics of the incremental information hypothesis measured by rolling Granger causality with window length 100

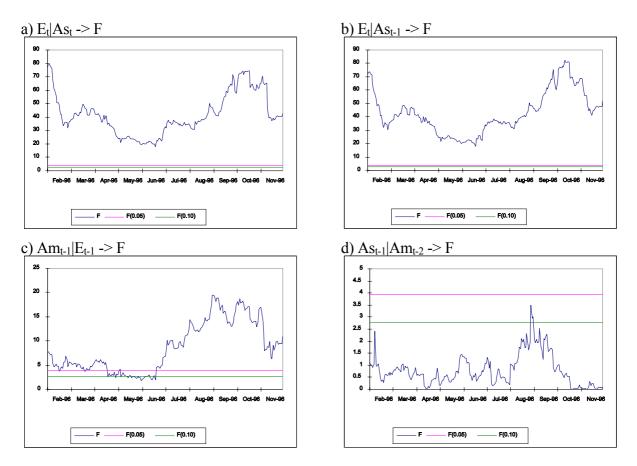
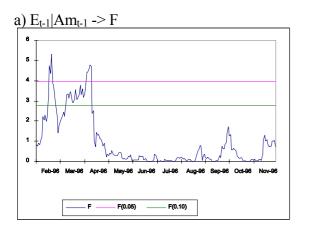
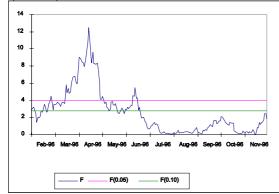
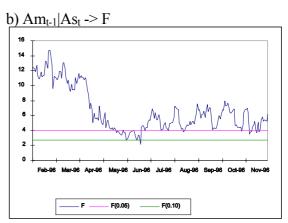


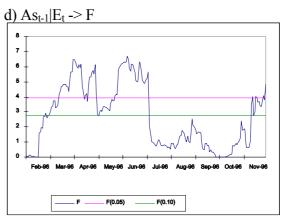
Figure 5.3: Dynamics of the timeliness hypothesis measured by rolling Granger causality with window length 100.



<u>c) $Am_{t-2}|As_{t-1} -> F$ </u>







<u>Trading Hours</u> Local Time GMT

 Helsinki: 9.00 - 14.30*
 +2.7.00 - 12.30

 Sydney:
 8.30 - 12.30 and 14.00 - 16.30 +9.23.30 - 3.30 and 5.00 - 7.30

 Tokyo:
 9.00 - 11.00 and 12.30 - 15.00 +9.0.00 - 2.00 and 3.30 - 6.00

 Toronto:
 9.30 - 16.00
 -5.14.30 - 21.00

* 1.11.1993 -> 17.00