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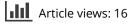
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The Effect of News on Return Volatility and Volatility Persistence: The Turkish Economy During Crisis

M. Nihat Solakoglu and Nazmi Demir

ABSTRACT: In this study, we investigate the effect of public information arrival on return volatility for Borsa Istanbul. New information arrival is measured by the number of daily news headlines for Turkey, the United States, and a sample of European countries with close trading ties with Turkey. We classify news headlines by country and type of news. Our findings indicate that, during a recessionary period, new information arrival causes return volatility mostly to decline. Moreover, both economic news and European news cause a significant decline in volatility persistence. However, when news is classified based on origin and type, a larger decline in persistence is observed.

KEY WORDS: news arrival, return volatility, Turkey, volatility persistence.

The effect of public information arrivals on asset prices and volatility has been investigated extensively in the past by researchers using different information arrival proxies. One strand of studies uses macroeconomic announcements to measure information arrivals and examine the effect on returns and return volatility (Almeida et al. 1998; Andersen and Bollerslev 1998; Ederington and Lee 1993; Kutan and Aksoy 2004a; Pearce and Roley 1985; Pearce and Solakoglu 2007). Another strand of studies focuses on trading volume as a measure of information arrival (Andersen 1996; Bollerslev and Domowitz 1993; Lamoureux and Lastrapes 1990; Locke and Sayers 1993).¹ The last line of studies measures new information arrival by the frequency of public news that arrives to the market. For the frequency of news, these studies use either the number of news reports obtained from newspapers (Berry and Howe 1994) or the number of news headlines released by companies that provide data services, such as Dow Jones & Company and Reuters (Baklaci et al. 2011; Chang and Taylor 2003; Janssen 2004; Kalev et al. 2004; Mitchell and Mulherin 1994).

We follow this last line of research by measuring information arrival via the daily number of news headlines provided by Foreks Data Terminal for the seven-month period September 1, 2008–March 31, 2009. Our aim is to investigate the effect of public information arrival on return volatility in Borsa Istanbul (BIST).² We believe this study differs from earlier studies on two fronts. First, we distinguish different types of economic news and different sources of news in our analysis. For different types of economic news, we use two classifications: news on real economy and news on inflation/money. For the classification of news sources, we differentiate among news originating from Turkey, the United States, and a subsegment of Europe that includes France, Germany, Italy, Spain,

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and the United Kingdom (which we call Europe in this study).³ Second, we compare the actual announcements to expectations and classify each news headline as a positive surprise, a negative surprise, or neutral. If markets are efficient, we should not find any relationship between the anticipated arrivals of news and return volatility, but we should observe a significant effect of unanticipated news arrivals on return volatility.⁴

Our findings show that return volatility is mostly declined by the arrival of news. Moreover, while we find that European and U.S. news on both real economy and inflation cause a large decline in volatility persistence, Turkish news on inflation, but not on real economy, leads to a decline in volatility persistence. Although the slowdown of the aggregate demands abroad and particularly in Europe hurt the Turkish trade balance and reduced its capital inflows during our sample period, Turkish investors/traders were not affected profoundly by the chain of events during the crisis years. This may be because both the Turkish government and news media downplayed the economic crisis abroad because the Turkish economy and particularly the financial sector, including the banking industry, were robust following the restructuring after the Turkish crisis of 2001. These actions, along with increased information disclosure, may be one reason why news arrival reduced return volatility.

Model Specification and Data

We use the daily number of all news headlines provided by Foreks Data Terminal to proxy new information arrival.⁵ We employ the standard model and use the number of daily news headlines (mostly utilizing the economic news) as our proxy for information arrival. News coverage only includes the weekdays, and hence there are no data available for weekends. Data on BIST100 and sectoral indexes are obtained from Matriks Data Terminal.⁶ Table 1 provides descriptive statistics on new information arrival by weekdays.⁷

Except for Fridays, the average daily number of news headlines stays above 600. The minimum number of daily news headlines is 373; the maximum is 837; and the overall average is 627. The highest number of news headlines arrives on Monday, as expected, with the lowest on Friday. An *F*-test of a null hypothesis of equality of all mean numbers of headlines for weekdays is rejected at the 1 percent level of significance. Moreover, the application of Tukey's honestly significant difference (HSD) test indicates that news arrival rates are significantly different from each other when comparing Monday and Friday only.

Although the average number of total news headlines is 627, the average number of economic news headlines is much smaller: 20.5 news items per day. In other words, only about one out of thirty news headlines is on economic news; investors must thus observe a flow of information and pick the relevant news to analyze and take action. With a correlation coefficient of 0.34, the total number of news headlines and the number of economic news headlines do not move together. Hence, we can assume that only informed traders, traders who have access to timely and accurate public information, are able to distinguish important news from the rest. If the market is controlled by informed investors/traders, then we should expect only surprise economic news to affect market volatility. Table 1 also presents the average number of daily economic news headlines per day. The lowest number of economic news arrivals occurs, on average, on Mondays, while the highest numbers occur on Thursdays and Fridays. Furthermore, Tukey's HSD test indicates a significant difference only between Monday and Thursday.

Variable	Weekday	Mean	Standard deviation	Minimum	Maximum
Total news	Monday	655	857	537	823
	Tuesday	633	897	461	837
	Wednesday	611	757	457	789
	Thursday	647	680	434	771
	Friday	585	929	373	753
Economic news	Monday	16	10	3	47
	Tuesday	20	8	4	37
	Wednesday	21	8	10	38
	Thursday	23	7	2	37
	Friday	23	12	5	53

Table 1. Descriptive statistics on daily news headlines

Since most of the news arriving in a day may not be relevant for investors, we separate economic news from all news and classify all economic news as (1) news on real economy, (2) news on inflation, or (3) other news. We exclude "other news" from the analysis as it includes news that may not be relevant a priori or news consisting of mixed signals.⁸ Furthermore, in segments (1) and (2), we classify news as a positive surprise, a negative surprise, or no surprise as the direction of the surprise may provide different signals for investors. Panel A of Table 2 provides summary statistics for the number of headlines under different segments of economic news.

As is clear from Table 2, most of the news is on the real economy, with an average of 10.3 news headlines per day. Moreover, as expected given the economic environment, we observe more bad news than good news on the real economy, with an average of 5.6 bad news items per day compared to 3.7 good news items per day. For news on inflation, there is not a significant difference between positive and negative surprise headline counts. To identify the surprises, we compare actual announcements to the expectations provided by Foreks Data Terminal within the same news headline.⁹

Not all foreign news is expected to affect return volatility in Turkey. In particular, news items related to the Turkish economy, to a subset of European economies with close trade relationships with Turkey, and to the U.S. economy are expected to be more vital for the return volatility in Borsa Istanbul. For that reason, we classify news arrival according to the country where the news originated. The segments we consider for analysis are news on the U.S. economy, news on the Turkish economy, and news on the European countries of France, Germany, Italy, Spain, and the United Kingdom. Since most of the news on Europe clusters around these countries, and the business relationships are stronger with this group of countries, we assume that this generalization does not cause much loss of information. Summary statistics on news counts for the country segments are provided in Panel B of Table 2. On average, 2.6 news headlines per day are observed for the U.S. economy, while the average for Europe is around 6.9. The average number of economic news headlines for the Turkish economy is 0.2, much lower than we expected. With a maximum of two economic news headlines per day, this indicates that there are many days with no economic news reported on the Turkish economy. The list of news on real economy and inflation that we consider, with a sample of news headlines, is provided in Table 3.

		Standard		
	Mean	deviation	Minimum	Maximum
Panel A: News by category				
All economic news	20.5	9.4	2	53
News on real economy	10.3	6.3	0	28
Stronger	3.7	3.2	0	18
As expected	1.0	1.3	0	6
Weaker	5.6	4.0	0	18
News on money/inflation	5.8	4.5	0	22
Higher	2.1	2.0	0	11
As expected	1.1	2.0	0	12
Lower	2.6	2.5	0	13
Other news	4.3	3.4	0	14
Panel B: News by origin				
News on U.S. economy				
All news	2.6	2.6	0	11
News on real economy	2.0	2.1	0	11
News on money/inflation	0.5	1.2	0	5
News on eurozone economy				
All news	6.9	4.9	0	20
News on real economy	4.5	4.0	0	15
News on money/inflation	2.4	3.1	0	15
News on Turkish economy				
All news	0.2	0.6	0	2
News on real economy	0.1	0.4	0	2
News on money/inflation	0.1	0.4	0	2

Table 2. Economic news and	news segments for France	, Germany, Italy, Spain,
and the United Kingdom		

Notes: Due to rounding, totals may deviate from the sum of subgroups. The list of economic news we use is provided in Table 3.

To estimate market volatility, we use several market indexes from Borsa Istanbul. The objective of utilizing several index returns is to check the robustness of our results and to observe any differences among sectoral indexes. The main index we use is the BIST100 index, which represents the market portfolio of 100 listed companies on Borsa Istanbul. We also include the Second National Market Index of Turkey, which is established to promote trading in stocks of small- and medium-sized companies. This index also includes companies delisted from the national market of the Borsa Istanbul. To capture differences in the sectoral response to new information arrival, the following sectoral indexes are also included in the analysis: banks, financials, industrials, services, and technology. For all index levels, we use daily closing prices, obtained from Matriks Data Terminal.

To test the effect of information arrival on return volatility, we use the conditional variance from a generalized autoregressive conditional heteroskedasticity (GARCH) process introduced by Bollerslev (1986) and Engle (1982). Under the mixture-of-distributions hypothesis (MDH), the arrival of new information drives the change in return volatility concurrently (Darrat et al. 2007; Kalev et al. 2004). Hence, when a proxy for new information arrival is included in the conditional variance equation, we expect most of the volatility persistence to disappear. Following Kalev et al. (2004), we estimate the model

News on real economy		
Retail sales	Productivity (overall or sector based)	Continuing claims
Manufacturing/service/ compound sector PMI (Purchasing Managers Index)	Capital expenditures	Initial jobless claims
Mortgage approvals	Change in Industrial Production	Industrial new orders
Construction approvals	Key current economic state index	IFO business climate survey
GDP growth	Investor confidence index	Existing home sales
ISM (Institute for Supply Management) manufacturing/ nonmanufacturing indicator	Leading indicators index	Factory good orders
The growth of construction expenditures	Housing starts	New motor vehicle sales
Business confidence index Factory orders	Building permits Capacity utilization rate	Manufacturing shipments ZEW (Centre for European Economic Research) survey indicator
Unemployment rate	Consumer confidence index	Employment change
Manufacturing orders	Fed manufacturing index	
News on inflation		
CPI (consumer price index) growth	Retail price index	Change in import prices
PPI (producer price index) growth	ISM prices paid	Change in GDP deflator
Money supply growth Producer prices: output or input	Labor cost index	Interest rates

Table 3. List of economic news

News on real economy

Sample of news headlines

09:45:23 The unemployment rate in France for the second quarter is 7.6 percent (expected 7.5 percent; previous 7.6 percent)

13:00:14 Manufacturing orders in Germany decreased by 1.7 percent in July (expected +0.4 percent; previous -2.9 percent)

14:00:03 Bank of England left interest rates unchanged at 5.0 percent (expected 5.0 percent) 15:15:12 The private sector employment declined by 33,000 in the United States (expected -30,000; previous -51,000)

Notes: The sample news is provided from September 4, 2008. The original news headlines were in Turkish. We have translated the sample from Turkish to English.

below for market return $R_{i,t}$, where the return is calculated as the change in the logarithm of the daily closing index level.¹⁰

$$R_{i,t} = \mu + \phi_1 D_1 + \phi_2 D_2 + \phi_3 D_3 + \phi_4 D_4 + \varepsilon_t, \text{ where } \varepsilon \sim N(0, \sigma_t^2)$$

$$\sigma_t^2 = \alpha_0 + \alpha_1 \varepsilon_{t-1}^2 + \beta \sigma_{t-1}^2 + \Sigma \lambda_j N_{jt}.$$
 (1)

In the mean equation, D_1-D_4 represent day dummies to capture the day-of-the-week effect. In the conditional variance equation, N_{jt} denotes measures of new information arrival at time *t* for measure *j*, and λ_j denotes the associated coefficients. If the new information arrival is important, we expect λ_j to be statistically significant for the *j*th news measure. Moreover, with the new information arrival, we expect a decline in volatility persistence, as defined by the sum of coefficient estimates α_1 and β .

Results Discussion

Estimation results for Equation (1) are presented in Table 4. The new information arrival is measured by using several different proxies, and market return is represented by several market indexes. To save space, we present only the coefficient estimates on new information proxies. Except for financial and Second National indexes, the total news does not influence return volatility significantly. With economic news used as our proxy, our findings show a decline in return volatility for most indexes. This result is puzzling as many of the earlier empirical works find news arrival causes the return volatility to increase (see, for example, Janssen 2004; Kalev et al. 2004). However, as Kutan and Aksoy (2004a) find, both lagged tourism announcements and gross national product (GNP) news influence return volatility negatively in Borsa Istanbul. Similarly, for the gold market returns in Turkey, Kutan and Aksoy (2004b) show that the volatility of returns declines in response to public news on industrial production. Moreover, for both bearish and bullish periods, Baklaci et al. (2011) provide some evidence that return volatility declines in response to news for several firms. Therefore, our results, although surprising, are not inconsistent with some of the earlier works that use data from Borsa Istanbul. Further, we use the number of total news headlines for the period January 2, 2013–May 31, 2013 and estimate Equation (1) with the total news as our proxy. The coefficient on news proxy is positive and insignificant for all indexes.¹¹ As a theoretical explanation for the negative effect of news arrival on return volatility, it can be argued that news arrival reduces information asymmetry, and reduced information asymmetry reduces the volatility (Diamond and Verrecchia 1991).¹² During our sample period, although investors/traders were not getting clear signals from the domestic and foreign events rolling simultaneously into the country, the Turkish government and news media were both heavily downplaying the economic crisis abroad because the Turkish economy and particularly the financial sector, including the banking industry, were robust following the restructuring after the Turkish crisis of 2001. Moreover, to lower the public unease, there was an increased public disclosure on firms' financial health and strength of the financial sector in Turkey. As a result, it is plausible to assume that reduced return volatility is caused by reduced information asymmetry due to arrival of new information during the crisis period.

In Table 4, our findings indicate that news on the Turkish and U.S. economies does not seem to affect return volatility in Turkey. News from Europe has a negative and significant effect on return volatility. During that time frame, perhaps the focus of investors was on crisis contagion through trade, and they thus paid more attention to news about major trading partners.

As discussed in the related literature, we cannot assume all news headlines send the same signal to investors/traders. News headlines that indicate a surprise output growth or a surprise rise in price levels can have a different interpretation than expected output growth or a surprise decline in price level. Therefore, we classify news on real economy

			News measure		
Index	Total	Economic	U.S.	European	Turkish
measure	news	news	news	news	news
BIST100	-0.00045	–0.00699***	–0.01790	-0.01440***	0.08820
index	(0.00032)	(0.00267)	(0.02190)	(0.00551)	(0.12900)
BIST30	–0.00055	–0.00809***	–0.02060	–0.01680***	0.09710
index	(0.00037)	(0.00295)	(0.02490)	(0.00624)	(0.15700)
Second	0.000484**	–0.00576	-0.00317	-0.01320***	0.09810
National	(0.00022)	(0.00399)	(0.01440)	(0.00519)	(0.09660)
Bank index	-0.00074	-0.01110***	–0.03950	-0.02370***	0.09990
	(0.00057)	(0.00422)	(0.06760)	(0.00894)	(0.46500)
Financials	-0.00063***	-0.01020***	–0.03620	-0.02170***	0.09140
index	(0.00003)	(0.00375)	(0.02920)	(0.00282)	(0.17800)
Industrial	-0.00012	-0.01020***	-0.01430	–0.02570***	0.10400
index	(0.00020)	(0.00758)	(0.01400)	(0.00770)	(0.07510)
Services	-0.00029	-0.00991***	–0.00238	-0.00851***	0.07550
index	(0.00018)	(0.00165)	(0.02820)	(0.00322)	(0.19600)
Technology	-0.00021	–0.00856***	–0.00447	–0.02930***	0.06120
index	(0.00030)	(0.00183)	(0.01770)	(0.01020)	(0.09740)
Notes: Standa	rd errors are in par	entheses. * Statisti	cally significant	at 10 percent: ** st	atistically

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Table 4. New information arrival and stock market volatility

Notes: Standard errors are in parentheses. * Statistically significant at 10 percent; ** statistically significant at 5 percent; *** statistically significant at 1 percent.

and money/inflation into three groups in order to distinguish surprises from expected news by comparing actual to expected news. For news on real economy, the subgroups defined are stronger than expected, as expected, and weaker than expected. For news on money/inflation, we use subgroups higher than expected, as expected, and lower than expected. Moreover, as discussed in Birz and Lott (2011) and Pearce and Solakoglu (2007), the same news headline can send a different signal depending on whether the economy is in a recession or is overheating. For example, in an overheating economy, news on surprise output growth can be interpreted as bad news, while the same can be interpreted as good news when the economy is in a recession. Given that our sample period corresponds to a crisis period, we believe that both surprise growth and surprise inflation news represent signals of good news, which may or may not lead to a decline in return volatility. However, we expect return volatility to decline after a good signal. This expectation is based on literature about herding (see, e.g., Chang et al. 2000; Christie and Huang 1995; Hwang and Salmon 2004).

Table 5 presents our estimation results under the aforementioned classification. Results indicate that higher-than-expected inflation leads to a decline in return volatility for all indexes, which is consistent with our expectation. Stronger-than-expected news on real economy causes return volatility to increase for services and bank indexes. Only for the BIST100 index, we find a negative and significant coefficient. Interestingly, weaker-than-expected news leads to a decline in return volatility.

In Tables 6, 7, and 8, we investigate the role of news arrival on return volatility for the United States, the European countries, and Turkey, respectively. As presented in Table 6, return volatility is mostly influenced by U.S. real economy news. The effect seems to be negative for both stronger and weaker real economy news. News on inflation from the

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Table 5. Economic news and volatility

	Ž	News on real economy			News on inflation	
Index measure	Stronger	Expected	Weaker	Higher	Expected	Lower
BIST100 index	-0.03130*	-0.09090	-0.02730	-0.03740***	0.02990	-0.00231
	(0.01470)	(0.10300)	(0.04030)	(0.00673)	(0.03410)	(0.03010)
BIST30 index	-0.03500	-0.10400	-0.03050	-0.04240***	0.00871	-0.05470***
	(0.05470)	(0.11700)	(0.01440)	(0.01240)	(0.02060)	(0.00130)
Second National	-0.03100	-0.05890	-0.02500	-0.05100**	-0.02970	-0.05020
	(0.04100)	(0.07720)	(0.02220)	(0.02490)	(0.02450)	(0.04650)
Bank index	0.03010***	0.02400	-0.04980***	-0.09940**	0.06800	-0.01080
	(0.00866)	(0.05190)	(0.00980)	(0.04860)	(0.07640)	(0.06710)
Financials index	0.00168	0.16400	-0.06260***	-0.05320***	0.04220	-0.03620
	(0.03130)	(0.10900)	(0.01950)	(0.01860)	(0.04430)	(0.03310)
Industrial index	-0.02230	-0.06210	-0.01830	-0.04320***	-0.01240	-0.02590
	(0.02750)	(0.05980)	(0.02130)	(0.01540)	(0.03430)	(0.03400)
Services index	0.01020*	0.02320	-0.01790***	-0.01230	0.01690	0.01360
	(0.00399)	(0.02820)	(0.00398)	(0.00829)	(0.01390)	(0.01470)
Technology index	0.00686	0.03490*	-0.02990**	-0.04730*	0.00204	-0.03660
	(0.00561)	(0.02070)	(0.00475)	(0.01690)	(0.03740)	(0.02550)
Notes: Standard errors are in parentheses. * Statistically significant at 10 percent; ** statistically significant at 5 percent; *** statistically significant at 1 percent	parentheses. * Statistics	ally significant at 10 p	bercent; ** statistically si	gnificant at 5 percent; **	** statistically signific	ant at 1 percent.

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Table 6. U.S. economic news and volatility

	Ň	News on real economy	y		News on inflation	
Index measure	Stronger	Expected	Weaker	Higher	Expected	Lower
BIST100 index	-0.18600***	0.05680	-0.08730	-0.06420	-0.52100	0.32500
	(0.06840)	(0.15100)	(0.06910)	(0.07570)	(0.35300)	(0.38600)
BIST30 index	-0.21400***	0.10800	-0.08990	-0.11200	-0.51500	0.23100
	(0.05970)	(0.20000)	(0.07420)	(0.13300)	(0.48100)	(0.33000)
Second National	-0.17100***	-0.08150	-0.11300**	0.03630	-0.13100	0.10500
	(0.04870)	(0.05430)	(0.04640)	(0.05270)	(0.11500)	(0.11600)
Bank index	-0.29900***	0.10900	-0.11900	-0.18900	-0.64300	0.17300
	(0.08820)	(0.25300)	(00260)	(0.12500)	(0.52000)	(0.34000)
Financials index	-0.03270	-0.28600	-0.04300	-0.09560	0.40200*	0.15900
	(0.07050)	(0.22000)	(0.04150)	(0.07760)	(0.20500)	(0.14300)
Industrial index	-0.11400^{***}	0.14000	-0.10000***	-0.03370	-0.01730	0.11600
	(0.03280)	(0.22900)	(0.03420)	(0.02720)	(0.11700)	(0.11100)
Services index	-0.09370***	-0.07320	-0.02900	-0.00354	-0.08190	0.04700
	(0.00626)	(0.09230)	(0.05850)	(0.07810)	(0.36900)	(0.26500)
Technology index	-0.04630	-0.18700	-0.04260*	-0.08470	-0.13900	-0.16500
	(0.04240)	(0.14900)	(0.02460)	(0.06200)	(0.15000)	(0.14000)
Notes: Standard errors are in parentheses. * Statistically significant at 10 percent; ** statistically significant at 5 percent; *** statistically significant at 1 percent	parentheses. * Statistica	ully significant at 10 ₁	percent; ** statistically si	gnificant at 5 percent; **	** statistically significa	unt at 1 percent.

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I	Ž	News on real economy	~		News on inflation	
Index measure	Stronger	Expected	Weaker	Higher	Expected	Lower
BIST100 index	-0.07870**	0.02300	-0.04370	-0.05270*	-0.07050***	0.08370
	(0.03240)	(0.08960)	(0.03100)	(0.02860)	(0.01850)	(0.08210)
BIST30 index	-0.09530*** (0.02630)	-0.01650 (0.07280)	-0.05820* (0.03360)	-0.04890* (0.02750)	-0.05080 (0.03930)	0.02030
Second National	-0.04030	0.02260	-0.02240	-0.10300***	-0.05690	-0.12200***
	(0.04820)	(0.07030)	(0.02550)	(0.03570)	(0.03580)	(0.02740)
Bank index	-0.17100***	0.04480	-0.03080	-0.00664	-0.10100***	0.00981
	(0.04280)	(0.09600)	(0.03390)	(0.05740)	(0.00277)	(0.06700)
Financials index	-0.12500**	0.06330	-0.06970)	-0.04990	-0.09380***	0.07230
	(0.05340)	(0.11000)	(0.05970)	(0.04930)	(0.02760)	(0.08150)
Industrial index	-0.04220***	-0.07430	-0.03950	-0.05750***	0.00874	0.00062
	(0.01180)	(0.09070)	(0.05050)	(0.01710)	(0.03400)	(0.05510)
Services index	-0.03570**	-0.04120	0.00773	0.01380	0.00957	-0.06390***
	(0.01430)	(0.05800)	(0.01350)	(0.06880)	(0.06050)	(0.02370)
Technology index	-0.05580	-0.05860**	-0.04640***	-0.06870***	0.02320	-0.03220
	(0.03700)	(0.02790)	(0.01650)	(0.01170)	(0.03630)	(0.03910)
Notes: Standard errors are in parentheses. * Statistically significant at 10 percent; ** statistically significant at 5 percent; *** statistically significant at 1 percent.	parentheses. * Statistic:	ally significant at 10 p	percent; ** statistically s	ignificant at 5 percent; **	** statistically significa	nt at 1 percent.

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Table 8. Turkish economic news and volatility

ExpectedWeakerHigher $-1.51300**$ $0.57300**$ -0.57000 $-1.51300**$ $0.57300**$ $-0.57000***$ $-1.60700**$ $0.57300**$ -0.16500 (0.53800) (0.28000) -0.16500 $0.63800)$ (0.28000) -0.19900 (0.39100) (0.28000) -0.19900 (0.39100) (0.28000) $-1.14300**$ (0.77200) (0.28000) $-1.14300**$ (0.77200) (0.28000) $-1.14300**$ (0.77200) (0.28000) $-1.14300**$ (0.77200) (0.28000) $-1.14300**$ (0.77200) (0.28000) $-1.14300**$ (0.77200) (0.28000) $-1.15400**$ (0.77200) (0.78100) (0.52200) (0.77200) (0.38100) -0.16600 (0.77200) (0.19100) (0.54000) (0.77200) (0.19100) (0.54000) (0.29900) (0.10100) (0.23600) (0.29900) (0.10100) (0.23600) $(0.23900)***$ $0.66200**$ -0.236900		-					
$\begin{array}{llllllllllllllllllllllllllllllllllll$	ndex measure	Stronger	Expected	Weaker	Higher	Expected	Lower
$ \begin{array}{llllllllllllllllllllllllllllllllllll$	IST100 index	-0.01280	-1.51300^{**}	0.57300**	-0.57000***	I	-0.15100
$\begin{array}{cccccccccccccccccccccccccccccccccccc$		(0.20100)	(0.59000)	(0.27700)	(0.14000)		(0.14900)
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	IST30 index	0.02050	-1.60700^{**}	0.58000**	-0.16500	I	-0.60300***
$\begin{array}{llllllllllllllllllllllllllllllllllll$		(0.24000)	(0.63800)	(0.28000)	(0.61000)		(0.10600)
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	econd National	-0.04810	-1.41700^{***}	0.46800^{*}	-0.19900	I	-0.05190
0.06550 -2.28700*** 0.75000* (0.34700) (0.77200) (0.39200) 0.06830 -2.19200*** 0.76800** 0.06660 -1.10200** 0.76800** -0.06660 -1.00400** 0.42100** 0.10400) (0.47600) 0.19100) -0.01380 -0.87900*** 0.39400*** -0.01380 -1.53900*** 0.66200**		(0.19500)	(0.39100)	(0.28000)	(0.27300)		(0.27200)
$\begin{array}{llllllllllllllllllllllllllllllllllll$	ank index	0.06550	-2.28700***	0.75000*	-1.14300^{**}	I	-0.88300***
0.06830 -2.19200*** 0.76800** 0.30000) (0.71800) (0.38100) -0.06660 -1.00400** 0.42100** 0.10400) (0.47600) (0.19100) -0.01380 -0.87900*** 0.39400*** 0.1000) (0.29900) (0.10100) -0.01380 -1.53900*** 0.66200**		(0.34700)	(0.77200)	(0.39200)	(0.53000)		(0.26100)
(0.30000) (0.71800) (0.38100) -0.06660 -1.00400** 0.42100** -0.10400) (0.47600) (0.19100) -0.01380 -0.87900*** 0.39400*** -0.01380 -0.87900*** 0.10100) -0.01380 -1.53900*** 0.66200**	inancials index	0.06830	-2.19200***	0.76800**	-1.15400^{**}		-0.74300**
-0.06660 -1.00400** 0.42100** (0.10400) (0.47600) (0.19100) -0.01380 -0.87900*** 0.39400*** -0.01380 -0.29900) (0.10100) -0.01380 -1.53900*** 0.66200**		(0.30000)	(0.71800)	(0.38100)	(0.52200)		(0.33600)
(0.10400) (0.47600) (0.19100) (-0.01380 -0.87900*** 0.39400*** - (0.10800) (0.29900) (0.10100) (-0.01380 -1.53900*** 0.66200** -	ndustrial index	-0.06660	-1.00400^{**}	0.42100**	-0.16600		-0.28700***
-0.01380 -0.87900*** 0.39400*** (0.10800) (0.29900) (0.10100) -0.01380 -1.53900*** 0.66200**		(0.10400)	(0.47600)	(0.19100)	(0.54000)		(0.05010)
(0.10800) (0.29900) (0.10100) (0. -0.01380 -1.53900*** 0.66200**	ervices index	-0.01380	-0.87900***	0.39400***	-0.36900		-0.18800
-0.01380 -1.53900*** 0.66200**		(0.10800)	(0.29900)	(0.10100)	(0.31100)		(0.29700)
	echnology index	-0.01380	-1.53900^{***}	0.66200**	-0.23600		-0.09530
0.41800) (0.28700) (0.28700)		(0.16300)	(0.41800)	(0.28700)	(0.57100)		(0.12200)

Type of new information	Change in persistence (percent)	
Total news	-1.9	
Economic news	-7.0	
U.S. news	-4.1	
U.S. real economy	-26.6	
U.S. inflation	-26.0	
European news	-7.6	
European real economy	-25.2	
European inflation	-16.3	
Turkish news	-3.9	
Turkish real news	-0.8	
Turkish inflation	-24.9	
Real economy news	-15.0	
Inflation news	-11.3	

Table 9. New information arrival and persistence

Notes: Persistence is measured as the sum of coefficients, excluding the constant, in the conditional variance equation. Persistence is calculated for models reported in Tables 5–8.

United States appears to be trivial for investors in Borsa Istanbul. For European news (Table 7), we have some evidence that stronger-than-expected news on real economy reduces return volatility. However, for inflation, most news leads to a decline in return volatility.

News on a stronger Turkish real economy (Table 8) does not affect return volatility. If there is no surprise, we observe a decline in volatility, whereas negative news causes return volatility to increase. Perhaps investors/traders do not put high expectations on positive news, but they value negative news at face value. Our results also indicate that market volatility responds negatively to news on inflation. Most of the responses are negative, regardless of whether the news indicates a positive surprise or a negative surprise. As discussed earlier, we may assume, as an explanation, that reduced return volatility is caused by reduced information asymmetry due to news arrival (Diamond and Verrecchia 1991). As another explanation, it is reasonable to assume that investors/traders cannot separate good news from bad news, and hence they choose the best action by, perhaps, acting together and following the market. That is, this finding might be related to "unintentional herding," where investors take independent but similar actions (Bikhchandani and Sharma 2001; Hwang and Salmon 2004). The results presented in Tables 6, 7, and 8 also indicate that return volatility is not influenced much by the expected news, which is consistent with the efficient market hypothesis.

According to MDH, if news arrival is important for volatility, we should expect to observe a decrease in volatility persistence. In Table 9, we present the average percentage change in volatility persistence with and without news. For instance, when we include total news as a measure of new information arrival in the conditional variance equation, volatility persistence reduces by an average of 1.9 percent. However, with economic news as the measure, we observe an average decline of 7.0 percent. The comparison of news arrival by country indicates that news from Europe leads to a larger decline in volatility persistence than news from Turkey or the United States.

When new information arrival is represented by the news categories in the conditional volatility equation, the decline in volatility persistence is about 15.0 percent for real economy news and about 11.3 percent for inflation news. Moreover, while real economy news from Europe and the United States has a greater effect on volatility persistence than does inflation news from Europe and the United States, inflation news on the Turkish economy has a greater effect on volatility persistence than does Turkish real economy news, which appears to have almost no effect on volatility persistence.

Concluding Remarks

In this study, we investigate the role of news on return volatility in Borsa Istanbul for the period September 1, 2008–March 31, 2009. The news arrival is measured by the number of daily news headlines, and news is classified according to the country of origin, the economic content, and the surprise component.

Our results show that the arrival of news mostly causes a decline in return volatility. Moreover, while both real economy and inflation news from Europe and Turkey affect return volatility, from the United States, only real economy news appears to effect return volatility. In most estimations, consistent with the efficient market hypothesis, return volatility is not influenced by expected news. Although some of the negative effects are expected (such as a decline in return volatility due to higher-than-expected inflation, which may signal economic growth and not inflationary pressures during a crisis), some of the negative effects are not expected (such as a fall in return volatility due to weakerthan-expected growth in the economy). This finding might be due to reduced information asymmetry or unintentional herding by investors in the Turkish stock market during the crisis.

As implied by the mixture-of-distributions hypothesis, our results show a decline in volatility persistence. In particular, economic news and specifically European news lead to a larger decline in volatility persistence. Furthermore, while only inflation news on the Turkish economy reduces volatility persistence, both types of news from Europe and the United States lead to a significant decline in persistence, indicating that Turkish news on real economy is not an important source for volatility clustering. Given the unusual negative effect of news arrival on return volatility, it would be useful to examine the effect of news on return volatility in a more recent, noncrisis period, taking into account the herding patterns.

Notes

1. Along with trading volume, other measures, such as floor transactions, number of price changes, and executed order imbalances, are also used (Locke and Sayers 1993).

2. In April 2013, all exchanges operating in the Turkish capital markets were brought together under Borsa Istanbul, and Istanbul Stock Exchange became part of this new entity. For more information, see the Borsa Istanbul Web site (www.borsaistanbul.com).

3. The share of Turkish exports to the subsegment of European countries is around 30 percent for both 2008 and 2009. If we look at the Turkish exports to EU countries, the share of exports increases to 62 percent and 65 percent, respectively, for 2008 and 2009 (www.economy.gov.tr).

4. Although political news can be important for emerging markets, our analysis includes political news counts only in the total news category; the focus is more on economic news. For the effect of political news on return volatility, see Önder and Şimga-Mugan (2006).

5. Foreks Information Communication Services is a company that specializes in providing real time financial data to banks and financial institutions.

6. Matriks Information Distribution Services is a company that specializes in providing real time financial data, mostly to banks and financial institutions.

7. Our sample includes several days with missing data. We replace missing values with the average number of news items for the relevant day. For example, if we have data missing from Monday, we use the Monday average, not the overall average.

8. Some examples of news that falls into this latter category are crude oil inventory, change in housing financing, and Central Bank net debt amount. These news items are excluded from the analysis for two reasons: first, their mixed signals make them difficult to categorize; further, we do not want to present results for individual news. Second, we use "other news" as a proxy for new information arrival; overall results indicate that other news does not influence volatility significantly.

9. There are days with no news arrival on real economy or inflation; hence, we have zero minimum values for these subsegments. However, in our sample, we do not have zero number of news arrivals on the same day for all subsegments, so Table 1, which provides descriptive statistics on overall economic news, reports positive values for the minimum economic news.

10. To check for the robustness of our results, we also estimate EGARCH(1,1) model, and the results are qualitatively the same. Moreover, the inclusion of lagged news arrival measures with and without contemporary news arrival measure in the conditional variance equation does not change the sign or significance of the effect significantly. Judging from the correlogram and Box-Pierce Q test results, we assume that there is no unaccounted information in the residuals and use GARCH(1,1) model in our estimations.

11. To test the robustness of our results, we calculate the correlation between squared residuals and total news in our sample data. The correlation is negative. Further, when we use lead and lag news to calculate the correlation coefficient, the sign stays negative. The calculation of the same correlation coefficient for the recent data shows a positive correlation coefficient.

12. We thank an anonymous referee for pointing out this explanation.

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