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MEASURING INFORMATION ASYMMETRY IN LARGE ACTIVE FIRMS ON THE TEHRAN STOCK EXCHANGE

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Abstract

In financial markets, transparency of financial information is one of the most effective variables of investment strategies. Information asymmetry can seriously affect firm performance on the stock exchange and firms with a poor informational environment can lose the interest of investors. Reducing information asymmetry can have an important effect on firm performance on the stock exchange. Firms may lack a clear informational environment in the market because of the emerging conditions governing the Tehran Stock Exchange. Because larger and more active firms on the Tehran Stock Exchange provide more information, measuring the informational environment of these firms provides an overview of information asymmetry. The present study calculated the information asymmetry in these firms using the PIN and FE indices. The inconsistent results provided by these indices prompted the authors to offer a new index that is a composite of the PIN and FE that can better explain information asymmetry in developing market such as Asian stock markets. The results show that the new composite index, by using the mechanisms of the PIN and FE indices, provides a better outcome. The new composite index shows that the Tosee Melli Inv (TMEL1), Mobarakeh Steel (FOLD1), Iran Mobil Tele (HMRZ1), Saipa (SIPA1) and I.N.C. Ind. (MSMI1) firms have a better informational environment on the Tehran Stock Exchange.

Keywords: Information asymmetry, Tehran Stock Exchange, Composite index, PIN, FE

JEL classification: D82, D83

1. INTRODUCTION

Information asymmetry is a fundamental issue in financial markets. The effect of information asymmetry is undeniable when determining market function and explaining its complexities. Trading stocks creates legal rights and commitments for firms and investors. Basically, firms must commit to repayment of the principal and interest in due time, but beyond all legal regulations, a stock contract is at risk in terms of economic considerations.

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The intrinsic uncertainty of each investment project challenges a firm's ability to cover published stock. This obstacle can affect stock value logically and by estimating probability. Another obstacle is breach of commitment in a stock contract by a firm. Firms can hide the nature of a project, deflect inputting capital from its original route, or hide actual results. Uneven distribution of information can cause different results. Before it is considered by individual users, determining the quality of the information distributed is important.

Researchers have developed measures to determine the level of information asymmetry because is not directly observable. These measures fall into three categories (Clarke and Shastri, 2000). The first category uses measures based on growth opportunities and argues that firms with more growth opportunities have a symmetrical information environment. The second category uses measures based on predicted profit and argues that an increase in the level of a firm's information leads to convergence of expectations of that firm's future profits. The third category uses a series of measures based on market microstructure literature that are commonly compared with other categories.

The present study measured information asymmetry in selected firms on the Tehran Stock Exchange using two indexes from the second and third categories. The PIN index (Probability of Informed Trade) falls into the third category; the FE index (Earning Forecast Error) falls into the second category and was used to verify the results. There are many criteria to measuring information asymmetry, but most of them are suitable for developed markets. In most of developing markets they are not able to measuring information asymmetry singly and significantly. Therefore, we tried to investigate this issue. The Tehran Stock Exchange is one of the best examples of developing markets. Differences in the results by the two indices led to the use of a composite measure of these indices that considers both profit forecast and market structure. This new index calculates both general information asymmetry at beginning of a period and detailed information asymmetry over time.

2. LITERATURE REVIEW

In the context of information asymmetry measures based on market microstructure literature, Demsetz (1968) introduced a range of bid-ask prices (spread). The spread has been commonly used as a proxy for information asymmetry in previous studies. Analytical models such as Kyle (1985) and Glosten and Milgrom (1985) predicted that information asymmetry increases the adverse selection risk of market makers (liquidity providers), which increases the spread and decreases market depth and, as a result, decreases liquidity. Higher levels of information asymmetry will combine to form a larger spread. Using the spread to control a firm's informational environment, Healy and Palepu (1995) and Welker (1995) found a negative (positive) relationship between spread and the quality of information disclosure (information asymmetry). Healy *et al.* (1999) and Leuz and Verrecchia (2000) showed that information asymmetry, the spread, and volatility of stock prices are negatively associated with quality disclosure.

Many measures of the bid-ask spread have been used. Stoll (2000) found that the spread protects market makers against losses from trade with informed investors; thus, they expand the spread in order to limit informed traders and limit the spread to attract liquidity traders. Hasbrouck (2009) found a relatively high correlation between various spread measures (~90%). Richardson (2000), Mohd (2005), Attig *et al.* (2006), Jayaraman (2008) and Bhattacharya *et al.* (2009) used different spread indices to measure information asymmetry. Easley and O'Hara (1987 and 1992) offered a model of order flow in the market

that can estimate the probability of random trading by an informed trader (PIN). Also Easley *et al.* (1997b) argued about information content between trades, Easley *et al.* (1997a) about trade size, Easley *et al.* (1998) about analyst coverage, Easley *et al.* (2001) about stock splits, and Easley *et al.* (2002) about asset pricing. Many studies used PIN index as proxy of information asymmetry, such as Heidle and Huang (2002), Vega (2006), Ascioglu *et al.* (2008), Brockman and Yan (2009), Kang (2010), Aslan *et al.* (2011), Abad and Yague (2012), Chen and Zhao (2012), Dey and Radhakrishna (2015), Lin *et al.* (2013), Sankaraguruswamy *et al.* (2013), Chang and Lin (2015), Agudelo *et al.* (2015) and Paparizos *et al.* (2016).

3. METHODOLOGY

Information asymmetry is a feature of many trades. Often, seller information about quality is greater than buyer information. On the stock market, information plays an important role and its acquisition cost is gradual. According to Crawford *et al.* (2015) we assume which i = 1, ..., I is visible investors of type k = 1, ..., K, and j = 1, ..., J is firms in stock markets m = 1, ..., M, and t = 1, ..., T term. Investors gain the utility for stock that is dependent on their demands as follows:

$$U_{ikjmt}^{D} = \bar{\alpha}_{0}^{D} + \alpha_{1}^{D}P_{jmt} + X_{jmt}^{\prime}\beta^{D} + \xi_{jmt}^{D} + \sigma^{D}\upsilon_{i} + Y_{i}^{\prime}\eta^{D} + \gamma_{K}^{D} + \varepsilon_{ikjmt}^{D}$$
(1)

Utility for the outside option is normalized to zero. Investors will choose firms that maximize their utility or avoid buying securities. Provided that they invest in stocks, they will choose a share of capital for which their utility is maximized as follows:

$$U_{ikmt}^{L} = \alpha_0^L + \alpha_1^L P_{jmt} + X_{jmt}' \beta^L + \xi_{jmt}^L + Y_i' \eta^L + \gamma_k^L + \varepsilon_{ikmt}^L$$
(2)

And investors will choose to default if the following utility is greater than zero:

$$U_{ikmt}^{F} = \alpha_0^F + \alpha_1^F P_{jmt} + X_{jmt}' \beta^F + \xi_{jmt}^F + Y_i' \eta^F + \gamma_k^F + \varepsilon_{ikmt}^F$$
(3)

In above equations, X_{jmt} are firms visible features, P_{jmt} are prices, ξ_{jmt} are firms invisible features, Y_i are investors visible feature and γ_k^F are type of fixed effects. Also assumed to be ε_{ikmt}^D are is distributed as a type of extreme value (Crawford *et al.*, 2015; Berry *et al.*, 1995). Crawford *et al.* (2015) assumed random coefficient of the demand's constant term $\alpha_{0i}^D = \overline{\alpha}_0^D + \sigma^D v_i$, with $v_i \sim N(0,1)$ that have normally distributed with ε_{ikmt}^L and ε_{ikmt}^F , so that:

$$\begin{pmatrix} \alpha_0^D \\ \varepsilon^L \\ \varepsilon^F \end{pmatrix} \sim N \begin{pmatrix} \sigma^{2D} & \rho D L \sigma^D \sigma^L & \rho D F \sigma^D \sigma^F \\ \rho D L \sigma^D \sigma^L & \sigma^{2L} & \rho L F \sigma^L \sigma^F \\ \rho D F \sigma^D \sigma^F & \rho L F \sigma^L \sigma^F & \sigma^{2F} \end{pmatrix}$$
(4)

There is a positive correlation between invisible demand by investors and default (ρDF) , that is evidence of information asymmetry. If there exists a positive correlation between invisible demand and invisible default, investors with higher risk will create a greater demand for stocks. The argument of correlation between α_0^D and ε^F is that if an investor who the model predicts should not be willing to buy stocks is observed to buy

stocks, then this denotes an investor with high α_0^D . A positive correlation between α_0^D and ε^F is evidence of information asymmetry. The interpretation of a positive correlation between stock size and invisible default (ρDF) is also evidence of information asymmetry.

Easley and O'Hara (1987 and 1992) offered a model of order flow in the market that can be used to estimate the probability of random trading by an informed trader (PIN). A higher the PIN value (0 to 1) indicates more confidential information or higher levels of information asymmetry (Wan, 2009). The basic assumption is that public information is directly reflected in prices without the need for trading activity, while confidential information is reflected in unusual orders (surplus Bid and Ask orders).

Figure no. 1 shows informed trade mechanism. Assuming that α is the probability of an informational event, then δ is the probability of bad news, and 1- δ is the probability of good news on a specific day. If no informational event (with probability 1- α) occurs, only uninformed traders (Liquidity) will trade in the market. On this day, the arrival rate of uninformed traders (what to buy or sell) will have an independent Poisson distribution with probability ϵ . Informed traders will be willing to trade only in the event of an informational event (with probability μ). If they receive good news, they will buy and, if they receive bad news, they will sell their stocks. For informed traders on a specific day, if the occurrence is a bad informational event (with probability $\delta \alpha$), the arrival rate of buy orders (α) will be less than that of sell orders ($\epsilon + \mu$). If the occurrence is a good informational event (with probability $\alpha(1-\delta)$), the arrival rate of buy orders ($\epsilon + \mu$) will be greater than the sell orders (ϵ). Easley and O'Hara (1992) stated that the probability of trading based on information for a specific stock (PIN_i) as defined by the estimated arrival rate of informed trading divided by the estimated arrival rate of all trades in a specific day is as follows (Vega, 2006):



Figure no. 1 – Tree diagram of the trading probability (Easley and O'Hara, 1992)

Easley *et al.* (1996) stated that, under these conditions, a market maker is assumed to be Bayesian if he uses the information to trade and update his expectations about good news, bad news, or no news. At time t, before the start of trading, market maker expectation for the probability of no news, good news, or bad news are as follows:

$$P_n(t) = 1 - \alpha, \tag{6}$$

$$P_{q}(t) = \alpha(1-\delta), \tag{7}$$

$$P_b(t) = \alpha \delta \tag{8}$$

Therefore if B and S Represent buy and sell orders at time t, $P(t|S_t)$ is Represents the market maker's updated expectation when a sell order arriving, $P_n(t|S_t)$ is expectation about no news when a sell order arriving, $P_b(t|S_t)$ is his updated probability of bad news when a sell order arrives, and $P_g(t|S_t)$ is his updated probability of good news when a sell order arriving at time t. Probabilities when a buy order arriving are represented in a similar way. According to Easley *et al.* (1996), using Bayes's rule, his posterior probability on no news when a sell order arrives at time t is:

$$P_n(t|S_t) = \frac{P_n(S_t|t)P_n(t)}{P(S_t)} = \frac{P_n(S_t|t)P_n(t)}{P_n(S_t|t)P_n(t) + P_g(S_t|t)P_g(t) + P_b(S_t|t)P_b(t)} = \frac{\varepsilon_s P_n(t)}{\varepsilon_s + \mu P_b(t)}$$
(9)

The posterior probability on bad news and good news are the as follows respectively:

$$P_b(t|S_t) = \frac{(\varepsilon_S + \mu)P_b(t)}{\varepsilon_S + \mu P_b(t)},$$
(10)

$$P_g(t|S_t) = \frac{(\varepsilon_S)P_g(t)}{\varepsilon_S + \mu P_b(t)}$$
(11)

The zero expected profit bid price is the expected value of the stock at time t. So, the bid is:

$$b(t) = \frac{\varepsilon_s P_n(t) V^* + (\varepsilon_s + \mu) P_b(t) \underline{V_i} + (\varepsilon_s) P_g(t) \overline{V_i}}{\varepsilon_s + \mu P_b(t)}$$
(12)

and the ask is:

$$a(t) = \frac{\varepsilon_b P_n(t) V^* + \varepsilon_b P_b(t) \underline{V}_i + (\varepsilon_b + \mu) P_g(t) \overline{V}_i}{\varepsilon_b + \mu P_g(t)}$$
(13)

Where V^* Represents the value of the stock when no news. So expected value of the stock is:

$$E[V_i] = P_n(t)V^* + P_b(t)V_i + P_g(t)\overline{V}_i$$
(14)

With substituting equation (14) into equation (12) and (13):

$$b(t) = E[V_i] - \frac{\mu P_b(t)}{\varepsilon_S + \mu P_b(t)} (E[V_i] - \underline{V_i})$$
(15)

$$a(t) = E[V_i] + \frac{\mu P_g(t)}{\varepsilon_b + \mu P_g(t)} (\overline{V}_i - E[V_i])$$
(16)

These equations considering the stock expected value, improve analysis of entering role of informed and uninformed investors in affecting trading prices. Now spread defined by the difference between ask and bid prices:

$$\Sigma(t) = a(t) - b(t), \tag{17}$$

$$\Sigma(t) = \frac{\mu P_g(t)}{\varepsilon_b + \mu P_g(t)} \left(\overline{V}_i - E[V_i] \right) + \frac{\mu P_b(t)}{\varepsilon_s + \mu P_b(t)} \left(E[V_i] - \underline{V}_i \right)$$
(18)

With these explanation, All four parameters that used in PIN formula ($\theta = \{\alpha, \delta, \mu, \varepsilon\}$), estimated by maximizing following likelihood function:

$$L(\theta|(B_i, S_i)) = \alpha(1 - \delta)e^{-(\mu + \varepsilon_b)} \frac{(\mu + \varepsilon_b)^{B_i}}{B_i!} e^{-\varepsilon_s} \frac{\varepsilon_s^{S_i}}{S_i!} + \alpha \delta e^{-\varepsilon_b} \frac{\varepsilon_b^{B_i}}{B_i!} e^{-(\mu + \varepsilon_s)} \frac{(\mu + \varepsilon_s)^{S_i}}{S_i!} + (19)$$
$$(1 - \alpha)e^{-\varepsilon_b} \frac{\varepsilon_b^{B_i}}{B_i!} e^{-\varepsilon_s} \frac{\varepsilon_s^{S_i}}{S_i!}$$

That for determine the direction of trading, used Lee and Ready (1991) algorithm (Vega, 2006). Also, in order to estimate parameter vector θ , for collection of data related to buy and sell ($\{M = (B_t, S_t)\}_{t=1}^T$), can be used the product of the daily likelihoods, as follow:

$$L(\theta|M) = \prod_{t=1}^{T} L(\theta|B_t, S_t)$$
(20)

As previously mentioned, according to Easley and O'Hara (1987 and 1992) the higher value of PIN (in the range of 0 to 1) shows higher levels of information asymmetry.

4. RESULTS

Several indices have been proposed to measuring information asymmetry. The present study used two indices to calculate information asymmetry on the Tehran Stock Exchange. The Tehran Stock Exchange is one of the best examples of developing markets and results of this market can be generalized to other developing stock markets. The PIN index was first calculated and then the forecast error (FE) was used to verify the PIN results.

4.1 Probability of informed trade (PIN)

Table no. 1 shows PIN results for selected firms on the Tehran Stock Exchange as calculated using R software. Some of the data required was not available, thus, some values were not calculated. The PIN values ranges from 0 to 1. For some years and firms, the PIN values approached zero and for others they approached 1. That means that, in some cases, information asymmetry was minimal and in others there was relatively total information asymmetry. On average, the firms Saipa (SIPA 1), Kharazmy Invest (IKHR1) and Bahman Group (BHMN1) had the lowest information asymmetry and Khorasan Steel Co. (FKAS1), IRI Marine Co. (KSHJ1) and Isf. Oil Ref. Co. (PNES1) highest during the study period.

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Symbola	Year							
Symbols	2010	2011	2012	2013	2014	2015		
BPAS1	NA	0.166704	0.139762	0.149106	0.271291	0.172495		
BPAR1	0.522269	0.180636	1.45E-06	0.113892	0.284433	0.0006		
BANS1	NA	4.06E-07	0.219249	0.586407	0.103655	1.19E-06		
NOVN1	0.497494	2.62E-05	0.158441	0.104573	0.159042	0.606513		
IKCO1	0.893349	0.288327	0.231418	0.119108	0.338615	2.03E-07		
HMRZ1	NA	NA	NA	0.198231	0.122192	0.001645		
AZAB1	0.106689	0.050791	0.113496	0.545859	0.349026	0.021220		
PRDZ1	NA	0.666667	6.56E-08	5.24E-07	0.193120	3.06E-06		
PNBA1	NA	NA	0.470499	3.12E-07	0.666667	0.154439		
PNES1	0.704277	0.000475	0.666667	0.413642	0.666667	0.031259		
BMLT1	0.106280	0.110235	0.705329	0.11969	0.296926	0.729434		
KRAF1	0.164573	0.096204	0.495496	0.064234	0.084361	8.29E-06		
BSDR1	0.338596	0.120153	0.199741	0.033096	0.078678	0.506873		
BTEJ1	0.082507	0.166919	0.136061	0.544487	0.126631	0.017420		
IKHR1	NA	NA	1.10E-06	0.133666	0.152341	1.90E-06		
TMEL1	1.44E-07	0.004674	0.131624	0.134457	0.119897	0.085609		
MADN1	0.096535	0.666667	0.107911	3.37E-07	0.305857	1.08E-07		
SIPA1	2.78E-07	0.133732	0.181860	0.0339	3.08E-07	0.034985		
INFO1	0.176605	0.179613	0.109613	0.533459	0.354425	0.091472		
CHML1	7.21E-06	0.131999	0.666667	0.046018	0.324008	0.230670		
PKHA1	0.666667	0.421011	2.06E-06	0.117356	0.089161	3.84E-07		
FKAS1	0.537248	0.457724	0.690347	0.848530	0.805267	0.999999		
FOLD1	2.20E-06	0.089611	0.222156	1.40E-05	0.223613	0.106802		
FKHZ1	0.462570	0.666667	3.47E-07	0.666667	0.273945	0.091927		
PKLJ1	NA	NA	NA	0.161692	0.189887	0.350816		
GDIR1	0.679264	0.020087	0.123194	0.471598	4.06E-07	0.356946		
SAND1	NA	NA	NA	0.666667	8.99E-08	NA		
DTIP1	NA	NA	NA	0.666667	0.144601	9.66E-08		
MSMI1	0.148367	0.003161	0.082847	0.081209	0.211537	0.070261		
MKBT1	0.104182	0.029881	0.182249	0.088111	0.179157	6.83E-06		
GOLG1	0.457707	0.244731	0.666667	3.24E-07	0.087778	0.242175		
PASN1	NA	0.283843	0.160343	0.135109	0.219111	0.079447		
MAPN1	0.411996	0.226090	2.47E-06	0.955160	0.159202	0.087237		
BHMN1	0.157715	0.134182	0.056006	0.085225	2.88E-07	1.37E-07		
KSHJ1	0.443594	0.813148	0.749914	0.717575	0.607927	8.60E-08		

Table no. 1 – PIN index for selected firms in Tehran Stock Exchange

The average PIN value for SIPA was 0.06, which suggests very low information asymmetry in the market. In reality, this firm has a good chance with investors on the Tehran Stock Exchange. Six firms had PIN values of less than 0.1 (SIPA1, IKHR1, BHMN1, TMEL1, MKBT1, MSMI1). The PIN value for Khorasan Steel Co. (FKAS1) was approximately 0.72, which is significant for a large and active firm on the Tehran Stock Exchange; only three firms recorded PIN values greater than 0.40 (FKAS1, KSHJ1 and PNES1).

Year	2010	2011	2012	2013	2014	2015
Average	0.310340	0.219102	0.247341	0.272440	0.233972	0.149125

Table no. 2 – PIN index in the years of study

Table no. 2 shows that the PIN index in 2010 had a maximum coefficient and in 2015 had minimum coefficient. In 2015, information asymmetry dropped considerably, which could have been in response to the increasing informational level of traders and good news in market that improved trader expectations about investing on the Tehran Stock Exchange. Positive political events played a strong role in this case. The PIN value has a direct relationship with instability on the Tehran Stock Exchange.

4.2 Earning forecast error (FE)

The FE index was used to calculate information asymmetry to verify the PIN results. Lang and Lundholm (1996) shows that increasing informational transparency improves forecast accuracy and decreases forecast dispersion and can be used as a measure of earnings forecast error to calculate information asymmetry. Using the method proposed by Christie (1987), the FE was measured as the ratio of absolute difference between forecast earnings (ACT_t) and actual earnings per share (EST_t) to calculate information asymmetry as:

$$FE = \frac{ACT_t - EST_t}{ACT_t}$$
(21)

Firms with a larger level of information asymmetry have higher FE (Krishnaswami and Subramaniam, 1999). Table no. 3 shows the FE for selected firms on the Tehran Stock Exchange. The FE value for the earnings for selected firms on the Tehran Stock Exchange were unusual; in 13 cases, the FE were more than 1000%. In 20 cases, the FE was 9% to 20%; these are the lowest FE values. Earnings data was not available for 2015 as the research was completed; therefore, the FE was not calculated for that year.

On average, Khalij Fars Industries (PKLJ1) recorded the maximum FE; however, this was based on published statistics for only two years (2013≅470 and 2014≅1). Tamin Daroo (DTIP1) ranks second at 370 with 1 FE value for two years. Iran Tele. Co. (MKBT1) and Pension Fund (SAND1) rank third and fourth and show high volatility in earnings FE. The average FE values for these four firms show a huge difference with those of other firms. On average, Tosee Melli Inv. (TMEL1) had the lowest FE; this firm has very stable values throughout the study period. Khouz. Steel (FKHZ1), Chadormalu (CHML1) and Khark Petr. (PKHA1) rank second, third and fourth after Tosee Melli Inv. (TMEL1) for the best FE values.

It is notable that the values presented by these indices do not match when the PIN and FE results are used as indicators of information asymmetry. For example, Khalij Fars (PKLJ1) (2013), Tamin Daroo (DTIP1) (2013) and Iran Tele. Co. (MKBT1) (2011) had the highest FE values, respectively, but ranked 84th, 14th and 149th rates for PIN. Conversely, Khark Petr. (PKHA1) (2014), Khouz. Steel (FKHZ1) (2010) and Iran Khodro (IKCO1) (2014) had the lowest FE values, but ranked 62th, 151th and 140th, respectively, with low PIN values.

Symbols	Year							
Symbols	2010	2011	2012	2013	2014			
BPAS1	-3.17460	-1.02454	-0.85714	-0.74771	-0.42249			
BPAR1	-1.17413	-1.36190	-0.61357	-0.54394	-0.32962			
BANS1	-0.36916	-0.36916	-1.16736	-0.92444	-0.46409			
NOVN1	-1.15880	-0.57307	-0.63743	-0.61694	-0.54211			
IKCO1	-0.89754	-0.28622	-0.63212	-0.33753	-0.11828			
HMRZ1	-0.99365	-0.35756	-3.08496	-0.34830	-0.31941			
AZAB1	-0.38053	-1.11111	-0.25126	-0.27083	-0.38636			
PRDZ1	NA	-0.35786	-0.47113	-0.36284	-0.78516			
PNBA1	0.245902	0.245902	-1.22388	-1.08600	-0.50023			
PNES1	-3.03448	-0.68269	-0.35051	-2.01127	-0.66791			
BMLT1	-0.26180	-0.33043	-1.00000	-1.00000	-0.48889			
KRAF1	-0.58867	-0.32558	-0.47737	-0.61472	-0.14085			
BSDR1	-0.18072	-0.27841	-0.71622	-0.98182	-0.47857			
BTEJ1	-0.41414	-0.71212	-2.43137	-0.91736	-0.64198			
IKHR1	NA	-0.65248	-4.20833	-4.20833	-0.80800			
TMEL1	-0.18182	-0.18182	-0.18182	-0.18182	-0.15385			
MADN1	-2.04473	-1.42045	-3.29255	-2.73377	-0.86141			
SIPA1	-2.66851	-1.59434	-0.32292	0.16000	-6.66667			
INFO1	-15.0590	-31.5926	-1.89107	-1.68209	-0.65069			
CHML1	-0.27061	-0.33645	-0.29078	-0.17532	-0.23200			
PKHA1	-0.55014	-0.19357	-0.14468	-0.32991	-0.09555			
FKAS1	-0.66953	-0.18956	-0.28514	-0.12220	-0.19789			
FOLD1	-0.13043	-0.41608	-0.42125	-0.13521	-0.22441			
FKHZ1	-0.10689	-0.33138	-0.25146	-0.20570	-0.19716			
PKLJ1	NA	NA	NA	-469.667	-0.99808			
GDIR1	-1.79231	-19.4286	-3.24537	-5.89535	-0.39162			
SAND1	-0.79245	-152.333	-42.7273	-59.0000	-0.94188			
DTIP1	NA	NA	NA	-370.500	-1.00602			
MSMI1	-1.07855	-0.30898	-0.46531	-0.61538	-0.22289			
MKBT1	-52.1250	-168.500	-58.4286	-15.0909	-1.02804			
GOLG1	-0.81166	-0.72078	-0.22617	-0.28709	-0.25086			
PASN1	NA	-1.00000	-1.68103	-1.27740	-0.98929			
MAPN1	-0.34896	-1.15315	-1.71978	-1.88304	-0.64271			
BHMN1	-0.25000	-1.26481	-5.93443	-5.64063	-0.87055			
KSHJ1	-27.5000	9.571429	-6.40000	-3.50000	-1.06338			

Table no. 3 – FE index for selected firms in Tehran Stock Exchange

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The difference between mechanisms of PIN and FE indicated that the FE index calculates information asymmetry at the beginning of a period and the PIN index calculates information asymmetry over time. If a logical relation between these indices can be found that covers both mechanisms, a new more accurate index can be developed. This would be more suitable for emerging markets such as the Tehran Stock Exchange in which firm performance does not show great stability and political issues and outside influences have a strong effect on stock trends.

4.3 New composite index (C-PIN-FE)

The PIN is used in many financial studies, but cannot solely represent information asymmetry in emerging markets such as the Tehran Stock Exchange. The results found in the previous sections illustrate this; firms with very low PIN values show relatively high FE values. This means the forecasts differed from actual earnings significantly, which denotes high information asymmetry. Also, earnings FE cannot represent information asymmetry in stock markets properly because it does not directly consider information traders. This suggests the need for a composite index of PIN and FE that can better measure information asymmetry in emerging stock markets, especially during turbulent years. This composite index (C-PIN-FE) can be calculated as follows:

$$C-PIN-FE = \frac{|FE|}{(Ln(PIN))^2} = \frac{|ACT_t - EST_t|[Ln(\frac{\alpha\mu}{\alpha\delta+2\varepsilon})]^{-2}}{|ACT_t|}$$
(22)

Because the PIN value falls between 0 and 1, but FE can be a number of any size, the PIN should be adjusted in the new index. The natural logarithm of PIN is, thus, obtained as a negative number and its value will be weighed as Ln(PIN) (Now larger numbers represent less information asymmetry). Because the goal is to measure information asymmetry and not to show the direction of important information, the absolute value of FE is used. This new composite index modifies the previous indices and offers a better measure of information asymmetry with emphasis on the characteristics of each index. It considers information, initial expectations, and adjusted information over time.

The results of this index for selected firms on the Tehran Stock Exchange are shown in Table no. 4. In the C-PIN-FE index, Tosee Melli Inv (TEMEL1) (2010), Mobarakeh Steel (FOLD1) (2010) and Khark Petr. (PKHA1) (2012) rank lowest; these firms in the low PIN values rank 7th, 25th and 24th, respectively, and in the low FE values rank 15th, 9th and 12th. In the C-PIN-FE, Tamin Daroo (DTIP1) (2013), MAPNA (MAPN1) (2013) and Pension Fund (SAND1) (2013) rank highly; in the high PIN values, they rank 13th, 2th and 14th and in the high FE values rank 2th, 33th and 5th, respectively. These results are more reasonable than the previous comparison between their PIN and FE values.

On average, the Tosee Melli Inv. (TMEL1) had the lowest value in the C-PIN-FE index, but in the previous indices, this firm had low values and showed volatility. The informational environment around this firm has been safer for investors because earnings forecasts and investor performance regarding this firm better match expectations. In the C-PIN-FE, Mobarakeh Steel (FOLD1), Iran Mobil Tele (HMRZ1), Saipa (SIPA1) and I. N. C. Ind. (MSMI1) rank second, third, fourth and fifth after Tosee Melli Inv. (TMEL1). On average, Tamin Daroo (DTIP1), Pension Fund (SAND1), MAPNA (MAPN1), IRI Marine Co. (KSHJ1) and Khalij Fars (PKLJ1) rank highest in this index, meaning there is a high level of information asymmetry about their performance in the market. On average, 2013 showed the highest level of information asymmetry; this was a turbulent year on the Tehran Stock Exchange, which justifies the high C-PIN-FE values for this year. In 2014, C-PIN-FE values decreased dramatically because of stagnation on the Tehran Stock Exchange and there was little good news to encourage investors to trade. The market conditions and PIN index indicates that the C-PIN-FE index for 2015 will have low values, although it was not calculated at the time of the study.

Symbols	Year					
	2010	2011	2012	2013	2014	
BPAS1	NA	0.319211	0.221352	0.206448	0.248250	
BPAR1	2.782668	0.465060	0.003395	0.115247	0.208527	
BANS1	NA	0.001705	0.506899	3.245037	0.090327	
NOVN1	2.377299	0.00515	0.187791	0.121016	0.160367	
IKCO1	70.56741	0.185052	0.295121	0.074555	0.100865	
HMRZ1	NA	NA	NA	0.132991	0.072279	
AZAB1	0.075986	0.125116	0.053064	0.738969	0.348708	
PRDZ1	NA	2.176708	0.001722	0.001735	0.290348	
PNBA1	NA	NA	2.152979	0.004839	3.042717	
PNES1	24.68897	0.011660	2.132030	2.581020	4.062691	
BMLT1	0.052099	0.067954	8.205818	0.221902	0.331573	
KRAF1	0.180803	0.059395	0.968132	0.081568	0.023037	
BSDR1	0.154099	0.062005	0.276055	0.084518	0.074039	
BTEJ1	0.066535	0.222192	0.611105	2.482312	0.150333	
IKHR1	NA	NA	0.022358	1.039150	0.228214	
TMEL1	0.000733	0.006315	0.044217	0.045160	0.034194	
MADN1	0.374113	8.640132	0.664215	0.012306	0.613820	
SIPA1	0.011712	0.393876	0.111145	0.013969	0.029651	
INFO1	5.009322	10.71692	0.386910	4.260027	0.604782	
CHML1	0.001930	0.082051	1.768715	0.018496	0.182663	
PKHA1	3.346306	0.258647	0.000844	0.071869	0.016353	
FKAS1	1.734494	0.310386	2.076537	4.529597	4.218709	
FOLD1	0.000769	0.071503	0.186133	0.001083	0.100026	
FKHZ1	0.179841	2.015644	0.001137	1.251202	0.117597	
PKLJ1	NA	NA	NA	141.4702	0.361620	
GDIR1	11.98289	1.272322	0.740137	10.43525	0.001808	
SAND1	NA	NA	NA	358.8765	0.003578	
DTIP1	NA	NA	NA	2253.623	0.269027	
MSMI1	0.296248	0.009323	0.075004	0.097622	0.092374	
MKBT1	10.19083	13.67267	20.16099	2.557425	0.347703	
GOLG1	1.32888	0.363786	1.375735	0.001286	0.042380	
PASN1	NA	0.630546	0.501724	0.318817	0.429217	
MAPN1	0.443792	0.521636	0.010320	894.6986	0.190338	
BHMN1	0.073286	0.313512	0.714336	0.930229	0.003837	
KSHJ1	41.62129	223.7174	77.26977	31.77689	4.292922	

Table no. 4 – C-PIN-FE index for selected firms in Tehran Stock Exchange

A review of the results of this new composite index and comparison with the previous indices show it presents a better measure for information asymmetry in markets such as the Tehran Stock Exchange. Where the PIN or FE are not able to explain and describe information asymmetry in markets on their own, the use of the C-PIN-FE index is recommended.

5. CONCLUSION

One reason for a breach in competition in the stock market is information asymmetry. In emerging stock markets with high volatility, trader information often differs. This holds true for the Tehran Stock Exchange. The present study measured information asymmetry in large and active firms on the Tehran Stock Exchange because investors have more information about these firms and a review their informational environment can be offered as a good indicator of the informational environment. The PIN and FE indices were initially used and it was shown that the PIN index, on average, showed information asymmetry of more than 20%. The PIN index predicts that information asymmetry will decrease considerably in 2015, because of political changes and improved market expectations.

To verify these results, the FE index was implemented, but presented unusual results. The FE values indicated that earnings forecasts were very different from actual earnings (in some cases more than 1000%). A comparison of the PIN and FE values reveals mismatched results when ranking firms based on information asymmetry. Because the FE index shows information asymmetry at the beginning of a period and the PIN index shows information asymmetry over time, a new composite index (C-PIN-FE) that uses the results and mechanism of both indices was developed. The C-PIN-FE index offers more reasonable results for ranking firms based on information asymmetry and corresponds with both PIN and FE results. It is suggested that the C-PIN-FE index more accurately measures information asymmetry in emerging, turbulent, and highly volatile stock markets showing large differences in earnings forecasts and dual functions.

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APPENDIX

Symbols, names and webpages of large active firms on the Tehran Stock Exchange

Symbol	Name	Webpage	Symbol	Name	Webpage
AZAB1	Azarab Ind.	*AZAB0001	INFO1	Inf. Services	*INFO0001
BANS1	Ansar Bank	*BANS0001	KRAF1	Karafarin Bank	*KRAF0001
BHMN1	Bahman Group	*BHMN0001	KSHJ1	IRI Marine Co.	*KSHJ0001
BMLT1	Mellat Bank	*BMLT0001	MADN1	Metals & Min.	*MADN0001
BPAR1	Parsian Bank	*BPAR0001	MAPN1	MAPNA	*MAPN0001
BPAS1	Pasargad Bank	*BPAS0001	MSMI1	I. N. C. Ind.	*MSMI0001
BSDR1	Saderat Bank	*BSDR0001	MKBT1	Iran Tele. Co.	*MKBT0001
BTEJ1	Tejarat Bank	*BTEJ0001	NOVN1	EN Bank	*NOVN0001
CHML1	Chadormalu	*CHML0001	PASN1	Parsian Oil&Gas	*PASN0001
DTIP1	Tamin Daroo	*DTIP0001	PKHA1	Khark Petr.	*PKHA0001
FKAS1	Khorasan Steel Co.	*FKAS0001	PKLJ1	Khalij Fars	*PKLJ0001
FKHZ1	Khouz. Steel	*FKHZ0001	PNBA1	B.A Oil Refinie	*PNBA0001
FOLD1	Mobarakeh Steel	*FOLD0001	PNES1	Isf. Oil Ref. Co.	*PNES0001
GDIR1	Ghadir Inv.	*GDIR0001	PRDZ1	Pardis Petr.	*PRDZ0001
GOLG1	Gol-E-Gohar.	*GOLG0001	SAND1	Pension Fund	*SAND0001
HMRZ1	Iran Mobil Tele	*HMRZ0001	SIPA1	Saipa	*SIPA0001
IKCO1	Iran Khodro	*IKCO0001	TMEL1	Tosee Melli Inv	*TMEL0001
IKHR1	Kharazmy Invest	*IKHR0001			

Note: * http://new.tse.ir/en/Instrument.html?IRO1 (repeated information)