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Probability of Price Reversal and Intraday Trading Activity on Tick Size-25 at Indonesia Stock Exchange

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Abstract

In this paper, I examine the probability of price reversal and the relationship with intraday trading activity in market microstructure context, focusing specifically on interrelate with stock's return, trading volume, frequency of transaction, volatility and liquidity on high tick size (IDR25). This research use purposive random sampling to get some observed samples and data online capturing. Trading fraction of tick-by-tick of trading activity is 15 minutes. To determine price reversal probability, I employ multiple logistic regressions in high frequency data. The conclusion of this research finds that stock's return, trading volume (V), frequency of transaction (F) and proxy V/F affect significantly on probability of price reversal on information uncertainty condition. These are substantial findings of intraday trading activity and overreaction concept at Indonesia Stock Exchange.

Keywords: probability, reversal, intraday, trading, IDX.

Abstrak

Dalam penelitian ini, diujikan hubungan probabilitas price reversal dengan aktivitas intraharian perdagangan saham di Bursa Efek Indonesia (BEI) dalam konteks mikrostruktur pasar. Riset berfokus kepada interrelate imbal hasil saham, volume perdagangan, frekuensi transaksi, volatilitas dan likuiditas pada tick size 25. Kajian ini menggunakan purposive random sampling untuk mendapatkan observed sampling dan online capturing. Fraksi perdagangan sebagai landasan tick-by-tick dari aktifitas perdagangan yang digunakan adalah 15 menit. Untuk mendapatkan probabilitas price reversal, saya menggunakan multiple logistic regression in high frequency data analysis. Konklusi dari riset ini

menunjukkan bahwa imbal hasil saham, volume perdagangan, frekuensi transaksi, dan proksi V/F memberikan efek signifikan terhadap price reversal.

Kata kunci: probabilitas, reversal, intra-harian, tick size, BEI.

1. Introduction

In this paper, I examine about probability of price reversal and the relationship with intraday trading activity in market microstructure context, focusing specifically on interrelate with return, trading volume, frequency of transaction, volatility and liquidity on high tick size. Price reversal issuance is subject as apart of intraday trading activity that often happened in any capital market around the world especially in emerging market like Indonesia Stock Exchange (IDX). This phenomenon caused by high price volatility and liquidity pressure of intraday trading that driven by capital market reaction to uncertainty of news or information arrivals. Higher level of probability price reversal can generate higher risk intraday trading because of price direction often change suddenly.

This circumstances may create more profit taking or cut loss activity because of the price direction is harder to predict by traders in uncertainty information condition. However, behind risky intraday trading activity there are plenty opportunity to make money from intraday capital gain consistently (Markman, 2000). The phenomenon of price reversal is often happens at any emerging market in the world especially at Indonesia Stock Exchange (IDX). This phenomenon potentially creates high risk trading which caused by overreaction and liquidity pressure. The purposes of this research are to examine the probability of price reversal and intraday trading activity based on high frequency data.

What do determines price reversal on trading activity? Experimental and survey evidence indicate that in probability revision problems, people show a tendency to over-react. The traders may have overweight recent information and underweight base rate data. Theoretically, the standard of researcher paradigm of stock price movement in capital market has reflected all informations which arrived into the market either public or private information.

The efficient market hypothesis is based on the belief where the past price patterns in security movement are consistent with a random walk. Two other efficiency-based hypotheses have been identified at the recent literature. The over-reaction hypothesis (OH) suggests that extreme movements in equity prices are followed by "corrective" movements in the opposite direction of the initial over-reaction (Rartner & Leal, 1998), meanwhile, stock price changes in one direction are followed by changes in opposite direction (reversal).

Kofman & Moser (2001) obtained a measure of the frequency with which prices may have over-reacted to new information. Over-reaction followed by price correction is a pattern which is consistent with what is termed *fad trading*. Fad trading is buying or selling equity based less on information about the value of assets than on the fact that buying or selling is the things to do. The idea of fad-motivated trading is described as prices being the result of traders "getting on the bandwagon" as opposed to independently arrived at judgments about the true value of these assets.

Practically, stock price at IDX is often in overvalued or undervalued condition because of some factors, such as over-reaction and uncertainty in the value of market fluctuating. The uncertainty makes traders and/or investors tend to over-reacting valuing stocks (assets). This circumstance makes the asset price is not in fair value (or intrinsic value). In other word, stock price in capital market tends to be mispriced. Stock prices which are in mispricing condition either in undervalued or overvalued will be followed by price correction process in the opposite direction. This price correction will create price reversal repeatedly.

Base on a descriptive observation, stocks of all tick sizes at IDX show probability of price reversal potency ranges from 7.7 to 24.3% in per 15 minute period. The general pattern is referred that potency of price reversal at equities market is a big enough and it will create equity trading risk in the short term period. The numbers of price reversal are around 14 times a day, either major or minor price reversal for each liquid stock. There is every stock create average more than 9 times price reversals that offer advantage from intraday trading of financial assets.

The volatility of trading volume was also high enough to create liquidity pressure to the market and increase the investment risk also. In this circumstances, capital market can be categorized not conducive for stock trading and short-term investment because the trend of price direction may change quickly (Lyons, 2001). However, for the traders that transact stock in very short time (hourly) on the same day (intraday trading), this market fluctuation may offer the plenty opportunity to get gain (Cassidy, 2002). This paper reframes the factors which determine probability of price reversal such as stock return and intraday trading activity variables. This research uses high frequency finance data at tick size IDR5 that represented by stock of Bank of International Indonesia, Tbk (BNII.IJ) as a sample. We ask whether stock return and micro-structure variables effect on the over-reaction associated with probability of price reversal.

This research is focused on main problems that are influenced by intraday trading activity in the market micro-structure context on probability of price reversal. The problems of formulation are conducted in order to every problem that can be discussed and solved that compiled in the stock return, trading volume, frequency of transaction, proxy trading volume per frequency, price volatility and liquidity influence on probability of price reversal of tick size-5 on intraday trading at IDX.

Zhang (2006) explains that substantial of short-term stock price continuation, which the prior literature often attributes to investor underreaction to new information. There are include positive serial correlation of returns at 3 to 12 month horizons (Jagadeh & Titman, 1993), and Daniel, Hirshleifer & Subrahmanyam (1998) review that post earnings announcement stock price drift in the direction indicated by earnings surprise, and the post-event return drift in the direction of announcement date return. Uncertainty information contributes to this probability of price reversal phenomenon in intraday trading activity. The ambiguity with respect to the implications of new information for value of the firm, that potentially stems from two sources: volatility underlying fundamentals value and poor analytical skill and poor information (Zhang, 2006).

Theoretically, movement of stock price in capital market has reflected all quality of informations that just arrived into market, either public or private information (O'Hara, 2005). Nevertheless, practically, stock price frequent experiences of overvalued or undervalued because of many factors, especially

uncertainty in the form of fundamental value and irrational market fluctuating. My main hypothesis is that if equities price formation does not represent its fair value (intrinsic value), stock price always in mispricing condition. Stockprice that experienced of mispricing will searching its fair value by price correction process that causing price reversal repeatedly. My hypothesis is motivated by prior results from the behavioral finance on microstructure theory especially O'Hara (2005) and Hirshleifer (2001).

This general pattern forming price reversal consistently and creates difference (gap) or mismatch between theory and empirical facts (Kofman & Mosser, 2001). ¹ This formulation of research problems are based on preliminary study on samples stock with high tick size at IDX. We examine the probability of price reversal and the relationship with intraday trading activity in market microstructure context, focusing specifically on interrelate with stock's return, trading volume, frequency of transaction, volatility and liquidity on tick size -25.

This research aims to explain how significant return and trading mechanism affected the price formation process and overreaction to make price reversal when the mispricing has to be corrected by market:

- a. Construct the framework of influence of return and intraday trading activity that consist of volume, frequency, proxy V/F, volatility and liquidity to probability price reversal on high tick size.
- b. Analysis the pattern intraday trading activity on high tick size and construct model in econometrics form as the intraday trading guidance and analysis technical tool for all traders, alternatively.

2. Related Literature and Research Paradigm

The price reversal process is the microstructure study of market efficiency. The overreaction process and price reversal is in market microstructure theory context, caused by the influence of order flow that carry new information content to move equity price on imbalance order effect. Kofman & Moser (2001) suggest that relation/link between conditional probability of price reversal and level of stock return and some other variables on logit model framework. They introduce a model in which prices are determined by two investing clienteles: *informed investors* and *noise-trading investors*.

Furthermore, Kofman & Moser (2001) explain that by studying the frequency with which stock price changes in one direction are followed by changes opposite direction (price reversals), they obtain a measure of the frequency with which prices may have overreacted to new information. Overreaction followed by price correction is a pattern that is consistent with what is termed *fad trading*.

To understand price discovery and reversal in context that how prices incorporate information properly, we have to consider about trading mechanism and price formation process seriously. In trading process—in particular the order flow—as a information filtration. Walsh (1997) argue that at any trading time, any 'news' that arises from a particular order or price will add to the existing information set.

¹ An observed signal (*s*) is characterized as a firm's fundamental value (*v*), such as future cash flow or dividend, plus noise term (*e*), that is $s=v+e$. The variance of signal measures information uncertainty: $var(s)=var(v)+var(e)$, where $var(v)$ is a firm's underlying fundamental volatility and $var(e)$ reflects the quality of information.

Based on the market microstructure and the respond time of intraday traders; price will respond to news or information that arrives into market as it changes the structure of the information set. In this circumstance, we focus on the speed of reaction of traders to news and extent of this price response that we considered in this research.

Summer (1986) explains that the presence of a fad component in the determination of stock prices implies that stock prices will reverse as fads dissipate. Stoll & Whalley (1990) enhance that behavior of price volatility are related to frequency of price reversal, determined also by level of capital market efficiency. High non trading probabilities would be encountered were trading activity concentrated in short time frames and, therefore, more likely motivated by similar information [(Kofman & Mosser (2001))]. This suggests that the higher stock return increase the probability of nontrading activity.

Kofman & Moser (2001) conducted three approaches to explain whether stock return level affects on trading activity, that produce:

- ☐ First: frequency of (in graphical) price reversal shows that percentage reversal have negative relation/link to level of stock return.
- ☐ Second: the mean time between price reversals also show negative relation/link to level of stock return.
- ☐ Third: the usage of model logit specification proves that probability of price reversal have negative relation/link to level of stock return. According to Chan *et al.*, (1995), the risk of price reversal is caused by volatility and liquidity pressure which market index change direction (rebound) suddenly after sharp increase or decrease. Bondt & Thaler (1985) explain that if stock prices systematically overshoot, then their reversal should be predictable from past return data alone, with no use of any accounting data such as earnings. Specifically, two hypotheses are suggested: (1) Extreme movements in stock prices will be followed by subsequent price movements in the opposite direction. (2) The more extreme the initial price movement, the greater will be the subsequent adjustment. Both hypotheses imply a violation of weak-form market efficiency.

Capital market with weak to semi strong form efficiency, often happens panic buying that stock price increase sharply, and then it can change direction becomes panic selling instantly, until stock price go down drastically (*noise trading*). On the contrary, after market index rebound, then panic selling changes become panic buying in just several hours period. Phenomenon of noise trading is frequent happened as consequence of information which entry to the market that countered redundantly (overreaction) [Black (1986); Kofman & Moser (2001)]. These opinions give contribution for this research that emerging market like IDX has huge price reversal potency caused by market microstructure noise.

Although trading trend direction fickle and difficult to predict, however trading activity remain to work. To know of this truly caused of this course is not easy job. The factors of this phenomenon include financial and non-financial. Beside behavior that include of mind set, knowledge, emotion, and perception of traders and investors. One factor that most significant is the arrival of news into market, especially when the news is unexpected information or dramatic news (Bowman & Iverson, 1998). In stock prices movement, market tends to too responsive and will be overreacted until disagree with new price prediction that implied in information that arrive into market [O'Hara (2005), Bowman & Iverson (1998)].

Nevertheless movement of stock price direction not only depend on content of news or information that accepted but also the market perception and analysis quality of traders/investor to get the intrinsic value of stock (Madhavan, 2000). If information is assumed as a good news, then price will go up, otherwise, the stock price will go down. Interesting perceived is how high the stock price can go up or downward fairly in accordance with information content that brought by news/public information. It is depend on market efficiency level.

According to Bowman & Iverson (1998), market tends to be overvalued or undervalued as consequence of market overreaction phenomenon. Bondt & Thaler (1987) conjectured that, a consequence because of investor overreaction to earnings, stock prices may also temporarily depart from their underlying fundamental values. Furthermore, stock price diverge from its fundamental value and the overreaction is an advantage for informed traders to make capital gain (Black, 1986). Because of investor overreaction to new information that arrives to the market, the stock price will be corrected (*price correction*) to the efficient value after create price reversal first. Thus the more often stock price are overpricing or underpricing in market, will be the more price reversal (Zhang, 2006).

Kofman & Mosser (2001) find that time period required to find the its fair value after conduct overreaction is vary, depend on quality analysis and rationality of traders it self. That means, price correction depend on level of market efficiency in instructing price matching with its fundamental value. In general, more efficient a market will make time requirement for price correction will shorten. Where as at the market that is emerging like IDX, price correction will take time longer than the efficient market like NYSE or TSE (Chan *et al.*, 1995). Price forming to its fair value influenced by infrequent trading or nontrading period of the market also (Lo & MacKinlay, 1990). Fama (1970) argued that price reversal was induced by imbalanced order and what kinds of market microstructure that instituted by exchange.

According to Easley & O'Hara (2003), the market information that can be public information only, not including private information who owned by informed strategic traders. Superior private information which owned by some of strategic traders was extracted by trading activity and leak to the uninformed traders. But the informed traders get the capital gain first because more strategic than uninformed traders anyway (Kyle, 1985). The informed trader will create price volatility; stimulate overreaction and price reversal in the right time because they know how the fair value is (Copeland & Galai, 1983). O'Hara (2005) explains the latter concept of trading gain, however, suggest otherwise: information cost will make this average investors actually loss money relative to the market return over time. This information loss arises because of the presence in the market of traders who have superior information. In particular, the market maker who is in the middle of all trades knows that some traders may have better information than he does.

These informed traders buy when they know the stock's current price is too low; they sell when they know it is too high. More over, these informed traders have the option not to trade, unlike the market maker. The market maker knows that when he is trading with informed traders he always loses. To remain solvent, the market maker must be able to offset these losses by making gain from uninformed traders. This gain arises from bid-ask spread. In consequence of strategic traders activity, the movement of stock price direction in general started by informed traders that make efficient price forming, then will followed by uninformed traders in *fad trading* activity that create overreaction and price reversal will be followed (Summers, 1986).

Despitefully, according to Kofman & Mosser (2001) there is also other market reaction which called *lag learning*, that is price reaction which is on generally "late" to respond the content of new information that arrives to the capital market. Its caused by many reason, for example because of the lack of information access and/or weaken its analysis. General illustration for this research that market reaction that emerge to the arriving of public information can be categorized into 3 types, : 1). overreaction, 2). fair value (efficient) and 3) lag learning (underreaction).

2.1. Research Paradigm

Research Paradigm is made arrangements for down process of analysis design that is by connect all independent variables research to probability of price reversal. Independent variables that used already accepted as microstructure and activity of stock trading variables. The compilation of framework idea is conducted base on prior of theory and journals that related to activity of stocktrading intraday and price reversal [Zhang (2006), Chordia *et al.* (2000), Easley & O'Hara (2003) and Chan & Faff (2003), Fama & French (1992), Kyle (1985), Kofman & Moser (2001) and Glosten & Milgrom (1985)]. Independent variables as used in analysis design covers volume trading, frequency, proxy volume per frequency (V/F), volatility and liquidity on high frequency data.

Analysis of regression equation logit is conducted at biggest tick size (Rp. 50) pass by sampel emiten stock high tick size-50. From model reggression analysis this logit are expected give independent pattern of variables influence public to probability price reversal. Analysis is is important conducted as the indication of introduction analysis that give instruction of continuation analysis in determining fit expected research model.

These independent variables are selected from some prior and recent research that relates to phenomenon of price reversal based on:

- According to Kofman & Moser (2001), Summers (1996): level stock return influences on probability of price reversal in the *long-run*.
- Intraday trading activity as independent variables that will be related to probability of price reversal based on:
 - Trading Volume [Chordia *et al.* (2001), Chan & Fong (2000), Odean (1998), Zarowin (1990), Admati & Pfleiderer (1988)]
 - Trading transaction frequency [Juh Lin *et al.* (1999), O'Hara (2005)]
 - Proxy volume/frequency [Juh Lin *et al.* (1999), Zhang (2006)].
 - Volatility of stock price [Koulikov (2004), Chan & Fong (2000), Chang *et al.* (1995) and French & Roll (1986)]
 - Stock Liquidity [Dataret *et al.* (1998) and Chordia *et al.* (2002), Hameed *et al.* (2010)]

Above mentioned researchers use secondary daily data that give explanation and finding in analysis framework of price reversal on a long-term capital market activity. Where as this research, we use high frequency data or tick-by-tick in time framework intraday trading so it's can give finding and different findings or conclusions. Beside the differences of its data character, market structure difference and rule of pricing from NYSE with IDX also give significant influence on analysis results also.

The analysis research model base on the framework above paradigm is designed in the form of *multiple logistic regressions* shall be as follows:

$$Log \left[\frac{\hat{P}_{i,t}}{1 - \hat{P}_{i,t}} \right] = \alpha_0 + \alpha_1 R_{i,t} + \alpha_2 Vol_{i,t} + \alpha_3 Freq_{i,t} + \alpha_4 (V / F)_{i,t} + \alpha_5 Volat_{i,t} + \alpha_6 Liq_{i,t} + u_{i,t} [1]$$

where:

- $R_{i,t}$: Stock return-i on period-t
- $Vol_{i,t}$: Trading volume stock-i on period-t
- $Freq_{i,t}$: Frequency of transaction stock-i on period-t
- $(V/F)_{i,t}$: Ratio volume to frequency of transaction stock-i on period-t
- $Volat_{i,t}$: Price volatility stock-i on period-t
- $Liq_{i,t}$: Liquidity stock-i on period-t

2.2. Hypothesis

The summaries of research hypothesis for high tick size on intraday trading at IDX are:

- 1.Hypothesis 1 (H1): Stock Return has positive affect on probability price reversal.
- 2.Hypothesis 2 (H2): Trading Volume has positive affect on probability price reversal.
- 3.Hypothesis 3 (H3): Frequency of Transaction has an positive affect on probability price reversal.
- 4.Hypothesis 4 (H4): Proxy V/F has positive affect on probability price reversal.
- 5.Hypothesis 5 (H5): Volatility has positive affect on probability price reversal.
- 6.Hypothesis 6 (H6): Liquidity (*proxy* turnover rate) has positive effect on probability price reversal.

3. Research Method

This microstructure research method uses *random purposive sampling* with a purpose to get stock samples at IDX that represents stock population with tick size-25. Sampel stock is required fulfill some criterions so its can reflects stock movement from its population referred [as]. Conditions sampel research covers: : (a) price volatility, (b) liquidity, (c) active stock more than 2 year, (d) big market capitalization (e) good fundamental reputation, (f) leader in its sector, and (g) most active stock list. Sampel research that obtained is PT Aneka Tambang, Tbk (ANTM), PT United Tractors, Tbk (UNTR), PT London Sumatera Plantations, Tbk (LSIP) and PT Timah, Tbk (TINS) as the leading companies on tick size-25 at IDX. These samples besides assessed quite active and liquid stock also represents of dominant mining sectors atIDX.

After sampel stock is got then primary data with drawal (by data capturing) directly from its source that is running trade active transaction. For dependent variable, probability price reversal has the form of data categorical 1 or 0 that counts directly from the intraday charts and the independent variables, which are return, volume, frequency, volume/frequency, volatility and liquidity. The period ot time fraction are 15 minutes as primary data that capyure directly when trading activity is running. Here in after, primary data archives (in primary archival data) transaction real time.This research are conducted February, 1st – April, 30th 2008 period.

²All transactions in the Exchange are processed in a facility called as Jakarta Automated Trading System (JATS). Only the Exchange Members, who also become the members of the Indonesian Clearing and Guarantee Corporation (KPEI), can input the orders into the JATS. The Exchange Members are responsible for every transaction they make in the Exchange and they have the responsibility to settle all the transactions they've made

3.1. Primary Data Sources and Capturing Technique

The Indonesia Stock Exchange (IDX) co-ordinates and employ trading mechanism on equities in Indonesia through a national trading system knowns Jakarta Automated Trading System (JATS)². JATS is a continuos electronic open limit order book for all equities at IDX through which brokers can enter bids and ask throughout the trading day from 09.30am to 04.00pm). Primary data source research that directly capture from real time trading from JATS in regular trading (RG board) in intraday activity.

We will present just a summary of step by step of data gathering here shall be followed:

- Organized a team that trained on how to capture primary data from running trade of IDX from JATS and other intraday trading data in real time. In this first step helped facility by PT Sarijaya Permana Sekuritas, Tbk. Terminal (PC) in the office of each broker allow instant access to this intraday trading activity (such as trading volume, frequency of transaction, bid-ask orders, price, etc) including price reversal charts of equities.
- Source data/information indigenous to running trade IDX that obtainedby real time StockWatch@services throughout day trading.
- Equipments preparation like PC, Modem, Internet network and others must in running well and electrics supply must maintain still on during primary data intake this. In the event of online trading equipments trouble (OLT) that caused technical problem then all previous datas will char.
- Data is captured from StockWatch@ in 15 minute period directly as primary data intraday. In this step calls 'zero error' caused if we make mistake then we have to do the data capturing from beginning again.
- Data subdividing into main file in accordance with stock sampel each. After succeed networked primary data grouping, and then conducts archival primary data from trading process during more-less 1059 data/variable/stock.

4. Analysis and Discussion

4.1. Descriptive Analysis

Descriptive analysis as reported on Table 1 below, shows that probability of price reversal of sampel is 0,2436 or around 24.36%. This condition indicates that stocks of tick size 25, has potency enough to create price reversal and that was supported by standard deviation price reversal around 0.4294. Standard deviation ofprice reversal indicates also that frequency of price reversal on intraday trading is high enough.

Table 1. Descriptive Analysis

	Price Reversal	Return	Volume	Frequency	V/F	Volatility	Liquidity
Mean	0.2436	-0.00024	7611.34	529.8165	38.23838	69.02933	253.7014
Median	0	0	3837.50	106.0000	35.36232	59.29061	156.6684
Maximum	1	0.04948	164543	94264.00	318.45	221.2550	3197.572
Minimum	0	-0.10041	0.00000	0.000000	0	22.31627	0.587707
Std. Dev.	0.4294	0.00944	12255.0	4056.513	24.65714	38.00419	334.4170
Skewness	1.1944	-1.17017	5.21258	16.37905	4.201521	1.391729	4.732394
Kurtosis	2.4267	19.0523	45.2821	321.8588	42.01871	4.940026	34.38255
Jarque-Bera	266.32	11611.7	84392.8	4572110.	70294.24	512.2548	47812.89
Probability	0	0	0.00000	0.000000	0	0.000000	0.000000
Observations	1059	1059	1068	1068	1067	1068	1068

This Table reports the descriptive analysis of that using tick-by-tick data during February 2008-April 2008 period on tick size-25at Indonesia Stocks Exchange (IDX). The variables, including price reversal, stock's return, trading volume, frequency of transactions, proxy trading volume per frequency of transaction, price volatility and proxy liquidity that used turnover rate. The period of tick-by-tick (*fraction*) are 15 minutes and data was captured directly from online intraday trading activity.

Despitefully, this elementary finding indicates that price forming of tick size 25 quite dynamic and efficient enough accompanied with high price reversal. The informed traders with block trading characteristic tends to take a abnormal return fromstock with tick size 25, because they know terminal value based on fundamental value of firm's value. Onhigh tick size stock (known as *blue chips*) and with high price, relatively.

This is exactly what might be expected by informed traders. On the other side, uniformed traders for small capitalization stocks should be compensated with greater expected return because they probably face the greater trading risk and higher transaction cost, anyway. Big market capitalization stocks with high tick size may have more complicated for uninformed traders because in each its transaction they will find difficulties to read and analysis about “the true information” that owned by informed traders.

Most of informed traders know how the terminal value of stocks are. Beside, the big capitalization stock like our sample tends to underreact relate to news that arrive into market because informed traders have its information and uncertainty information to stock return.

Table 1 shows decriptive analysis of high tick size stock of IDX that tradingwith low return, medium volume (lot) and low frequency of trasaction. These findings indicates that intraday trading activity for blue chips dominated by block trading system which the trading volume more priority than transaction number. The low return may have caused by essentially and primarily long-term investment oriented from informed investor with underreact to response new public information that released by regulator or corporate action.

However, the volatility is medium scale for intraday trading. This medium volatility may give plenty opportunity to intraday traders both informed and uninformed. This volatility supported by liquidity that shows good enough for intraday trading when traders dealing with these medium liquid stocks.

4.2. Preliminary Analysis of Logit Regression on Tick Size-25

To explore whether the influence of stock return and intraday activity variables such as trading volume, frequency of transaction, ratio V/F, price volatility and liquidity to probability of price reversal, this research use hypothesis testing shall folows:

- H₀: No independent variables that influence to probability price reversal on high tick size
- H₁: There are independent variables that influence to probability price reversal on high tick size

Test criteria: Reject H₀ if probability value<5%, do not reject otherwise.

In Table 2, influence of all independent variables to probability price reversal:

Table 2. Preliminary Analysis of Logit Regression on Tick Size-25

Variable	Coefficient	Std. Error	z-Statistic	Prob.	Explanation
C	-2.472333	0.226184	-10.93062	0.0000	Significant
R _{i,t}	27.43632	8.229632	3.333845	0.0009	Significant
Vol _{i,t}	6.98E-05	1.30E-05	5.376918	0.0000	Significant
Freq _{i,t}	-8.47E-05	2.77E-05	-3.061723	0.0022	Significant
(V/F) _{i,t}	0.026655	0.003986	6.687387	0.0000	Significant
Volat _{i,t}	0.002179	0.001951	1.116898	0.2640	Not Significant
Liq _{i,t}	-0.000205	0.000400	-0.513161	0.6078	Not Significant

Significant at α=10% respectively

This Table shows that the results analysis of logistic regression model of the following form where probability of price reversal (in logit) are estimated:

$$Log \left[\frac{\hat{P}_{i,t}}{1 - \hat{P}_{i,t}} \right] = \alpha_0 + \alpha_1 R_{i,t} + \alpha_2 Vol_{i,t} + \alpha_3 Freq_{i,t} + \alpha_4 (Vol / Freq)_{i,t} + \alpha_5 Volat_{i,t} + \alpha_6 Liq_{i,t} + u_{i,t}$$

Where R_{it} is the return to stock-i in fraction period t, Vol_{it} is the trading volume (lot) in natural logarithm of stock-i in fraction-t, the Freq_{it} is the natural logarithm of frequency of transaction of stock-i in fraction-t, the ratio V/F_{it} is a proxy of intraday trading activity, Volat_{it} is the volatility of stock's price-i in fraction-t that measured by t-7 standard deviation, and Liq_{it} is proxy liquidity of stock-i in fraction-t that measured by turnover rate (t-3). U_{it} is residual value. This logit regression without normality, multicollinearity, heteroskedasticity test. Level of significance at α= 5%.

Table 2 presents the result of affect analysis of independent variables such as return, volume, frequency, V/F, volatility and liquidity on logit probability of price reversal of samples. Based on the result analysis above, we can know that probability value of variable return, trading volume, frequency and V/F<0.05, and then H₀ is rejected. These results indicate that variable return, volume, frequency, and V/F have significant effect on logit probability of price reversal whereas volatility and liquidity hasn't effect on probability of price reversal.

4.4. Logit White Reggression on Tick Size-25

To explore whether the influence of stock return and intraday activity variables such as trading volume, frequency of transaction, ratio V/F, price volatility and liquidity to probability of price reversal, this research use hypothesis testing shall folows:

- H₀: No independent variables influence to probability price reversal on high tick size.
 - H₁: There is independent variables influence to probability price reversal on high tick size.
- Test criteria: Reject H₀ if probability value<5 %, do not reject otherwise.

Table 3 presents the result of influence analysis of independent variables such as return, volume, frequency, V/F, volatility and liquidity to logit probability of price reversal on tick size-25 at IDX. Based on the result analysis above can be known that probability value for variable return, volume, frequency, and V/F, have value of test probability<0.05, then H₀ are rejected. These results indicate that variable return, frequency, V/F and volatility have significant influence to logit probability of price reversal where as volatility and liquidity are no effect on probability of price reversal.

The results analysis of logit "fit model" explain that coefficient of independent variables such as stock return is more stronger effect than coefficient of volume, frequency of transaction and proxy V/Fon probability of price reversal. I also document a second findings that complements logit analysis results. I find volatility and liquidity has no effect on probability of price reversal, surprisingly. Where as trading volume and liquidity are most important atributes in intraday trading activity and specifically, price volatility has dynamic relationship with stock return.

Table 3. Logit White Regression onTick Size-25.

Variable	Coefficient	Std. Error	z-Statistic	Prob.	Explanation
C	-2.47233	0.284604	-8.68691	0	Significant
R_{it}	27.43632	9.681287	2.833953	0.0046	Significant
Vol_{it}	6.98E-05	2.36E-05	2.952124	0.0032	Significant
$Freq_{it}$	-8.47E-05	2.44E-05	-3.4694	0.0005	Significant
$(V/F)_{it}$	0.026655	0.007055	3.778428	0.0002	Significant
$Volat_{it}$	0.002179	0.001961	1.111596	0.2663	Not Significant
Liq_{it}	-0.00021	0.000458	-0.44844	0.6538	Not Significant
McFadden R-squared:	0.2475				

Significant at $\alpha=10\%$, respectively

This Table 3 explains that the results analysis of logistic regression model of the following form where probability of price reversal (in logit) is estimated (with Normality test, Multicollinearity test, and White Heteroskedasticity treatment)

$$\text{Log} \left[\frac{\hat{P}_{i,t}}{1 - \hat{P}_{i,t}} \right] = \alpha_0 + \alpha_1 R_{i,t} + \alpha_2 Vol_{i,t} + \alpha_3 Freq_{i,t} + \alpha_4 (Vol / Freq)_{i,t} + \alpha_5 Volat_{i,t} + \alpha_6 Liq_{i,t} + u_{i,t}$$

Where R_{it} is the return to stock-i in fraction period t, Vol_{it} is the trading volume (lot) in natural logarithm of stock-i in fraction-t, the $Freq_{it}$ is the natural logarithm of frequency of transaction of stock-i in fraction-t, the ratio V/F_{it} is a proxy of intraday trading activity, $Volat_{it}$ is the volatility of stock's price-i in fraction-t that measured by t-7 standar deviation, and Liq_{it} is proxy liquidity of stock-i in fraction-t that measured by turnover rate (t-3). The results are not quite different from Table 2 above that no normality test, multicollinearity test and white heteroskedasticity treatment. However, trading volume still has effect on price reversal by proxy V/F. In facts, intraday trading on tick size-25 at IDX has been controlled by informed traders, especially foreign investors with huge fund (including hot money), have excellent analytical skill both fundamental and technical and have information access.

4.5. Discussion

Analysis of multiple logit regression on high tick size stocks find some interesting research results. Regression results (with step wise)shows important equation to explain various of probability of price reversal on intraday trading activity at IDX. I document new patterns in dynamics between probability of price reversal and some independent variables such as stock return, volume, frequency of transaction, and proxy V/F. However, volatility and liquidity as most important atributte in intraday trading activity are not significant. These findings prove that some variables inconsistent with some recent research [Kofman & Moser (2001), Chordia *et al.* (2003), Chan & Ffaf(2003), Zhang (2006)].

This model gives new dynamics views and pattern that this research variables are strong enough to explain probability of price reversal with theory of microstructure and some recent research. The re-frames of research paradigm is good enough also. The independent variables which show best effect for this research is stock return and proxy V/F.

In general, study finds the fit model that represent tick size-25 stocks shall be as follows:

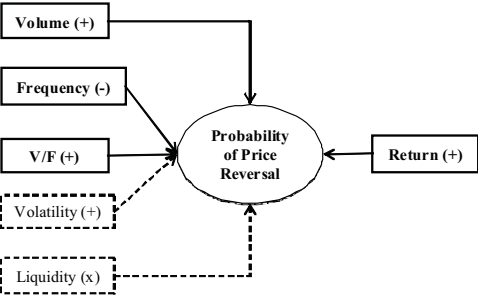


Figure 1. Dynamics Pattern of Probability of Price Reversal on Tick Size-25

As noted above, this research finds a positive relationship between stock return and probability of price reversal and the likelihood of non-intraday trading fraction. The stock return driven by price changes that incorporated with volatility, more over, the component of price change that is unrelated to predictable effects like turnover rate and bid-ask bounce (liquidity effects).

To properly understand price formation, overreaction, price correction and then price reversal in one frame, we need to know how prices incorporated information flow. Walsh (1997) explains that at any trading time, any 'news' that arises from a particular order or price will add to the existing information set. Based on the market structure and the reaction times of traders, prices will respond to 'news' as it changes the structure of the information set. It is the speed and extent of this price response that we consider here. The logit regression result indicate that high return extent of non intraday trading period and that return positively related to probability of price reversal.

Inspection of Table 3 shows that positive coefficient of stock return may not surprisingly. This is probably due a combination of momentum effects about corporate action and earning report meet analyst forecast in the first two quarters during this study. These momentum may have microstructure implication relates to the friction and the microstructure noise. Related to coefficient of return, this condition can be just happened in intraday trading framework because the price up-down around the bid-ask spread only like "bouncing effect". This bouncing effect relates to stock return behavior that generated under continous trading mechanism that the stock price bounce between bid-ask price because of fad trading that based on less information (Summers, 1986).

According to Rhee & Wang (1997), the bid-ask error can induce two types of measurement errors: 1) the bid-ask bounce and 2). the spread size error. Each of these errors is likely to cause negative autocorrelation in observed returns. For the stock return analysis, the optimal return series, which is characterized by minimum bid-ask error, can be defined as the return calculated by averaging the bid-ask prices. The alternative suggestion, the bouncing effect was generated on high tick size stock because of the friction and noise microstructure. The friction and noise may have make distortion on the price formation process and generate autocorrelation. The most bouncing and noise microstructure implication of this logit specification model, confirm by the empirical results.

This price reversal phenomenon repeats consistently because of asymmetric information where the informed traders "care of" in order to price on positive trend as long as research duration. The price reversal is more sensitive when the price go down than price go up because of the rational expectation. These circumstances, especially the friction and noise tend to have greater short run variability of stock prices. The evidences of research show consistence with order imbalance temporary and microstructure dynamic under the assumption of poisson arrival of traders also. However, the increasing of price pressure that exploited by smart informed trader which conducts trading with buy or sell order manipulation may have not optimized the fit model prediction.

Consider how trading volatility and liquidity unrelated to probability of price reversal depend on the arrival of new orders to market is processed. When a new order arrives at the market, market participants will know the size and direction of the order, particularly those participants who informed and closely follow the market, like informed traders and may be brokers. Volatility and frequency of transaction was tend controlled by informed traders consistently with big transaction value and its transaction frequency that always keep balanced proporsionally. This findings is not relevant with Easley & O'Hara (2003) paper.

This finding was proven that variable V/F most significant, because the ratio between volume trading and transaction frequency always proporsional. The value of private information who owned by strategic traders is protected with V/F ratio proporsionally that difficult to extract to become public information. This finding supports Lin *et al*, (1999). However strategic traders have conditional expectation of the order about its terminal value and some forecast error can be involved hence the price will be overvalued because deviate from rational expectation. In any incident, most uninformed traders participate in trading activity based less information, off course, so price volatility will be higher than expected by informed traders. The orders process like this repeats for every news arrives into market and the always create price reversal.

5. Conclusion

The logit specification results indicate that on high tick size at IDX, the extent of intraday trading, and that stock return positively related to the probability of price reversal. Another finding noted that volume, and proxy V/F, positively related to probability of price reversal except frequency of transaction. However, volatility and liquidity have no influence on probability of price reversal. These results are not consistent with the conclusion of Kofman & Mosser (2001), Summers & Summers (1989) and O'Hara (2005), so far. At the high level of tick size for big size stocks (*blue chips*), uninformed traders hard to enter the capital market that can't increases the overreaction of these stocks price which will diverge from its fundamental level. As we know, price reversal occur when price return to its fundamental like stock reverse hypothesis. The research findings suggest that informed traders are more sensitive to the fundamental value and less to its transaction cost.

Autocorrelation of the return for intraday trading are higher in the observation periods when the sentiment of corporate action is low. This implies an increase of the probability of price reversal and is suggestive of a positively correlation between return and price reversal caused by higher market participation of informed traders. The price will be bounced around the bid-ask spread only if no news arrives into market.

The bounces create high trading volume and transaction number but the price do not change, significantly. Hence, the trading volume unrelated to probability of price reversal and transaction frequencies are positively correlated to probability of price reversal in large range movement of price.

The proxy V/F is positively correlated to probability of price reversal more substantially than other independent variables. These results are consistent with the conclusion of descriptive analysis. This implication shows that ratio of volume-frequency suggest eliminate price manipulation of stock transaction in intraday trading. In other hand, ratio volume- frequency can reduce the gap of transaction between informed traders who conduct transaction in block trading and uninformed traders that rely on retail trading. The other variables such as price volatility and liquidity are positively related to probability of price reversal. The volatility is indicated that its pattern has link with stock price movement that implies an increase of the probability of price reversal. The authority of IDX implements stock trading system with continous auction method, without the role of market maker or specialist. The behavior of return volatility is higher on open-to-open than close-to-close. This findings can be happened because occur abnormal return in market on close.

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