

# Cluster Analysis on PEAD Using SUE Model with Quarterly Data

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## ABSTRACT

The paper constructs SUE model with Quarterly Report Data of the A-share listed companies for the first time, to analyze Post-Earnings-Announcement Drift (PEAD) with a higher data frequency. A further cluster analysis is made to examine PEAD phenomenon under different market caps and report periods. The study finds that large cap stocks show stronger PEAD than the small cap. It also firstly finds that PEAD on Q1 is more obvious than others. These findings have practical significance for the investment management decisions and the quantitative trading financial products developments.

**Keywords:** Post-Earnings-Announcement Drift; PEAD; SUE Model; Cluster Analysis

## 1. Introduction

Ball and Brown (1968) firstly found “Post-Earnings-Announcement-Drift” (shorts for PEAD); that is, stock prices show continuous and enduring reaction to the latest information, which means stable excess return can be obtained through quantitative investment model based on excavating on report data. Bernard and Thomas (1989) demonstrated that the PEAD is attributed to the CAPM model’s mispricing to the market. Bhushan (1994) considered PEAD, directly or indirectly, driven by the transaction cost. Although Fama (1998) insisted on the efficient market hypothesis, he admitted that obtaining excess return through PEAD is “over suspicion”.

After Fama’s approach, most of the scholars are inclined to recognize the existence of PEAD, developing depth discussions of the explanation of PEAD. Standardized unexpected earnings (SUE) model was firstly presented by Rendleman, Hones and Latane (1982), and they find that 2% abnormal return could be achieved through buying top 10% SUE portfolio during 1971-1980. Foster, Olsen, Shelvin (1984) did the same test and found that small cap stocks are more sensitive to the unexpected earning announcements.

In recent years, Chinese scholars did various researches on PEAD phenomenon from different angles. Wu and Wu (2005) constructed SUE model with EPS data from semi-annual reports and annual reports of A-share listed companies in china. Kong and Ke (2007) analyzed characteristics between institutions and person-

al investors. Lu (2012) commented PEAD on systematic mispricing.

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This paper constructs SUE model with quarterly report data of A-share market for the first time, and use cluster analysis to estimate PEAD of different market caps and periods. The study makes up for the low sampling frequency deficiency of the previous research, and also finds out two new features of PEAD in A-share market. These findings not only help to deepen the understanding of the nature of A-share market, but also provide the basis and reference for the development of quantitative financial innovative products.

The structure of this paper is organized as follows: after the introduction, the second part is the model design; third, fourth, fifth parts are the experiments and the results. The third part is the whole experiment for A-share

market; the fourth part is the cluster analysis for different market cap; the fifth part is the cluster analysis for different report periods. The last part is the conclusion.

## 2. Model Design

Different from the previous research, the paper includes quarterly report data in the scope of the study, using it to construct the SUE model, which is shown as below:

$$SUE = \frac{(E_{i,t} - E_{i,t-4}) / |E_{i,t-4}|}{\sigma_{i,t}} \quad (1)$$

Here,  $E_{i,t}$  represents the Earnings of company  $i$  from quarterly report  $t$ .  $E_{i,t-4}$  represents Earnings from the same quarter but of the previous year;  $\sigma_{i,t}$  is the standard deviation of numerator of Equation (1) for the recent eight quarters. A large SUE means high unexpected earnings.

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$$CAR_{jt} = R_{jt} - R_{pt} \quad (2)$$

China Securities Regulatory Commission requested that listed companies should disclose quarterly report since 2002. In this paper, the data are collected since 2002Q1 to 2011Q3 for all available A-share stocks, 39 periods and 65,372 earnings data in total. The available SUE data is 38,333, since 2005Q2 to 2011Q2. After filtering some stocks that are not available for investing, the final data set is 32,308. Report data are collected from Wind database, and the experiments are coded on Tinysoft database.

The standards of data restrictions are set as follows:

- Exclude the delisting stocks within 60 trading days

after the announcement date.

- Exclude the stocks that suspension period over 1 year during [-60,60] trading day period.
- Exclude the stocks that prices change over 50% between two consecutive trading days.

## 3. General Experiment Results

Figure 1 shows the behavior of CAR over 60 days before and after the announcement date for SUE model. Decile 10 represents the largest group of SUE, which means the highest unexpected earnings; conversely, Decile 1 represents the smallest group of SUE. 0 on abscissa represents the announcement date.

Some conclusions can be drawn through the experiments:

- All groups show early response to the unexpected earnings. Take Decile 10 as an example. CAR surges up before the announcement date, and the magnitude is over 7%.
- After the announcement date, high SUE value groups still show obvious positive CAR, which means positive unexpected earnings leading positive drifting phenomenon. To be contrast, low SUE groups show negative CAR and downside drifting phenomenon as well. The SUE model with quarterly report data validates PEAD driven by unexpected earnings.
- All groups' CAR show a slight fall on the announcement date. The interpretation is that stocks will be facing a higher selling pressure on the announcement date, no matter good or bad for the report.
- Considering the Cross-section results of different periods, Decile 10 shows an enduring trend throughout the timeline (see Figure 2). PEAD is more obvious before 2007, and after that, the drifting magnitude has been weakened but still remarkable.

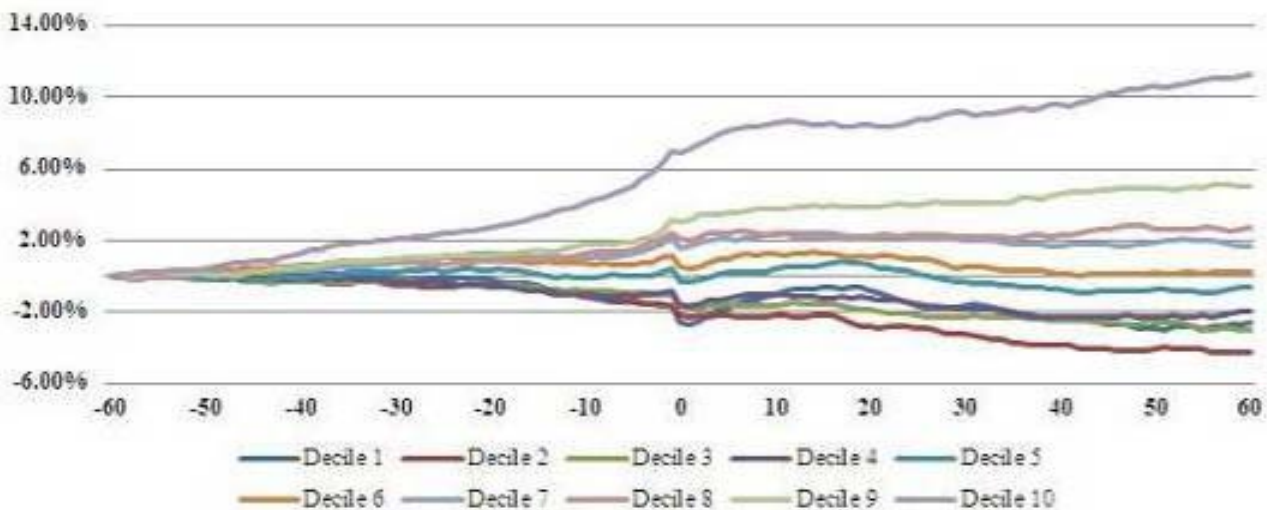


Figure 1. Behavior of CAR over [-60,60] trading period for SUE model.

#### 4. Cluster Analysis and Results of Different Market Cap

This paper combines SUE model with cluster analysis to examine the characteristics of PEAD of the large, medium and small cap stocks in A-share market. (see **Figure 3**).

Cluster Analysis shows that obvious drifting phenomenon exist in the large and medium market cap groups, especially in the large group. Relatively, comparing the drifting magnitude and the degree of differentiation between decile groups, small cap group does not show PEAD as obviously as the large and medium groups.

#### 5. Cluster Analysis and Results of Different Report Periods

The paper conducts cluster analysis on different report periods with SUE model. Results (see **Figure 4**) show the first quarter (Q1) has the strongest PEAD, which is reflected by the remarkable drifting magnitude of Decile

10, and the differentiation among deciles. The PEAD effect of Q2 (semi-annual report) appears weakened and Q3 and Q4 (annual report) show unregulated ups and downs rather enduring trend. In general, the result shows a feature that the effect of PEAD, which is driven by unexpected earnings, diminishes quarter by quarter.

#### 6. Conclusion

The paper constructs SUE model with quarterly report data, covering all the available earnings report period of A-share stocks for the first time. With higher frequency data, the study verifies the existence of PEAD in A-share market again, and also finds the ‘drifting phenomenon’ during the pre-announcement period. A further cluster analysis shows the A-share large and medium cap groups have stronger drifting effect than small cap group, which is different from the foreign stock market. What’s more, cluster analysis also suggests that unexpected earning is a more obvious driven force in Q1 than in other periods.

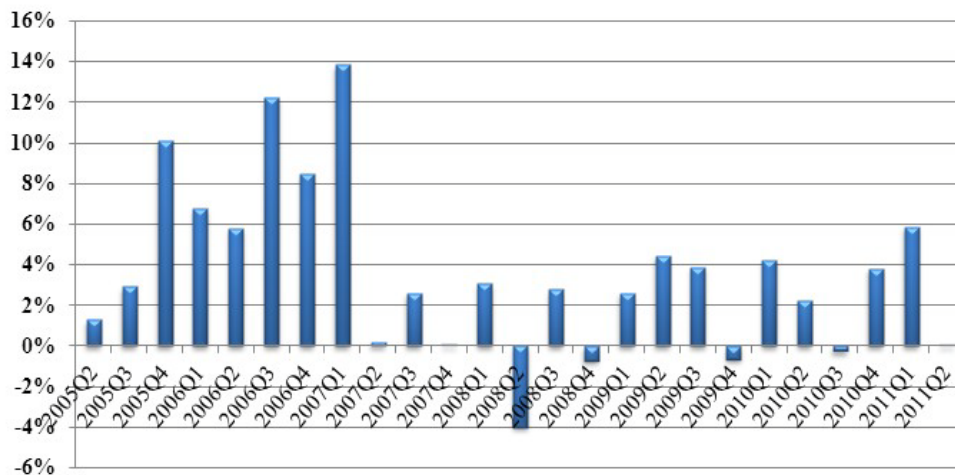


Figure 2. Behavior of Decile 10's CAR at 60 trading days for SUE model.

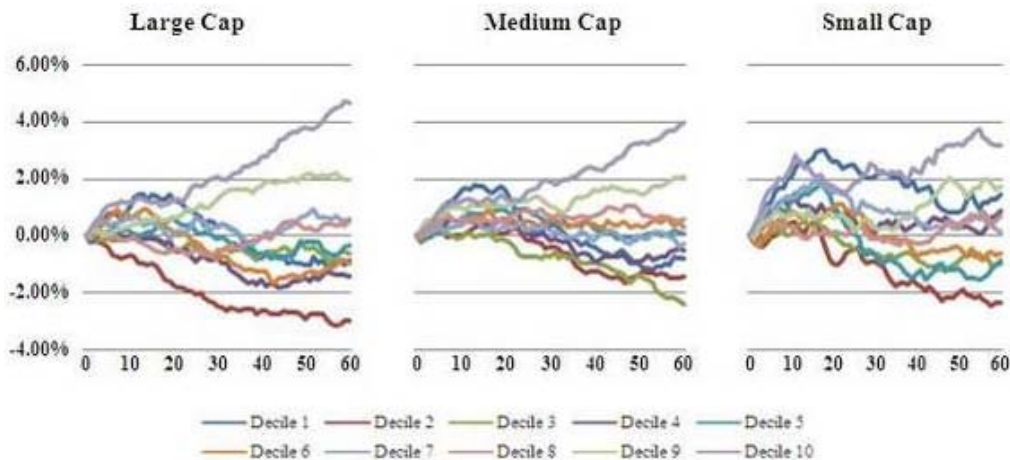


Figure 3. Behavior of Cap Groups' CAR over [0,60] trading period for SUE model.

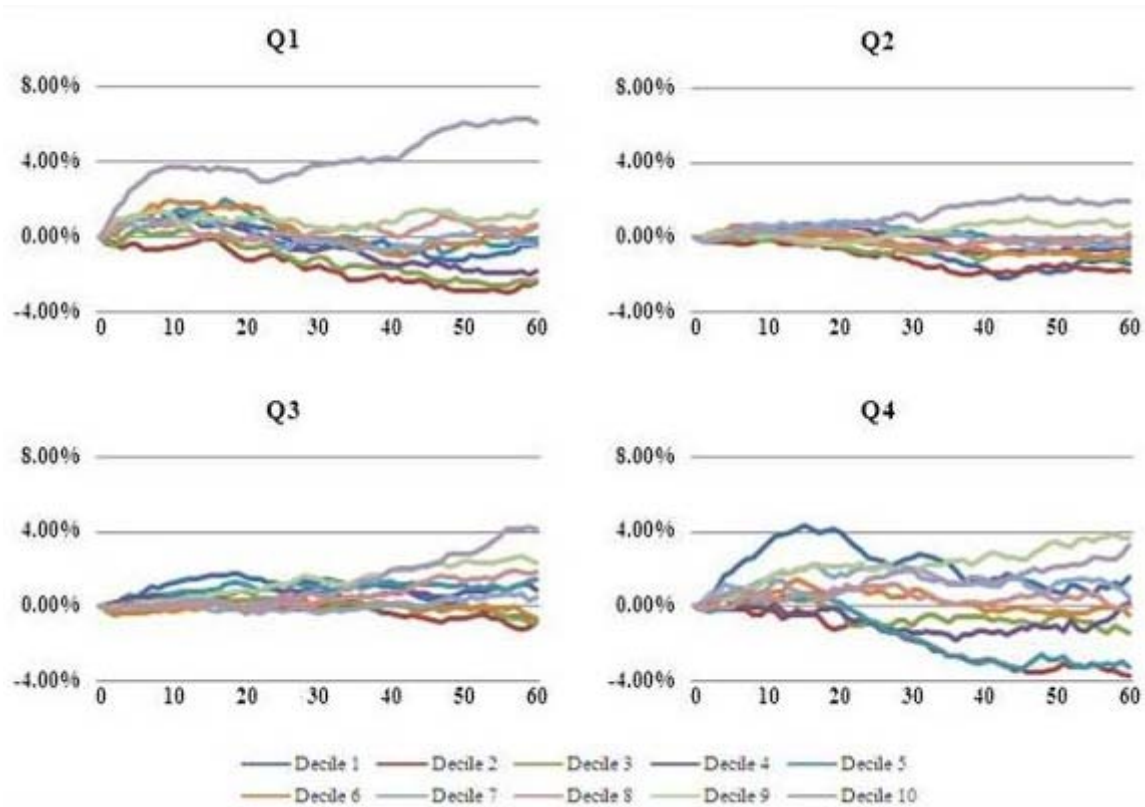


Figure 4. Behavior of Quarter Groups' CAR over [0,60] trading period for SUE model.

The securities market has massive transaction data and listed company information. The use of data-mining in securities information research has a far-reaching theoretical and practical significance. The new findings of the PEAD phenomenon in this study not only enrich our knowledge and understanding of the nature of A-share market, but also provide specific guidance for the development and improvement of the quantitative investment model and the financial innovation products as well. For example, strong drift effects in the large-cap stocks give better liquidity support for the SUE model based quantitative investment strategy and ensure the operability of the derivative portfolio management. The different drift effects of report periods provide an empirical basis and practical way to improve the model performance. Due to the A-share market has not yet matured, further empirical researches to explore more features and laws are ongoing.

## REFERENCES

- [1] Ball, Brwon, 1968, An Empirical Evaluation of Accounting Income Numbers, *Journal of Accounting Research* 6, 159-177.
- [2] Rendleman, Jones, Latane, 1982, Empirical anomalies based on unexpected earnings and the importance of risk adjustments, *Journal of Financial Economics* 10, 269-287.
- [3] Fama, French, 1992, The Cross Section of Expected Returns, *Journal of Finance* 47, 427-465.
- [4] Fama, 1998, Market Efficiency, Long-term Returns, and Behavioral Finance, *Journal of Finance Economics* 49, 283-306
- [5] Bernard, Thomas, 1989, Post-Earnings-Announcement Drift: Delayed Price Response or Risk Premium?, *Journal of Accounting Research* 27, 1-36.
- [6] Bhushan, 1994, An informational efficiency perspective on the post-earnings announcement drift, *Journal of Accounting and Economics* 18, 45-65.
- [7] Foster, Olsen, Shelvin, 1984, Earnings releases, anomalies, and the behavior of security returns, *Accounting Review* 59, 574-603
- [8] Wu, Wu, 2005, A Study on Measures of Earnings Information, Market Reaction and Investors' Framing Dependence Bias, *Economic Research Journal*, No. 2
- [9] Kong, Ke, 2007, Who Drives the PEAD in China?, *Journal of Financial Research*, No. 10
- [10] Lu, 2012, Systematic Mispricing:A Research on Post Earnings Announcement Drift in China's A-share Stock Market, *Journal of Financial Research*, No. 3