

DO INSTITUTIONS FOLLOW THE HERD? EVIDENCE FROM IRAN

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The Question

- The ongoing research centers around examining the trading patterns of both Institutional and Individual investors within the Tehran Stock Exchange market.
- Specifically, I am asking whether these institutions and individuals tend to follow momentum or anti-momentum strategies when they trade.
- The focus is on institutions because they're major players in the market, and naturally we expect them to make trades based on fundamentals.
- I am exploring momentum and anti-momentum strategies because of how they can influence market herding. This might push prices far from their real values and even lead to market bubbles.

Data

- ① TSE Stocks' Returns, *Daily*
- ② TSE Overall Index Returns (Market Return), *Daily*
- ③ CBI Risk-Free Rate, *Monthly*
- ④ All Codal News (Letters), *Secondly*
 - Both before 1389 (1382 - 1389) and after 1389
- ⑤ TSE Individual-Institutional Trade, *Daily*
- ⑥ TSE Stocks' Nominal Prices, *Daily*
- ⑦ TSE Stocks' Market Capitalization, *Daily*

Methodology

This research has two stages:

- 1 The first stage revolves around the categorization of company-specific news as either positive or negative. These categorizations are needed based on their influence on trading patterns of both institutions and individuals.
- 2 In the subsequent stage, the labels assigned during the first phase are utilized to analyze the trading behavior of institutions and individuals within distinct and non-overlapping time frames.

More details are in the following slides.

Stage 1: News Labeling

Stage 1: Why News Are Important?

- Abundant evidence suggests that news pertaining to individual stocks, characterized as either "positive" or "negative," significantly impacts the trading choices made by investors.
- Any study or research design intending to analyze the trading behavior of investors of any type must take into account information related to companies.
- Regrettably, we currently lack a database containing company-specific news curated through assessments and opinions from expert capital market analysts.
- Creating such a dataset would significantly enhance the research landscape within the Tehran Stock Exchange market.

Stage 1: How I Solve The News Challenge?

- In the first stage, I suggest a method for categorizing stock-specific news as either "positive" or "negative".
- The labels generated during this initial stage are then employed in the second stage as dummy variables to account for the presence of "positive" and "negative" news associated with each individual stock.

Stage 1: Source of Firm-Specific News

- I rely on Codal for firm-specific news.
- I perceive every letter posted on Codal as a news item.
- In the context of Codal, any financial statement, report, PDF, or Excel document that is shared is regarded as a "Codal letter" within the Codal framework.
- Thankfully, each company-specific update on Codal is consistently linked to the respective company's ticker symbol. Each piece of content includes a ticker field.

Stage 1: How I Label News as Good or Bad?

To categorize news and systematically determine whether each stock-specific news is positive or negative, I follow this methodology:

- Using the adjusted returns data, for each firm-day pair I define a two month period prior to that day.
- Within this moving window, I assess the CAPM model (utilizing a rolling CAPM approach).
- By utilizing the CAPM betas, I compute the anticipated (expected) return for each specific firm-day.

Stage 1: How I Label News as Good or Bad? (Cont'd)

- Based on the calculated expected return in the previous step, I ascertain the Abnormal Returns for each given day.
- I combine the computed abnormal return data with the dataset containing news articles (letters sourced from Codal).
- Employing a symmetric threshold, I make determinations whether news should be classified as favorable or unfavorable.

A few Remarks:

- The outlined steps yield a dataset comprised of trios: firm, day, and corresponding news.
- I haven't detailed the precise procedure and meticulous considerations within this process.

Stage 2: Trading Behavior

Stage 2: How to Measure Trading Activity?

To assess investors' purchasing and selling activities for stock i at time t , we analyze their excess buying ($XB_{i,t}$) and excess selling ($XS_{i,t}$) of the stock individually.

I define the excess buying and selling for the group G as follows:

$$XB_{i,t}^G = NB_{i,t}^G - \mathbb{E} \left[NB_{i,t}^G \right]$$

$$XS_{i,t}^G = NS_{i,t}^G - \mathbb{E} \left[NS_{i,t}^G \right]$$

Where:

$$G \in \{Institution, Individual\}$$

Stage 2: How to Measure Trading Activity?

The definition of $NB_{i,t}^G$ and $NS_{i,t}^G$ are as follows:

$$NB_{i,t}^G = \frac{Buy_{i,t}^G - Sell_{i,t}^G}{Buy_{i,t}^G + Sell_{i,t}^G}$$

$$NS_{i,t}^G = \frac{Sell_{i,t}^G - Buy_{i,t}^G}{Buy_{i,t}^G + Sell_{i,t}^G}$$

Referring to the existing literature, I regard the expected value of the $NB_{i,t}^G$ and $NS_{i,t}^G$ equal to average of $NB_{i,t}^G$ and $NS_{i,t}^G$ for all stocks that investor group G trades at time t , respectively.

Stage 2: The Model

We utilize fixed-effects ordinary least squares (OLS) regressions on panel data with t-statistics adjusted for panel-corrected standard errors (PCSE).

$$y_{i,t} = \alpha + \beta R1 + \gamma R2 + \delta R6 + \zeta R28 + \sum_i \kappa_i * d_i + \epsilon$$

where:

$$y \in \{XB_{i,t}^{\text{Ins}}, XS_{i,t}^{\text{Ins}}, XB_{i,t}^{\text{Ind}}, XS_{i,t}^{\text{Ind}}\}$$

Stage 2: The Model, Independent Variables

$$y_{i,t} = \alpha + \beta R1 + \gamma R2 + \delta R6 + \zeta R28 + \sum_i \kappa_i * d_i + \epsilon$$

where:

$R1 \triangleq R(-1)$: The one-day holding period prior to trading day.

$R2 \triangleq R(-2, -5)$: The 2-5 days holding period prior to trading day.

$R6 \triangleq R(-6, -27)$: The 6-27 days holding period prior to trading day.

$R28 \triangleq R(-28, -119)$: The 28-119 days holding period prior to trading day.

Note that:

- The holding period returns are market-adjusted returns. The adjusted return minus the market return.
- The market return is the TSE Overall Index. All days are working days.

Stage 2: The Model, Control Variables

In my analyses, I incorporate various control variables that have previously been identified as influencing investor trading behavior.

$$y_{i,t} = \alpha + \beta R1 + \gamma R2 + \delta R6 + \zeta R28 + \sum_i \kappa_i * d_i + \epsilon$$

The control variables are:

- To control for news, I include dummies that identify whether the contemporaneous, one-day and two-day lagged stock-specific news announcements are “good” or “bad”.

Stage 2: The Model, Control Variables

$$y_{i,t} = \alpha + \beta R1 + \gamma R2 + \delta R6 + \zeta R28 + \sum_i \kappa_i * d_i + \epsilon$$

The control variables are:

- To control for news, I include dummies that identify whether the contemporaneous, one-day and two-day lagged stock-specific news announcements are “good” or “bad” (6 dummies).
- I also include dummies to capture day-of-the-week effects. (4 dummies)
- And two dummy variables for ‘reference point’ effects, which equal one if the stock price is at the monthly highest or lowest level and zero otherwise.

Results

First Stage: Results

- The first stage of the research is to label the news as good or bad.
- The following table shows the distribution of news articles that have been labeled as good or bad by the explained method.

News Type	Freq.	Percent	Cum.
Bad	25,703	44.78	44.78
Good	25,621	44.64	89.42
Neutral	6,075	10.58	100.00
Total	57,399	100.00	

Second Stage Results: Main Result

	Institutions		Individuals	
	Buy	Sell	Buy	Sell
$R(-1)$	-1.842*** (0.08)	1.755*** (0.09)	0.550*** (0.03)	-0.550*** (0.03)
$R(-2, -5)$	0.046* (0.03)	-0.083*** (0.02)	-0.006 (0.01)	-0.003 (0.01)
$R(-6, -27)$	0.053*** (0.01)	-0.049*** (0.01)	-0.028*** (0.00)	0.024*** (0.00)
$R(-28, -119)$	0.007* (0.00)	-0.012*** (0.00)	-0.006*** (0.00)	0.005*** (0.00)
Mentioned FEs	YES	YES	YES	YES
N. of Unique Firms	742	742	742	742
R^2	0.9%	0.9%	0.5%	0.4%
NObs	1,133,473	1,133,473	1,133,473	1,133,473

Main Result Interpretation

- The findings suggest that institutions tend to go against momentum trading in the context of a one-day holding period. Conversely, for the three longer time horizons, institutions seem to adopt a momentum-oriented strategy.
- On the other hand, the results reveal that individuals engage in momentum trading within a one-day holding period but shift towards an anti-momentum approach for longer horizons.
- Overall, the trading behaviors of institutional and individual traders exhibit a notable counteractive nature.

Conculsion

- The present study introduces an approach to categorize stock-specific news into positive and negative classifications.
- In the context of a one-day timeframe, institutional traders appear to adopt an anti-momentum approach, whereas they seem to embrace a momentum strategy over longer periods.
- The findings highlight systematic distinctions between institutions and individuals regarding their responses to historical price trends, as well as their varying degrees of adherence to momentum and contrarian strategies.