

Discussion of:
Revisiting the Cross-section of Expected Stock
Returns: Evidence from a Textual Analysis of
Buy Recommendations

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ABFER
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Leading explanations for anomaly returns

1. Risk framework

- Stochastic discount factor
- CAPM, consumption CAPM

2. Irrational framework

- Overconfidence (Daniel, Hirshleifer, Subrahmanyam, 1998)
- Extrapolation (Barberis, Shleifer, Vishny, 1998)

3. Non-standard preferences framework

- Prospect theory
- Cumulative prospect theory (Kahneman and Tversky)

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Other frameworks not considered here: Underreaction (Hong and Stein, 1998), social transmission bias (Han, Hirshleifer, Walden, 2022), Saliency, etc.

Research question

Which of these framework(s) explains anomaly returns?

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Introduce a new way to address this question:

- **Elicit** investors' investment framework using textual analysis on sell-side analyst reports & Seeking Alpha articles.
- **Key assumption:** These writings either influence or reflect investors' beliefs/preferences.

Empirical approach

Data: A large corpus of investment articles

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New Method:

- Categorize the reasoning behind each recommendation into the three frameworks:
 1. Risk: "Safety"
 2. Irrational: "Exuberance" (Growth)
 3. Prospect theory: "Lottery"

Mapping textual content to investment frameworks

Create a new bag-of-words from textual analysis using a self-created surveys of:

- 100 institutional investors (CoreData Research)
- 303 retail investors (Prolific platform)

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Example of words:

- Risk: *reliable, safe, secure, stable, steady*
- Exuberance: *consistent, excellent, growth, innovative, winner*
- Lottery: *exciting, gamble, potential, speculative, volatile*

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Analyze stocks on the short-side of the anomaly

- Value factor: Glamour/Growth decile of stocks
- Gross profitability: Extremely unprofitable decile
- Idiosyncratic volatility: Extremely volatility decile

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Makes sense to look here:

- Short side of the anomaly usually generates a large fraction of the total return predictability
- Frameworks tend to explain the short side
- Requires enough buy recommendations

Main finding (1): Lottery reasons are the most common

	Risk Framework	Irrational Beliefs Framework	Non-Traditional Preferences Framework	Inconclusive
<i>Panel A: Institutional Investors' Survey-Based Wordlists</i>				
Sell-Side Analyst Reports	9% [7%]	29% [17%]	69% [58%]	19%
Seeking Alpha Articles	6% [6%]	15% [10%]	58% [54%]	29%
<i>Panel B: Retail Investors' Survey-Based Wordlists</i>				
Sell-Side Analyst Reports	21% [6%]	24% [14%]	70% [66%]	14%
Seeking Alpha Articles	9% [8%]	22% [13%]	64% [57%]	22%
<i>Panel A: More Likely to be Irrational Beliefs Based / Less Likely to be Non-Traditional Preferences Based</i>				
Sell-Side Analyst Reports	16% [12%]	47% [27%]	55% [37%]	23%
Seeking Alpha Articles	10% [10%]	24% [19%]	32% [26%]	45%
<i>Panel B: Less Likely to be Irrational Beliefs Based / More Likely to be Non-Traditional Preferences Based</i>				
Sell-Side Analyst Reports	1% [1%]	11% [6%]	82% [79%]	14%
Seeking Alpha Articles	2% [2%]	6% [2%]	84% [82%]	13%

Lottery is much stronger for small stocks

Main finding (2): By anomaly/factor

Firm Characteristic	Do Sell-Side Analyst Buy Recommendations of Short-Leg Securities Disproportionately Frequently Use		
	Safety Words?	Exuberance Words?	Lottery Words?
Accrual			7% (10.79)
Difference of Opinion			12% (16.49)
Asset Growth			15% (32.72)
Value		5% (14.36)	14% (31.65)
Gross Profitability			43% (68.73)
Investment			8% (14.65)
Composite Equity Issuance			11% (23.05)
Idiosyncratic Volatility			20% (36.98)
Expected Idiosyncratic Skewness		12% (36.80)	6% (14.51)
Long-term Reversal		10% (27.98)	9% (19.61)
Maximum Daily Return			18% (37.66)
Momentum			9% (13.64)
Net Stock Issuance			32% (56.76)
External Finance			32% (60.38)
Net Operating Assets			7% (12.76)
O-Score			35% (39.17)
Organizational Capital			
Failure Probability			11% (3.39)
Return On Assets			42% (66.17)
Post-Earnings Announcement Drift			
Short-Term Reversal			14% (29.32)
Market Capitalization	7% (14.65)	5% (19.34)	

Overall impressions

Unique research question & approach:

- Horseshoe different theories
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The results are thought-provoking

- Interesting data point
- Prospect theory might matter a lot more than my prior
- Characteristics versus covariances

Comment 1: Why does sell-side research exist?

Why spend time writing research reports?

Career incentives

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Suggestion: Placebo and different benchmark

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1. Placebo

- Find characteristics that are not associated with future returns.
- Create a long-short portfolio on a random characteristic (e.g., revenues, depreciation)
- Repeat your exercise in the extreme stocks
- Null hypothesis: No frameworks should be significant

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2. Use the results from part 1 as the benchmark level of key words

- Currently the benchmark is the average of all articles

Comment 1b: Analyst and anomalies

Investment professionals seem to trade/recommend investments in the wrong direction of anomalies

- Analysts: Engelberg, Mclean, Pontiff (2020), Li, Li, and Wei (2020)
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Open ended questions: Can we learn more from your exercise?

Comment 2: Validation test

Lottery-like characteristics exhibit positive skewness in returns

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Suggestion:

- Validate your elicited results
- Prediction: Positive correlation of lottery word % with realized skewness

Comment 3: Exploit the cross-section of stock returns

Not sure if this is possible, but a suggestion...

Use the cross-section of stock returns:

- Within the short-leg, double sort:
- Identify stocks that have high versus low lottery words %
- Is the anomaly return concentrated in the high lottery word % sub-sample?

Final remarks

Addresses a fundamental question in asset pricing:

How to differentiate between various investor frameworks?

Introduces creative new methods

- Data selects the theory
- Surveying for bag of words
- Authors acknowledge the limitations of their approach