

# Accounting Data and Stock Market Prediction

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# Outline

The Paper

The Predictors

The Predictability Evidence

The Investors' Inattention Evidence

Final Remarks

## The Predictors in a Nutshell...

- Aggregate 28 accounting variables
- Separate them into 8 groups (1 variable per group)
  - Assets, Equity, Liabilities, Income, Revenue, Taxes, Extraordinary Items, and Dividends.
- Detrend the aggregates
- Create indexes (EW, PCA and PLS)
- Study the predictability of such indexes on future returns

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## The Empirical Evidence in a Nutshell...

### 1. There is (out of sample) predictability:

- (a) PCA or PLS ( $R^2$  up to 3.62%)
- (b) Indexes are crucial
- (c) Predictability is higher than with standard variables
- (d) It is really about the short-term (less than 1 year)
- (e) Large economic significance

### 2. The predictability is related to investors' inattention



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## The Contribution in a Nutshell...

1. Add to the out of sample predictability evidence after Welch and Goyal (2008)
  - Rapach et al. (2010); Ferreira and Santa-Clara (2011); Kelly and Pruitt (2013); Huang et al. (2015); Rapach et al. (2016)
2. Add to the literature linking slow information diffusion to predictability
  - Hong and Stein (1999); Peng and Xiong (2006); Cohen and Franzini (2008); Dellavigna and Pollet (2009); Hirshleifer et al. (2009); Hong et al. (2007); Loh (2010); Hirshleifer et al. (2011); Li and Yu (2012).

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  - Should we be looking at detrended variables or ratios? (some theories suggest ratios)
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  - Could calculate  $\sum F / \sum ME$  and use  $\sum ME_t / \sum ME_{t-1}$  (for firms available at  $t-1$  and  $t$ ) to get aggregate  $F$

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## Could Improve the Comparison to Recent Evidence ▸

This Paper						Rapach et al. (2010)			
Trend		1990	1995	2000	2005	Period 1		Period 2	Period 3
EW	Linear	-0.1%	-0.0%	-0.0%	2.3%	Mean	3.6%	1.2%	3.0%
	Stochastic	-1.3%	-1.1%	-0.9%	3.1%	Median	3.0%	1.5%	1.6%
PCA	Linear	0.8%	1.1%	1.7%	2.6%	Trimmed	3.5%	1.2%	3.0%
	Stochastic	0.7%	1.3%	1.4%	3.6%	DMSPE(1)	3.5%	1.1%	2.6%
PLS	Linear	1.2%	2.6%	3.0%	2.5%	DMSPE(0.9)	3.5%	1.0%	2.7%
	Stochastic	-1.3%	2.3%	2.0%	0.3%	Mean, CT	3.2%	1.2%	2.4%
Average	Linear	1.4%	1.9%	1.9%	1.3%	Average	3.4%	1.2%	2.5%
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## Volatility Impacts Forecasting Accuracy

	Low Attention	(3)	(5)	(7)	High Attention	Average
Small	4.7%	1.3%	3.4%	4.6%	3.25%	3.8%
(5)	4.7%	8.6%	4.5%	5.9%	5.3%	5.9%
Large	15.3%	5.2%	6.3%	3.0%	5.3%	6.3%
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- Control = size within firm. Should sort on  $ME_{it} - \overline{ME}_i$
- Volatility impacts forecasting accuracy
- If  $\rho(\text{Attention}, \text{Volatility}) > 0$
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$$\circ \sigma^2 = \mathbb{E} \left[ (R - \mathbb{E}[R])^2 \right], \text{ but } y = \log \text{ of } (R - \mathbb{E}[R])^2$$

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Large	15.3%	5.2%	6.3%	3.0%	5.3%	6.3%
	6.6%	5.1%	4.4%	5.0%	4.8%	

- Control = size within firm. Should sort on  $ME_{it} - \overline{ME}_i$
- Volatility impacts forecasting accuracy
  - $R^2 = \sigma^2(\mathbb{E}[R]) / \sigma^2(R)$
- If  $\rho(\text{Attention}, \text{Volatility}) > 0$ 
  - $R^2$  lower for high attention stocks ( $\neq$  mechanism)
- Same logic applies to firm level analysis
  - $\sigma^2 = \mathbb{E}[(R - \mathbb{E}[R])^2]$ , but  $y = \log$  of  $(R - \mathbb{E}[R])^2$

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# Outline

The Paper

The Predictors

The Predictability Evidence

The Investors' Inattention Evidence

Final Remarks

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# The Empirical Evidence

1. There is (out of sample) predictability:
  - (a) PCA or PLS ( $R^2$  up to 3.62%)
  - (b) Indexes are crucial
  - (c) Predictability is higher than with standard variables
  - (d) It is really about the short-term (less than 1 year)
  - (e) Large economic significance
2. The predictability is related to investors' inattention
  - (a) Higher for "low attention" stocks (Google searches)
  - (b) Inattention predicts "forecasting accuracy"

## Autocorrelation is Typical of Long-Run Predictors [▶ Back](#)

	This Paper			Welch and Goyal (2008)					
	EW	PCA	PLS	dp	ep	bm	dfy	ntis	svar
$\rho(-1)$	0.96	0.98	0.91	0.98	0.99	0.99	0.98	0.98	0.63

- Some of these variables tend to have predictive over longer horizons (higher than 1 year)
- The indexes in this paper tend to predict up to 6 months
- It is interesting that the autocorrelations are similar
- Can we learn anything from this?

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