



THE OHIO STATE UNIVERSITY

FISHER COLLEGE OF BUSINESS

# The Cross-Section of Subjective Expectations: Understanding Prices and Anomalies

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Discussant: **Andrei S. Gonçalves**

2024 Emerging Voices in Finance (Notre Dame)

# Outline

The Paper

My Comments

Final Remarks

# This Paper in a Nutshell

- Literature:
- This paper:
- Empirical Setting

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  - FIRE:  $dr$  "drives" time variation in aggregate equity prices
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$$\begin{aligned}
 \tilde{p}x_{i,t} &\approx \sum_{j=1}^h \rho^{j-1} \cdot E_t^*[\Delta x_{i,t+j}] - \sum_{j=1}^h \rho^{j-1} \cdot E_t^*[r_{i,t+j}] + \rho^h \cdot E_t^*[\tilde{p}x_{i,t+h}] \\
 &\downarrow \\
 1 &\approx \frac{\text{Cov}\left(\sum_{j=1}^h \rho^{j-1} \cdot E_t^*[\Delta x_{i,t+j}], \tilde{p}x\right)}{\text{Var}(\tilde{p}x)} - \frac{\text{Cov}\left(\sum_{j=1}^h \rho^{j-1} \cdot E_t^*[r_{i,t+j}], \tilde{p}x\right)}{\text{Var}(\tilde{p}x)} + \frac{\text{Cov}\left(\rho^h \cdot E_t^*[\tilde{p}x_{i,t+h}], \tilde{p}x\right)}{\text{Var}(\tilde{p}x)}
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$$1 \approx CF_h + DR_h + FPX_h$$

- Need to measure log prices ( $p$ ) and log earnings ( $x$ )
- FIRE: Realized earnings and prices (CRSP+COMPUSTAT)
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**Table I** Decomposition of dispersion in price-earnings ratios

	One-year horizon ( $h = 1$ )		One-to-four year horizon ( $h = 4$ )	
	FIRE	Expected		
$CF_h$	0.103***	0.331***		
$DR_h$	0.143***	0.033***		
$FPX_h$	0.746***	0.620***		

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Table III

$$E_{t+1}^*[\Delta \tilde{x}_{i,t+2}] - E_t^*[\Delta \tilde{x}_{i,t+2}] = \beta (\Delta x_{i,t+1} - E_t^*[\Delta x_{i,t+1}]) + u_{t+1}$$

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Panel B: Revisions after Surprises

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Main Sample 1999-2020	-0.863***
Long Sample 1982-2020	-0.786***

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# Model: Slow Moving Learning + Preference for LT CFs

$$x_{i,t} = x_t^{SEE} + \tilde{x}_{i,t}$$

$$x_t^{SEE} = \phi \cdot x_{t-1}^{SEE} + u_t$$

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- Economic Assumptions:

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# Model: Slow Moving Learning + Preference for LT CFs

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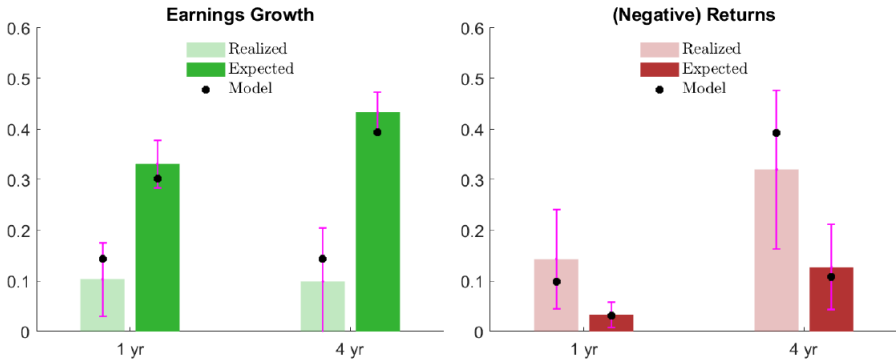


Figure 3. Empirical decomposition and model decomposition.

# Outline

The Paper

My Comments

Final Remarks





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**Table 2**  
Return decomposition using ICC approach

	Horizons (Quarters)								
	1	2	4	8	12	16	20	24	28
Panel B: Firm-level									
	Decomposition								
CF	0.19	0.32	0.48	0.63	0.68	0.68	0.67	0.66	0.62
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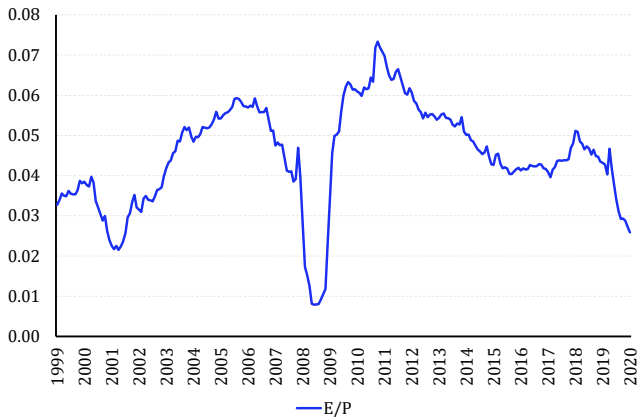
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- The objective is to decompose prices (proxied by  $P/E$ )
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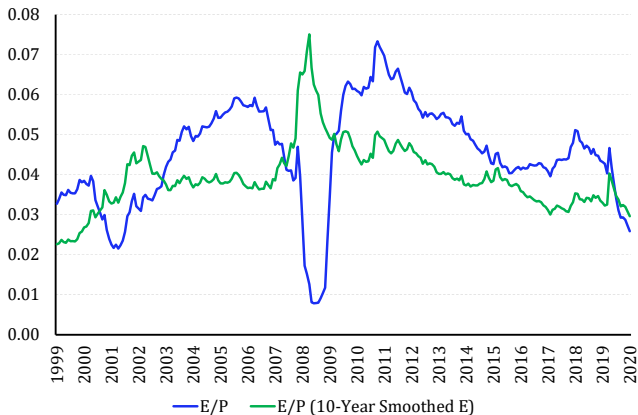
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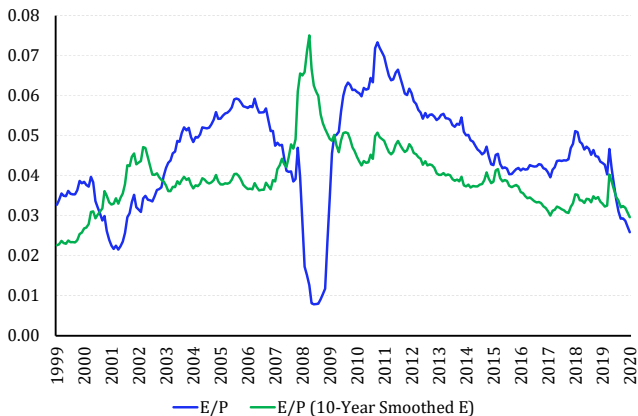
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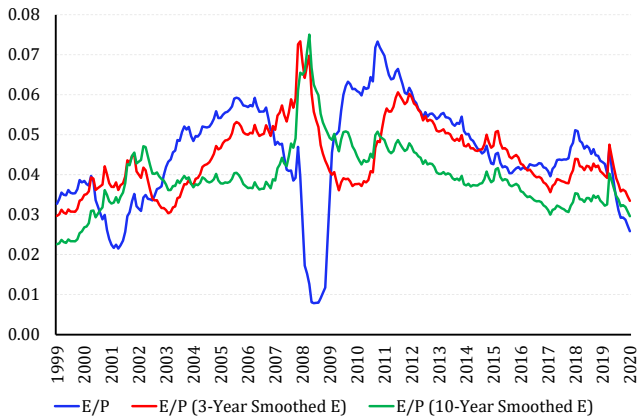
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  - $E^*[r]$  from individual investors are procyclical (or acyclical)  
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(Muller (2018), Bastianello (2024))
  - $E_t^*[r]$  from analysts (cross-section of stocks)  
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  - $E^*[g]$  from analysts have a small pass through to investors ( $B$ )
  - $E^*[g]$  from investors have a small pass through to prices ( $M_g$ )
- Your model is consistent with  $B \approx 0$  ( $E_A^*[r]$  proxy for  $E_I^*[r]$ )
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  - Show that your model is consistent with low  $M_g$  or
  - Argue for a large  $M_g$  from general investor belief shocks

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## Some Other Comments

- 1) We disagree on why FIRE models fail
  - You think they have too little variation in risk premia (e.g., second paragraph of page 16)
  - I think they have too much variation in expected cash flows
- 2) Bayesian Learning vs Constant-Gain Learning
  - Suppose we let  $g_i$  differ across firms (Internet Appendix H.3)
  - Is the  $\beta$  value the only difference between learning models?
  - If so, how far is the  $\beta = 1.8\%$  from the Bayesian  $\beta$ ?
- 3)  $E^*[g]$  based on EPS while realized earnings are not per share
- 4) I think  $w_{i,t,1}$  in Equation 18 should have a  $E_t^*[\Delta\tilde{x}_{i,t+1}]$  term

# Outline

The Paper

My Comments

Final Remarks

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- Good luck!

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## Final Remarks

- Absolutely a great paper (expect to see it in a top journal)
  - Link valuation differences across firms to  $E^*[r]$  and  $E^*[g]$
  - Identifies a mismatch between FIRE and SubE
  - Proposes a (very) parsimonious model to explain results
- It would be useful to:
  - Further discuss connections to prior (and subsequent) literature
  - Add an analysis of  $P/D$
  - Think about the  $b \approx 1$  in  $r_{t+1} = a + b \cdot E_t^*[r] + \varepsilon_{t+1}$
  - Think about the weak effect of beliefs on asset prices
- Good luck!