



THE OHIO STATE UNIVERSITY

FISHER COLLEGE OF BUSINESS

From Numbers to Words: Breaking Down Institutional Beliefs

Andrea Andolfatto and Federico Bastianello

Discussant: **Andrei S. Gonçalves**

Outline

The Paper and My Main Comment

My Comments on the Limitations

Final Remarks

Capital Market Assumptions (CMAs)

Capital Market Assumptions (CMAs)



2026 Long-Term Capital Market Assumptions

30th Annual Edition

Time-tested projections to
build stronger portfolios

J.P.Morgan
ASSET MANAGEMENT

Capital Market Assumptions (CMAs)

Annualized Volatility (%)				U.S. Inflation	U.S. Cash	U.S. Intermediate Treasuries	U.S. Long Treasuries	TIPS	U.S. Aggregate Bonds	U.S. Securitized	U.S. Short Duration Government/Credit	U.S. Long Duration Government/Credit	U.S. Inv Grade Corporate Bonds	U.S. Long Corporate Bonds	U.S. High Yield Bonds	U.S. Leveraged Loans	World Government Bonds hedged	World Government Bonds	World ex-U.S. Government Bonds hedged	World ex-U.S. Government Bonds	Emerging Markets Sovereign Debt	Emerging Markets Local Currency Debt	Emerging Markets Corporate Bonds
Arithmetic Return 2026 (%)																							
Compound Return 2026 (%)																							
Fixed Income	U.S. Inflation	2.50	2.52	1.77	1.00																		
	U.S. Cash	3.10	3.10	0.67	-0.03	1.00																	
	U.S. Intermediate Treasuries	4.00	4.06	3.48	-0.27	0.20	1.00																
	U.S. Long Treasuries	4.90	5.69	13.02	-0.22	0.05	0.84	1.00															
	TIPS	4.30	4.47	5.88	-0.03	0.02	0.63	0.61	1.00														
	U.S. Aggregate Bonds	4.80	4.91	4.76	-0.24	0.10	0.85	0.85	0.76	1.00													
	U.S. Securitized	5.20	5.28	4.19	-0.22	0.10	0.79	0.75	0.70	0.94	1.00												
	U.S. Short Duration Government/Credit	4.00	4.01	1.63	-0.29	0.32	0.85	0.61	0.63	0.82	0.77	1.00											
	U.S. Long Duration Government/Credit	5.20	5.81	11.45	-0.22	0.04	0.75	0.91	0.70	0.94	0.82	0.67	1.00										
	U.S. Inv Grade Corporate Bonds	5.20	5.46	7.39	-0.19	0.02	0.55	0.62	0.73	0.87	0.77	0.67	0.86	1.00									
	U.S. Long Corporate Bonds	5.40	6.10	12.28	-0.19	0.01	0.54	0.69	0.68	0.87	0.75	0.61	0.92	0.97	1.00								
	U.S. High Yield Bonds	6.10	6.46	8.74	-0.01	-0.03	0.03	0.04	0.51	0.41	0.41	0.31	0.36	0.67	0.61	1.00							
	U.S. Leveraged Loans	6.60	6.88	7.70	0.15	-0.04	-0.34	-0.26	0.18	0.05	0.06	-0.04	0.05	0.36	0.32	0.76	1.00						
	World Government Bonds hedged	4.30	4.38	4.02	-0.29	0.12	0.87	0.87	0.64	0.88	0.80	0.74	0.85	0.68	0.70	0.17	-0.17	1.00					
World Government Bonds	4.30	4.55	7.27	-0.18	0.13	0.77	0.68	0.69	0.82	0.74	0.76	0.76	0.71	0.70	0.35	-0.03	0.75	1.00					
World ex-U.S. Government Bonds hedged	4.20	4.27	3.95	-0.30	0.12	0.74	0.74	0.59	0.79	0.71	0.65	0.77	0.66	0.67	0.24	-0.07	0.96	0.68	1.00				
World ex-U.S. Government Bonds	4.30	4.69	9.06	-0.17	0.12	0.68	0.58	0.66	0.76	0.69	0.71	0.69	0.70	0.68	0.42	0.04	0.67	0.98	0.63	1.00			
Emerging Markets Sovereign Debt	6.30	6.66	8.83	-0.16	0.03	0.34	0.38	0.63	0.67	0.61	0.51	0.64	0.83	0.79	0.76	0.46	0.47	0.62	0.49	0.65	1.00		
Emerging Markets Local Currency Debt	6.70	7.38	12.12	-0.06	0.11	0.26	0.22	0.47	0.50	0.44	0.44	0.46	0.63	0.60	0.62	0.35	0.30	0.63	0.31	0.69	0.81	1.00	

Muni 1-15 Yr Blend

Muni 1-15 Yr Blend

Capital Market Assumptions (CMAs)

		Annualized Volatility (%)			U.S. Inflation	U.S. Cash	U.S. Intermediate Treasuries	U.S. Long Treasuries	TIPS	U.S. Aggregate Bonds	U.S. Securitized	U.S. Short Duration Government/Credit	U.S. Long Duration Government/Credit	U.S. Inv Grade Corporate Bonds	U.S. Long Corporate Bonds	U.S. High Yield Bonds	U.S. Leveraged Loans	World Government Bonds hedged	World Government Bonds	World ex-U.S. Government Bonds hedged	World ex-U.S. Government Bonds	Emerging Markets Sovereign Debt	Emerging Markets Local Currency Debt	Emerging Markets Corporate Bonds
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	U.S. Aggregate Bonds	4.80	4.91	4.76	-0.24	0.10	0.85	0.85	0.76	1.00														
	U.S. Securitized	5.20	5.28	4.19	-0.22	0.10	0.79	0.75	0.70	0.94	1.00													
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	World Government Bonds hedged	4.30	4.38	4.02	-0.29	0.12	0.87	0.87	0.64	0.88	0.80	0.74	0.85	0.68	0.70	0.17	-0.17	1.00						
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Emerging Markets Sovereign Debt	6.30	6.66	8.83	-0.16	0.03	0.34	0.38	0.63	0.67	0.61	0.51	0.64	0.83	0.79	0.76	0.46	0.47	0.62	0.49	0.65	1.00			
Emerging Markets Local Currency Debt	6.70	7.38	12.12	0.06	0.11	0.26	0.22	0.47	0.50	0.44	0.44	0.46	0.63	0.60	0.62	0.35	0.30	0.63	0.31	0.69	0.81	1.00		

Capital Market Assumptions (CMAs)

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3. Muni 1-15 Yr Blend

Literature: Properties of CMAs and Link to Allocations

- CMAs have “sensible” $\mathbb{E}[r]$ properties and are linked to allocations

(e.g., Delgado, Iltis (2024, 2026), Costa, Gonçalves, Lellis (2023), Costa, Gonçalves, Iltis, Lellis (2024))

Literature: Properties of CMAs and Link to Allocations

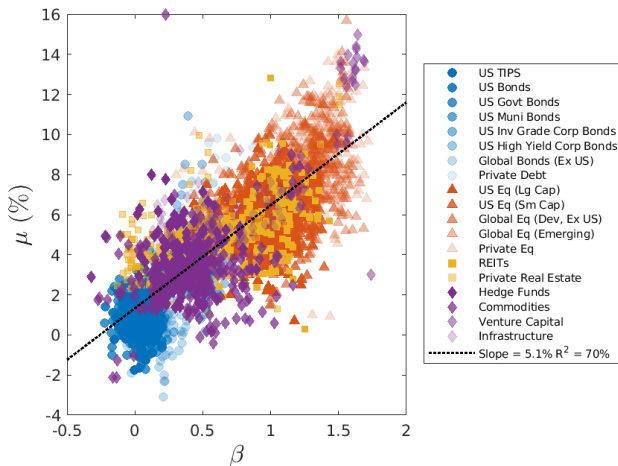
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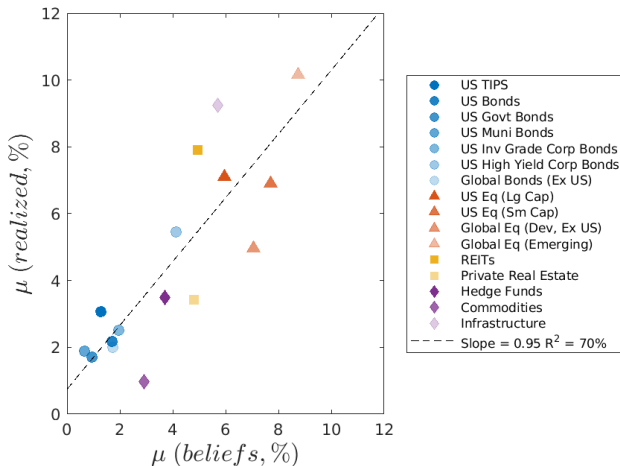


Source: Coutts, Gonçalves, Loudis (2023)

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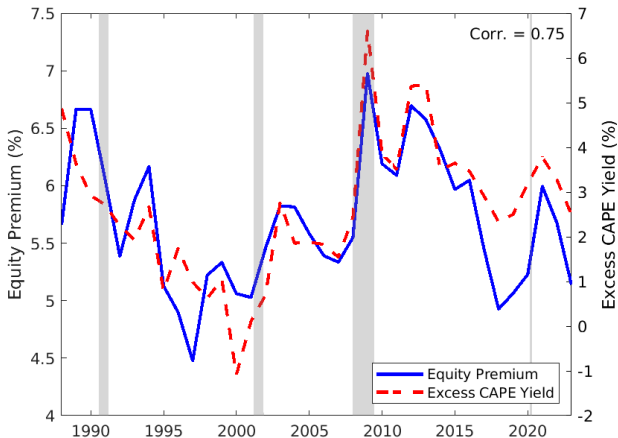
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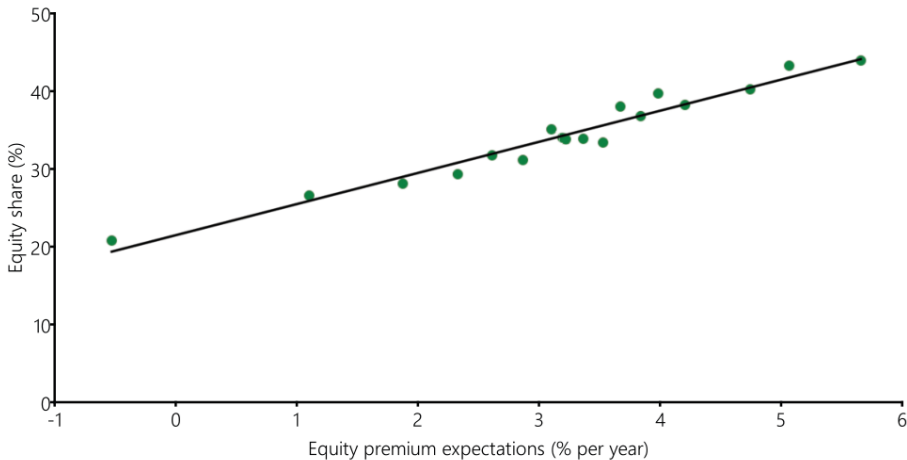


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Source: Dahlquist, Ibert (2024) (see also Andonov, Coutts, Gonçalves, Loudis, Rossi (2026))

This Paper: CMAs have Information Beyond Forecasts

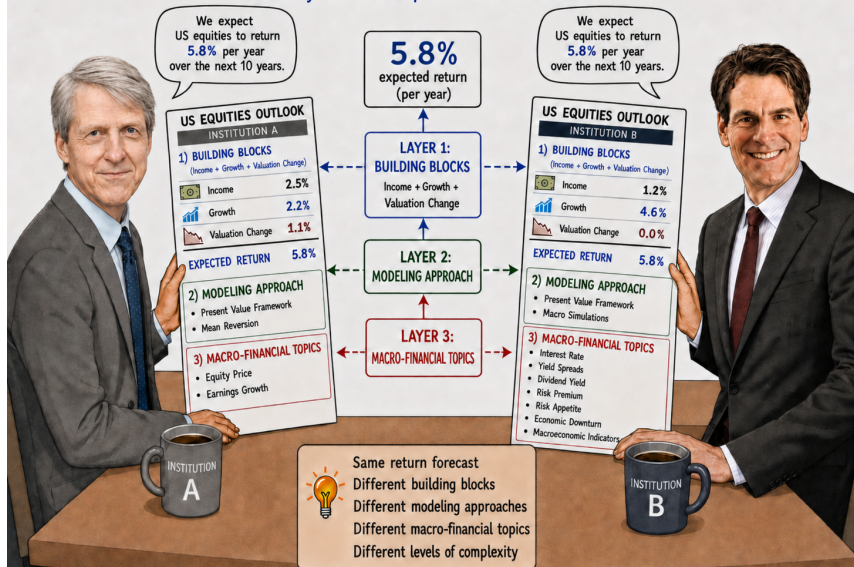
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This Paper: CMAs have Information Beyond Forecasts

SAME NUMBER. DIFFERENT UNDERLYING INFORMATION.

Three layers behind reported return forecasts



Outline

The Paper and My Main Comment

My Comments on the Limitations

Final Remarks

The Main Limitations

1) The Results on Cyclicalities & Forecast Errors

- The sample period is too short to infer about on these properties

2) The Anchoring Result

- If I was building CMAs, I would anchor on consensus (to get it right)

3) The Centrality of Topics

- Topics that require more discussion may look more “central”

4) The Backward Looking Second Moments

- CMA second moments contain important forward looking information

(Limitation 1) The Results on Cyclicalty & Forecast Errors

- Because these are long-term expectations, they are persistent
- So, 2008-2026 period is short to study cyclicalty
- Specially given that the data concentrate in the 2016-2025 period

(Limitation 1) The Results on Cyclicalty & Forecast Errors

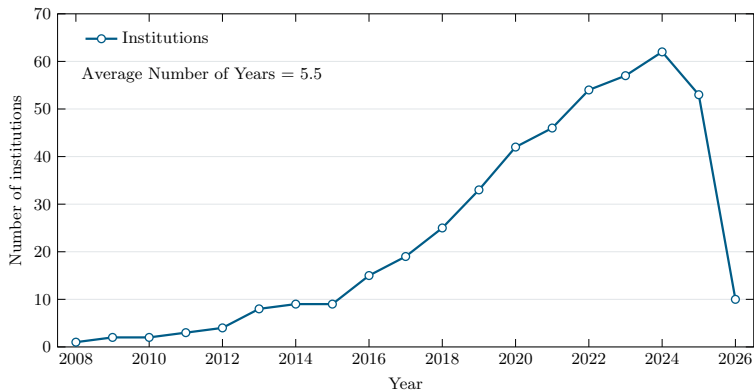
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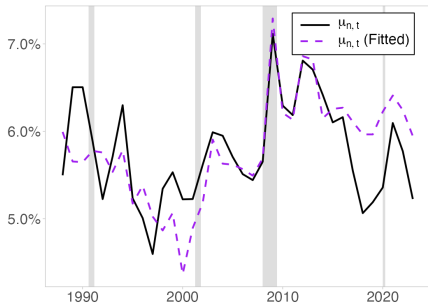
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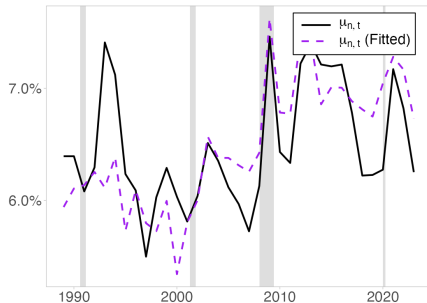
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US Equity



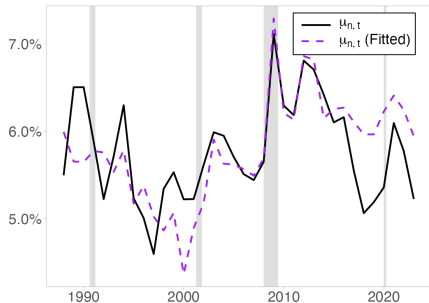
Ex US Equity



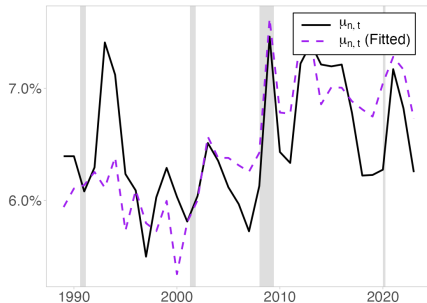
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US Equity



Ex US Equity



(Limitation 2) The Anchoring Result

$$\Delta F_{i,t} = \alpha_i + \beta \cdot (F_{i,t-1} - \bar{F}_{t-1}) + \delta \cdot \Delta \bar{F}_t + \epsilon_{i,t}$$

- $\beta < 0$ has two possible interpretations:
 - You argue for (1): “their forecasts should tend to lie on the same side of the machine-learning benchmark as the recent consensus.”
 - But $\beta > 0$ here can also be a “rational response to information”

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$$G_t - F_{i,t} = \alpha_i + \beta \cdot (G_t - \bar{F}_{t-1}) + \epsilon_{i,t}$$

- But $\beta > 0$ here can also be a “rational response to information”
 - Suppose $\bar{F}_t \neq G_t$, but it is better than any $F_{i,t}$ (e.g., $F_{i,t} = G_t + u_{i,t}$)
 - Then, CMAs would want their $F_{i,t}$ to “anchor” on $\bar{F}_{t-1}$
 - $\bar{F}_t$ is still not full information rational, but the response is rational

(Limitation 2) The Anchoring Result

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$$G_t - F_{i,t} = \alpha_i + \beta \cdot (G_t - \bar{F}_{t-1}) + \epsilon_{i,t}$$

- But $\beta > 0$ here can also be a “rational response to information”
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 - Then, CMAs would want their $F_{i,t}$ to “anchor” on $\bar{F}_{t-1}$
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Source: Coutts, Gonçalves, Loudis (2023)

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Variation = Return Horizon =	Across Asset Classes		Over Time	
	1-Year	3-Year	1-Year	3-Year
Slope	0.79**	0.84***	1.74***	1.11*
(se)	(0.32)	(0.17)	(0.58)	(0.60)
R^2_{Oos}	10.6%	14.9%	11.0%	11.9%

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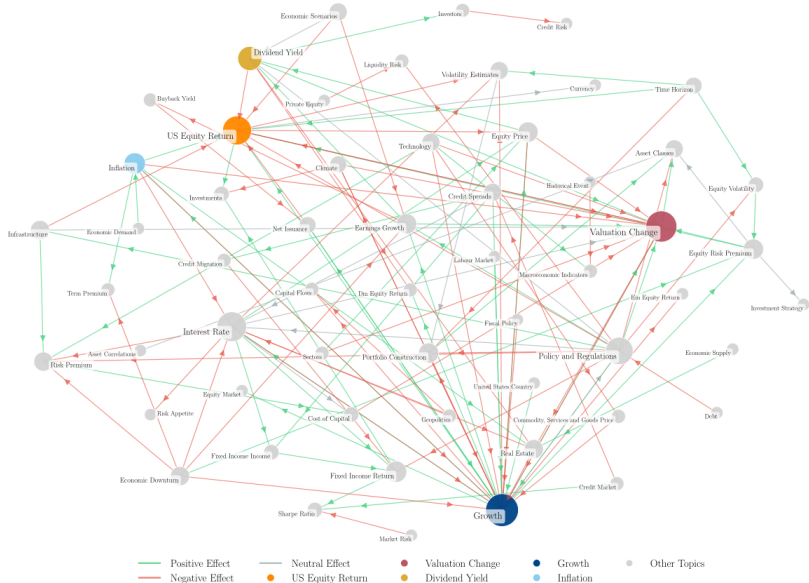
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(Limitation 3) The Centrality of Topics

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(Limitation 4) The Backward Looking Second Moments

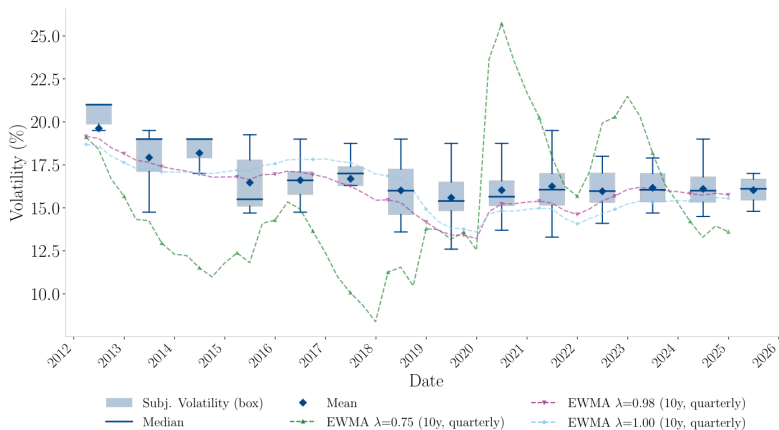
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(Limitation 4) The Backward Looking Second Moments

PANEL A - Variation Across Asset Classes						
	σ^2			β		
	$\kappa = 0.5$	$\kappa = 0.7$	$\kappa = 0.9$	$\kappa = 0.5$	$\kappa = 0.7$	$\kappa = 0.9$
Past Return Moment	0.76*** (0.17)	0.91*** (0.16)	1.10*** (0.11)	0.71*** (0.07)	0.73*** (0.07)	0.75*** (0.07)
R^2	61.6%	73.3%	86.4%	80.4%	82.9%	86.1%

PANEL B - Variation Over Time						
	σ^2			β		
	$\kappa = 0.5$	$\kappa = 0.7$	$\kappa = 0.9$	$\kappa = 0.5$	$\kappa = 0.7$	$\kappa = 0.9$
Past Return Moment	0.12*** (0.03)	0.22*** (0.04)	0.63*** (0.08)	0.32*** (0.08)	0.44*** (0.09)	0.65*** (0.10)
R^2	12.5%	22.8%	41.2%	15.1%	24.0%	37.4%

Source: Couts, Gonçalves, Loudis (2023)

(Limitation 4) The Backward Looking Second Moments

PANEL C - Variation Over Time for US Equities (Large Cap)

	σ^2			β		
	$\kappa = 0.5$	$\kappa = 0.7$	$\kappa = 0.9$	$\kappa = 0.5$	$\kappa = 0.7$	$\kappa = 0.9$
Past Return Moment	0.22*	0.41*	1.31**	0.50***	0.56***	0.57***
	(0.12)	(0.21)	(0.58)	(0.09)	(0.09)	(0.10)
	R^2	24.6%	36.1%	34.7%	33.0%	25.6%
Excess CAPE Yield	0.22***	0.22***	0.22***	5.88***	5.88***	5.88***
	(0.08)	(0.08)	(0.08)	(1.97)	(1.97)	(1.97)
	R^2	36.7%	36.7%	31.6%	31.6%	31.6%
Past Return Moment	0.14*	0.30*	1.00**	0.41***	0.45***	0.43***
	(0.08)	(0.16)	(0.47)	(0.07)	(0.06)	(0.11)
	R^2	45.8%	53.0%	52.8%	51.8%	45.4%
Excess CAPE Yield	0.18***	0.16***	0.14***	4.61***	4.68***	4.83***
	(0.06)	(0.06)	(0.05)	(1.37)	(1.46)	(1.70)
	R^2	45.8%	53.0%	52.8%	51.8%	45.4%

(Limitation 4) The Backward Looking Second Moments

$$\widehat{Risk}_{n,t \rightarrow t+3} = \text{Fixed Effects} + \text{Slope} \cdot \overline{Risk}_{n,t} + \epsilon_{n,t}$$

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Variation =	Across Asset Classes		
Risk Measure =	σ^2	β	
Slope	1.04**	1.12***	
(se)	(0.23)	(0.04)	
R^2_{Oos}	20.2%	32.6%	

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Slope (se)	1.04** (0.23)	1.12*** (0.04)	-0.40 (0.66)	0.75*** (0.28)
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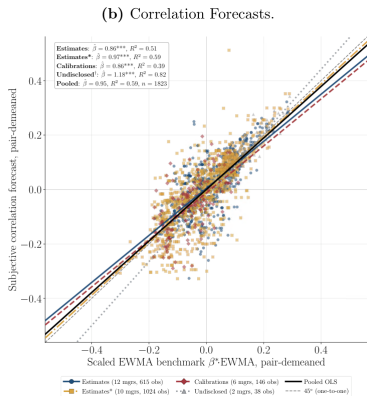
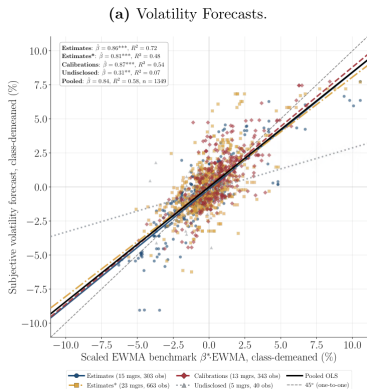
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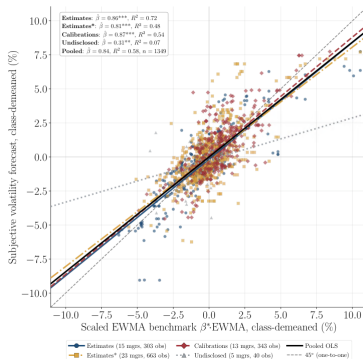


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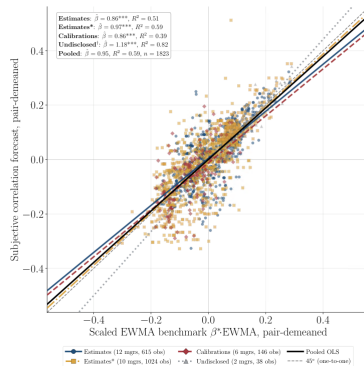
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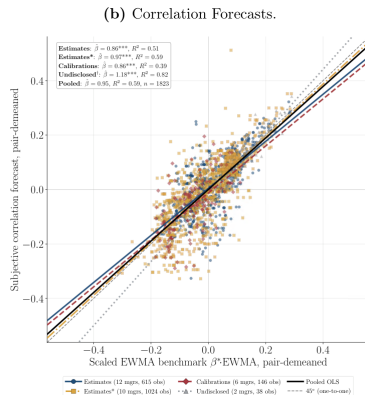
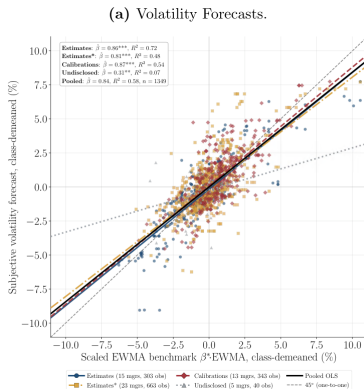
(b) Correlation Forecasts.



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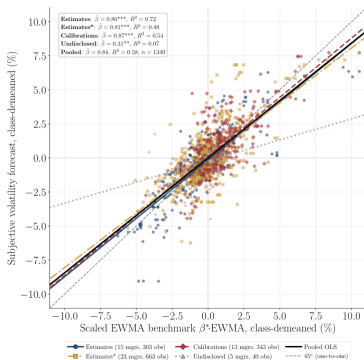


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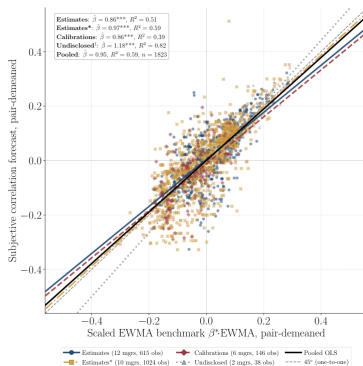
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Some Generic Comments

1. In anchoring regressions: do we want α_i or α_t or $\alpha_i + \alpha_t$?
we can even have $\alpha_{i,t}$ if combining asset classes or blocks
2. I would replace Type II ANOVA with Shapley Decomposition
(Shapley properly considers all combinations)
3. Objective beliefs should always be estimated out-of-sample
(sometimes they are, sometimes I am not sure)
4. Section 3.2 really needs to be simplified
(Tables 9 to 14 have related messages, probably some way to simplify)
5. The discussion of measurement error in Section 4 is wrong
(the institution-level estimation implies the error correlates with the y variable)

Outline

The Paper and My Main Comment

My Comments on the Limitations

Final Remarks

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 - Each of these layers is very informative about belief formation
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