

**“Idiosyncratic financial risk and a reevaluation of the
market risk-return tradeoff”**

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Discussion by:

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Boston College

AFA, San Antonio

January 7, 2024

¹I would like to thank Jade Peng for excellent research assistance.

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There is a risk-return trade-off **after all** ☆

Eric Ghysels^a, Pedro Santa-Clara^b, Rossen Valkanov^b  

← This title tells how much we have tried!

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↪ A realistic marginal investor also considers other wealth (e.g., labor) that contribute to her consumption. **“Omitted factor”**

Run of show

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More positive and significant
GARCH-in-mean coefficient.

$$r_{t+1} = \phi_0 + \phi_1 \cdot r_t + \gamma \cdot \sigma_t^2 + u_{t+1},$$

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Volatility (σ_t)	γ	0.19 (0.18)	0.68*** (0.22)	0.56*** (0.20)
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<i>Panel B: GJR-GARCH</i>				
Variance (σ_t^2)	γ	1.01 (1.88)	4.17** (1.67)	5.29*** (1.94)
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Volatility (σ_t)	γ	0.15 (0.19)	0.44*** (0.16)	0.49*** (0.17)
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<i>Panel C: E-GARCH</i>				
Variance (σ_t^2)	γ	1.61 (1.88)	6.17*** (2.08)	5.78** (2.26)
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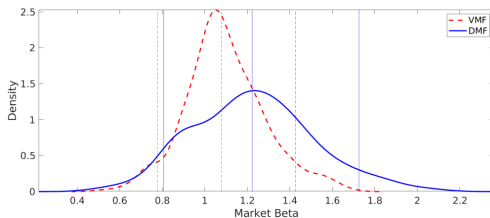
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2. Cross section:

Wider distribution of DMF-based market betas; explains more variation in the CS.



Nice paper!

▶ **What I like:**

- ↪ New, intuitive empirical approach to resolve the risk-return trade-off puzzle.
- ↪ Empirics are very carefully done and explained, involving multiple methodologies to examine the puzzle.

▶ **My extending thoughts:**

1. Interpretations & links to economics
2. Empirics
3. Streamlining

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- ▶ ↪ **Even more interesting (might be a separate paper):** Start with market ret; more directly estimate a part of it that “maximally” explains future changes in the economic conditions / financial conditions / non-real variables (up to h).

Empirics: On the main result

- **Main empirical result, Table 3:** Is the empirical benchmark too easy to reject anyways?

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→ The empirical literature has used other **CV proxies** (other than GARCH), other **horizons** (other than monthly), and other **conditioning variable** (other than econometricians' perspective) to prove γ can be significant and positive. Is GARCH, e-GARCH, or GJR-GARCH (which has been rejected and improved upon) still the best empirical benchmark to have?

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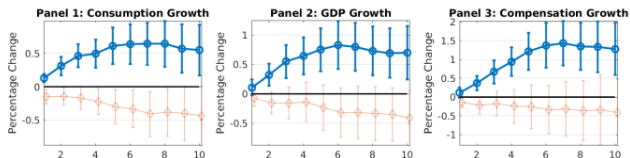
☞ Suggestion (2): Validate the σ_t^2 estimate from the GARCH-system. If the macro interpretation is solid, one would expect the $\sigma_{t|DMF}^2$ performs better than $\sigma_{t|VMF}^2$ in predicting future cumulative IP growth, conditional on VIX etc.

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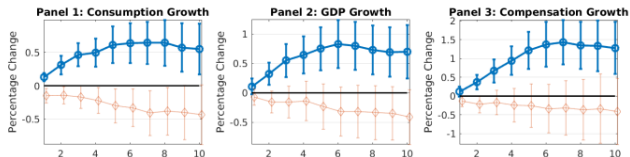
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→ For now, I see emphasis on “common” (↑). Some ideas on “risk”:

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☞ For now, I see emphasis on “common” (↑). Some ideas on “risk”:

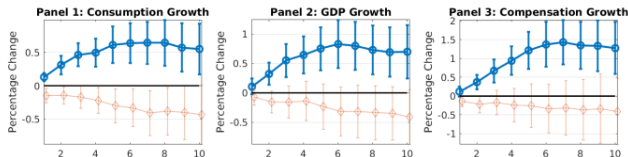
▶ Economic interpretation of DMF:

Intuitively, one almost wanted to test the other way around as well – How much do 1-period ahead (or cumulative) excess returns (based on DMF or PCs) look like given 1 SD increase in real economic volatility or risk shocks?

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▶ Compare good v. bad DMF and other PCs:

☞ Given what we know about the behaviors of RP and CV, in this plot, one should expect that bad DMF absorbed more non-linearity.

Streamlining the paper (put on a “Referee 2” hat)

► Theoretical frameworks:

1. Section 4.1 (stylized model with human capital and financial capital) to show labor share could change the composition of risk premium
2. Section 4.2 (dynamic reduced-form asset pricing model) to show an affine solution of equity risk premium that contains a true market factor compensation, and et al.;
3. Section 5.1 derives the conditional variance, based on Section 4.2's setup;
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- ▶ **Clarification:** Explain (and cite) the difference between this paper and *“Uncovering the Risk-Return Relation in the Stock Market”* by Hui Guo and Robert F. Whitelaw (2006, JF) – they also estimate cleanse market returns into a pure risk component and find relative risk aversion is positive and significant.

Conclusion

▶ **Highly recommend:**

The idea that, “vw-ret is not a good market return proxy and let’s improve on it,” is cool! Time to provide another perspective to revisit risk-return trade-off. I am very sympathetic to this direction.

▶ **Concrete suggestions:**

Streamlining the paper with clearer interpretations, fewer “models,” and more versions of empirical benchmarks (parameteric or not).

Thank You!

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