

# When is Growth at Risk?

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# Background

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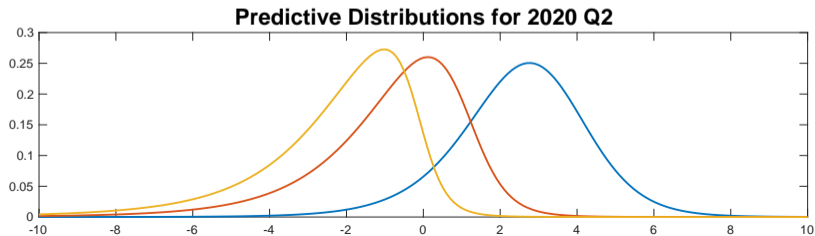
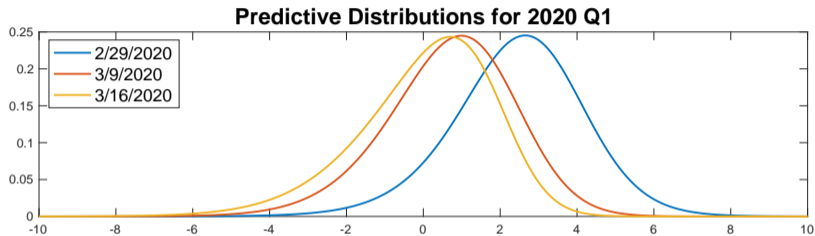
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(Adrian, Grinberg, Liang and Malik 2018, Adrian et al. 2019)
- ▶ **Empirical implication: financial variables** may be helpful for predicting moments other than the mean and **signal risks for growth**

# The Economic Impact of COVID-19

## Predictive distributions of GDP growth using financial and real information



# Research Questions

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2. Do financial variables contain **advance information** for higher moments of the predictive distribution of GDP growth?

## Contributions:

- ▶ Disentangle **financial specific information** from information that is **contemporaneous and common** to macro and financial variables
- ▶ We assess **uncertainty** in different forms:
  - ▶ Out-of-sample at one and four quarters
  - ▶ In sample uncertainty in estimation of moments
  - ▶ Data uncertainty – real time and calendar



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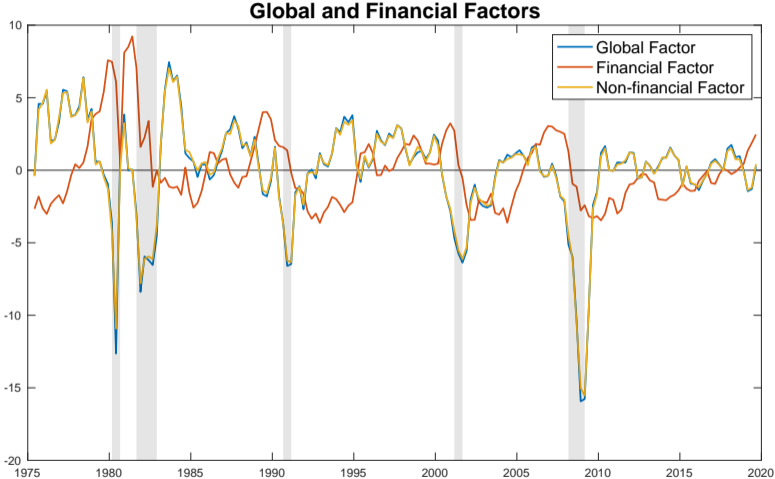
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⇒ Results

1. **Global** factor and **non-financial** factor are **almost identical**
2. The orthogonal financial factor could potentially capture some **leading information**

# Global, Financial and Real Factors



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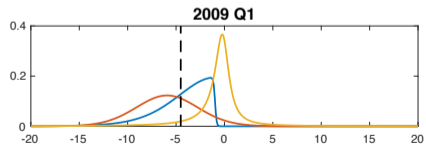
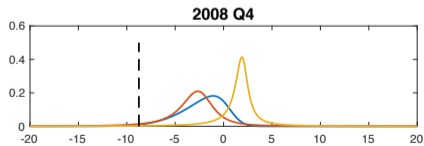
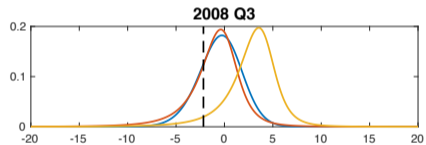
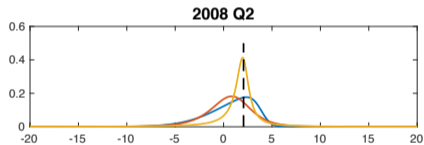
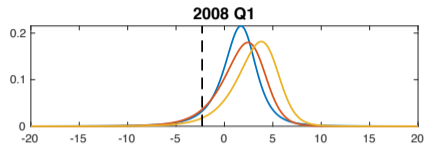
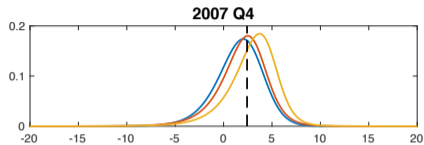
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### ⇒ Results:

1. The global factor helps predict the mean – no additional predictive power from the financial factor
2. Higher moments difficult to predict
3. Limited predictability at the nowcasting horizon for the mean

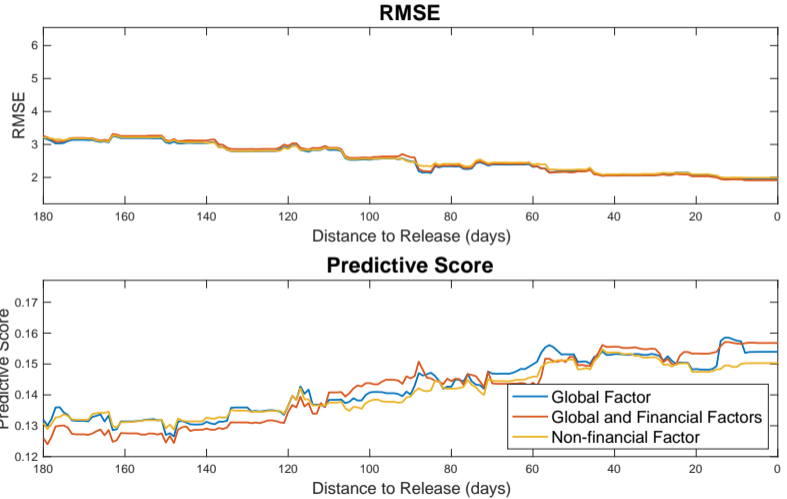
# Forecasting the GDP Distribution of the Great Recession

## One quarter ahead



— Global factor, Financial factor, and GDP — Global factor and GDP — GDP only

# Nowcasting in Real-Time: RMSFE and Predictive Score



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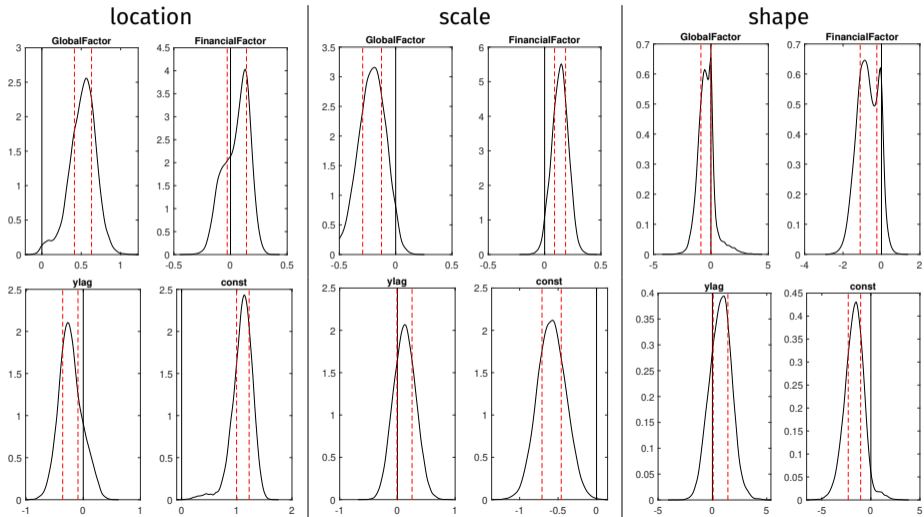
### ⇒ Results:

1. The **global factor** helps predict the **mean** – **no** additional **predictive power** from the **financial factor**
2. **Variance and mean** are **negatively correlated** via the global factor
3. Financial-specific information has a small, positive association with the variance of the predictive distribution
4. But the effects on the **variance** and **skewness** are **imprecisely estimated**



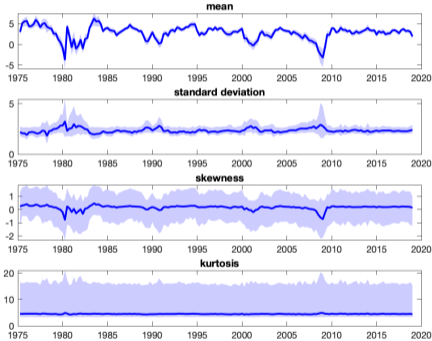
# Posterior Distribution of Moment Coefficients

## Posterior distributions

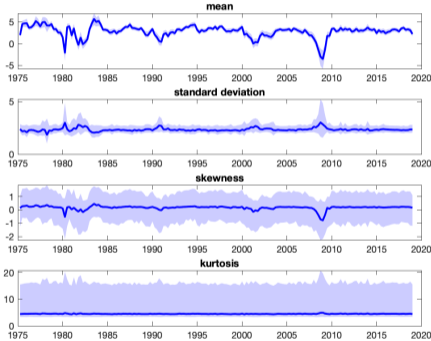


# Dynamics of Estimated Moments

**Global & Financial Factor**



**Only Global Factor**



## Step 4: Contribution of Individual Variables

### Objectives

- ▶ Evaluate whether we can **identify** some **key individual financial predictors**

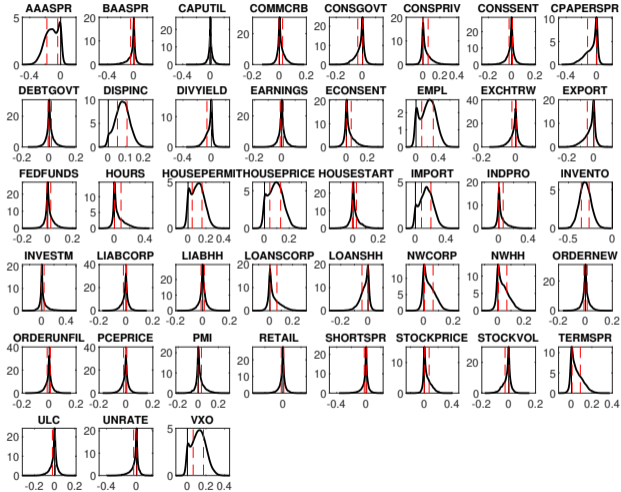
### ⇒ Method and data

- ▶ Bayesian parametric model (as in Step 3) but combined with a variable selection algorithm
- ▶ Data: quarterly US and 12 other OECD countries

### ⇒ Results:

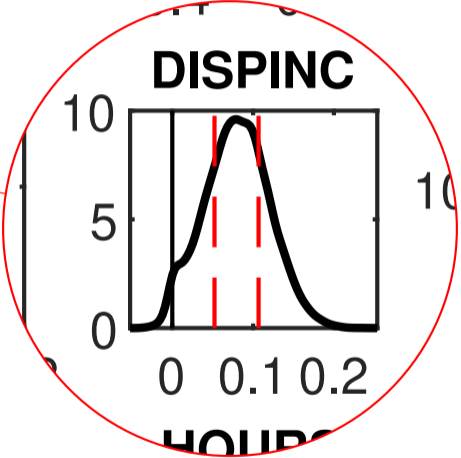
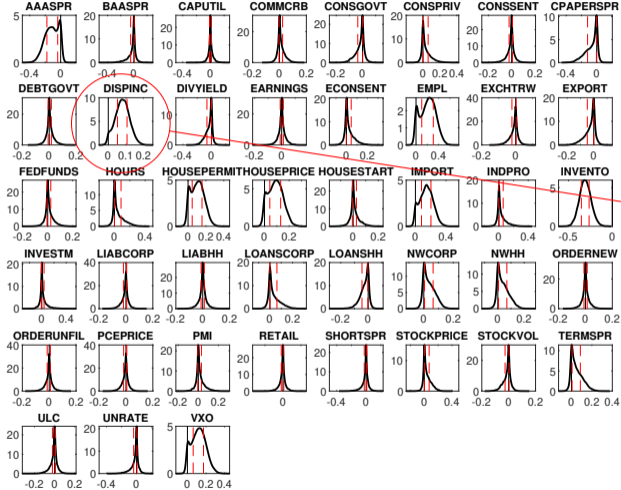
1. For the US, **few real and financial variables** have **predictive power** for the mean
2. Very limited predictability for higher moments
3. **No variables robustly predict** any **moment** across more than a handful of the 13 countries

# US: Predictors of the Mean

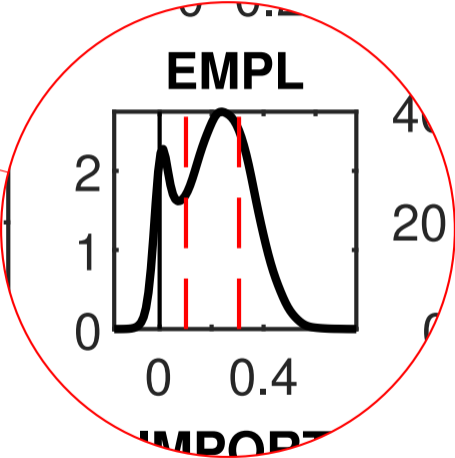
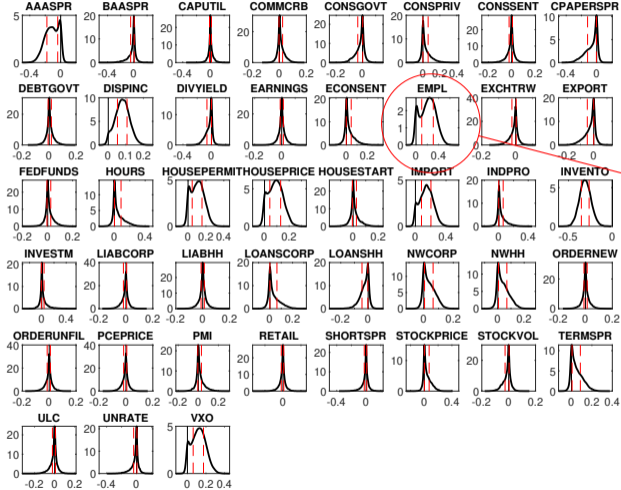




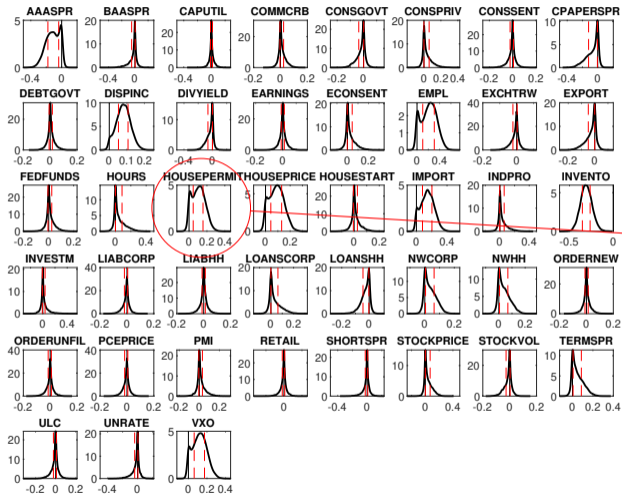
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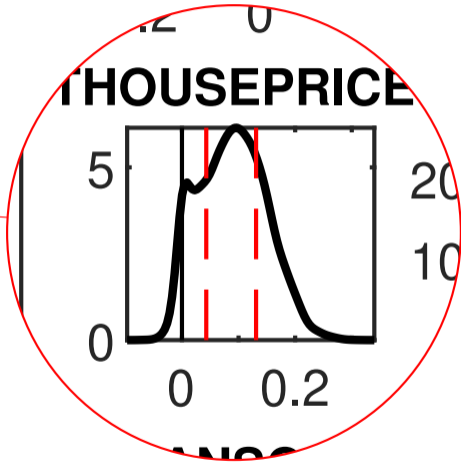
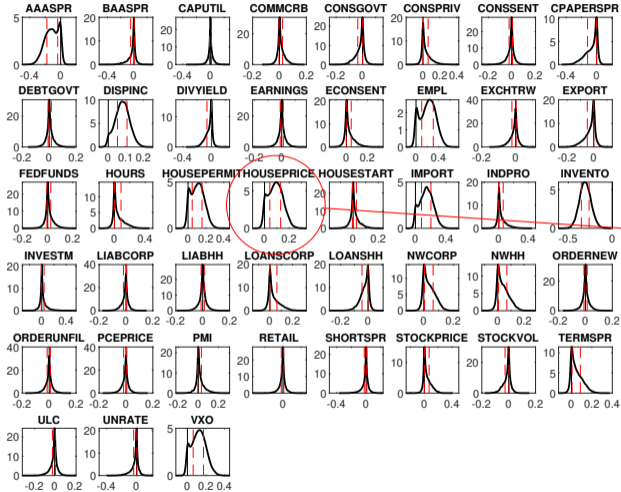


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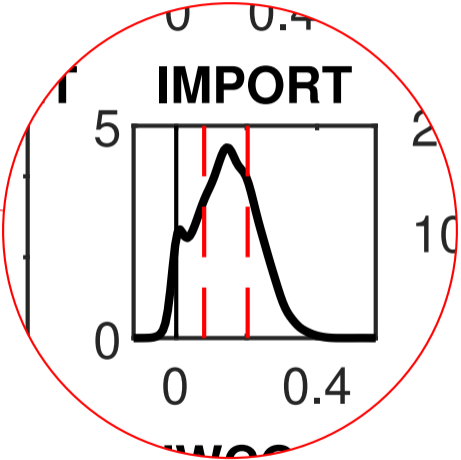
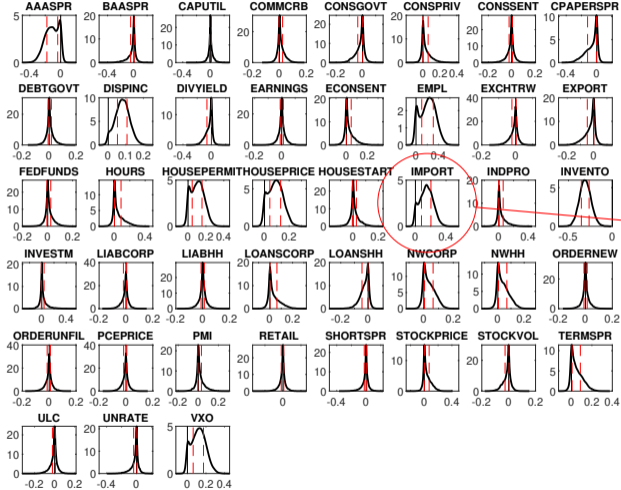




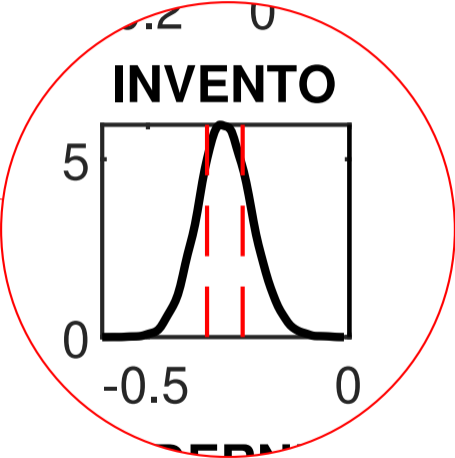
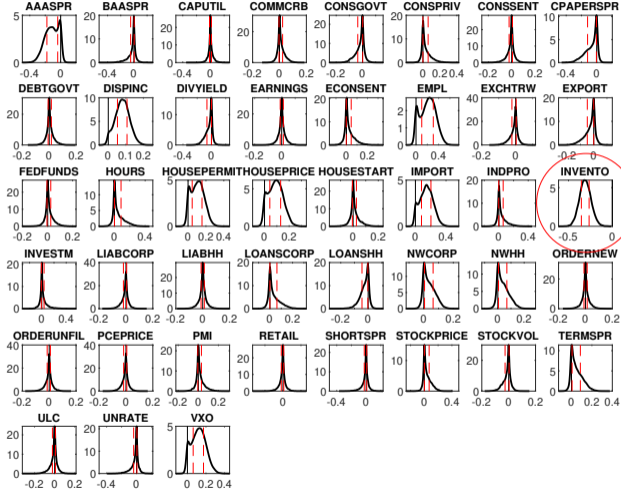
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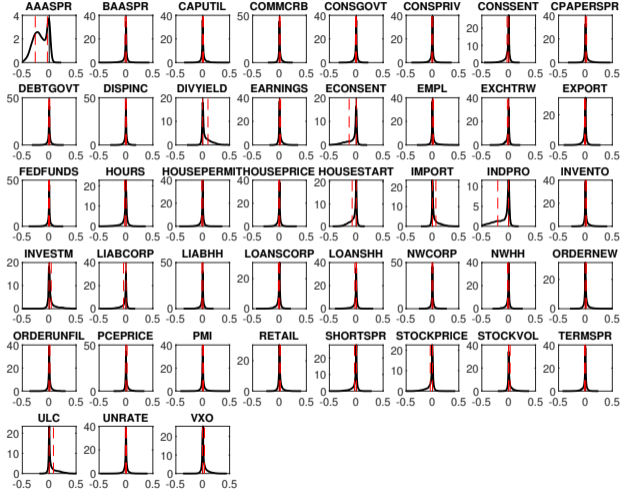
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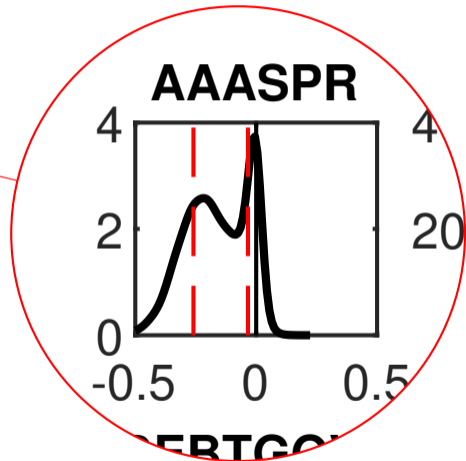
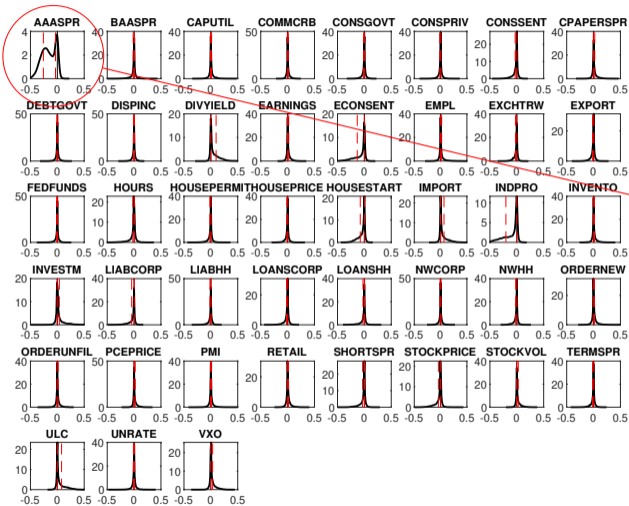
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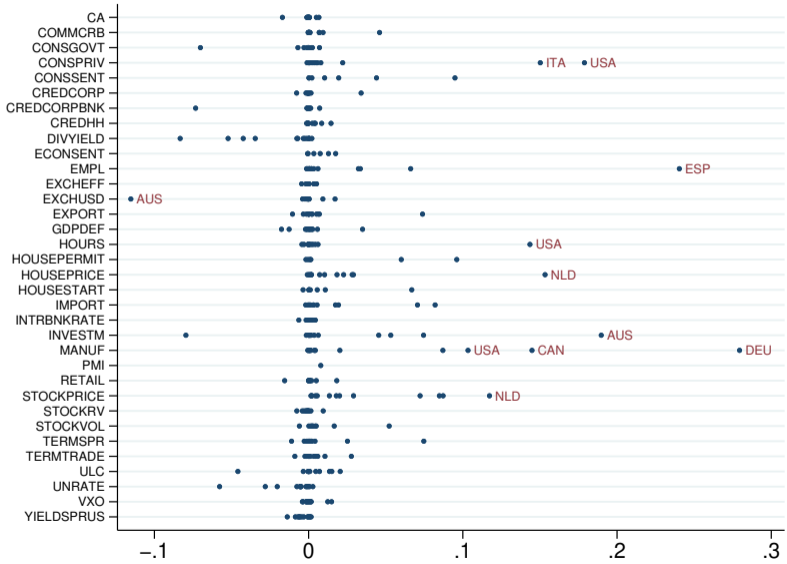
# US: Predictors of the Variance



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# Cross-Country Data: Predictors of the Mean



# Conclusions & Open Questions

1. **Markets** have **limited forward information** – most information is **contemporaneous**
2. Most of the **action is in the mean**, not the tails
3. **Non-linear relationships hard to capture**, either because of the small number of recessions/crises or because the magnitude of the non-linearity is small
4. However, some indication of **negative correlation between mean and variance**, although imprecisely estimated
5. **Be cautious when drawing structural conclusions** for modeling macro-finance link

⇒ Future research: Lots to understand!

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**Now!**