

The Real Effects of Financial Uncertainty Shocks: A Daily Identification Approach

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Introduction

- **Financial market uncertainty** raises in recession. Why?
 - Ludvigson et al. (2018): uncertainty causes recessions
 - Berger et al. (2020): recessions cause uncertainty

This Paper

- We examine the puzzle using a **novel identification strategy** that exploits **daily data** to disentangle the interactions between stock prices, bond spreads and volatility.

A 3-stage Identification Strategy

1. Identify the structural shock of interest in a daily VAR
2. Average the estimated daily shocks to the monthly frequency
3. Use monthly averages as an instrument in a monthly VAR

- By using daily data, we can identify uncertainty shocks accounting for the fact that **financial markets respond to many 'macro news'** on a monthly or quarterly basis \Rightarrow High-frequency data help in **disentangling exogenous shocks and endogenous responses**.

Does the empirical strategy work?

- In theory, if temporal aggregation is obtained by skip-sampling or averaging, then **an average of HF shocks correctly recovers the initial responses of the LF variables**.

– Intuition: in a linear model, $\sum_{t \in \tau} \varepsilon_t^y \approx \varepsilon_\tau^y$

- This strategy eliminates **temporal aggregation bias** if the DGP is a daily VAR.
- **Monte Carlo experiments** support this conclusion:

– Simulate data from a known high-frequency DGP assuming that one or more variables are not observed every period.

– We compute the Mean Absolute Distance (MAD) between the true and the estimated IRFs.

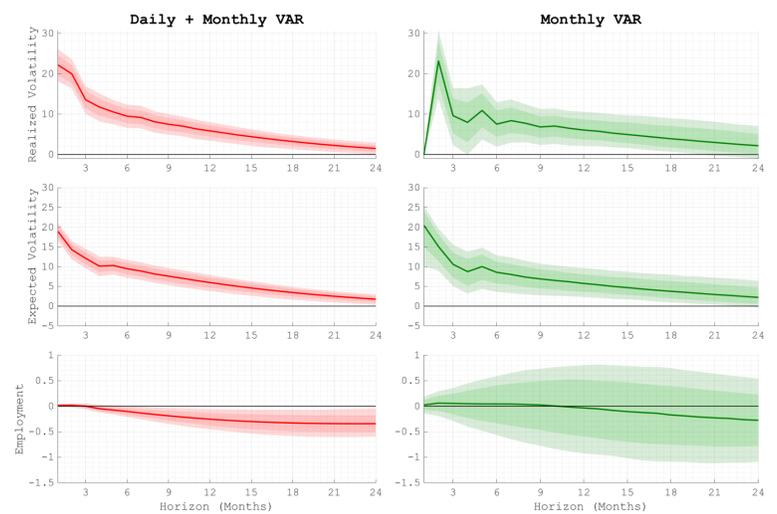
– We compare our strategy (**HF+LF VAR**) to the **standard approach (LF-VAR)** in recovering the true IRFs under alternative frequencies/temporal aggregation schemes.

	MAD ratio over HF-VAR							
	Monthly-Quarterly Case (3)				Daily-Monthly Case (30)			
Frequency Mismatch	Skip-sampling		Averaging		Skip-sampling		Averaging	
Temporal Aggregation	100		1000		100		1000	
Sample Size	100	1000	100	1000	100	1000	100	1000
LF VAR	2.86	7.69	5.88	20.03	5.00	14.29	33.33	10.01
HF+LF VAR	1.17	1.15	4.76	15.80	2.85	3.00	22.67	2.10

– We also compare it to established **mixed-frequency VARs** (Schorfeide & Song, 2015; Ghysels, 2016), and find it a valid alternative.

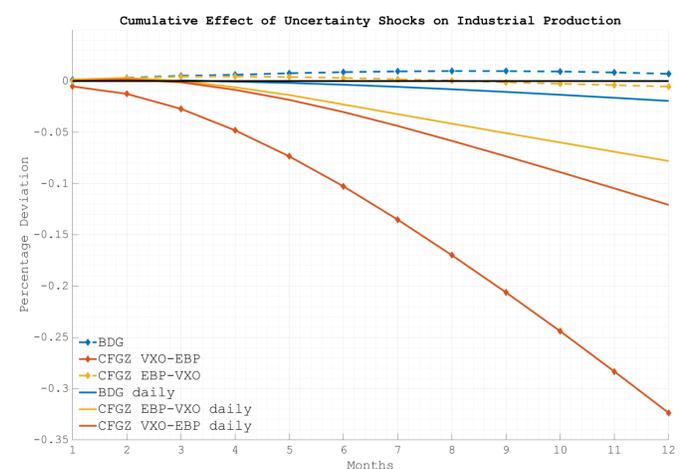
Impact of Financial Uncertainty Shocks

- **Berger et al. (2020)** \Rightarrow realized volatility (RV) versus option-implied volatility ($V \approx VXO$). Uncertainty shocks identified as the linear combination of residuals that maximizes the 2Y-ahead FEV of V but do not affect RV within a month.
- The theory says there is no *contemporaneous* impact of V on RV . Does that mean that RV does not respond for *one month*?
- We can answer using our 3-step procedure:
 - Estimate **daily VAR** with RV , V and financial covariates.
 - Assume uncertainty shocks maximize the 2Y-ahead FEV of V but do not affect RV *within a day*.
 - **Average** the daily series of uncertainty shocks and use them as external instruments in a **monthly VAR**.



- We also use the identification proposed by **Caldara et al. (2016)** but at daily frequency: identifying uncertainty shocks as the linear combination of *daily* residuals that maximizes the 6 month-ahead VXO response subject to orthogonality w.r.t. financial shocks.

- **The shocks turn out to have similar effects in the two setups:**



Conclusions

- We use daily data to identify uncertainty shocks with a more accurate information set and looser restrictions on LF variables.
- **Temporal aggregation matters:** uncertainty shocks reduce economic activity, and their impact is similar in Berger et al. (2020) and Caldara et al. (2016).