

# Synthetic Volatility Forecasting and Other Aggregation Techniques for Time Series Forecasting

## Preliminary Exam

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# A seemingly unprecedented event might provoke the questions

- 1 What does it resemble from the past?
- 2 What past events are most relevant?
- 3 Can we incorporate past events in a systematic, principled manner?

# When would we ever have to do this?

- Event-driven investing strategies (unscheduled news shock)
- Pairs trading strategies
- Structural shock to macroeconomic conditions (scheduled news possibly pre-empted by news shock)
- Biomedical panel data subject to exogenous shock or interference

Example: weekend of March 7th and 8th, 2020

# Punchline of the paper

Forecasting is possible under structural shocks, so long as we incorporate external information to account for the nonzero errors.

# Background and related methods

## Volatility Modeling

- GARCH is slow to react (Andersen et al. 2003)
- Asymmetric GARCH models catch up faster but need post-shock data
- Realized GARCH (Hansen, Huang, and Shek 2012), in our setting, would require post-shock information and/or high-frequency data in order to outperform, and the model is highly parameterized

# Background and related methods

## Forecast Augmentation

- Clements and Hendry 1998; Clements and Hendry 1996 laid the groundwork for modeling nonzero errors in time series forecasting
- Guerrón-Quintana and Zhong 2017 use a series' own errors to correct the forecast for that series
- Dendramis, Kapetanios, and Marcellino 2020 use a similarity-based procedure to correct linear parameters in time series forecasts
- Foroni, Marcellino, and Stevanovic 2022 adjust pandemic-era forecasts using intercept correction techniques and data from Great Financial Crisis
- Lin and Eck 2021 use distanced-based weighting (a similarity approach) to aggregate and weight fixed effects from a donor pool

# Outline

- 1 Introduction
- 2 Setting
  - Model Setup
  - Volatility Profile of a Time Series
- 3 Post-shock Synthetic Volatility Forecasting Methodology
- 4 Properties of Volatility Shock and Shock Estimators
- 5 Real Data Example
- 6 Numerical Examples
- 7 Discussion
- 8 Future directions for Synthetic Volatility Forecasting
  - Signal Recovery
  - Synthetic Impulse Response Functions
- 9 Supplement

# The news has broken but markets are closed

- $y \in \mathbb{R}^n$ , a mean-zero, real-valued response to be predicted



# A Primer on GARCH

# Our Model is Nested Within GARCH-X

# Volatility Profile

In this particular setting, excess risk of an estimator  $\theta$  has the form

$$R(\theta)$$

# What's the method here?

$$2 = 3$$

# Minimum Norm Estimator

# Key Conceptual Innovation: Effective Rank

# Main Result: Existence Proof, Dichotomy, and Bounds

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Remarks



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## Remarks

- 1 A gives us a (high probability) upper bound on the excess risk.

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# Some things to think about with papers like this

# Simplest Simulation Setup

# A more benign example

Example (Coverging at the slowest rate possible)

Fix  $\alpha = 1, \beta > 1$ . Let  $\lambda_i = \frac{1}{i \log^\beta(i+1)}$ .



# How noise is hidden just right

After all of this waiting, we formalize the notion under discussion.

Definition (Asymptotically Benign)

We analyze the real-world example with Brexit included.

# Bibliography

-  Andersen, Torben G et al. (2003). “Modeling and forecasting realized volatility”. In: *Econometrica* 71.2, pp. 579–625.
-  Clements, Michael and David F Hendry (1998). *Forecasting economic time series*. Cambridge University Press.
-  Clements, Michael P and David F Hendry (1996). “Intercept corrections and structural change”. In: *Journal of Applied Econometrics* 11.5, pp. 475–494.
-  Dendramis, Yiannis, George Kapetanios, and Massimiliano Marcellino (2020). “A similarity-based approach for macroeconomic forecasting”. In: *Journal of the Royal Statistical Society Series A: Statistics in Society* 183.3, pp. 801–827.
-  Forni, Claudia, Massimiliano Marcellino, and Dalibor Stevanovic (2022). “Forecasting the Covid-19 recession and recovery: Lessons from the financial crisis”. In: *International Journal of Forecasting* 38.2, pp. 596–612.
-  Guerrón-Quintana, Pablo and Molin Zhong (2017). “Macroeconomic