

A paper

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somewhere

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- ▶ N forecasts, $\{f_{it}\}_{i \in [N]}$, available at time t

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- ▶ N forecasts, $\{f_{it}\}_{i \in [N]}$, available at time t
- ▶ Question:
 - ▶ What weight $\mathbf{w} \in \{\mathbf{w} \in \mathbb{R}^N : \sum_{i \in [N]} w_i = 1\}$ minimizes the forecast error variance $E \left[(y_{t+1} - \mathbf{w}'\mathbf{f}_t)^2 \right]$

Empirical Example (Macro)

- ▶ European Central Bank's surveys of professional forecasters. CPI, 1-year-ahead or 2-year-ahead. Data: 1999Q1–2018Q4 (20 years), about 120 forecasters. Unbalanced, 30 forecasters of complete record. 40-quarter rolling window.

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Section 1

Problem and Algorithm

Bates and Granger (1969)

- ▶ Forecast error $\mathbf{e}_t = (e_{1t}, \dots, e_{Nt})'$ with $e_{it} = y_{t+1} - f_{it}$.
- ▶ Sample variance-covariance (VC): $\widehat{\Sigma} := T^{-1} \sum_{t=1}^T \mathbf{e}_t \mathbf{e}_t'$.

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- ▶ Bates and Granger: (**Problem-BG**)

$$\min_{\mathbf{w} \in \mathbb{R}^N} \frac{1}{2} \mathbf{w}' \widehat{\Sigma} \mathbf{w} \quad \text{subject to} \quad \mathbf{w}' \mathbf{1}_N = 1.$$

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- ▶ When $\hat{\Sigma}$ is invertible,

$$\hat{\mathbf{w}}^{\text{BG}} = \hat{\Sigma}^{-1} \mathbf{1}_N / \mathbf{1}_N' \hat{\Sigma}^{-1} \mathbf{1}_N.$$

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- ▶ When N is close to T , the inverse is ill-conditioned.
- ▶ When $N > T$, sample VC is singular.

Summary

- ▶ Constrained quadratic optimization formulates many economic and financial questions.
- ▶ We propose L2-relaxation to stabilize the numerical solution.
- ▶ A new machine learning algorithm to improve the estimation.
- ▶ Asymptotic optimality under group structure.