Asset Pricing in the Frequency Domain: Theory and Empirics

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Discussed by Jaroslav Borovička

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- 'myopic' and 'hedging' demand
- innovation to the SDF

$$\Delta E_{t+1}\left[m_{t+1}\right] = -\left(\sum_{k=0}^{\infty} z_k g_k\right) \varepsilon_{t+1}$$

• correlation between $\{z_k\}$ and $\{g_k\}$

Discrete-time Fourier transform

representation in the frequency domain

$$G(\omega) = \sum_{k=0}^{\infty} g_k e^{-i\omega k} \qquad Z(\omega) = \sum_{k=0}^{\infty} z_k e^{-i\omega k}$$

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Parseval's theorem (frequency domain representation of a correlation)

$$\sum_{k=0}^{\infty} z_k g_k = \frac{1}{2\pi} \int_{-\pi}^{\pi} Z(\omega) G(\omega) d\omega$$

• $Z(\omega)$ operates as a filter over macroeconomic risk $G(\omega)$ at different frequencies

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This paper: What can we learn from the spectral decomposition of preferences $Z(\omega)$

- Estimate $\{g_k\}$ (G (ω)) from data (VAR)
- Estimate different specifications for $Z(\omega)$
 - Some are linearizations of conventional preferences
 - Others have more statistical basis: aversion to risk at different frequencies

Goals

Why do we do all this?

- 1. Intuition
 - How do preferences load on different frequencies?
- 2. Estimation
 - Spectral decomposition cannot bring in any new information.
 - What if models are misspecified?
 - Estimating reduced form preference specification in the frequency domain.

Intuition



Aversion to / preference for persistence



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 - What if we take the model seriously and start fishing for cash flows which are underpriced/overpriced?
 - Very similar cash flows with frequencies concentrated around the steps should be priced quite differently.
 - Security design: spuriously attractive investment opportunities with very high Sharpe ratios.

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 - relative volatilities of the permanent and transitory component, etc.

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- Approximation errors
 - logarithms vs levels
 - ▶ loglinear approximation \implies bandpass filter \implies what happens to the SDF in levels?