

# ASSET PRICING IN THE FREQUENCY DOMAIN: THEORY AND EMPIRICS

by Ian Dew-Becker and Stefano Giglio

Discussed by Urban Jermann

# Contribution

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- Theoretical results: Frequency domain representation for SDF with applications
- Empirical results: Estimation of risk prices by frequency from cross-section, find significantly priced low frequency risk
- Analysis suggests EZ closer to data than Habits

# Theory

- Focus on innovations in SDF with time-non-separabilities

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$$\Delta E_{t+1} m_{t+1} = - \sum_j \left( \sum_{k=0}^{\infty} \underbrace{z_k}_{\text{Price of risk}} \cdot \underbrace{g_{j,k}}_{\text{Imp rsp.}} \right) \varepsilon_{j,t+1}$$



# Result 1

$$\begin{aligned}\Delta E_{t+1} m_{t+1} &= - \sum_j \left( \sum_{k=0}^{\infty} z_k g_{j,k} \right) \varepsilon_{j,t+1} \\ &= - \sum_j \left( \frac{1}{2\pi} \int_{-\pi}^{\pi} Z(\omega) G_j(\omega) d\omega \right) \varepsilon_{j,t+1}\end{aligned}$$

with

$$Z(\omega) \equiv z_0 + 2 \sum_{k=1}^{\infty} z_k \cos(\omega k)$$

$$G_j(\omega) \equiv \sum_{k=1}^{\infty} g_{j,k} \cos(\omega k)$$

# Derivation

- #1: Discrete-time Fourier Transform, and Inverse

$$\tilde{G}_j(\omega) \equiv \sum_{k=0}^{\infty} e^{-i\omega k} g_{j,k}, \text{ and } g_{j,k} = \frac{1}{2\pi} \int_{-\pi}^{\pi} \tilde{G}_j(\omega) e^{-i\omega k} d\omega$$

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- #2: Using fact that  $g_{j,k} = 0$ , for  $k < 0$ , and algebra

$$\sum_{k=0}^{\infty} z_k \left( \frac{1}{2\pi} \int_{-\pi}^{\pi} \tilde{G}_j(\omega) 2 \cos(\omega k) d\omega \right)$$

- $$\sum_{k=0}^{\infty} z_k g_{j,k} = \frac{1}{2\pi} \int_{-\pi}^{\pi} \tilde{G}_j(\omega) \underbrace{\left[ z_0 + \sum_{k=1}^{\infty} z_k 2 \cos(\omega k) \right]}_{\equiv Z(\omega)} d\omega,$$

$Z(\omega) \equiv$  'weighting function'

- #3: A lot more algebra

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- Frequency response function:  $H(\omega) = |H(\omega)| \cdot e^{-i\psi(\omega)}$
- Need Inverse "Dew-Becker-Giglio" Transform

$$z_k = \begin{cases} \frac{1}{2\pi} \int_{-\pi}^{\pi} Z(\omega) d\omega & \text{for } k = 0 \\ \frac{1}{\pi} \int_{-\pi}^{\pi} Z(\omega) \cos(\omega k) d\omega & \text{for } k > 0 \end{cases}$$

# Alternative derivation

- #1: Parseval's theorem

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

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- $Z(\omega)$  is complex, because  $\sum_{k=0}^{\infty} z_k L^{-k}$

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- Need Inverse D-B-G Transform, split weights



# Habits versus Epstein-Zin

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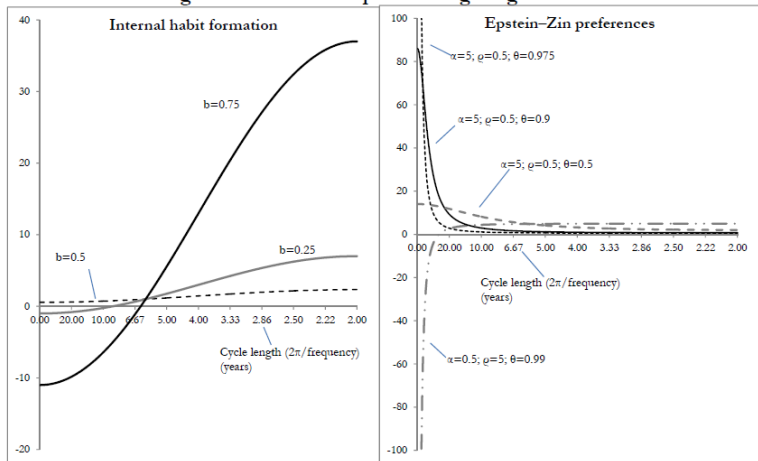
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Figure 2. Theoretical spectral weighting functions



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- External habits, only innovations in  $\Delta c_{t+1}$ , spectrum is flat

$$\exp(m_{t+1}) = \beta \frac{(C_{t+1} - b\bar{C}_t)^{-\alpha}}{(C_t - b\bar{C}_{t-1})^{-\alpha}}$$

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- Bandpass basis

$$Z^{BP}(\omega) = q_1 Z^{(\infty, 32\text{quarters})} + q_2 Z^{(32, 6)} + q_3 Z^{(6, 2)}$$

- Apply Inverse Transform

$$\begin{aligned}
 \Delta E_{t+1} m_{t+1} &= -\bar{q}' \left( \sum_{j=0}^{\infty} \underbrace{H_j^{U,BP}}_{\text{Freq.func.}} B_1 \Phi^j \right) \underbrace{(X_{t+1} - \Phi X_t)}_{\text{VAR state var.}} \\
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- Moment conditions

$$E \left( R_{i,t} - R_{t-1}^f \right) = -E \left( -\bar{q}' u_{t+1}, r_{i,t} - r_{t-1}^f \right)$$



# Results

Portfolios:		FF25			
Basis:		Bandpass	t-stat	Utility (0.975)	t-stat
Consumption growth	q1	269	2.47 **	555.47	1.66 *
	q2	-431	-1.17	-442.65	-0.44
	q3	138	0.33	616.12	0.32

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- Epstein-Zin Risk Aversion coefficient implied by Utility basis

$$q_1 = (\alpha - \rho) 2$$

$$\text{Risk Aversion: } \alpha = \frac{555.47}{2} + \rho > 277.7$$

# Other comments

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- Cross-section: Drivers of results, and literature (ex: Parker and Julliard (2005), Hansen, Heaton, Lee & Roussanov (2007))

# Conclusion

Nice paper!