Smooth Diagnostic Expectations: A Discussion

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Empirical Evidence

Overreaction to News

Survey forecasts show consistent overreaction (Bordalo Gennaioli Ma Shleifer 2020).

Horizon-dependent Overreaction

Overreaction intensifies at longer forecast horizons.

Overconfidence

3

Subjective uncertainty often underestimates true uncertainty.

Uncertainty Amplification

Novel finding: High uncertainty environments amplify overreaction.

What do Smooth DE bring to the table?

State-dependent overreaction

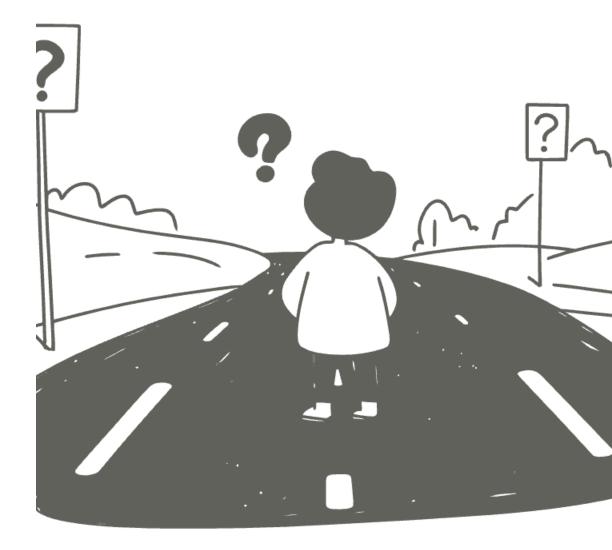
Extends literature on distorted conditional mean forecasts to higher-order moments.

Methodological advancement

Incorporates distortions to
higher-order distribution
moments in dynamic GE models.
models.

Uncertainty distortions

Explores interactions with the "news shock" channel in economic modeling. modeling.



Business Cycle Dynamics with Smooth DE

	Belief Distortions
1	Connects with time-series dynamics
2	Asymmetric Cycles Sharper recessions than expansions
3	Countercyclical Volatility At micro and macro levels
4	Policy Insights Reducing idiosyncratic uncertainty stabilizes macroeconomy

Smooth DE vs. Standard DE: A Comparison



Flexibility

Smooth DE adapts dynamically to uncertainty uncertainty levels, unlike standard DE.



Uncertainty Interaction

Smooth DE accounts for interaction with uncertainty, which standard DE ignores.



Overconfidence Explanation

Smooth DE provides a comprehensive explanation for overconfidence in forecasts.

$$\frac{|X|N}{R} = \frac{|X|N}{2} + 2|X|/(2n) \qquad \qquad \frac{|Z|N}{2} + 2|N|, \qquad \frac{|Z|N}{N} + \frac{|Z|N}{2} + \frac{|Z|N}$$

Key Equations

$$R_{t+h|t,t-J} = \frac{\sigma_{t+h|t}^2}{\sigma_{t+h|t-J}^2}$$

$$E_t^{\theta}(x_{t+h}) = \mu_{t+h|t} + \underbrace{\theta \frac{R_{t+h|t,t-J}}{1 + \theta(1 - R_{t+h|t,t-J})}}_{\equiv \tilde{\theta}_t} (\mu_{t+h|t} - \mu_{t+h|t-J})$$

$$V_t^{\theta}(x_{t+h}) = \frac{\sigma_{t+h|t}^2}{1 + \theta(1 - R_{t+h|t,t-J})}$$

$$\frac{|x|^{2}}{|x|} = \frac{3\sqrt{7} + \frac{1}{2}\sqrt{(x)}}{4\sqrt{x}} \frac{2\sqrt{x}}{|x|} \frac{|x|^{2}}{|x|} \frac{|x|^{2}}{|$$

- ▶ Overreaction increases with relative uncertainty: $\frac{\partial \tilde{\theta}_t}{\partial R_{t+h|t,t-J}} > 0$.
- ▶ Overconfidence when $R_{t+h|t,t-J}$ < 1.
- ▶ Underconfidence when $R_{t+h|t,t-I} > 1$.

Insights from an AR (1) example

- ▶ Smooth DE applied to an AR(1) process: $x_{t+1} = \rho x_t + \epsilon_{t+1}$.
- Effective overreaction ($\tilde{\theta}_t$) depends on the ratio of conditional variances:

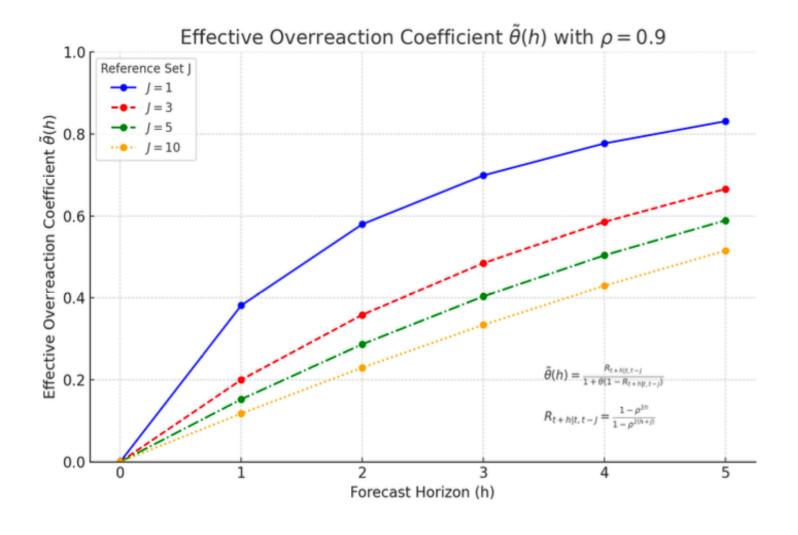
$$R_{t+h|t,t-J} = \frac{1-\rho^{2h}}{1-\rho^{2(h+J)}}.$$

- Predictions:
 - Overreaction increases with forecast horizon h.
 - Subjective overconfidence ($V_t^{\theta} < V_t$) typical in stationary settings.
 - Explains survey findings: stronger overreaction for long-term forecasts.

$$\frac{|X|}{80} = \frac{5x7}{24(6n)} = \frac{7x-2h}{4x} = \frac{5x-2h}{25(n)} = \frac$$

$R_{t+h|t,t-J}$ AS A FUNCTION OF J AND h

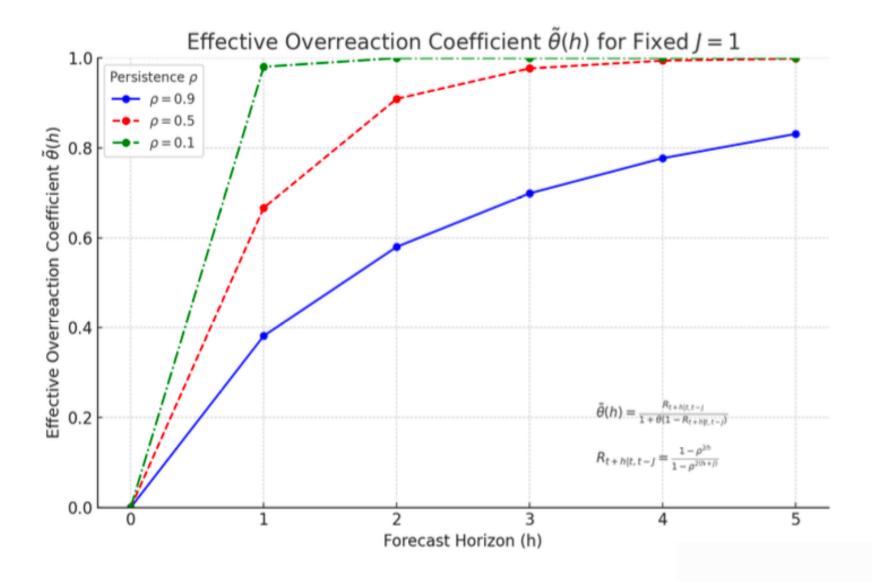
- ► Vary J (e.g., J = 1, 3, 5, 10).
- Fix $\rho = 0.9$ and compute $\tilde{\theta}_t^{h,J}$.



$$\frac{|X|^{N}}{|X|} = \frac{5}{5}X^{T} + \frac{1}{2}2A(|x_{0}|) \frac{|X|}{|x_{0}|} \frac{1}{|x_{0}|} \frac{1}{|x_{0}|} \frac{|X|}{|x_{0}|} \frac{1}{|x_{0}|} \frac{1}{|x_{0}|}$$

$R_{t+h|t,t-J}$ as a Function of ρ and h

- ightharpoonup Fixed J=1.
- ▶ Vary ρ across [0.1, 0.9] and compute $\tilde{\theta}_t^{h,l}$.



$$\frac{\sum_{K \in S, K} x_{i}}{L_{i}(K)} = \frac{\sum_{K \in S, K} x_{i}}{L_{i}(K)} \frac{1}{L_{i}(K)} \frac{1}{L_{i}(K)$$

Sign of $\frac{\partial R}{\partial \rho}$ for Fixed J

- ▶ For J = 1:
 - $ightharpoonup \frac{\partial R}{\partial \rho} > 0$: R increases with ρ .
 - Sensitivity is moderate.
- ► For J = 3:
 - $ightharpoonup rac{\partial R}{\partial
 ho} > 0$: *R* increases more steeply with ho.
 - Sensitivity grows as *J* increases.
- ► General trend:
 - ► Higher *J* amplifies sensitivity of *R* to ρ.
 - For fixed h, $\frac{\partial R}{\partial \rho} > 0$ for all J.

effective overreaction is decreasing in ρ



Conclusions

Unified Framework Overreaction and overconfidence **Empirical Support** Lot of Applications Comment(s) Explore a bit more on the simple applications What does it do for a simple NK model with J=1? Managing micro uncertainty matters for macro-stabilization!!! Huge!

Show that micro uncertainty matters for macro-stabilization with

Great paper!!

Smooth DE --- Huge!