## 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

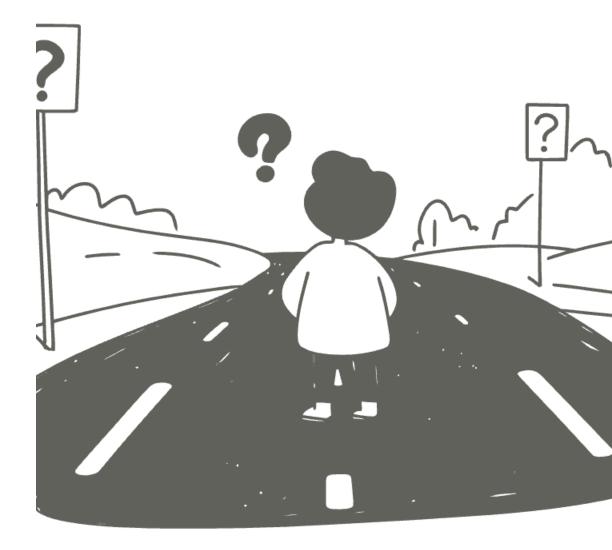
Extends literature on distorted conditional mean forecasts to higher-order moments.

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:

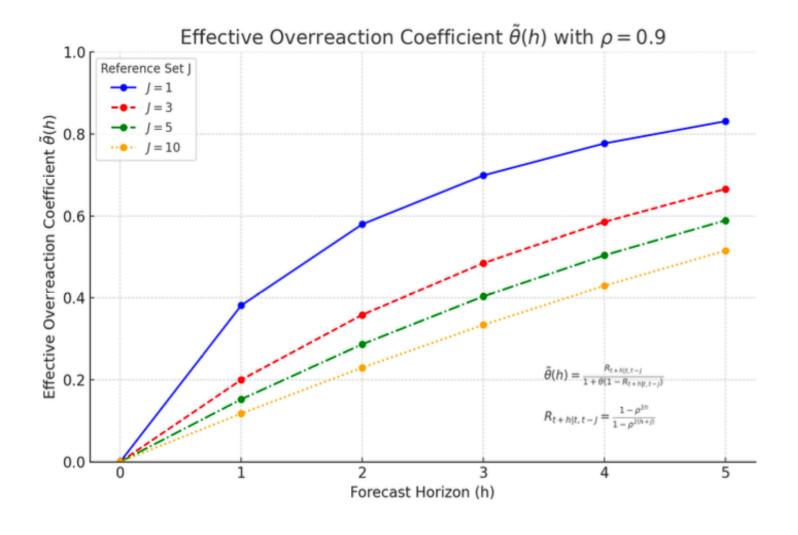
$$R_{t+h|t,t-J} = \frac{1-\rho^{2h}}{1-\rho^{2(h+J)}}.$$

- 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.

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

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

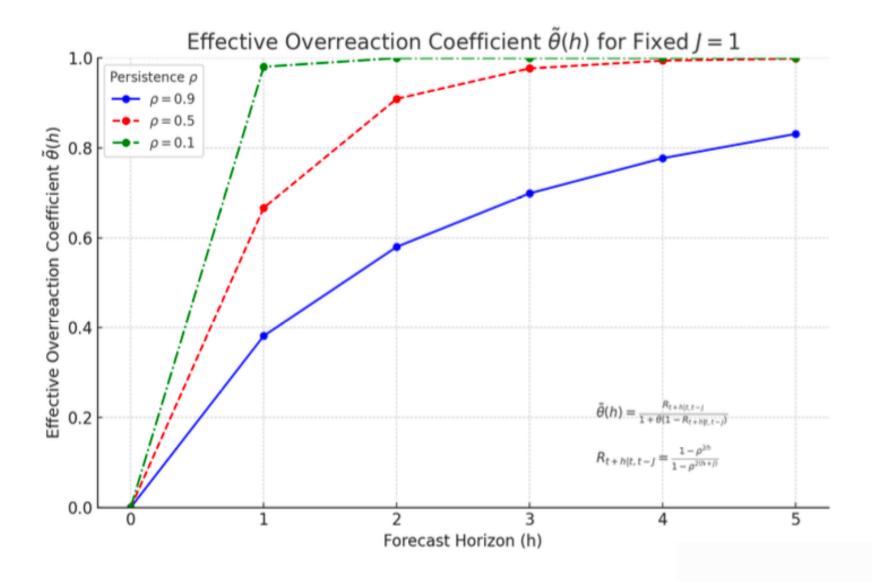
- ► Vary J (e.g., J = 1, 3, 5, 10).
- Fix  $\rho = 0.9$  and compute  $\tilde{\theta}_t^{h,J}$ .



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

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

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



Sign of 
$$\frac{\partial R}{\partial \rho}$$
 for  $h=1$ 

- For J = 1:
  - $ightharpoonup rac{\partial R}{\partial \rho}$  < 0: *R* decreases with  $\rho$ .
  - Sensitivity decreases at higher  $\rho$  values.
- For J = 2:
  - $ightharpoonup rac{\partial R}{\partial \rho} < 0$ : R continues to decrease with  $\rho$ .
  - $\triangleright$  Negative derivative becomes more pronounced for small  $\rho$ .
- ► For J = 3 and J = 4:
  - $\frac{\partial R}{\partial \rho}$  < 0: Similar trends observed.
  - ▶ Higher *J* values amplify the rate of decrease.
- ▶ Implication: For small horizons h, R is consistently decreasing in  $\rho$  across all J.

effective overreaction is decreasing in  $\rho$ 



#### 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!