

# Discussion of “Individual Uncertainty and Attentiveness to Macroeconomic Conditions”

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## Review of Main Results

- ▶ **How do different types of individual uncertainty correlate with attention to inflation?**
- ▶ Proxy of inattention: Size of forecast errors for consumer price and home price inflation in NY Fed SCE microdata

## Review of Main Results

- ▶ **How do different types of individual uncertainty correlate with attention to inflation?**
- ▶ Proxy of inattention: Size of forecast errors for consumer price and home price inflation in NY Fed SCE microdata
- ▶ Smaller forecast errors when conditional\* income uncertainty is high
  - \* Uncertainty about future income if hypothetically in same job at same employer
    - ▶ Precautionary attention
- ▶ Larger forecast errors when unemployed
  - ▶ Job search effort crowding out attention

# Comments

- ▶ Measures of attention
- ▶ Other moments
- ▶ Role of sentiment
- ▶ Theory?

# Measures of attention

Does size of forecast errors reflect attention?

$$\pi_{t+1} = \rho\pi_t + v_{t+1}, \quad v_{t+1} \text{ iid normal}$$

$$s_t = \pi_t + \epsilon_t, \quad \epsilon_t \text{ iid normal and indep of } v$$

$$\pi_{t+1} - \pi_{t+1|t} = \rho [(1 - m_t)(\pi_t - \pi_{t|t-1}) + m_t\epsilon_t] + v_{t+1}$$

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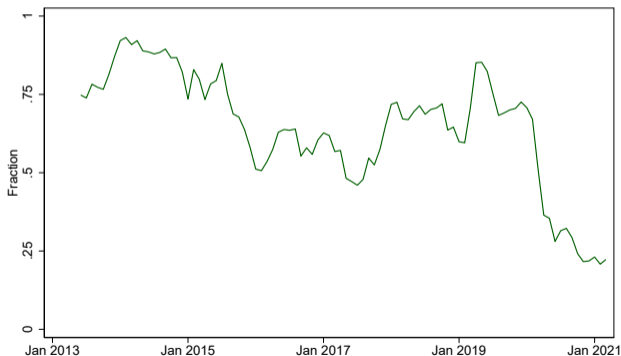
- ▶ No clear monotonic relationship between size of forecast error and attention ( $m_t$ )
- ▶ Persistence of forecast errors may provide a better test (Bracha and Tang 2022)
- ▶ Can also use CG (2015) regressions to elicit properties of attention
- ▶ Or inflation forecast uncertainty directly since, under rational inattention,  
$$\text{Var}(\pi_{t+1} - \pi_{t+1|t}|t) = \rho^2(1 - m_t)\sigma_{t|t-1}^2 + \sigma_v^2$$

# Other moments?

## Inflation forecast errors and mean forecasts

- ▶ Given consumers' bias in forecasting inflation, larger forecast errors are often just higher mean forecasts

Fraction of Inflation Forecasts Above Actual



## Other moments?

### Inflation forecast errors and mean forecasts

- ▶ Given consumers' bias in forecasting inflation, larger forecast errors are often just higher forecasts

HDFE Linear regression		Number of obs	=	67,615		
Absorbing 4 HDFE groups		F( 1, 8783)	=	395.87		
Statistics robust to heteroskedasticity		Prob > F	=	0.0000		
		R-squared	=	0.6707		
		Adj R-squared	=	0.6179		
		Within R-sq.	=	0.2586		
Number of clusters (userid) = 8,784		Root MSE	=	2.7133		
(Std. err. adjusted for 8,784 clusters in userid)						
absinflerr	Robust		t	P> t	[95% conf. interval]	
	Coefficient	std. err.				
inflexp	.4110188	.0206578	19.90	0.000	.3705248	.4515129
_cons	2.014446	.0744966	27.04	0.000	1.868416	2.160477

- ▶ Similar results for house price inflation.



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Income uncertainty vs “optimism” about income

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HDFE Linear regression		Number of obs	=	67,615		
Absorbing 4 HDFE groups		F( 1, 8783)	=	721.32		
Statistics robust to heteroskedasticity		Prob > F	=	0.0000		
		R-squared	=	0.8176		
		Adj R-squared	=	0.7884		
		Within R-sq.	=	0.2491		
Number of clusters (userid) = 8,784		Root MSE	=	1.9685		
(Std. err. adjusted for 8,784 clusters in userid)						
income_iqr	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
income_cent75	.3072904	.0114416	26.86	0.000	.2848622	.3297186
_cons	1.788638	.0499475	35.81	0.000	1.690729	1.886547

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- ▶ Given downward nominal wage rigidity, income uncertainty may be mainly driven by increased probability of higher income

HDFE Linear regression		Number of obs	=	67,615		
Absorbing 4 HDFE groups		F( 1, 8783)	=	34.98		
Statistics robust to heteroskedasticity		Prob > F	=	0.0000		
		R-squared	=	0.7614		
		Adj R-squared	=	0.7232		
		Within R-sq.	=	0.0178		
Number of clusters (userid) = 8,784		Root MSE	=	2.2515		
(Std. err. adjusted for 8,784 clusters in userid)						
income_iqr	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
income_cent25	-.0938503	.0158678	-5.91	0.000	-.1249548	-.0627458
_cons	3.24603	.0196021	165.60	0.000	3.207606	3.284455

## Other moments?

- ▶ In general, important to control for first moments when looking at relationships involving uncertainty.
- ▶ Particularly important in this context because of relationships between inflation forecasts and absolute forecast errors and between upper tail of income distribution and income uncertainty.

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- ▶ Partly due to relationship between variables examined and other moments of the distribution.
  - ▶ A pessimistic consumer may expect a higher probability of unemployment, a lower chance of large income increases, and higher inflation, which yield larger absolute forecast errors due to positive inflation expectation bias

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- ▶ Partly due to relationship between variables examined and other moments of the distribution.
  - ▶ A pessimistic consumer may expect a higher probability of unemployment, a lower chance of large income increases, and higher inflation, which yield larger absolute forecast errors due to positive inflation expectation bias
- ▶ Broader point: Sentiment may drive relationships between not just mean expectations of variables, but also uncertainty
- ▶ Complicates causal inference based on relationships between different survey responses

# Theory?

- ▶ What would theory predict about inflation/macro vs income uncertainty?
- ▶ Reis (2006) predicts that attentiveness rises with volatility of the exogenous income shock
  - ▶ But in reality, income uncertainty may be endogenous and could be lowered with more attentiveness
  - ▶ Could potentially reduce endogeneity by looking at income uncertainty increases due to job changes
- ▶ Is this more about attention to individual versus aggregate conditions (Mackowiak and Wiederholt 2009)?



## Conclusions

- ▶ Paper has very interesting suggestive results so far
- ▶ Consider other measures/tests of attention
- ▶ Control for other moments to isolate relationship with “pure” uncertainty
- ▶ Think more about theoretical predictions