

Fundamental Disagreement

Philippe Andrade
(Banque de France)

Richard Crump
(FRBNY)

Stefano Eusepi
(FRBNY)

Emanuel Moench
(Bundesbank)

“Price-setting and inflation”
Paris, Dec. 17 & 18, 2015

The views expressed here are the authors' and are not representative of ones of the Banque de France, the Bundesbank, the Eurosystem, the Federal Reserve Bank of New York or the Federal Reserve System.

Disagreement About Future Economic Outcomes

- Well documented in every survey of financial analysts, households, professional forecasters, FOMC members. . .
- At odds with full information rational expectation setup.
- Heterogeneous beliefs: important role in models with info. frictions
 - Macro: Mankiw-Reis (2002), Sims (2003), Woodford (2003), Lorenzoni (2009), Mackowiak-Wiederholt (2009), Angeletos-Lao (2013), . . .
 - Finance: Scheinkman-Xiong (2003), Nimark (2009) . . .
- Empirical properties of disagreement informative about such models?

This Paper

- Document some new facts about forecasters' disagreement.
- Evaluate models' ability to match the facts.
- Consider two popular models of imperfect info:
 - **Noisy information model:** Sims (2003), Woodford (2003).
 - **Sticky information model:** Mankiw and Reis (2002).
- Consider setup where forecasts do not impact data generating process.

New Facts

- Use Blue Chip Financial Forecasts (BCFF) survey:
 - Forecasts about output growth, inflation, fed-funds rate.
 - Up to 6-11 years ahead horizon.
- We document:
 - 1 Forecasters disagree about **fundamental** (long-horizon) outcomes.
 - 2 **Term-structure of disagreement** differs markedly across variables.
 - 3 LT disagreement varies over time and co-varies across variables.

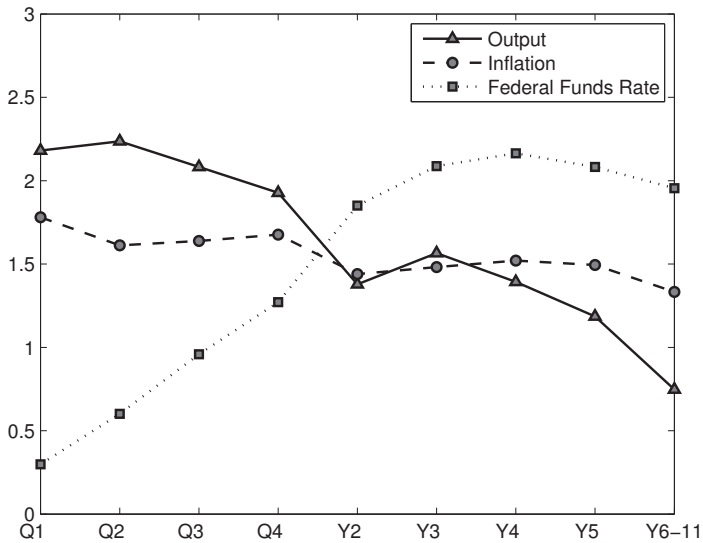
Replicating the Facts

- Independent of the model of info. frictions, key ingredients:
 - ① Current state of the economy *imperfectly observed*.
 - ② Unobserved *permanent* and *transitory* shocks.
 - ③ *Dynamic interactions* between variables.
- Minimal departure from full information setup: *symmetric* agents
 - No information advantage (consistent with evidence that hard to beat the consensus).
 - Forecasters agree on the parameters of the 'true' model but disagree about the unobserved states.

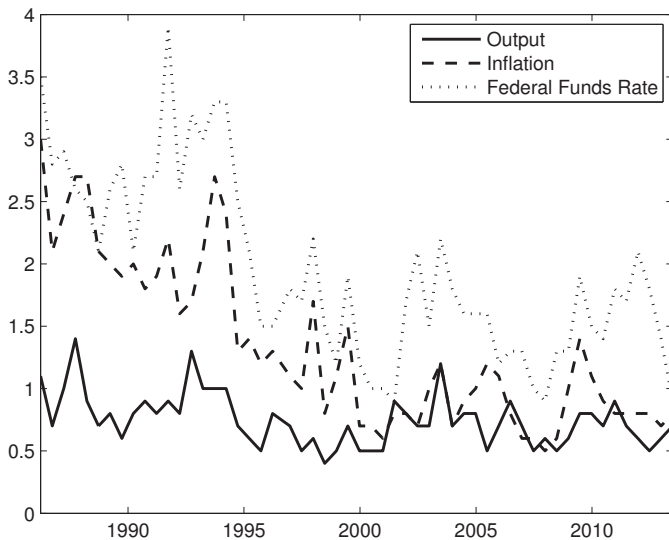
The Blue Chip Financial Forecasts Survey

- ~ 50 professional forecasters.
- We look at forecasts for RGDP growth (g), CPI inflation (π), FFR (i).
- Sample period is 1986:Q1-2013:Q2.
- For 1Q, 2Q, 3Q, 4Q: observe individual forecasts.
- For 2Y, 3Y, 4Y, 5Y and long-term (6-to-11Y): observe average forecasts, top 10 average forecasts, and bottom 10 average forecasts.
- Our measure of **disagreement**: top 10 average – bot 10 average.
 - strongly correlated with cross-section stdev or IQR at short horizons.

The Term Structure of Disagreement in the BCFF



The Time Series of Long Run Disagreement



Model

Underlying state

- True **state** $z = \{g, \pi, i\}$ where

$$\begin{aligned}z_t &= (I - \Phi)\mu_t + \Phi z_{t-1} + v_t^z, \\ \mu_t &= \mu_{t-1} + v_t^\mu,\end{aligned}$$

with $v_t^z \sim iid N(0, \Sigma^z)$ and $v_t^\mu \sim iid N(0, \Sigma^\mu)$.

- Parameters: $\theta = (\Phi, \Sigma^z, \Sigma^\mu)$

Model

Information Friction: Noisy Information

- Forecaster j observes:

$$y_{jt} = z_t + \eta_{jt}$$

with $\eta_{jt} \sim iid N(0, \Sigma^\eta)$, Σ^η diagonal.

- Individual j 's optimal forecast computed using the Kalman filter.
- Model parameters: (θ, Σ^η) .
- Disagreement driven by variance of observation errors Σ^η .

Model

Information Friction: Sticky Information

- At each date, a forecaster j observes k^{th} element of y_t with a fixed probability λ_k ; otherwise sticks to latest observation of that variable.
- Individual j 's optimal forecast computed using the Kalman filter with missing observations.
- Disagreement driven by different individual sequences of observations.
- Same number of parameters as in noisy info with λ 's instead of Σ^η .

Calibration via Penalized MLE

Principle

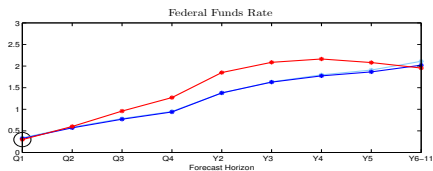
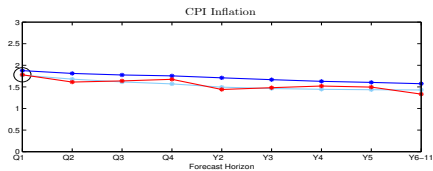
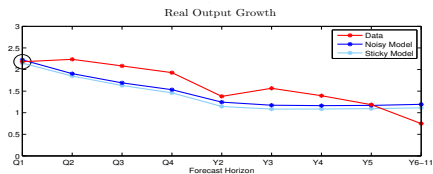
- Sample: realizations 1955Q1-2013Q2; survey 1986Q1-2013Q2.
- We minimize the **Likelihood** associated to true state + ...
- ... a **penalty function** measuring the distance between model implied moments (simulations) and their survey data counterpart.
- We choose 15 moments:
 - Std-dev of consensus forecasts for Q1, Q4, Y2 and Y6-11.
 - Disagreement about Q1 forecasts **only**.

Summary of Parameter Estimates

- Common parameters (θ) robust to type of info. friction considered.
- Long-run vol. (Σ^μ) **much lower** than short-run vol. (Σ^z).
- FFR is **perfectly observed**:
 - *Noisy*: observation error (Σ_η) for FFR is zero.
 - *Sticky*: probability of observing FFR (λ_i) is one.
- RGDP/CPI: degree of info frictions consistent with previous studies.

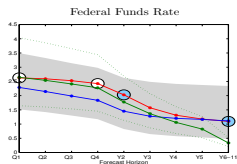
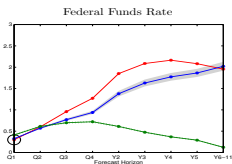
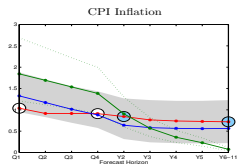
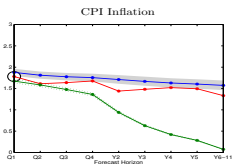
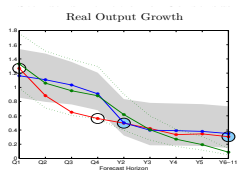
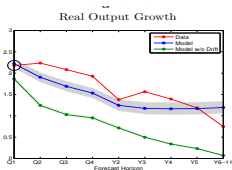
Data and Model-implied Term Structures of Disagreement

Noisy and Sticky



Disagreement and Consensus Volatility

Noisy (very similar patterns for sticky version)



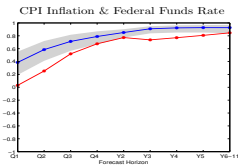
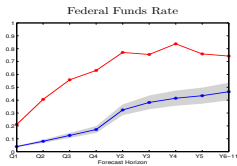
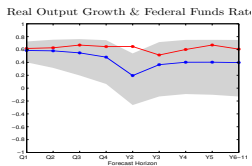
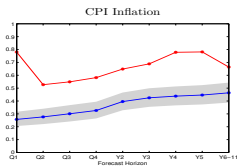
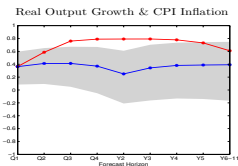
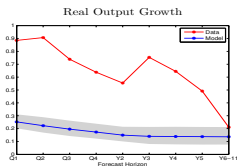
33

Role of Key Ingredients

- Decomposition into permanent and transitory components:
 - Needed to explain disagreement beyond the short/medium term.
- Multivariate model:
 - Needed to explain disagreement about future FFR even though perfectly observed.
 - Needed to generate upward-sloping disagreement.

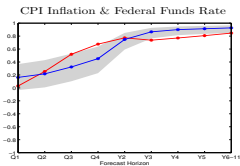
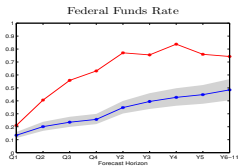
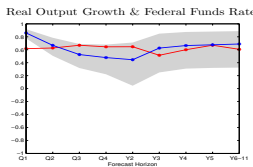
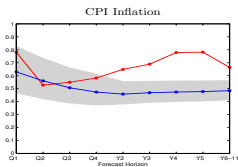
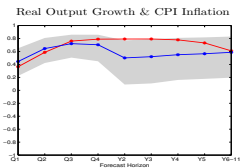
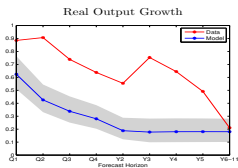
Time Variation & Co-movement in Disagreement

Noisy



Time Variation & Co-movement in Disagreement

Sticky



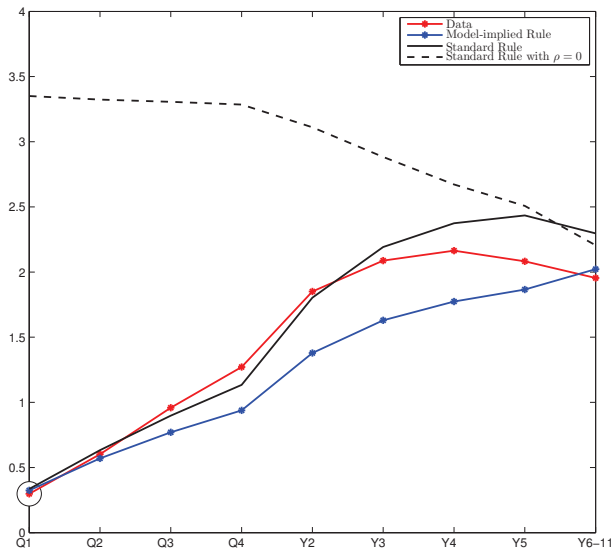
Is Disagreement about FFR Consistent with a Taylor Rule?

- Generate ind. FFR forecasts given model ind. forecasts of g and π

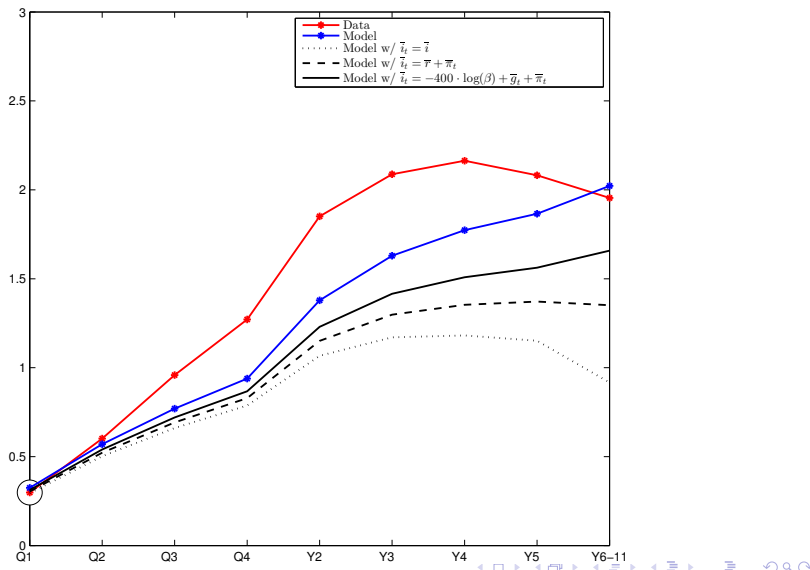
$$\begin{aligned}i_t &= \rho \cdot i_{t-1} + (1 - \rho) \cdot i_t^* + \epsilon_t \\i_t^* &= \bar{i}_t + \varphi_\pi \cdot (\pi_t - \bar{\pi}_t) + \varphi_g \cdot (g_t - \bar{g}_t)\end{aligned}$$

- Find parameters $\{\rho, \varphi_\pi, \varphi_g\}$ giving best fit of FFR term structure of disagreement obtained with previous reduced form model.
 - $\rho = 0.98$, $(1 - \rho)\varphi_\pi = .26$, $(1 - \rho)\varphi_g = 0.30$.
- Compare with various parametric restrictions.
 - Std Taylor rule parameters: $\rho = 0.9$, $\varphi_\pi = 2$, $\varphi_g = 0.50$.
 - No policy smoothing $\rho = 0$.
 - Restrictions on time varying components of \bar{i}_t .

'Standard' Taylor Rule



Role of Components in the Time-Varying Intercept



Conclusion

- Present new facts about forecaster disagreement.
 - Identify key features needed to replicate them.
- Models of imperfect info can account for complex facts for sound parameter values.
 - Minimal departure from REH: agents know and agree on true model/params.
- Disagreement informative about both degree of imperfect info and underlying DGPs.
 - Results underline importance of uncertain long-run component in macro-variable (implications for macro & finance).
 - Informative about perceived structural relationships (e.g. Taylor rule).

Extensions / Future Research

- Accounting for time variance of disagreement:
 - Stochastic volatility.
 - State-dependent imperfect info / inattention.
- More structure...

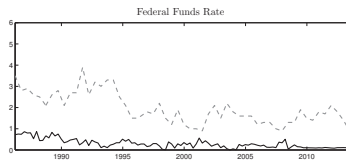
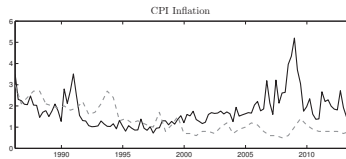
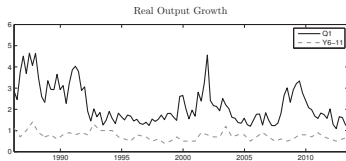
Noisy Information Model

Φ	Σ^z	$\text{sqrt}(\text{diag}(\tilde{\Sigma}^n))$
$\begin{bmatrix} 0.378 & -0.503 & -0.153 \\ 0.125 & 0.974 & -0.033 \\ 0.147 & 0.104 & 0.924 \end{bmatrix}$	$\begin{bmatrix} 3.419 & -0.019 & 0.561 \\ -0.019 & 0.645 & 0.365 \\ 0.561 & 0.365 & 0.632 \end{bmatrix}$	$\begin{bmatrix} 2.592 \\ 1.429 \\ 0.000 \end{bmatrix}$
$ \text{eig}(\Phi) $	Σ^μ	$\text{sqrt}(\text{diag}(\Sigma^n))$
$\begin{bmatrix} 0.920 \\ 0.711 \\ 0.646 \end{bmatrix}$	$\begin{bmatrix} 0.008 & 0.014 & 0.026 \\ 0.014 & 0.024 & 0.045 \\ 0.026 & 0.045 & 0.085 \end{bmatrix}$	$\begin{bmatrix} 4.317 \\ 2.731 \\ 0.000 \end{bmatrix}$

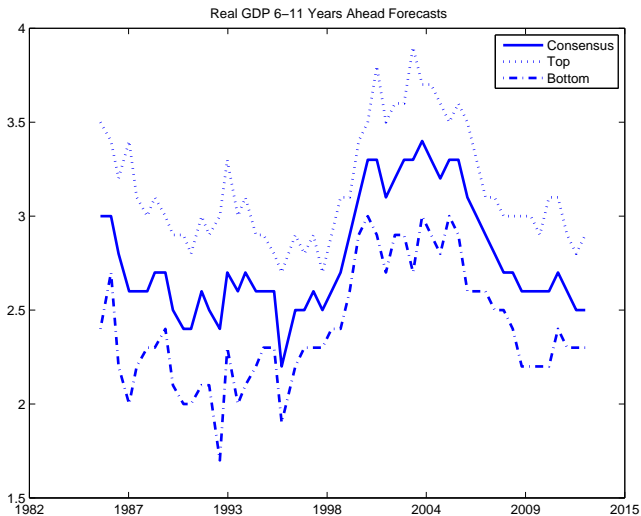
Sticky Information Model

Φ	Σ^z	$\text{sqrt}(\text{diag}(\tilde{\Sigma}^\eta))$
$\begin{bmatrix} 0.392 & -0.478 & -0.142 \\ 0.122 & 0.939 & -0.024 \\ 0.146 & 0.087 & 0.931 \end{bmatrix}$	$\begin{bmatrix} 3.736 & -0.065 & 0.564 \\ -0.065 & 0.911 & 0.347 \\ 0.564 & 0.347 & 0.635 \end{bmatrix}$	$\begin{bmatrix} 2.586 \\ 1.355 \\ 0.000 \end{bmatrix}$
$ \text{eig}(\Phi) $	Σ^μ	λ
$\begin{bmatrix} 0.920 \\ 0.674 \\ 0.674 \end{bmatrix}$	$\begin{bmatrix} 0.007 & 0.012 & 0.022 \\ 0.012 & 0.021 & 0.039 \\ 0.022 & 0.039 & 0.073 \end{bmatrix}$	$\begin{bmatrix} 0.260 \\ 0.260 \\ 1.000 \end{bmatrix}$

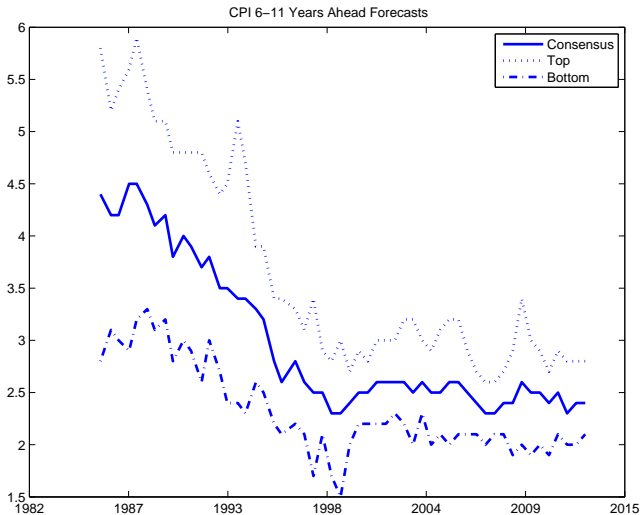
ST/LT Disagreement



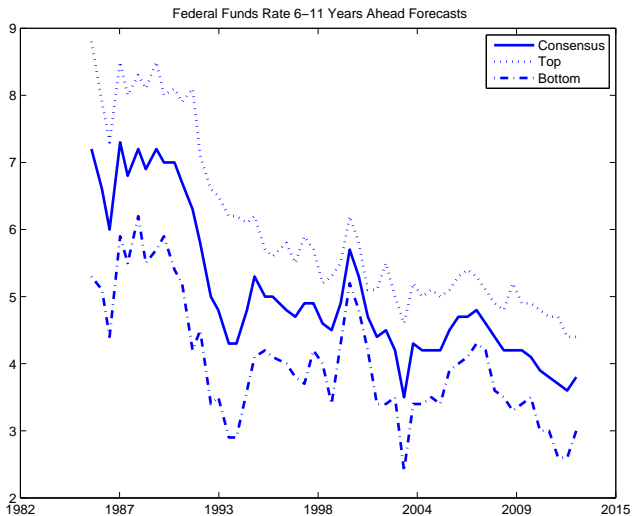
Long-Term Forecasts: GDP/GNP



Long-Term Forecasts: CPI



Long-Term Forecasts: FFR



Calibration via Penalized MLE

Details (1/2)

- Consider realizations as signals about z_t : $\mathcal{Y}_t = z_t + \tilde{\eta}_t$ with $\tilde{\eta}_t \sim iid N(0, \tilde{\Sigma}^\eta)$.
- $-\mathcal{L}(\mathcal{Y}_1, \dots, \mathcal{Y}_T; \theta, \tilde{\Sigma}^\eta) =$ likelihood obtained with Kalman filter.

Calibration via Penalized MLE

Details (2/2)

- Given (θ, Σ^η) , we generate individual forecasts f_{jt}^h and minimize distance between simulated moments $S(\theta, \Sigma^\eta)$ and their survey data counterparts \mathcal{S}_t .
- distance between model implied expectation moments and their survey data counterpart, $\mathcal{P}(\mathcal{S}_1, \dots, \mathcal{S}_T; \theta, \Sigma^\eta)$

$$\mathcal{P} = \left[S(\theta, \Sigma^\eta) - \frac{1}{T} \sum_t \mathcal{S}_t \right]' W \left[S(\theta, \Sigma^\eta) - \frac{1}{T} \sum_t \mathcal{S}_t \right]$$

- We minimize the penalized likelihood:

$$\mathcal{C}(\theta, \Sigma^\eta, \tilde{\Sigma}^\eta) = \mathcal{L}(\mathcal{Y}_1, \dots, \mathcal{Y}_T; \theta, \tilde{\Sigma}^\eta) + \alpha \mathcal{P}(\mathcal{S}_1, \dots, \mathcal{S}_T; \theta, \Sigma^\eta).$$

Calibration in Practice

- Sample: realizations 1955Q1-2013Q2; survey 1986Q1-2013Q2.
- Simulate $R = 100$ histories of shocks ϵ_t and observation noises η_t^j with $T = 120$ (nb of dates) and $N = 50$ (nb of forecasters).
- We choose 15 moments:
 - Std-dev of consensus forecasts for Q1, Q4, Y2 and Y6-11.
 - Disagreement about Q1 forecasts **only**.
 - Strong weight on Q1 disagreement in the penalty criterion (“normalisation”).
- Penalty parameter $\alpha = 50$.

Robustness

- Various penalty parameters $\alpha = 1$, $\alpha = 10$.
- Initial conditions.
- Econometrician perfectly observing the state ($\tilde{\Sigma}^\eta = 0$).