

Understanding the Sources of Macroeconomic Uncertainty

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Objective of the Paper

- Recent economic events (great recession, unconventional monetary policy, fiscal cliff, etc.) sparked great interest in **understanding uncertainty** and its macroeconomic impact.
 - Stock and Watson (2012) suggests the liquidity-risk and uncertainty shocks to be the most important contributor to the decline in the U.S. GDP during the Great Recession
 - 2/3 of the recession's decline in GDP and employment
- There has been increased emphasis in trying to **characterize** uncertainty, which is inherently unobserved.
- There are **many measures** of uncertainty
 - ex-ante, ex-post, disagreement, mean-squared forecast errors, forecast error distributions, etc.
- Our paper proposes to **reconcile** the various measures.

The Measure of Uncertainty Matters

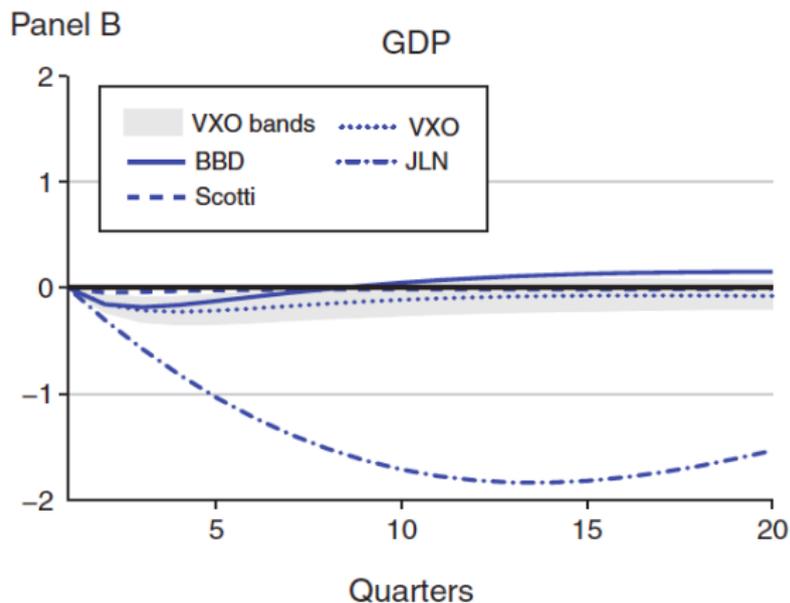
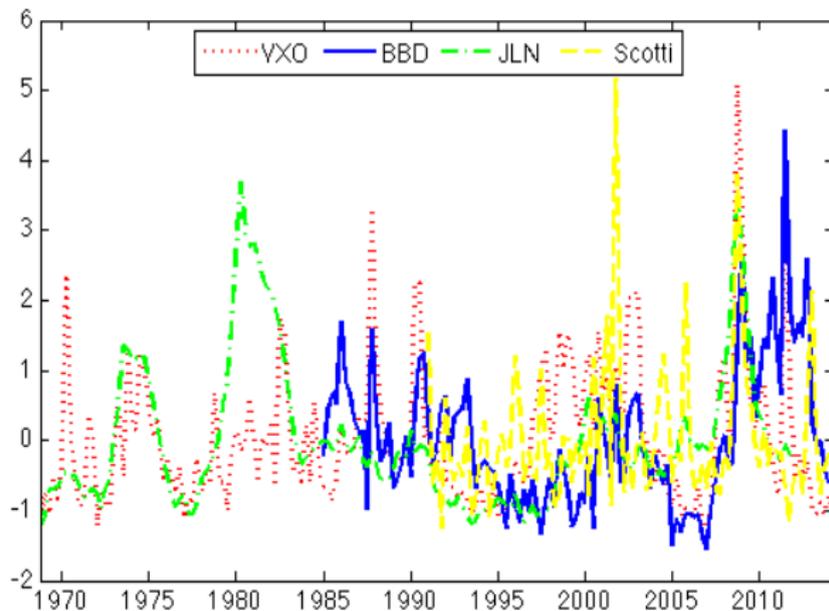


FIGURE 3. IMPACT OF UNCERTAINTY ON GDP

Note: The figure depicts impulse responses of GDP to uncertainty shocks measured by various indices.

The Measure of Uncertainty Matters



Summary of Various Measures

- Based on some observables
 - realized volatility, implied volatility (VIX, VXO, Bloom, 2009), Baker, Bloom & Davis (2015) index
- Measures of ex-ante uncertainty or perceived uncertainty
 - typically based on surveys
 - disagreement as a special case
 - Clements (2015), Leduc & Liu (2015), D'Amico & Orphanides (2014), Patton & Timmermann (2010), etc.
- Ex-post measures of uncertainty
 - based on forecast errors
 - Has the notion that “What matters for economic decision making is whether the economy has become more or less predictable; that is, less or more uncertain.”
 - Jurado, Ludvigson & Ng (2015), Rossi & Sekhposyan (2015), Scotti (2013), etc.

Our Contribution

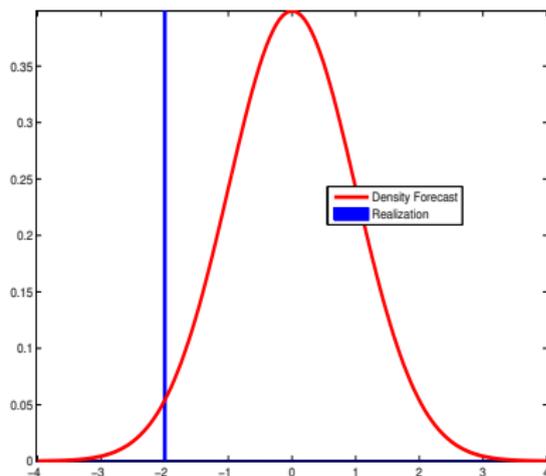
- We propose a **predictive distribution-based** uncertainty measure.
- We can further decompose this measure to
 - measures of **aggregate uncertainty** and **disagreement**
 - measures of bias (**Knightian**) and realized variance (**risk**)
 - measures of **ex-ante** and **ex-post** uncertainty
- We provide evidence of differential macroeconomic impact.
- Provide simulation experiments documenting the evolution of the channels of the various measures of uncertainty.

Risk versus Knightian Uncertainty

- **Risk** - uncertainty stemming from the fact that a realization of the state of nature is not known in advance even if **all possible states of nature and their likelihoods could be reasonably contemplated**.
 - various measures of volatility
- **Knightian Uncertainty** - uncertainty stemming from the fact that **it is not possible to assign correct probabilities to future outcomes or agree on the probabilities**.
 - depends on the realization or the disagreement among forecasted probabilities

Uncertainty Index based on Density Forecasts

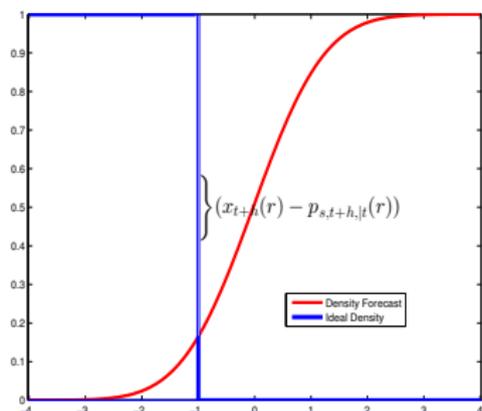
- At a particular point in time you have
[1.] the density forecast [2.] realization



Uncertainty Index based on Density Forecasts

Work with a binary variable and cdf instead. Let

- $x_{t+h}(r) = 1\{y_{t+h} < r\}$
- $p_{s,t+h|t}(r) = P(x_{t+h}(r) = 1 | \Omega_{s,t})$



- For a given threshold r , s -th forecaster's uncertainty is:

$$u_{s,t+h|t}(r) = E \left[(x_{t+h}(r) - p_{s,t+h|t}(r))^2 \mid \mathfrak{S}_{t-R}^t \right].$$

- Has the spirit of a forecast error for a particular quantile.

The Uncertainty Index

- The measure of uncertainty is defined as the average of the individual uncertainty measure across forecasters:

$$\begin{aligned}
 u_{t+h|t}(r) &= \frac{1}{N} \sum_{s=1}^N u_{s,t+h|t}(r) \\
 &= \frac{1}{N} \sum_{s=1}^N E \left[(x_{t+h}(r) - p_{s,t+h|t}(r))^2 \mid \mathcal{S}_{t-R}^t \right]
 \end{aligned}$$

- Similar to Lahiri & Sheng (2010), Zarnowitz & Lambros (1987) for a particular point in a distribution
- Uncertainty

$$U_{t+h|t} = \int_{-\infty}^{+\infty} u_{t+h|t}(r) dr$$

Decomposition I: Aggregate Uncertainty & Disagreement

$$\begin{aligned}
 u_{t+h|t}(r) &= \frac{1}{N} \sum_{s=1}^N E_t \left[\left(x_{t+h}(r) - p_{t+h|t} + p_{t+h|t} - p_{s,t+h|t}(r) \right)^2 \right] \\
 &= E_t \left(x_{t+h}(r) - p_{t+h|t}(r) \right)^2 \\
 &+ \frac{1}{N} \sum_{s=1}^N E_t \left[\left(p_{t+h|t}(r) - p_{s,t+h|t}(r) \right)^2 \right] \\
 &= u_{t+h|t}^A(r) + d_{t+h|t}(r),
 \end{aligned}$$

$$\begin{aligned}
 \underbrace{U_{t+h|t}}_{\text{"Uncertainty"}} &= \int_{-\infty}^{\infty} \underbrace{u_{t+h|t}^A(r)}_{\text{"Aggregate Uncertainty"}} dr + \int_{-\infty}^{\infty} \underbrace{d_{t+h|t}(r)}_{\text{"Disagreement"}} dr \\
 &= \underbrace{U_{t+h|t}^A}_{\text{"Aggregate Uncertainty"}} + \underbrace{D_{t+h|t}}_{\text{"Disagreement"}}
 \end{aligned}$$

Decomposition II: Aggregate Uncertainty as Risk and Knightian Uncertainty

$$\begin{aligned}
 u_{t+h}^A(r) &= \left(\left[E(p_{t+h|t}(r) | \mathfrak{S}_{t-R}^t) - E(x_{t+h}(r) | \mathfrak{S}_{t-R}^t) \right]^2 \right) \\
 &+ V(x_{t+h}(r) | \mathfrak{S}_{t-R}^t) + V(p_{t+h|t}(r) | \mathfrak{S}_{t-R}^t) \\
 &- 2\text{Cov}(x_{t+h}(r), p_{t+h|t}(r) | \mathfrak{S}_{t-R}^t),
 \end{aligned}$$

$$U_{t+h|t}^A \approx \underbrace{B_{t+h|t}}_{\text{"Mean-Bias"}} + \underbrace{V_{t+h|t}}_{\text{"Dispersion"}} + \underbrace{Vol_{t+h|t}}_{\text{"(Realized) Risk"}}$$

Putting things together

$$U_{t+h|t} \approx \underbrace{Vol_{t+h|t}}_{\text{"(Realized) Risk"}} + \underbrace{B_{t+h|t} + D_{t+h|t}}_{\text{"Knightian Uncertainty"}}$$

Decomposition III: Aggregate Uncertainty as Ex-Ante and Ex-Post Uncertainty

- Let $\hat{y}_{t+h|t} \sim N(\mu_{t+h|t}, \sigma_{t+h|t}^2)$.
- This is the density forecast.

$$\begin{aligned}
 U_{t+h|t}^A &= E|Y - y_{t+h}| - 0.5E|Y - Y'| = \\
 &\underbrace{\left[2\sigma_{t+h|t}\phi\left(\frac{y_{t+h} - \mu_{t+h|t}}{\sigma_{t+h|t}}\right) + (y_{t+h} - \mu_{t+h|t})\left(2\Phi\left(\frac{y_{t+h} - \mu_{t+h|t}}{\sigma_{t+h|t}}\right)\right) \right]}_{\text{"Ex-Post"}} \\
 &\underbrace{-\sigma_{t+h}/\sqrt{\pi}}_{\text{"Ex-Ante"}}
 \end{aligned}$$

Recap

We propose

- to look at total uncertainty as an average squared **distributional** forecast error
- still has the notion that the **unpredictable elements** constitute to uncertainty

We can distinguish between

- aggregate uncertainty and disagreement
- realized risk and Knightian uncertainty
- ex-ante and ex-post uncertainty

Empirical Implementation

- Density forecasts from the Survey of Professional Forecasters provided by the Philadelphia Fed
 - assign a probability value over pre-defined intervals for a variety of variables
 - forecasts are for the current year and next year year-over-year growth rates

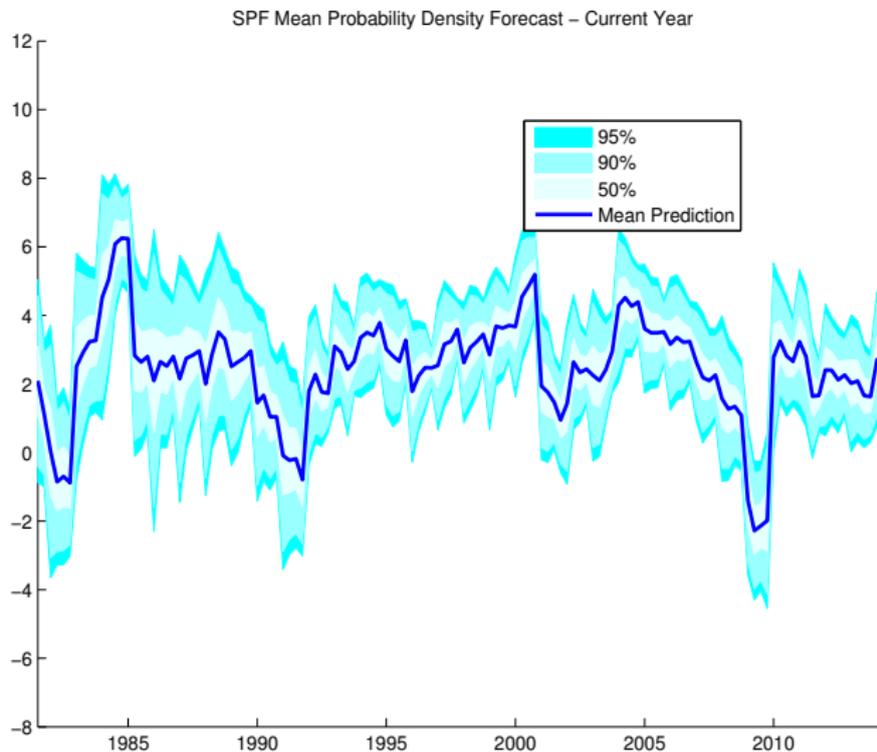
- Use Doovern et al. (2012) re-weighting scheme to get 4-step-ahead forecasts:

$$\widehat{f}_{t+4|t}^{FH} = \frac{k}{4} \widehat{f}_{t+k|t}^{FE} + \frac{4-k}{4} \widehat{f}_{t+k+4|t}^{FE}$$

- 4-quarter-ahead growth of “Advance” release in real time
- Empirical counterparts with 4-quarter-moving averages

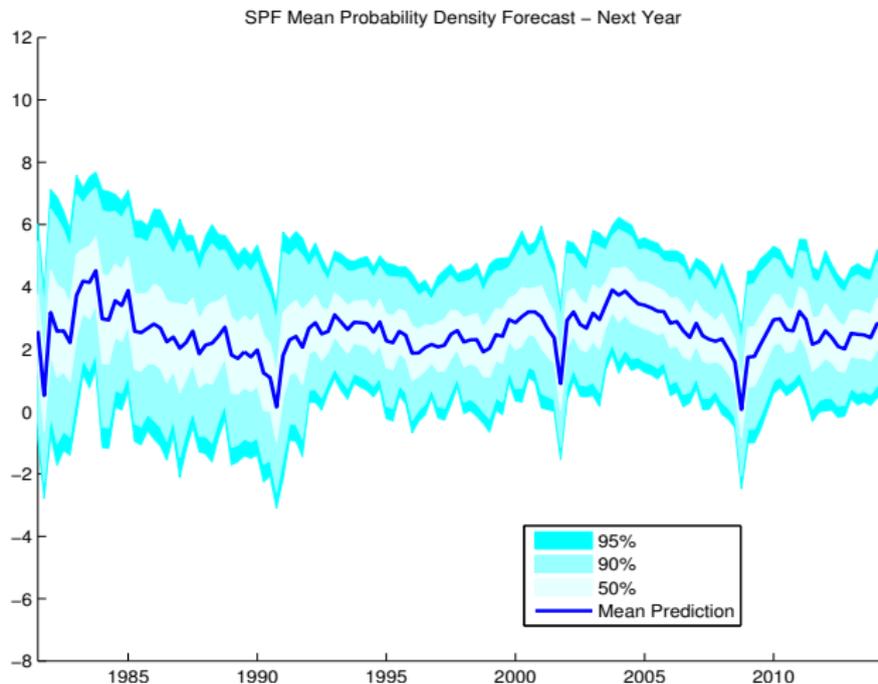
Data

- Predictive quantiles of SPF



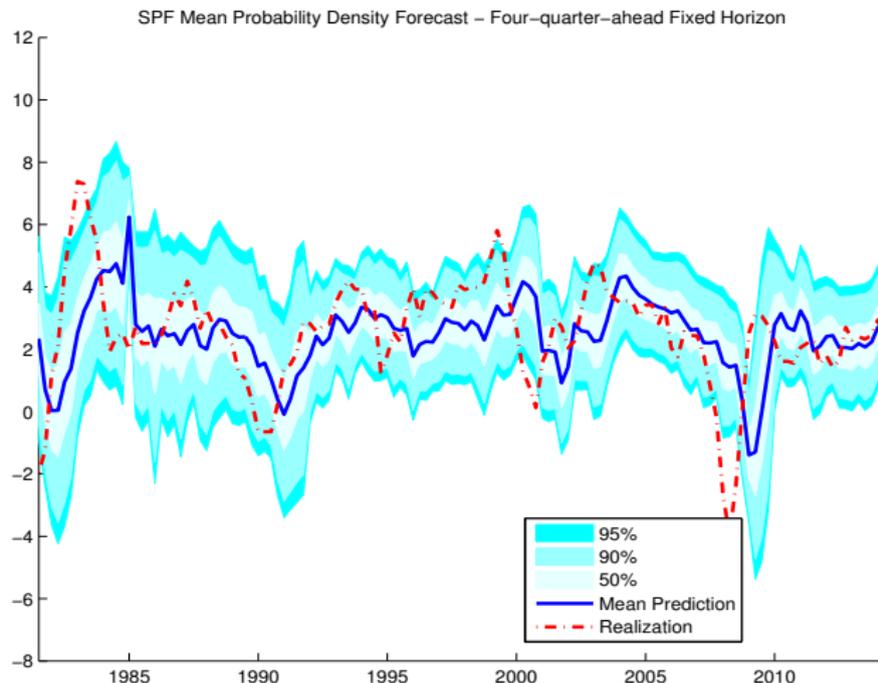
Data

- Predictive quantiles of SPF

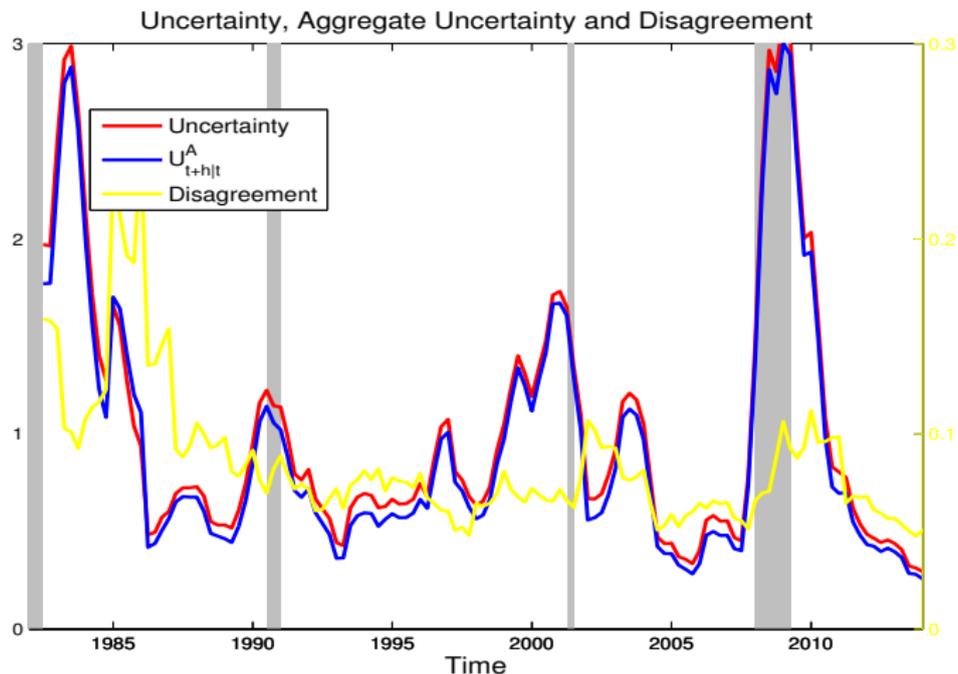


Data

- Predictive quantiles of SPF versus the realizations



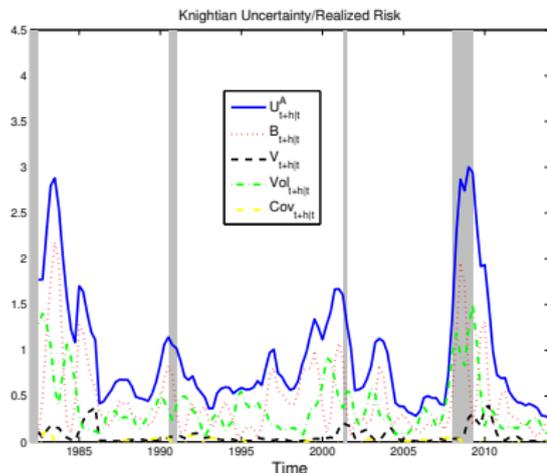
Results: Decomposition I



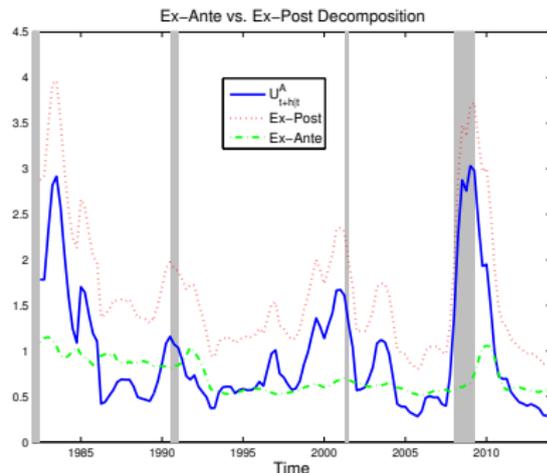
- The role of disagreement is very small
- Disagreement lags the aggregate measure

Results: Decompositions II and III

Knightian vs. Realized Risk

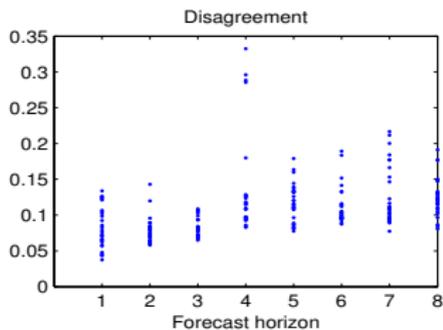
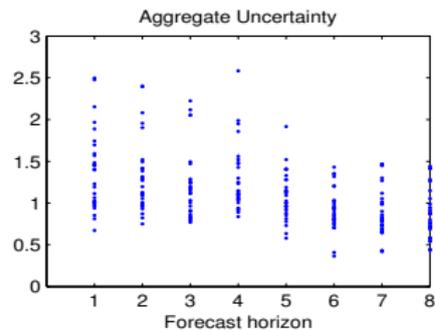
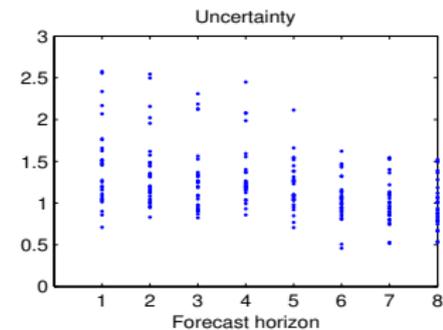


Ex-Ante vs Ex-Post

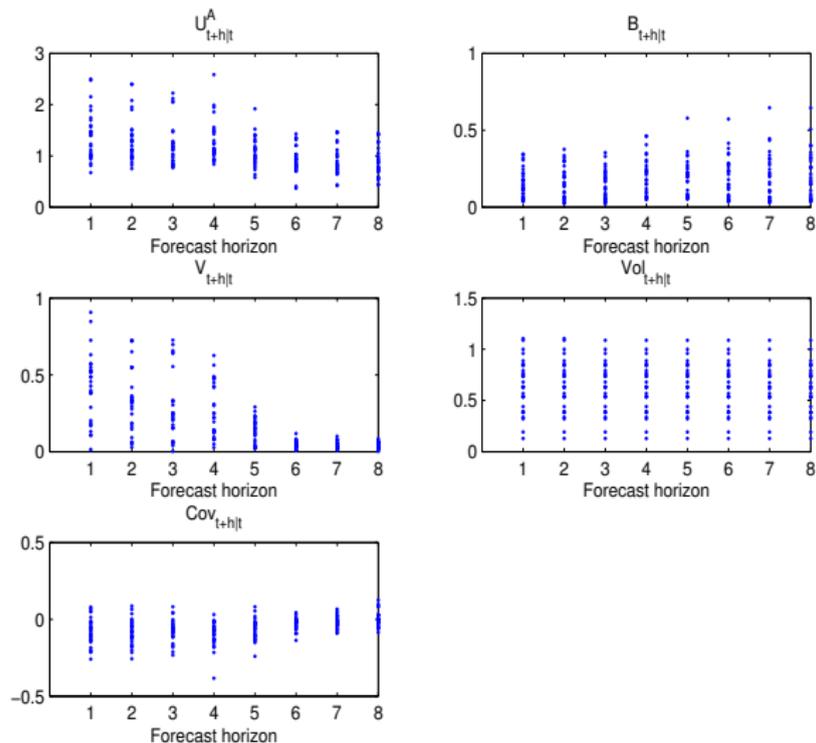


- Ex-ante volatility larger than the realized volatility
- Ex-ante volatility is smoother than the realized one
- Knightian uncertainty and ex-post are more important for the aggregate uncertainty

Resolution of Uncertainty over Time

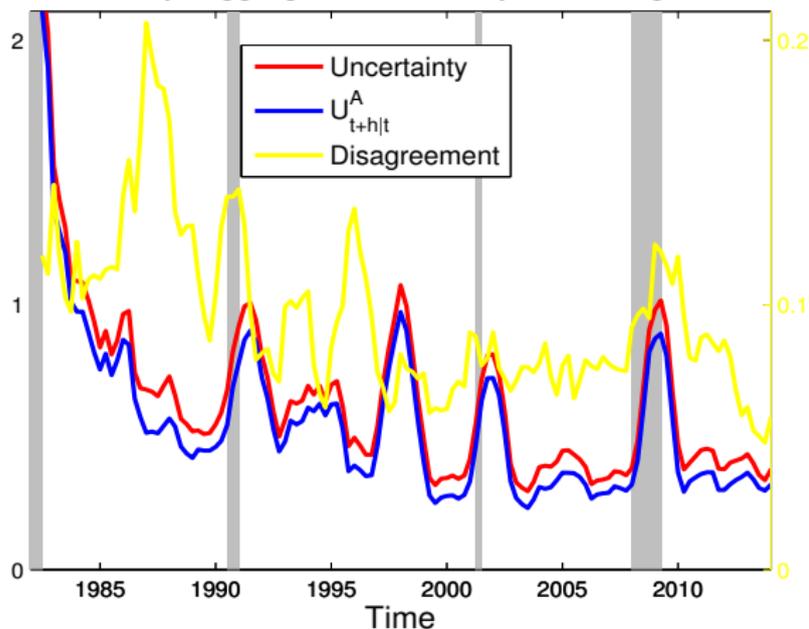


Resolution of Uncertainty over Time



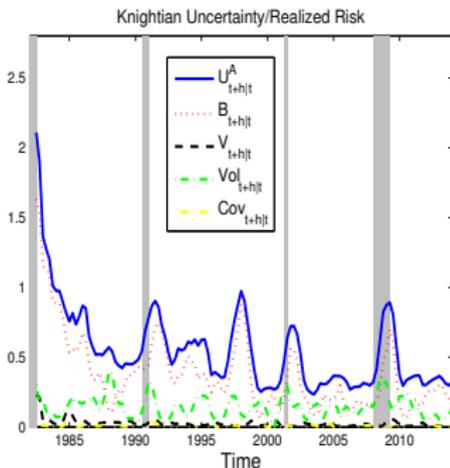
Results: Decomposition I for Inflation

Uncertainty, Aggregate Uncertainty and Disagreement

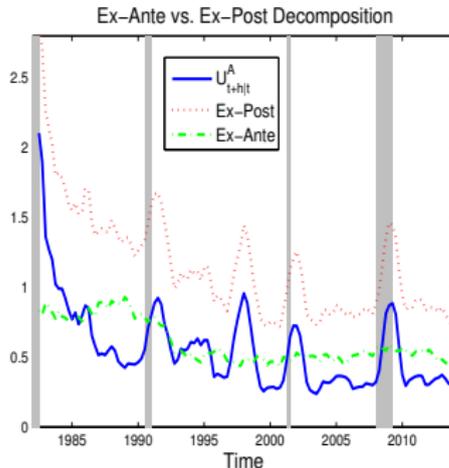


Results: Decompositions II and III for Inflation

Knightian vs. Risk

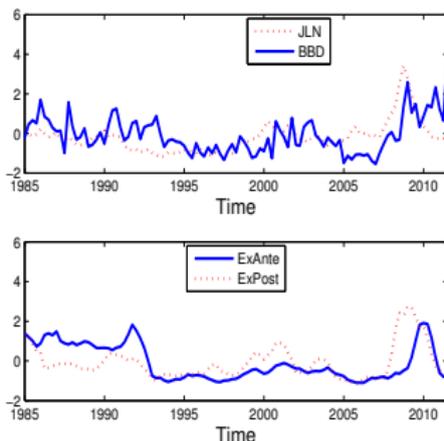


Ex-Ante vs Ex-Post



- Ex-ante volatility larger than the realized volatility
- Ex-ante volatility is smoother than the realized one
- Bias and ex post are more important for the aggregate uncertainty, though the latter more for dynamics

Comparison with some Existing Measures

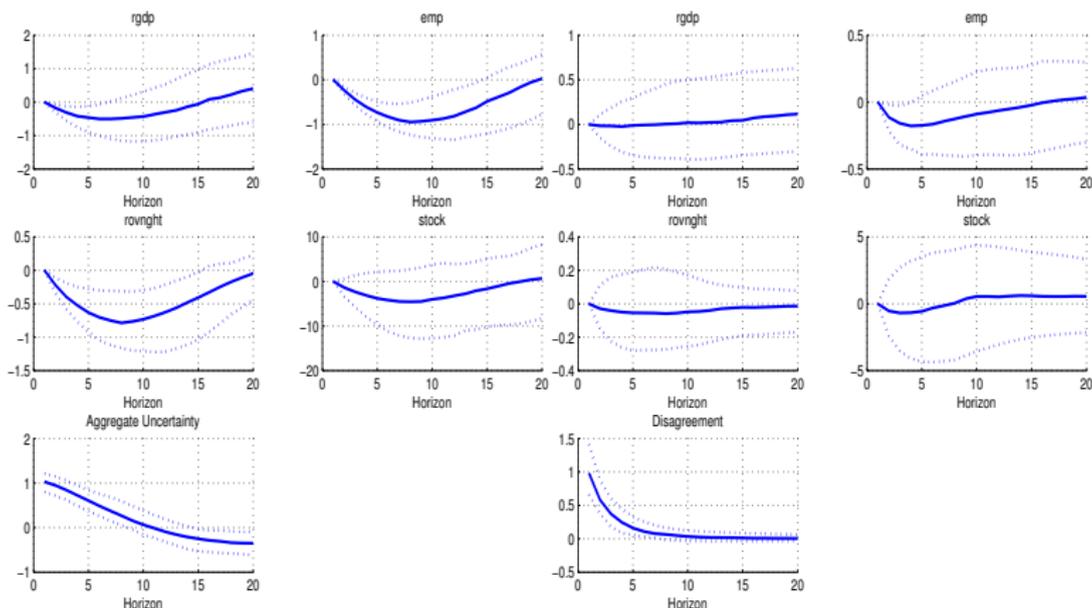


- Jurado et al. (2015) similar to ex-post
- Baker et al. (2015) similar to ex-ante
- Roughly similar patterns

Macroeconomic Impact

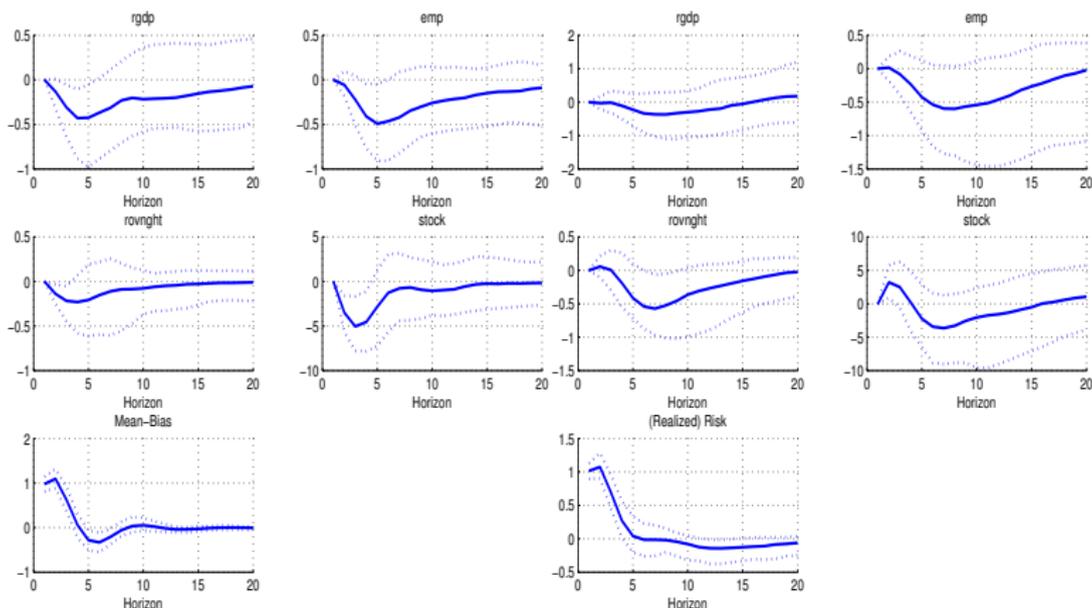
- Based on (the log of) real GDP, (the log of) employment, the Federal Funds rate, (the log of) stock prices and uncertainty indices + const
- Uncertainty indices are standardized
- Identification according to recursive ordering
- Lag length is selected via BIC
- Robust to an 11 variable specification

Decomposition I



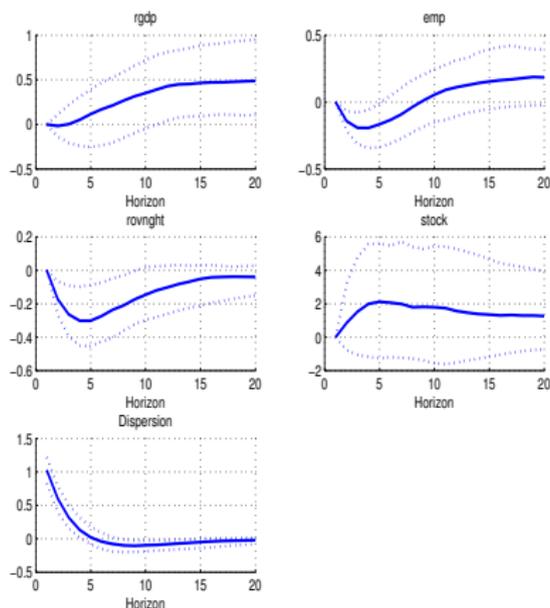
- Insignificant response to disagreement

Decomposition II



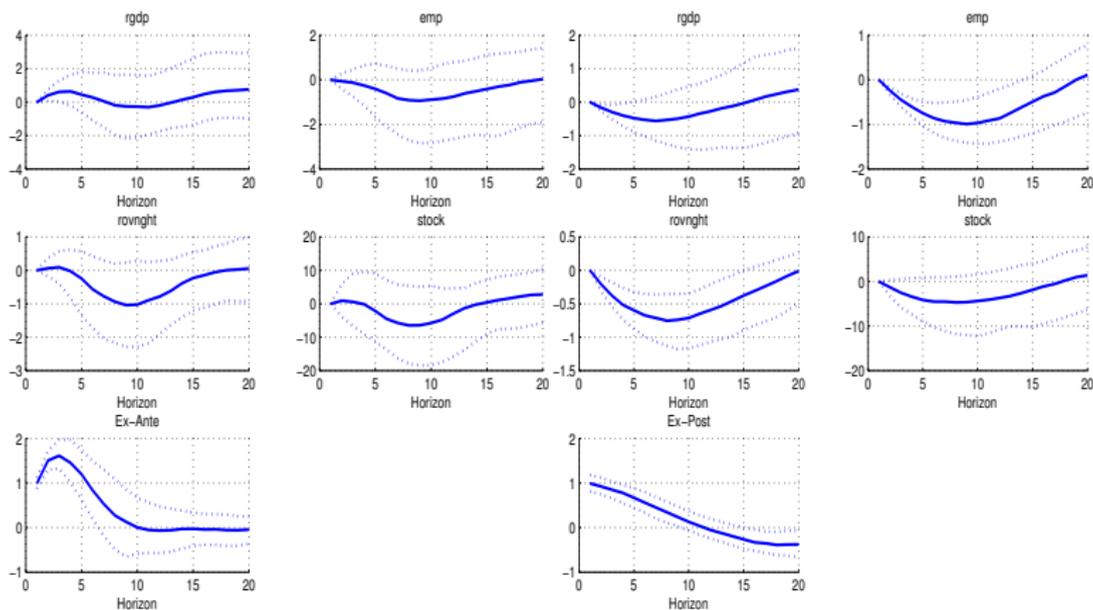
- Insignificant response to realized risk

Decomposition II



- Effects of dispersion are expansionary

Decomposition III



- Effects of ex-ante uncertainty are insignificant

Glancing through a lens of a model

- Model and parameter values inspired by Ilut and Schneider (2014)

- Data is generated by

$$Z_{t+1} = \rho_z Z_t + \mu_t^* + u_{t+1}$$

- μ_t^* ambiguous component, $\mu_t^* \sim iidN(0, \sigma_z - \sigma_u)$
 - lack confidence to assign probabilities to all relevant events
- u_{t+1} random component (capturing risk), $u_t^* \sim iidN(0, \sigma_u)$
 - can assign probabilities to all relevant events
- Agents get noisy signals about μ_t^*
 - conflicting news reports, disagreement among experts, poor information, etc.

Glancing through a lens of a model

- Data is generated by

$$Z_{t+1} = \rho_z Z_t + \mu_t^* + u_{t+1}$$

- Agents get noisy signals about μ_t^*
- Their beliefs set is $\mu_t \in [-a_t, -a_t + 2|a_t|]$
- They choose $\mu_t^{**} = \min[-a_t, -a_t + 2|a_t|]$, **worst case scenario**
- While the agents get signals according to

$$a_{t+1} - \bar{a} = \rho_a (a_t - \bar{a}) + \sigma_a \epsilon_{t+1}^a.$$

- $\bar{a} = n\sigma_z$ and $\sigma_a = \sigma_n\sigma_z$ for $n \in (0, 1)$

Glancing through a lens of a model

General notions about the model:

- Ambiguity is about the mean.
- It yields a perceived law of motion that is misspecified in the mean.
- The shocks to risk (σ_u) not only affect the second moment dynamics, but can propagate through the mean.
 - It affects the width of the confidence set.
- Different than the news shocks since the signal does not need to be validated by a realization.

Glancing through a lens of a model

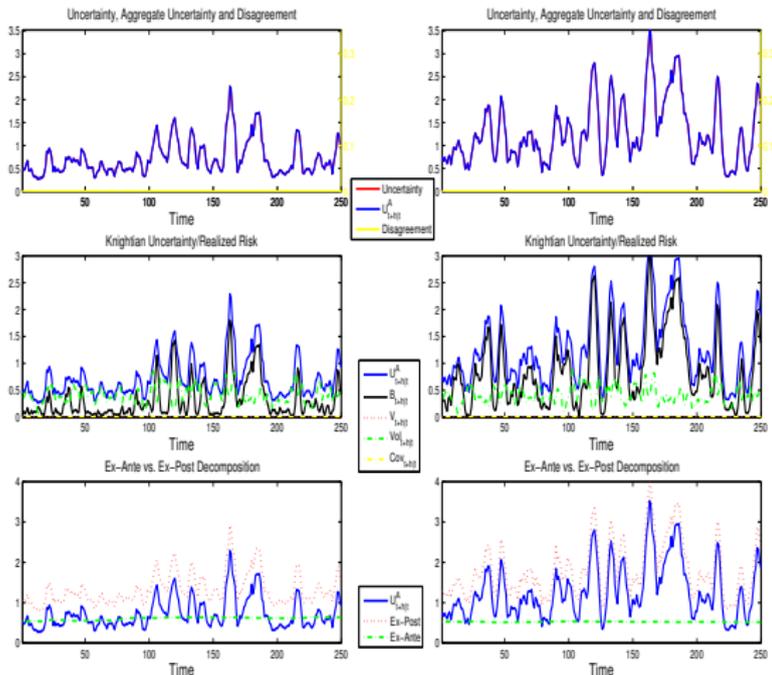
- Baseline Parameter Values

ρ_z	0.625	estimated
ρ_a	0.887	IS mode
n	0.995	IS mode
σ_u	0.780	estimate
σ_μ	0.500	arbitrary
σ_n	0.134	IS mode

- Simulate for 254 periods, with a burn in of 100.
- Scenarios
 - changing level of ambiguity - change in the quality of the mean signal
 - changing level of risk - implies change in the ambiguity mean and variance
 - changing risk in a model with no ambiguity

Simulation Results

1. Changing the Level of Ambiguity

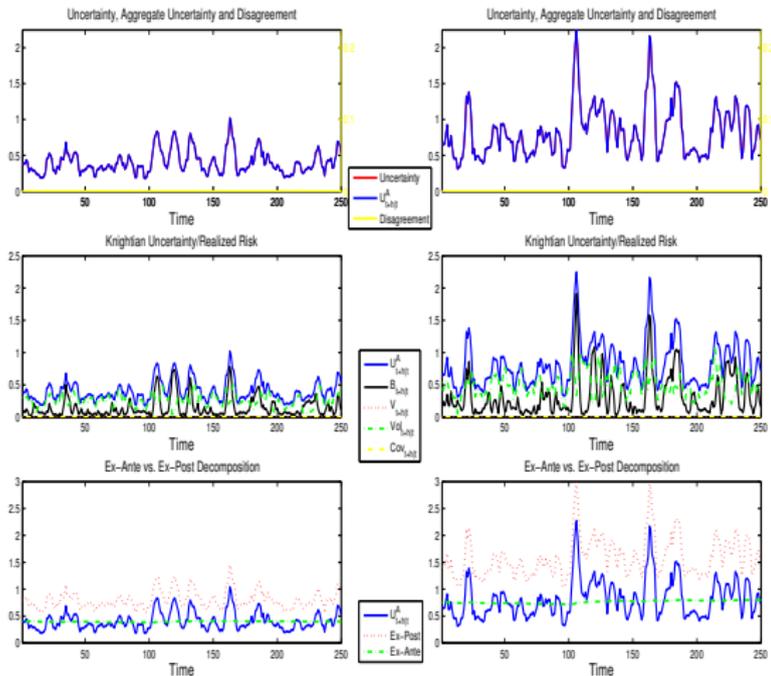
 $n = 0.2$ $n = 0.8$ 

Simulation Results

3. Changing the Level of Risk, no Ambiguity

$$\sigma_u = 0.3$$

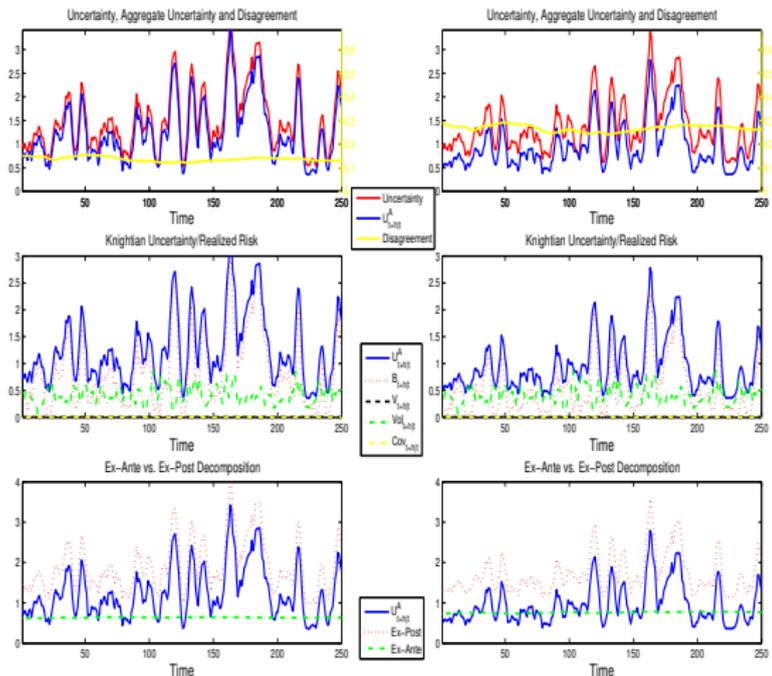
$$\sigma_u = 1$$



Scenario 4: Increasing Cross-Sectional Dispersion in Ambiguity

$$\sigma_{n,l} = 0.5$$

$$\sigma_{n,l} = 1$$



Conclusions

- Propose a way to reconcile various measures of uncertainty.
- They differ with their business cycle dynamics, as well as macroeconomic impact.
- One can reconcile the dynamics of the various measures of uncertainty with a model with ambiguity.