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A Bayesian Interpretation of the Federal Reserve's dual mandate and the Taylor Rule

R/Finance 2013

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Samantha Azzarello Economist, CME Group May 2013

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Outline

- Objective
- Taylor Rule
- Econometrics Why Bayesian Statistics?
- Model
- Other Applications



Contribution

• To look at the Fed's interest rate policy through the lens of the Taylor Rule.

• Using a Dynamic Linear model with Bayesian Inference to update parameters over time.

 Using only the information available to the Fed at each point in time.

Original Taylor Rule

Analyzes interest rate policy in the context of the Fed's Dual Mandate:

Promoting price stability and encouraging full employment

Target Federal Funds Rate =

Actual Inflation Rate

- Short-Term Real Rate Assumption
- + 0.5 X (Actual Inflation Desired Inflation)
- + 0.5 X (Output Gap in Percent)

Relative Influence of Inflation and Economic Growth

Adjusted Level of Federal Funds Rate =

- + β₀ x Constant
- + β1 x (Inflation Desired Inflation)
- + β₂ x (Output Gap)

Broke down analysis into the Inflation and Output components

- Are estimated β1and β2 stable over time?
- Close to their expected values of 0.5 given the Taylor Rule?



Federal Reserve Policy Original Taylor Rule and Level of FFR





Estimated Beta Coefficient for the Target Rate Given by the Taylor Rule



Taylor Rule Adherence Analyzed by Components

Inflation, Output Gap, and Drift Components 2.0 1.5 **Estimated Coefficient Values Output Gap** 1.0 0.5 Drift 0.0 -0.5 Inflation -1.0 -1.5 1954 1960 1963 1966 9661 2008 1957 1969 1972 1975 1978 1981 1984 0661 1993 1999 2005 2011 1987 2002 Source: Data from the Federal Reserve Bank of St. Louis FRED Database, Bayesian Dynamic Linear Model Estimation by CME Economic Research.

Decomposition of Taylor Rule Attirbution into

Why Bayes? Thought process is the same as in finance.

Develop a hypothesis based on available information/theories and assess one's confidence.

Receive new information, evaluate errors in terms of previous hypothesis, then develop new hypothesis and new confidence assessment.



Photo Source: Public (Unknown)



Motivation Dynamic Linear Model

- Dynamic allows time varying parameter estimates.
- Sequential analysis allows for updating of estimates as new observations are observed.
- Compare:
 - OLS One set of Beta Estimates for whole time period.
 - DLM Impact of variable X on Y can change and vary over time. DLM estimates capture this change.
- Modified Kalman Filter



Classical vs. Bayesian Statistics How uncertainty is treated

Classical:

Uncertainty about quantities or parameters estimated is captured by looking at how estimates would change in repeated sampling from the same population.

Bayesian:

Uncertainty is addressed by updating *prior* opinions about quantities and parameters estimated as new data is observed.

Bayesian Analysis

PRIOR x LIKELIHOOD \rightarrow POSTERIOR

- Prior Initial probability distribution of parameters
- Likelihood Joint probability of observing the data given the parameters estimated
- Posterior Probability of parameters given the data
- The process of moving from Prior to Posterior is called Bayesian Learning



Priors

Non-informative Priors:

- Variance of 10,000 and mean of 0
- Reflection of lack of knowledge

Informative Priors:

 Controversy regarding use and variability of results



DLM Equations

The DLM is a two equation system estimated as:

Observation Equation

 $\mathbf{Y}_{t} = (\mathbf{F}_{t})' \boldsymbol{\beta}_{t} + \mathbf{V}_{t} \qquad \mathbf{v}_{t} \sim \mathbf{N}[\mathbf{0}, \mathbf{V}_{t}]$

- Where F are the explanatory factors and β are the Beta parameter estimates

State Equation

 $\beta_t = \mathbf{G}_t \ \beta_{t-1} + \mathbf{W}_t \qquad \mathbf{w}_t \sim \mathsf{N}[0, \mathsf{W}_t]$

 DLM Iteration Steps

- 1. Posterior at t-1 $(\beta_{t-1}|D_{t-1}) \sim N[m_{t-1}, C_{t-1}]$ 2. Prior at t $(\beta_t|D_{t-1}) \sim N[a_t, R_t]$
- 3. Next-step ahead forecast (Y_t|D_{t-1})~N[f_t,Q_t]
- 4. Posterior at t $(\beta_t | D_t) \sim N[m_t, C_t]$



DLM Output

• Update the Distribution of parameters, and hence update the 2 moments which characterize each distribution.

From the Posterior Distribution at time t:

Mean → Beta Coefficient

Standard Deviation → Standard Error

Applications

- Federal Reserve Policy
- Natural Gas Price and Power Price
- GDP Forecasting
- Trade Volume Models



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