

Discussion of  
**Is Applied Monetary Policy Analysis Hard?**  
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*Macroeconomics and Reality, 25 years later*

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Question: Is any change in monetary policy analysis progress or regress?

Answer: We don't know

Argument:

Monetary policy analysis is HARD (no guarantee for the sequence of formal solutions of "simple" problems to converge to the "ideal" )

All of our models are *grossly deficient* with respect to the "ideal"

Proposal: Listing and recognizing the *gross deficiencies* is progress

## Strategy 1

Discuss the epistemologic foundations of the economic science

Go back to Popper:

*"the scientific status of a theory is its testability"*

Use this criterion to discuss the thesis of the paper

## **Strategy 2**

Talk about my work

(and its relation with some aspect of the paper)

## DSGE models: progress or regress

2 problems:

1. **Identification** - different frictions generate the same fit
2. **Many classes of model** - different models for different tasks  
forecast vs. policy analysis

## 1. Identification

Canova-Sala, 2005

In DSGE models, parameters are often non-identified

Identification may depend on the objective function chosen (likelihood vs. distance)

$$x_t = \frac{\gamma}{1+\gamma}x_{t-1} + \frac{1}{1+\gamma}E_t x_{t+1} - \frac{1}{\varphi}(i_t - E_t \pi_{t+1}) + e_{1t}$$

$$\pi_t = \frac{\omega}{1+\omega\beta}\pi_{t-1} + \frac{\beta}{1+\omega\beta}E_t \pi_{t+1} + (\varphi + \vartheta_n) \frac{(1-\zeta_p)(1-\beta\zeta_p)}{(1+\omega\beta)\zeta_p} x_t + e_{2t}$$

$$i_t = \phi_r i_{t-1} + (1 - \phi_r)(\phi_\pi \pi_t + \phi_x x_t) + e_{3t}$$

where:

$\gamma$ : habit formation

$\omega$ : indexation to past inflation in price setting

$\zeta_p$ : price stickiness

$\varphi$ : CRRA

$\vartheta_n^{-1}$ : labor supply elasticity

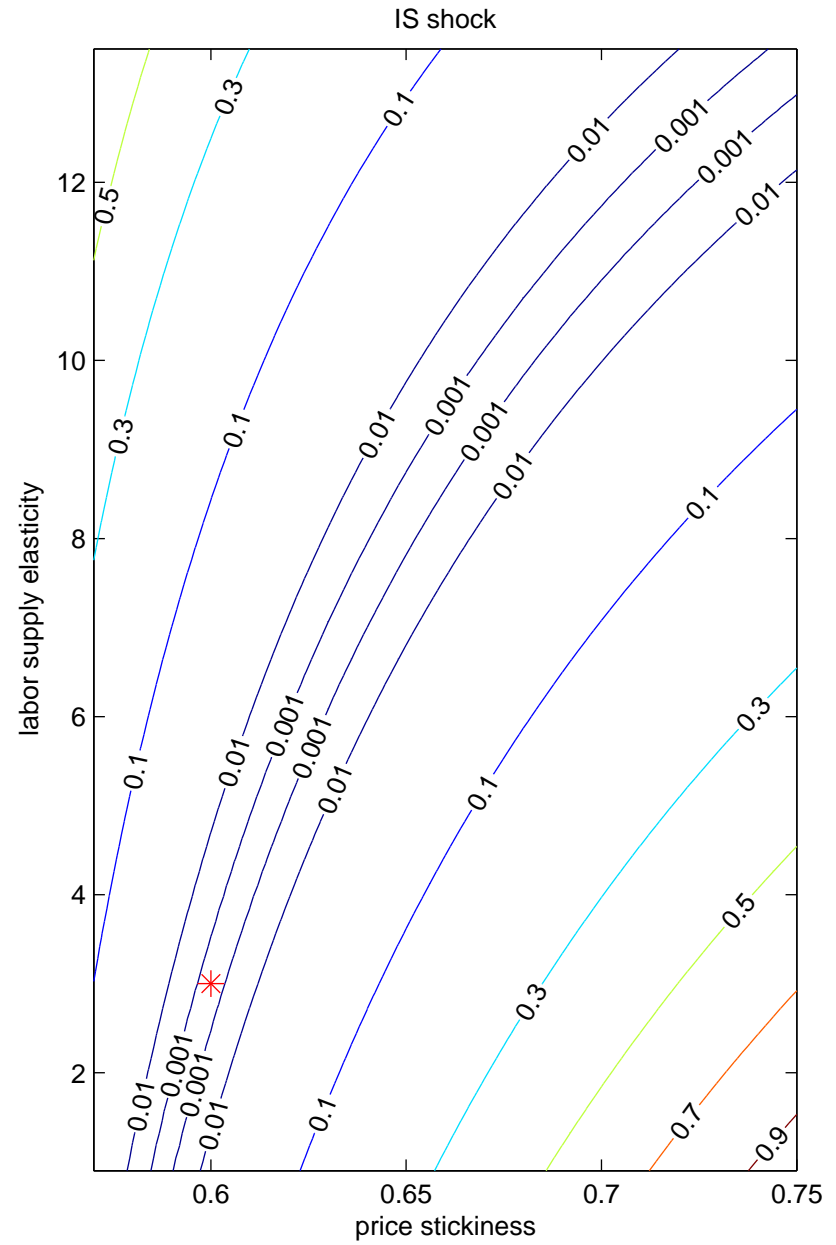
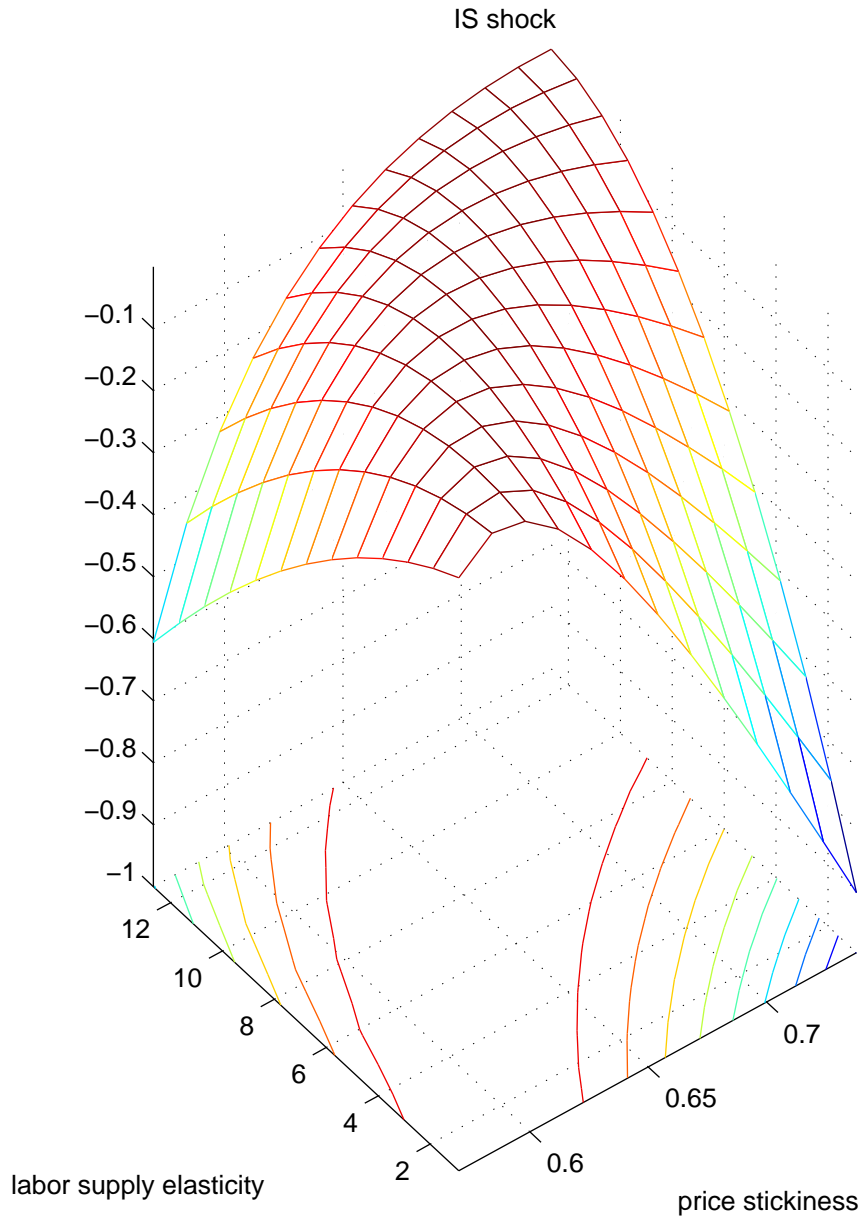
## Computing the distance function

1. Calibrate the model with  $\theta_0$
2. Generate theoretical impulse responses:  $A(\theta_0)$
3. Compute the distance:  $D(\theta) = \|(A(\theta_0) - A(\theta))\|$
4. OK if:  $D(\hat{\theta}) = 0 \iff \hat{\theta} = \theta_0$
5. If one estimates  $\hat{\theta}$  as:

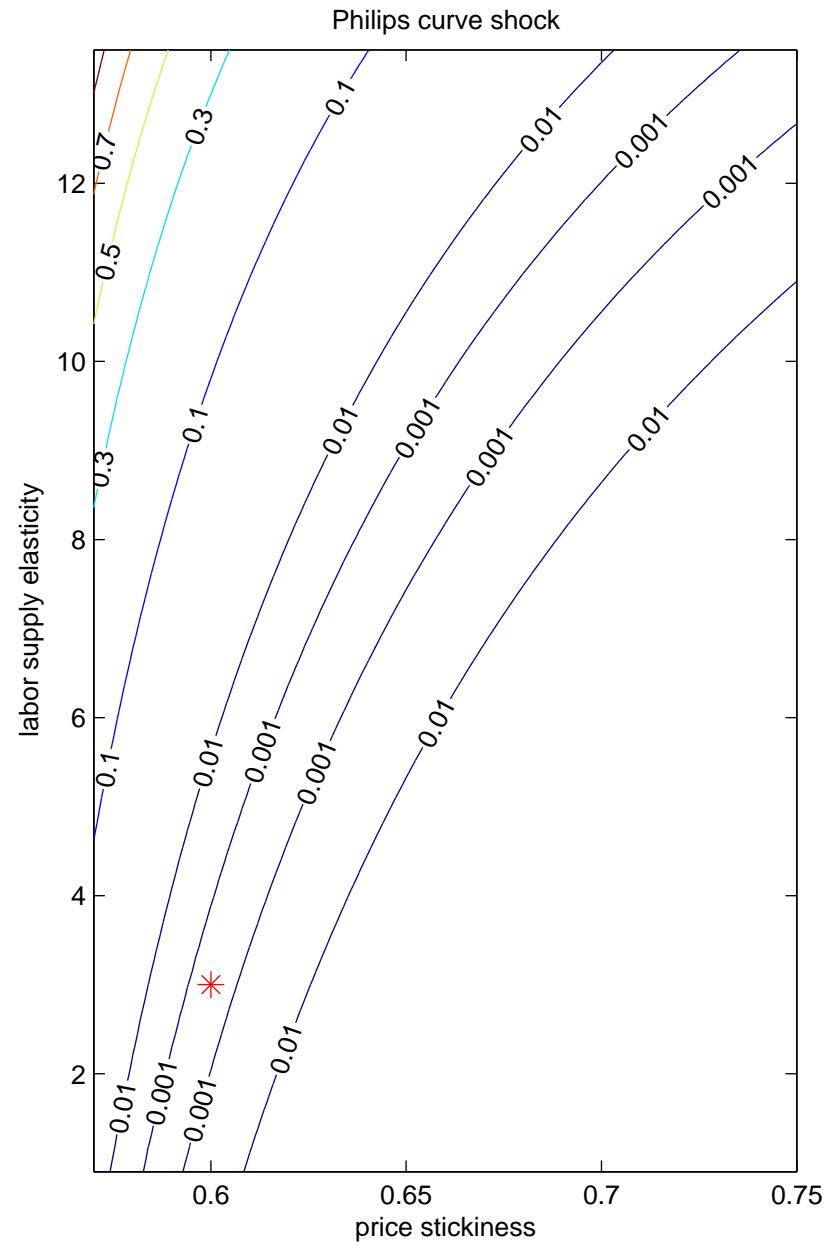
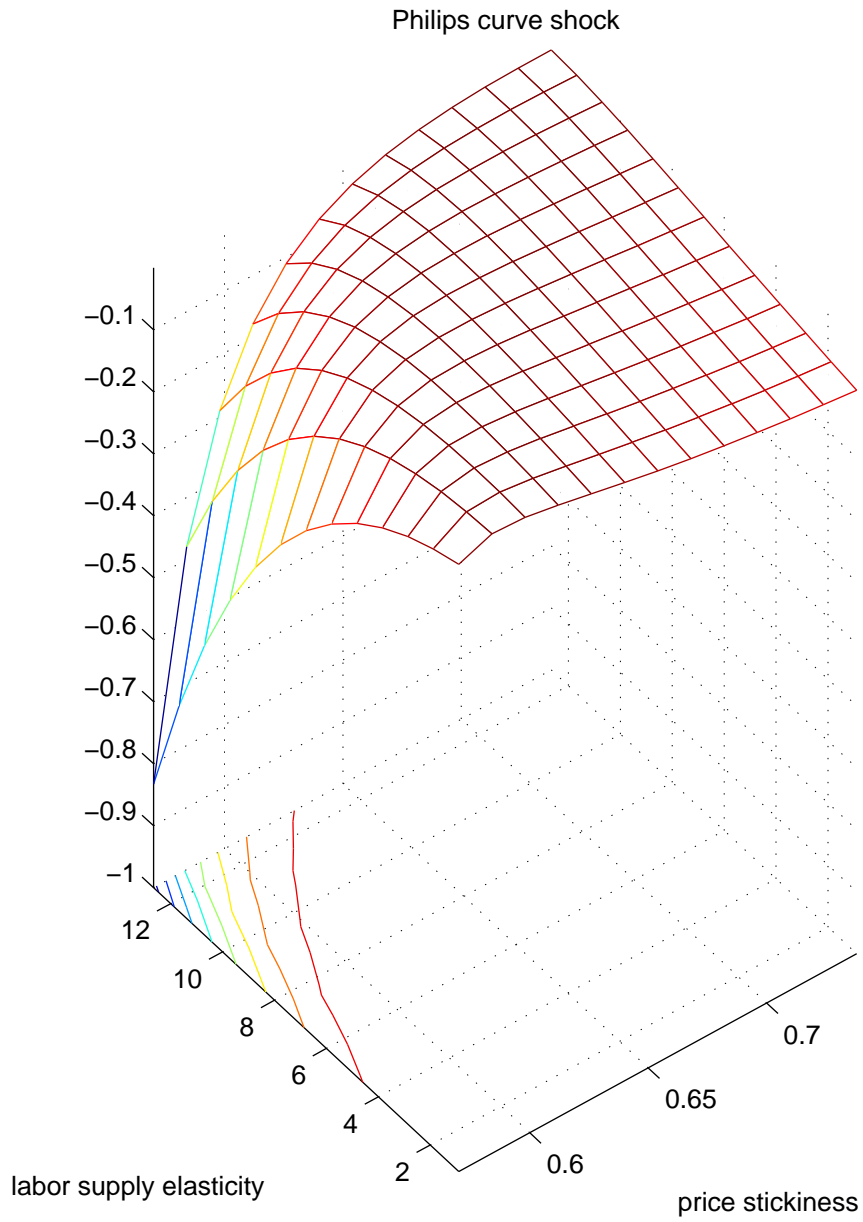
$$\hat{\theta} = \operatorname{argmin} \|A(\theta_0) - A(\theta)\|$$

one obtains:  $\hat{\theta} \neq \theta_0$

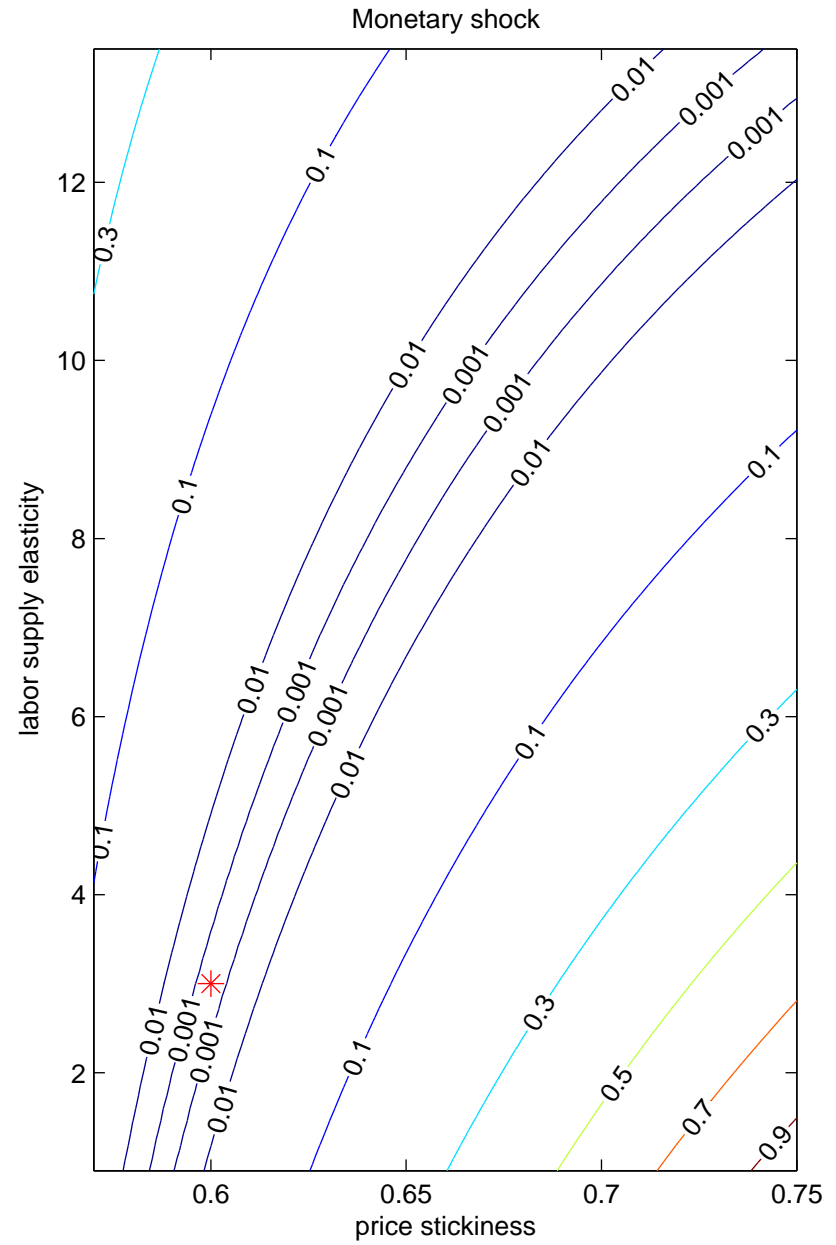
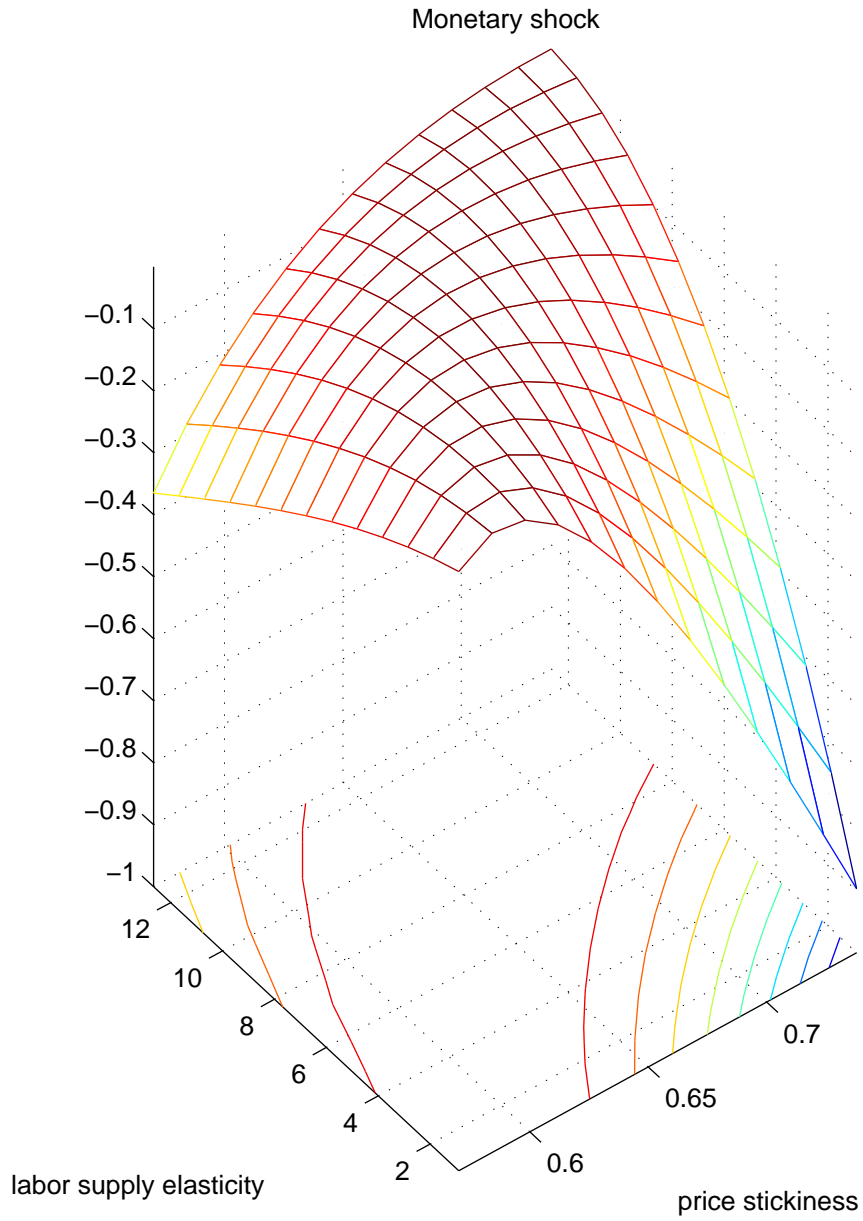
# Minimizing the distance



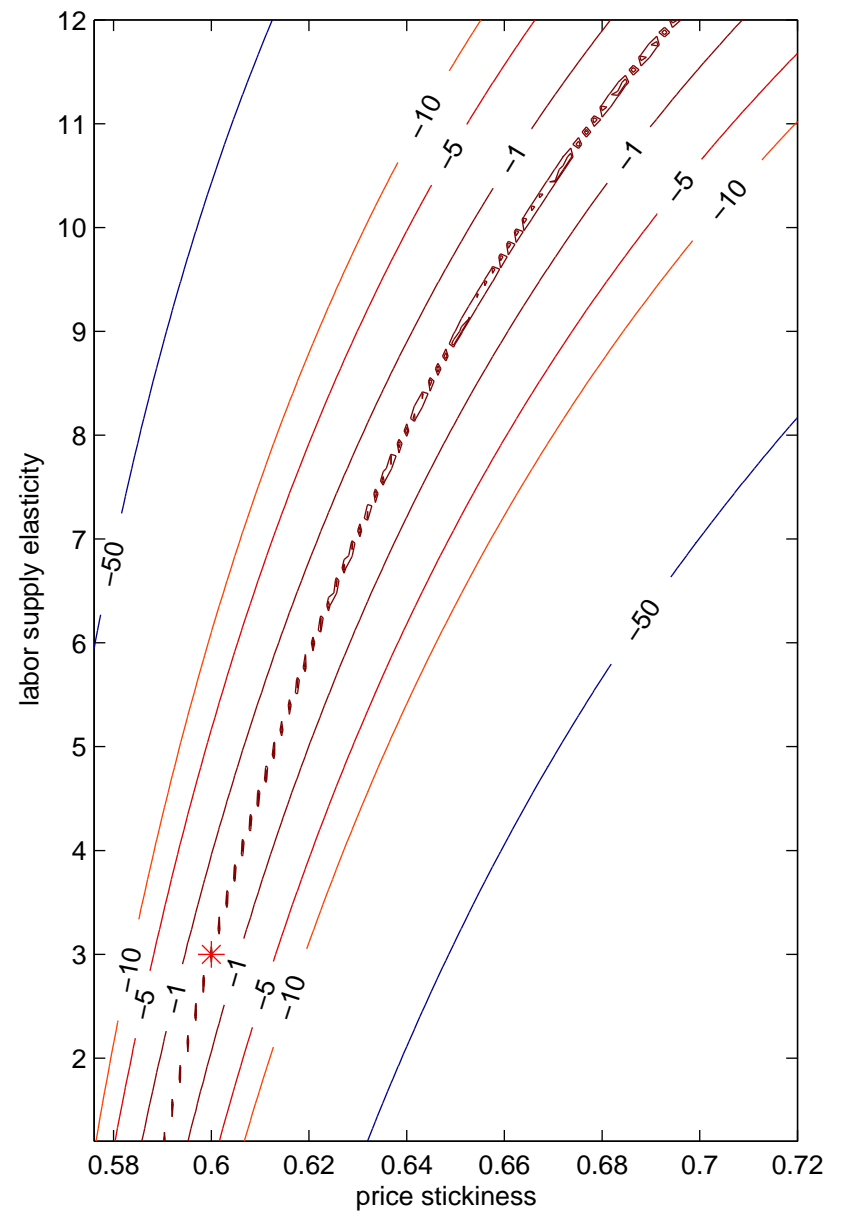
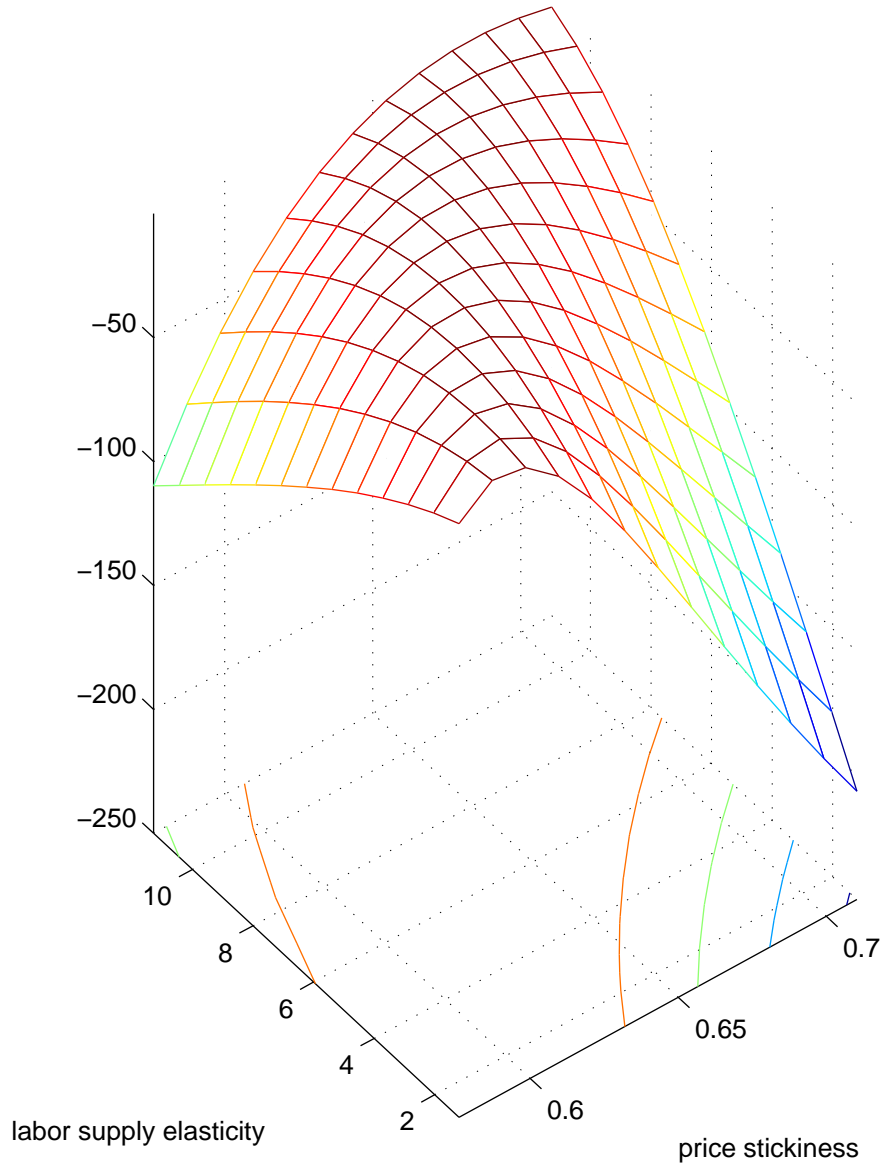
# Minimizing the distance



# Minimizing the distance



Likelihood Function



## 2. The Structural Aspect of Forecasting

Factor models (large dataset + few pervasive shocks) are good forecasting models (Stock-Watson and Giannone, Reichlin and Sala)

Good reduced form of the economy

Use these results to improve structural modelling?

Discipline device for modelling: start with few pervasive shocks

### 3. My Preferred Weakness

Liu (*Econometrica*, 1960)

*"An AR scheme for an error term as a part of a structural estimation is clearly unsatisfactory. [...] it amounts to a confession that an explanation for a systematic part of the movement in the variable to be explained has not been found."*