

# Evaluating An Estimated New Keynesian Small Open Economy Model

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Central Bank Macroeconomic Modelling Workshop,

Banco Central de Chile, Santiago

September 29, 2006

# Introduction

- Open economy DSGE models often criticized for not being able to:
  - account for the transmission mechanism of monetary policy
  - generate enough intrinsic persistence and volatility in the real exchange rates

Adolfson et al. (2005); Justiniano and Preston (2005); Lubik and Schorfheide (2005); Rabanal and Tuesta (2005)

- Typically embed a UIP condition which has received little empirical support from
  - VAR evidence: Eichenbaum and Evans (1995), Faust and Rogers (2003)
  - Forward discount puzzle: Fama (1984), Froot and Frankel (1989)

# What we do

- Consider a modified UIP condition in a DSGE model for a small open economy
  - Allow the risk premium to be negatively related to the expected depreciation of the nominal exchange rate
- Explore the consequences of this modification
  - Implications for policy (impulse responses and actual forecasts)
  - Overall empirical coherence (posterior marginal density, forecasting performance, misspecification)

# Framework

- The model is an open economy version of CEE, 2005 (see Adolfson et al., 2005a)
  - Consumption and investment baskets (domestic and imported goods)
  - Incomplete exchange rate pass-through
  - Stochastic unit-root technology shock (Altig et al., 2003)
- Bayesian estimation following Smets and Wouters (2003)
  - Use Swedish data
  - Foreign output, inflation and interest rates are exogenous
  - From fixed exchange rate regime to explicit inflation targeting in 1993
    - # Allow for a discrete shift in the policy rule

# UIP condition

- “Standard” specification of UIP condition in a small open economy features:

$$\widehat{R}_t - \widehat{R}_t^* = \mathbf{E}_t \Delta \widehat{S}_{t+1} - \widetilde{\phi} \widehat{a}_t + \widehat{\phi}_t$$

- Risk premium is a function of net foreign assets (endogenous) and exogenous disturbance  
⇒ cannot account for the forward premium puzzle
- If  $\widetilde{\phi}$  small, this specification cannot generate a hump-shaped response to a monetary policy shock

# Modified UIP condition

- Allow risk premium to be negatively related to the expected depreciation of the nominal exchange rate

– Modified risk premium

$$\Phi(a_t, \hat{\phi}_t, \frac{S_{t+1}}{S_{t-1}}) = \exp \left\{ -\tilde{\phi}_a a_t + \hat{\phi}_t - \tilde{\phi}_s \left( \frac{S_{t+1}}{S_t} \frac{S_t}{S_{t-1}} - 1 \right) \right\}$$

– Resulting modified UIP condition

$$\hat{R}_t - \hat{R}_t^* = (1 - \tilde{\phi}_s) \mathbf{E}_t \Delta \hat{S}_{t+1} - \tilde{\phi}_s \Delta \hat{S}_t - \tilde{\phi}_a \hat{a}_t + \hat{\phi}_t$$

or

$$\hat{S}_t = (1 - \tilde{\phi}_s) \mathbf{E}_t \hat{S}_{t+1} + \tilde{\phi}_s \hat{S}_{t-1} - (\hat{R}_t - \hat{R}_t^*) - \tilde{\phi}_a \hat{a}_t + \hat{\phi}_t$$

# Main findings

- Modified UIP condition strongly preferred to standard UIP specification
  - Induces hump-shaped impulse responses and intrinsic persistence in the RER
  - Important for the model's forecasting performance (in particular RER and interest rate)
- Forecasting performance of both DSGE specifications typically better than VARs and BVARs
- Both DSGEs are useful as a prior for a VAR model  
DSGE-VAR( $\lambda$ ) analysis; Del Negro and Schorfheide, 2004
  - But data supports relaxation of the cross-equation restrictions in both models

# Structure of talk

- DSGE model overview
- Estimation results
- Effects of modifying the UIP-condition
  - Impulse responses
  - Forecasting performance
- Misspecification analysis
- Concluding remarks

## DSGE model (closed economy aspects)

- Households
  - Utility from consumption, leisure and real cash balances
  - Habit persistence
  - Capital accumulation (investment adjustment costs)
  - Wage stickiness: Calvo, partial indexation
- Domestic goods firms
  - Cobb-Douglas (capital and labor)
  - Working capital
  - Price stickiness: Calvo, partial indexation
- Central bank (Taylor rule)
- Exogenous fiscal policy (VAR for taxes and gov. expenditures)

## DSGE model (open economy aspects)

- Consumption and investment baskets
  - CES composites of domestic and imported goods
- Importing (consumption, investment) and exporting firms
  - Brand naming technology
  - Local currency price setting
  - Price stickiness: Calvo, partial indexation
  - ⇒ Incomplete pass-through
- Trade in foreign bonds with risk-premium
- Foreign economy exogenous (VAR for  $\pi^*$ ,  $y^*$ ,  $R^*$ )

# Monetary policy

- Inflation targeting instrument rule (Taylor type):

$$\begin{aligned}\hat{R}_t = & \rho_R \hat{R}_{t-1} + (1 - \rho_R) \left( \hat{\pi}_t^c + r_\pi \left( \hat{\pi}_{t-1}^c - \hat{\pi}_t^c \right) + r_y \hat{y}_{t-1} + r_x \hat{x}_{t-1} \right) \\ & + r_{\Delta\pi} \Delta \hat{\pi}_t^c + r_{\Delta y} \Delta \hat{y}_t + \varepsilon_{R,t}\end{aligned}$$

- Policy prior to adopting the inflation target in 1993Q1:

- No break
- Fixed exchange rate rule:  $\hat{R}_t = r_S \Delta \hat{S}_t$ ,  $r_S = 10^6$
- Semi-fixed exchange rate rule
- Discrete break in the policy parameters  $\theta_{R,t}$ :

$$\theta_{R,t} = \begin{cases} \theta_{R,1} & \text{if } t < 1993Q1 \\ \theta_{R,2} & \text{if } t \geq 1993Q1 \end{cases}$$

## Estimated shocks

- Technology shocks  $(\mu_{z,t}, \epsilon_t)$ , investment specific  $(\Upsilon_t)$ , asymmetric  $(\tilde{z}_t^*)$
- Markup shocks  $(\lambda_t^d, \lambda_t^{m,c}, \lambda_t^{m,i}, \lambda_t^x)$  - i.i.d.
- Preference shocks  $(\zeta_t^c, \zeta_t^h)$
- Risk premium shock  $(\tilde{\phi}_t)$
- Monetary policy shocks  $(\varepsilon_{R,t}, \bar{\pi}_t^c)$

# Estimation

- Estimate 43 (51 with break) parameters
  - Price stickiness, technology growth, habit formation, policy rule, persistence and std of shocks, etc.
  - Calibrate most parameters pertaining to the steady state (“great ratios” etc.)

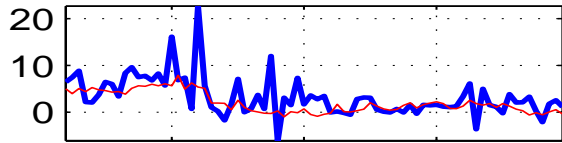
- Swedish data 1980Q1 – 2004Q4

- Match large set of variables (facilitate identification of parameters)

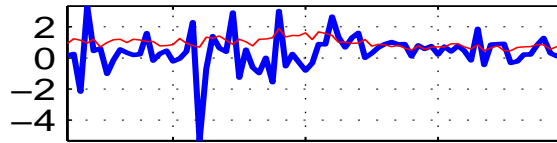
$$\tilde{Y}_t = \begin{bmatrix} \pi_t^d & \Delta \ln(W_t/P_t) & \Delta \ln C_t & \Delta \ln I_t & \hat{x}_t & R_t & \hat{H}_t & \Delta \ln Y_t \dots \\ & \Delta \ln \tilde{X}_t & \Delta \ln \tilde{M}_t & \pi_t^{cpi} & \pi_t^{def,i} & \Delta \ln Y_t^* & \pi_t^* & R_t^* \end{bmatrix}'.$$

- No detrending - work with raw data (except imports and exports)

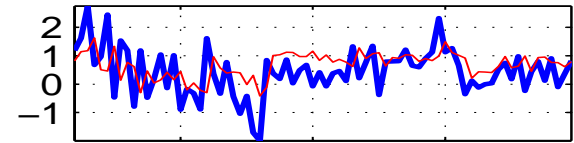
Domestic inflation



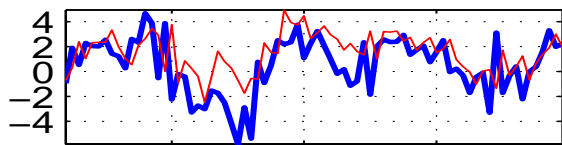
Real wage



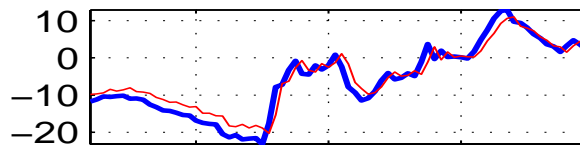
Consumption



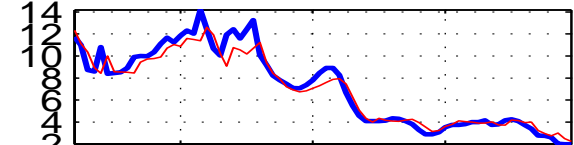
Investment



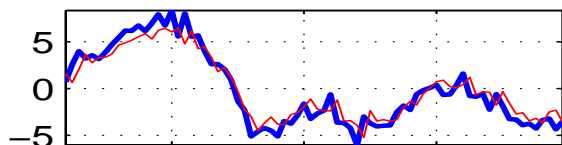
Real exchange rate



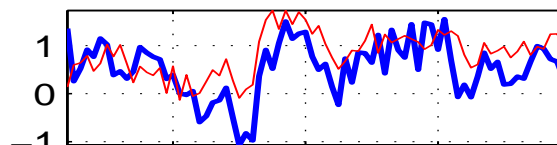
Interest rate



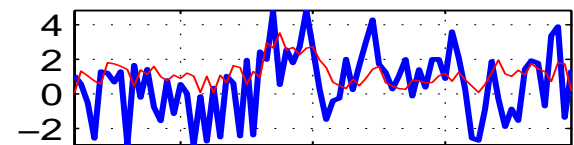
Hours worked



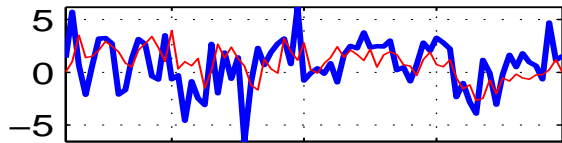
Output



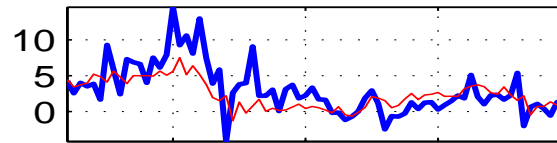
Export



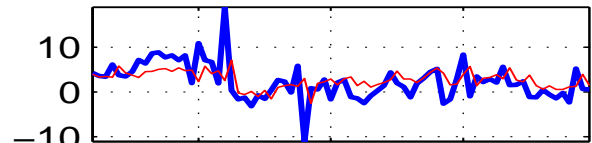
Import



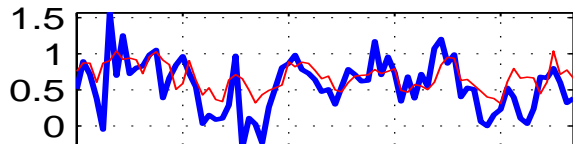
CPI inflation



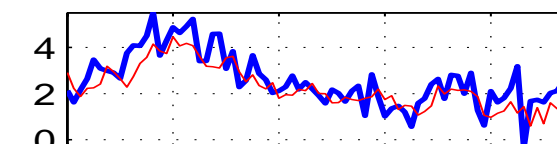
Invest. defl. infl.



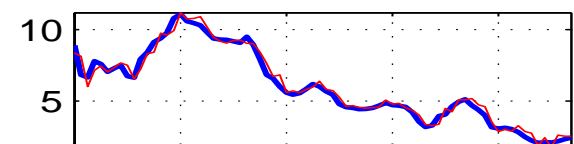
World output



World inflation



World interest rate



1986 1990 1994 1998 2002

1986 1990 1994 1998 2002

1986 1990 1994 1998 2002

## Estimation results: model specification

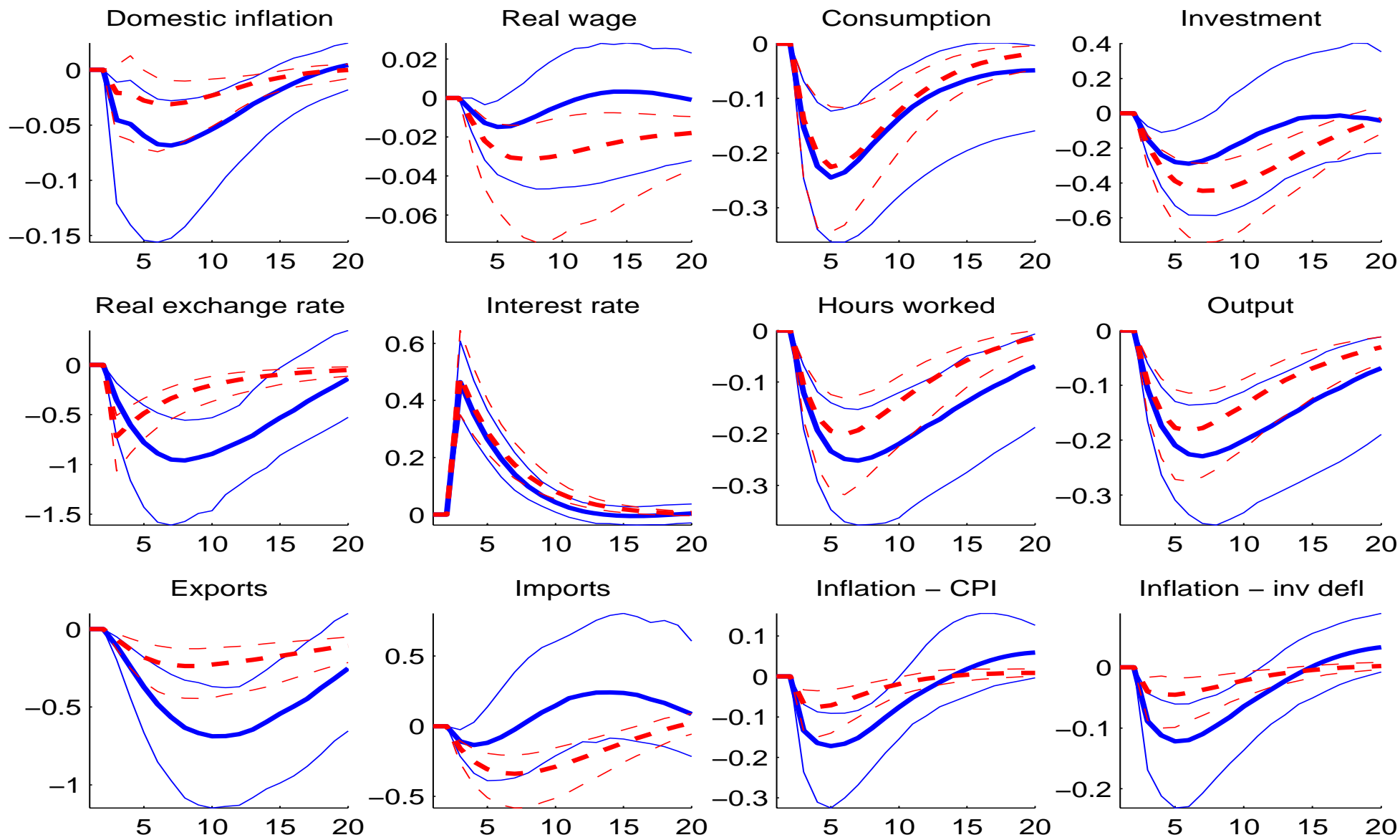
- Bayesian posterior odds in favor of a break in policy rule
  - Parameters similar except for standard deviation of policy shock (shrinking)
- Bayesian posterior odds in favor of the modified UIP specification
  - Deep parameters similar across all specifications
  - Risk premium shock estimates different

Exert from Table 1: Prior and posterior distributions

Parameter		Prior distribution			Posterior distributions									
		type	mean	std. dev.	Instrument rule without policy break		Fixed exchange rate rule		Semi-fixed exchange rate rule		Instrument rule with policy break			
					median	std.	median	std.	median	std.	UIP		Modified UIP	
										median	std.	median	std.	
UIP modification	$\tilde{\phi}_s$	beta	0.500	0.15									0.611	0.063
Risk premium shock persistence	$\rho_{\tilde{\phi}}$	beta	0.850	0.100	0.932	0.025	0.954	0.013	0.959	0.016	0.929	0.026	0.679	0.101
Risk premium shock	$\sigma_{\tilde{\phi}}$	invg.	0.050	2	0.372	0.079	0.319	0.035	0.297	0.036	0.376	0.079	0.805	0.221
Interest rate smoothing	$\rho_{R,1}$	beta	0.800	0.050	0.909	0.016			0.827	0.039	0.879	0.022	0.883	0.025
Inflation response	$r_{\pi,1}$	normal	1.700	0.100	1.664	0.099			1.710	0.098	1.679	0.100	1.679	0.100
Diff. infl response	$r_{\Delta\pi,1}$	normal	0.300	0.050	0.095	0.030			0.288	0.048	0.133	0.049	0.156	0.056
Real exch. rate resp.	$r_{x,1}$	normal	0.000	0.050	0.046	0.027					0.039	0.030	0.018	0.027
Nominal exch. response	$r_s$	normal	100	10			$10^6$	calib.	2.0	0.8				
Output response	$r_{y,1}$	normal	0.250	0.050	0.129	0.046			0.216	0.051	0.113	0.044	0.138	0.048
Diff. output response	$r_{\Delta y,1}$	normal	0.125	0.050	0.152	0.036			0.142	0.050	0.127	0.041	0.120	0.046
Monetary policy shock	$\sigma_{R,1}$	invg.	0.150	2	0.249	0.024			2.335	0.778	0.398	0.060	0.398	0.066
Inflation target shock	$\sigma_{\bar{\pi}^*,1}$	invg.	0.050	2	0.116	0.041			0.083	0.054	0.148	0.067	0.248	0.085
Interest rate smoothing	$\rho_{R,2}$	beta	0.800	0.050			0.884	0.018	0.864	0.021	0.896	0.018	0.874	0.022
Inflation response	$r_{\pi,2}$	normal	1.700	0.100			1.725	0.090	1.747	0.089	1.709	0.099	1.718	0.097
Diff. infl response	$r_{\Delta\pi,2}$	normal	0.300	0.050			0.127	0.023	0.143	0.025	0.104	0.026	0.120	0.027
Real exch. rate response	$r_{x,2}$	normal	0.000	0.050			0.022	0.019	-0.001	0.003	0.038	0.026	-0.023	0.020
Output response	$r_{y,2}$	normal	0.250	0.050			0.269	0.040	0.274	0.039	0.107	0.041	0.106	0.041
Diff. output response	$r_{\Delta y,2}$	normal	0.125	0.050			0.099	0.031	0.107	0.030	0.104	0.030	0.105	0.030
Monetary policy shock	$\sigma_{R,2}$	invg.	0.150	2			0.102	0.013	0.094	0.011	0.104	0.013	0.103	0.013
Inflation target shock	$\sigma_{\bar{\pi}^*,2}$	invg.	0.050	2			0.065	0.030	0.069	0.035	0.080	0.038	0.077	0.038
Log marg. likelihood					-2285.8		-2636.72		-2348.24		-2268.33		-2252.57	

# Impulse response functions

- Monetary policy shock
  - From jump to hump-shaped response for the real exchange rate with modified UIP condition
  - Negative correlation between the risk premium and expected exchange rate changes  $\implies$  increased persistence in RER response



# Forecasting performance

- Compare RMSEs for various models, 1999Q1 – 2004Q4 (1-8 quarters ahead)
  - Modified UIP condition outperforms on forecasts for  $x_t$ ,  $R_t$ , and  $\pi_t^{cpi}$ , less successful for  $Y_t$  and  $H_t$
  - DSGE well in line or better than alternative forecasting tools

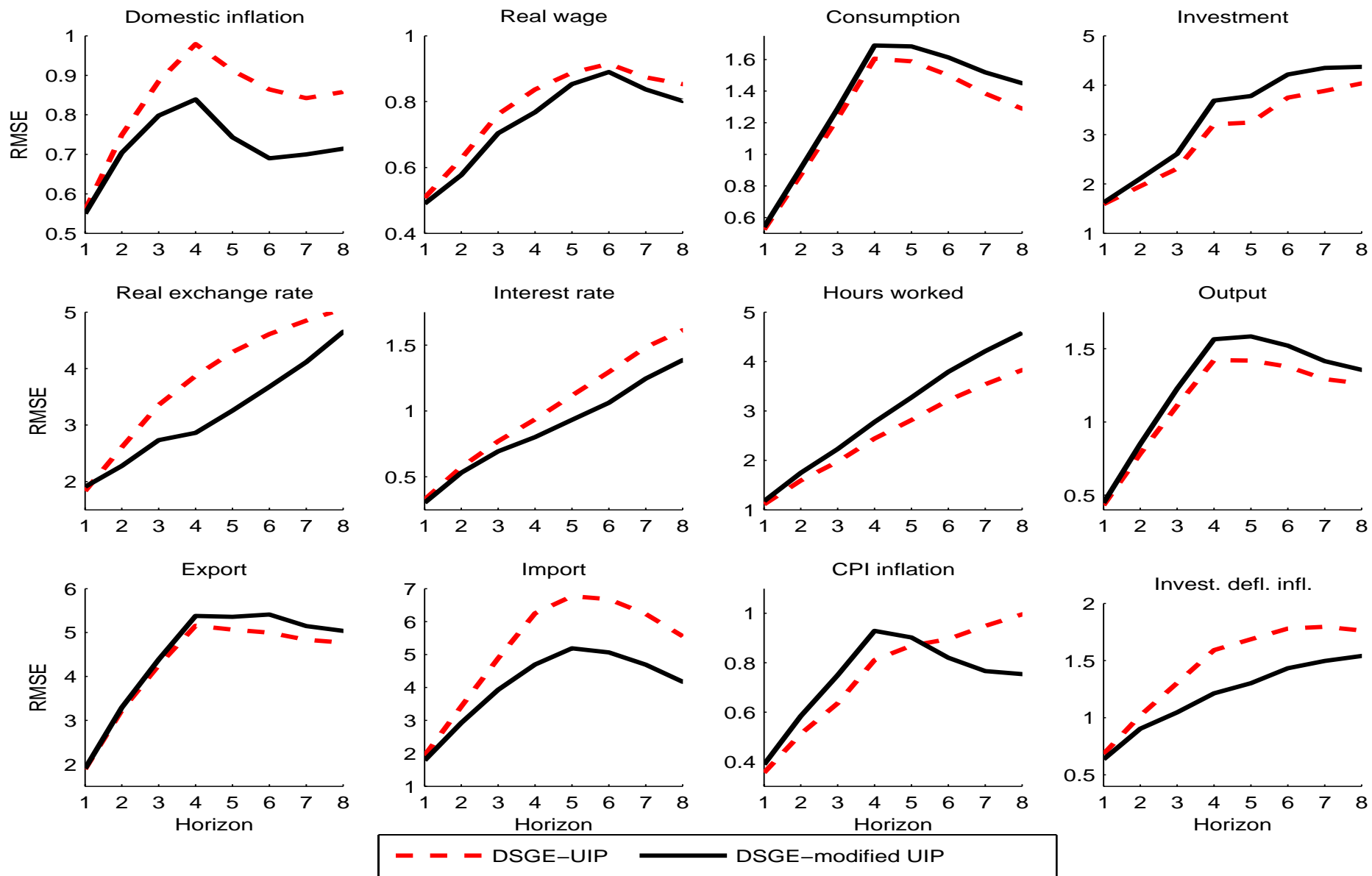
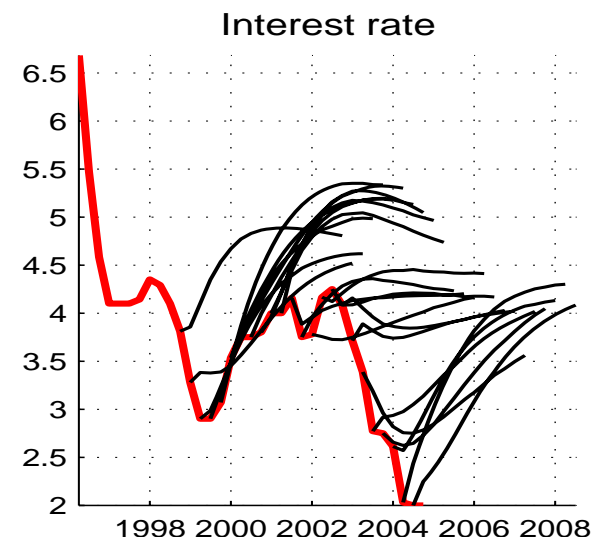
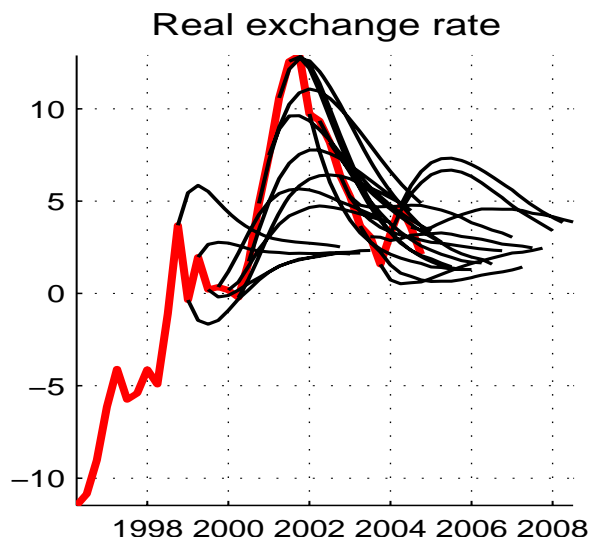
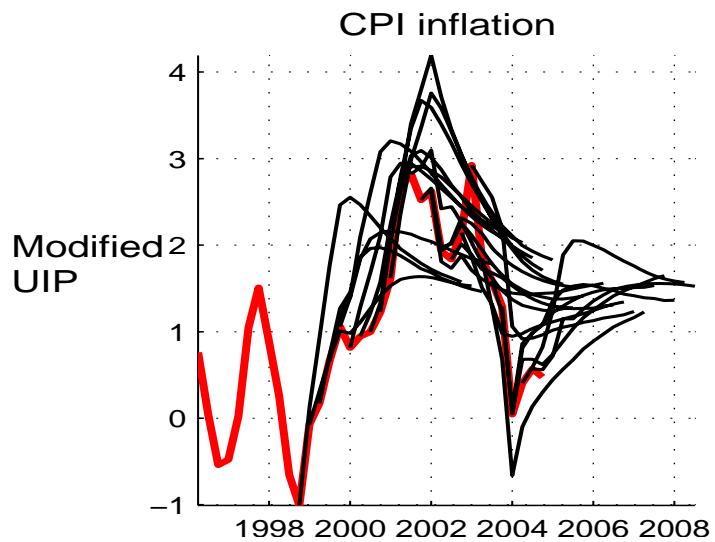
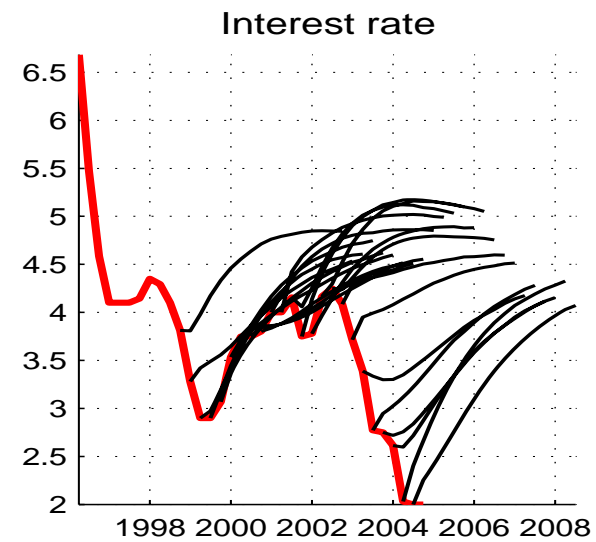
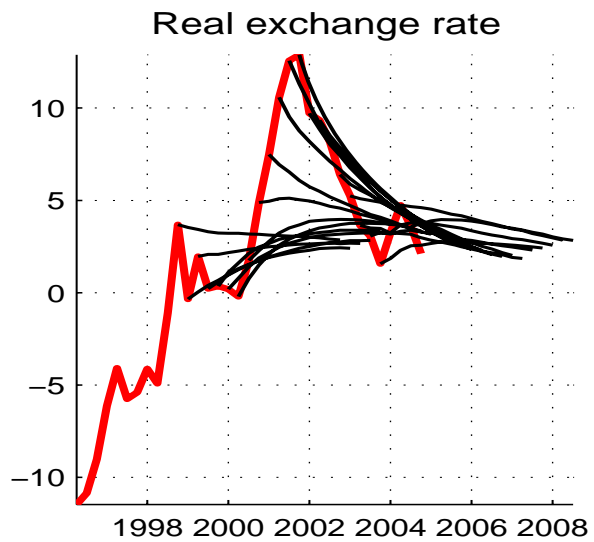
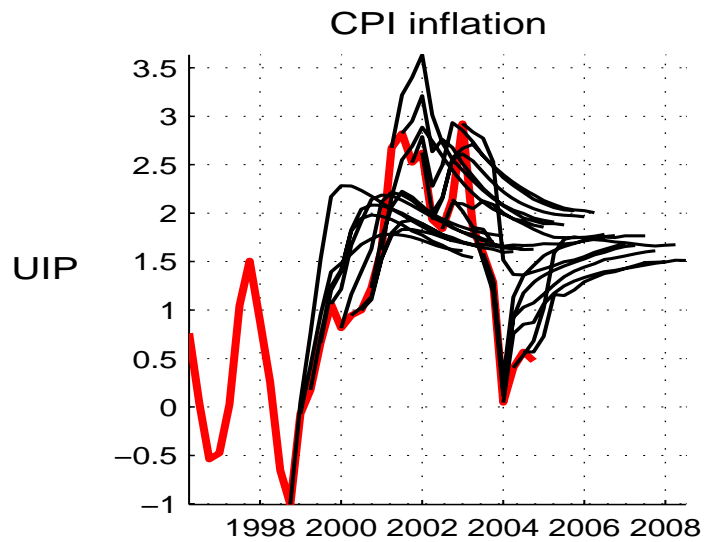


Figure 1:



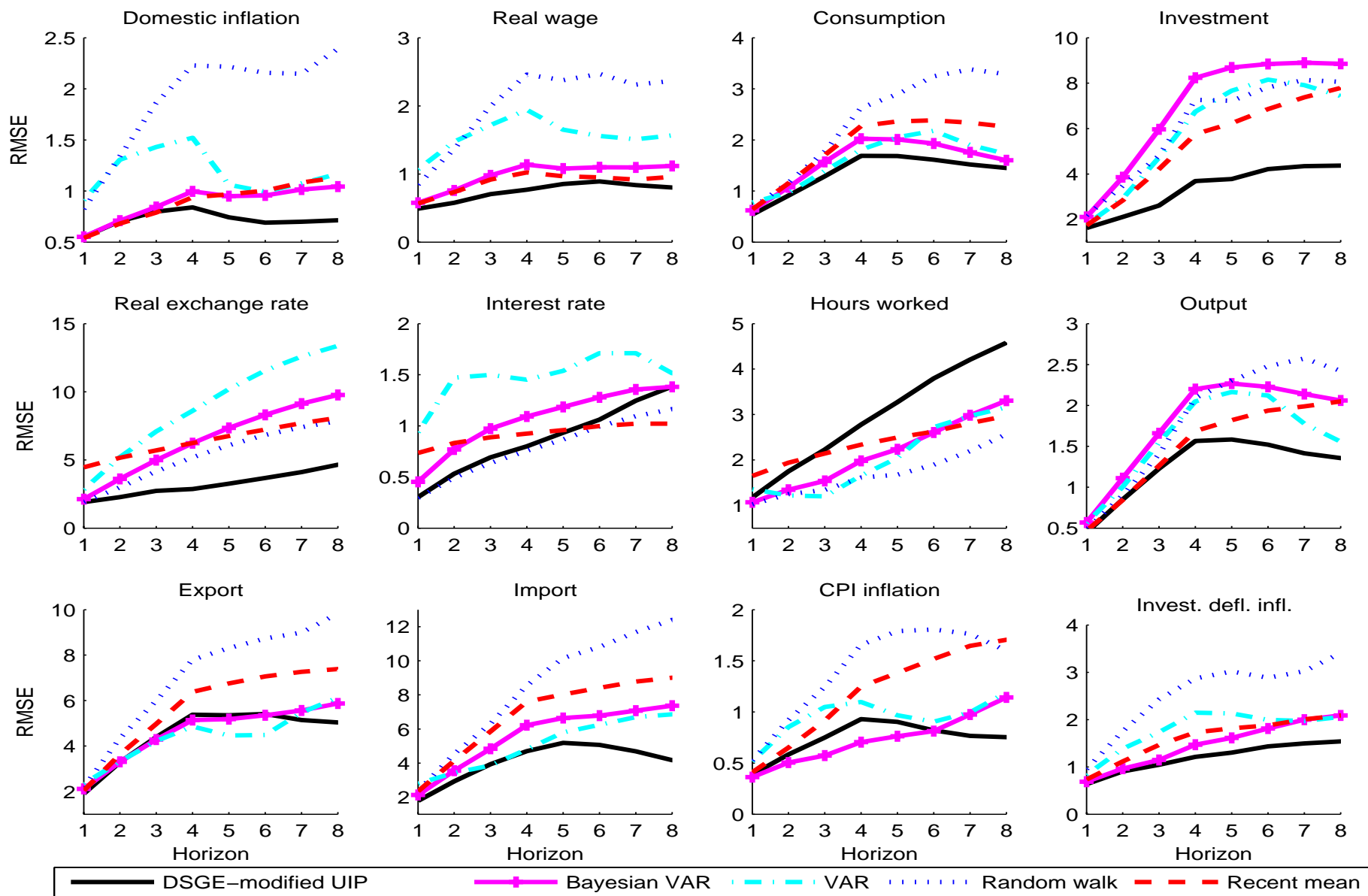


Figure 2:

# Misspecification analysis

- Do the data support the DSGE models' cross-equation restrictions?
  - Use Del Negro and Schorfheide (2004) and Del Negro, Schorfheide, Smets and Wouters (2004) to measure the degree of misspecification
- Compare marginal likelihoods of DSGE and a VAR with prior centered on the DSGE (i.e, DSGE-VAR( $\lambda$ ))
  - Tightness in the prior  $\lambda$  from diffuse ( $\approx 3$ ) to very informative ( $\infty$ ) governs to what extent the cross-equation restrictions are enforced
  - Let data determine preferred relaxation ( $\hat{\lambda}$ )
- Check implications by comparing vector autocovariance functions in the various specifications

## Misspecification - results

- How well does the DSGE-VAR/VECM( $\infty$ ) approximate the DSGE?
  - DSGE-VECM better approximation than DSGE-VAR (see VACF; fig. 4a)
- Both UIP specifications show signs of misspecification
  - $\hat{\lambda} = 5-7 \ll \infty$  and about the same for both specifications
  - $\hat{\lambda}$  implies substantial relaxation of the cross-equation restrictions
  - Hybrid model improves considerably in terms of marginal likelihood
- Weak support for model based cointegration
  - DSGE-VECM has zero posterior probability compared to DSGE-VAR

Table 2: Log marginal likelihood of VAR/VECM with DSGE prior. With regime change in the Taylor-type instrument rule.

$\lambda$	DSGE-VAR		DSGE-VECM	
	UIP	Modified UIP	UIP	Modified UIP
2.714	-2164.78	-2163.19		
2.929			-2190.79	-2187.77
4	-1998.27	-1996.23	-2028.20	-2025.89
4.5	-1980.51	-1979.88	-2002.60	-2000.46
5	<b>-1964.96</b>	-1970.43	-1987.80	-1985.78
5.5	-1970.78	<b>-1965.86</b>	-1979.47	-1977.53
7	-1965.98	-1966.72	<b>-1974.12</b>	<b>-1972.28</b>
7.5	-1968.72	-1969.46	-1975.74	-1973.91
8	-1972.05	-1972.79	-1978.24	-1976.39
10	-1988.94	-1987.02	-1992.72	-1990.83
25	-2083.09	-2081.13	-2093.68	-2091.13
50	-2143.73	-2140.50	-2164.52	-2160.00
$\infty$	-2232.06	-2227.65	-2270.78	-2265.08
DSGE	-2268.33	-2252.57	-2268.33	-2252.57

Note: The table displays laplace approximations of the log marginal likelihood.  $\lambda = 2.714$  and  $\lambda = 2.929$  are the minimal tightnesses for the VAR and VECM, respectively. Bold numbers indicate the  $\lambda$  with the maximal log marginal likelihood.

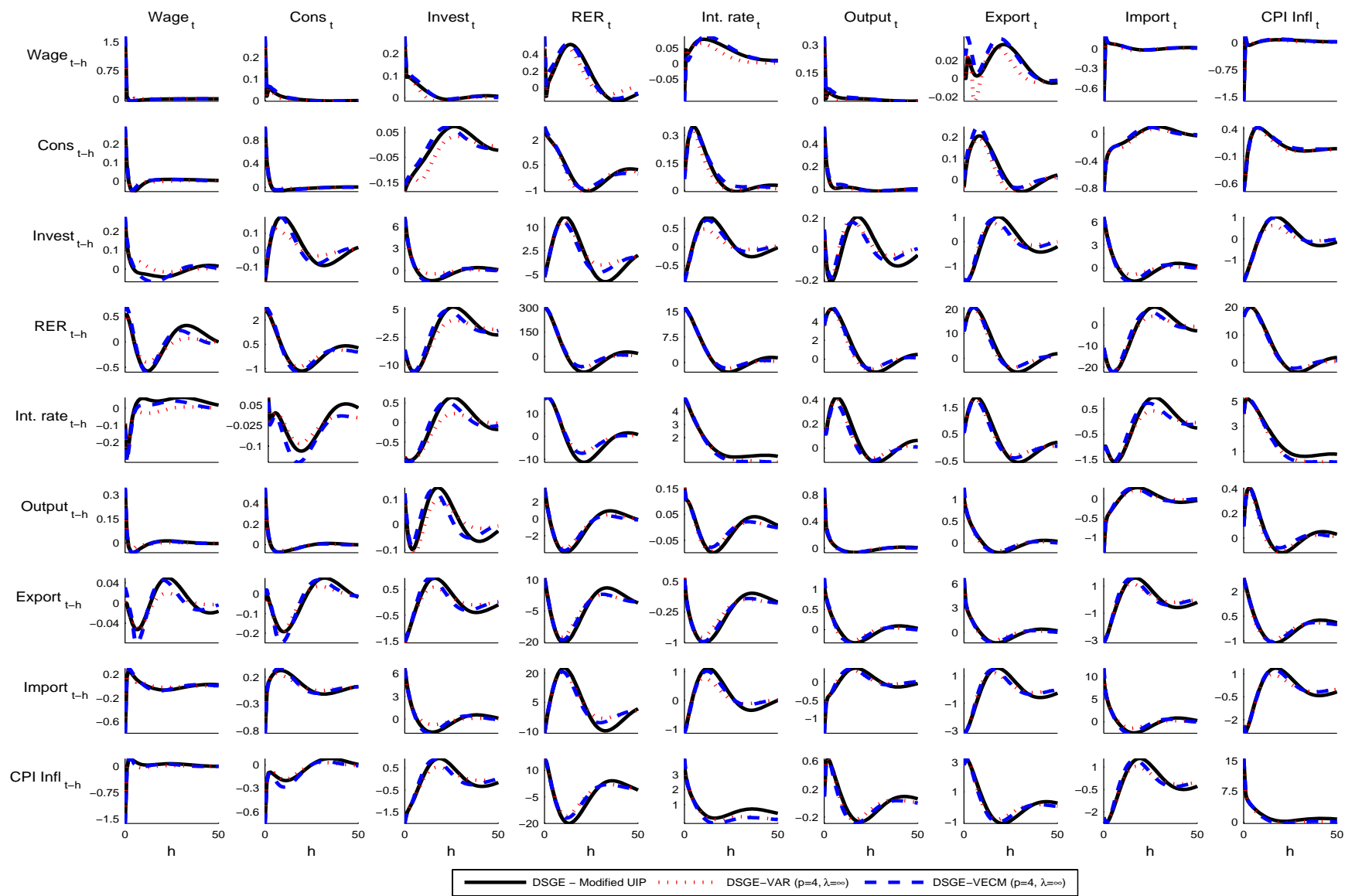


Figure 3:

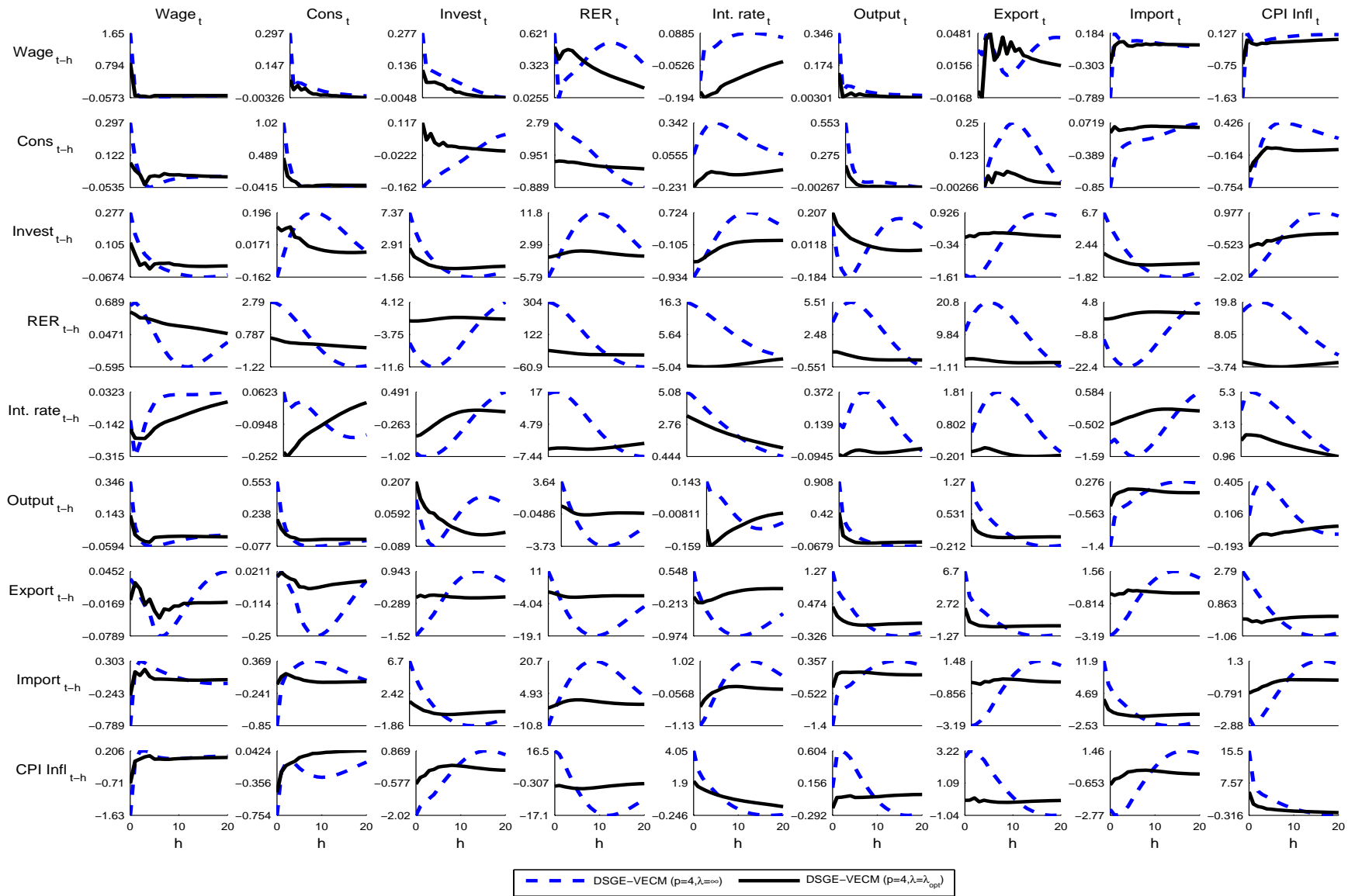


Figure 4:

# Concluding remarks

- For a DSGE to be successful in a policy environment, it needs to be empirically coherent:
  - Comply with the CB's view of the monetary transmission channel
  - Forecast based policy environment  $\Rightarrow$  forecasting performance
- Modified UIP condition allowing for a negative correlation between the risk premium and expected exchange rate changes
  - Induces intrinsic persistence in the real exchange rate
  - Improves on the model's forecasting properties
- DSGE model is useful as a prior for a VAR model
  - DSGE misspecified in the sense that data relaxes the cross-equation restrictions

# Identification

- Are key parameters identified?

- Consider the modified UIP condition

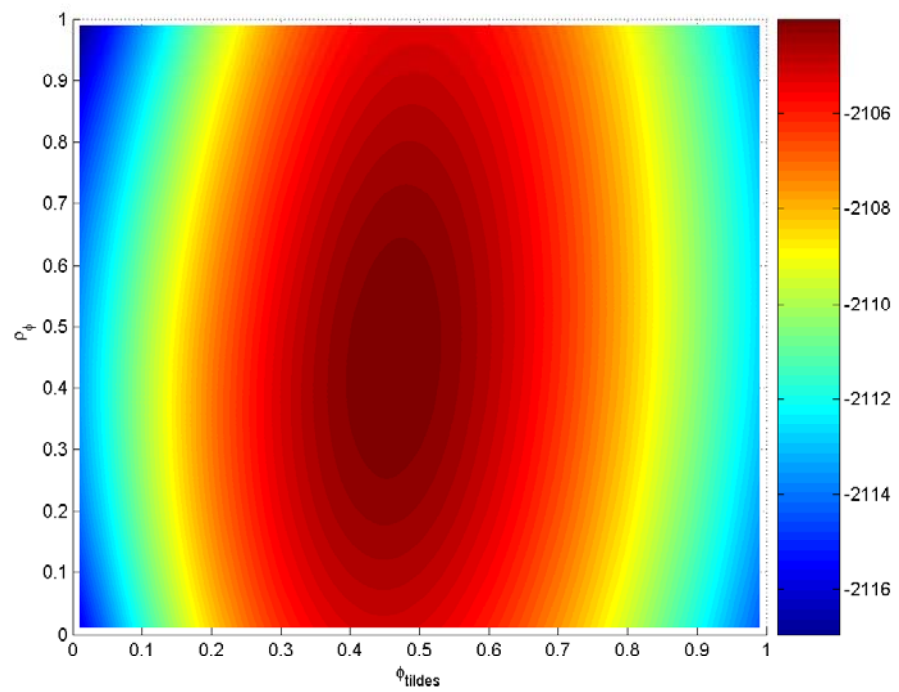
$$\hat{S}_t = (1 - \tilde{\phi}_s) E_t \hat{S}_{t+1} + \tilde{\phi}_s \hat{S}_{t-1} - (\hat{R}_t - \hat{R}_t^*) - \tilde{\phi}_a \hat{a}_t + \hat{\phi}_t$$

- Can  $\tilde{\phi}_s$  be identified when simultaneously allowing for correlated risk premium shocks  $\hat{\phi}_t$

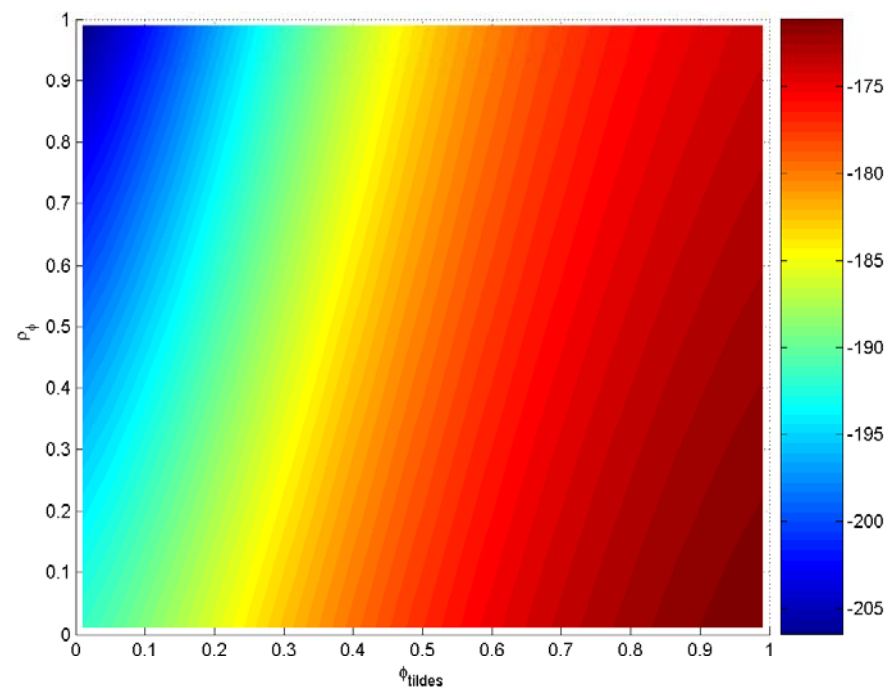
- Contour plots of the log-likelihood in  $\{\tilde{\phi}_s, \rho_{\tilde{\phi}}\}$ -space

- Both parameters well identified when using all variables to compute likelihood

15 observables



Only RER



## Identification (cont.)

- Small sample properties of ML estimation using simulated data from the model (Adolfson and Lindé, 2006)
  - Distribution of estimates located around the true parameters
  - Convergence in distribution when sample size increased

