

Probability forecasts of deflation for the Euro area and Japan: an evaluation using scoring rules for binary outcomes

Inske Pirschel (Swiss National Bank), Christian Schumacher (Deutsche Bundesbank)

Motivation

Why do we care about deflation?

- ▶ In light of the debate on deflation risks after the Great Recession in the Euro area, there is a renewed interest in deflation phenomenon
- ▶ Academic literature is relatively silent on this matter

How can we measure deflation?

- ▶ Definition by Cogley and Sargent (2015) identifies a deflationary period, if the cumulative rate of price changes over the forecast horizon is negative, i.e. $\pi_{T+h}^h = \ln(HICP_{T+h}/HICP_T) = \sum_{i=1}^h \pi_{T+i} < 0$

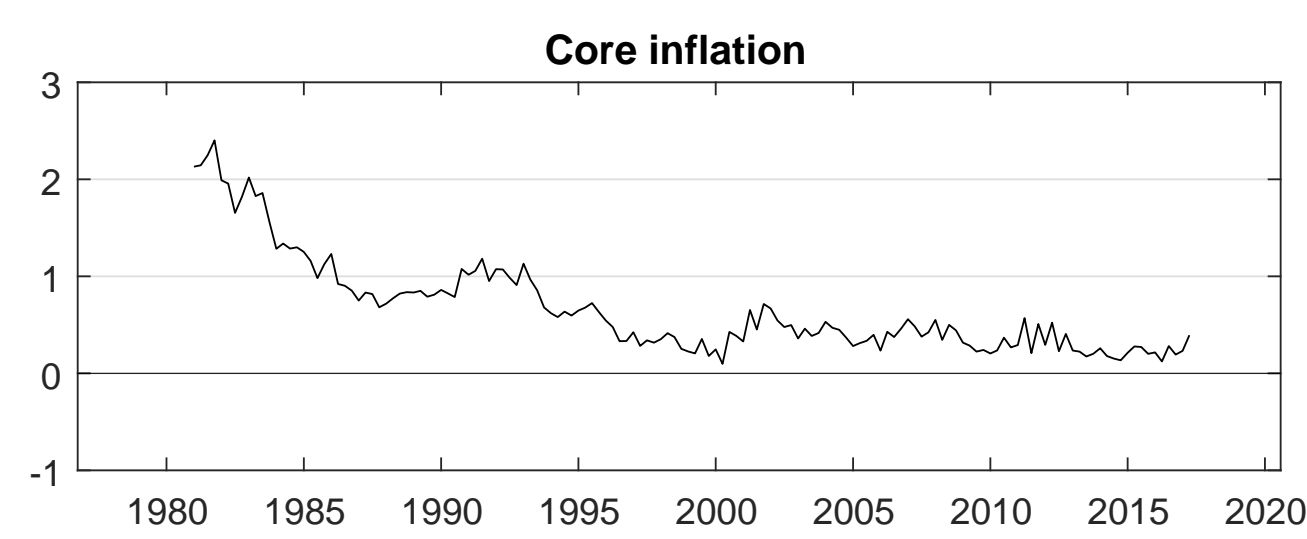
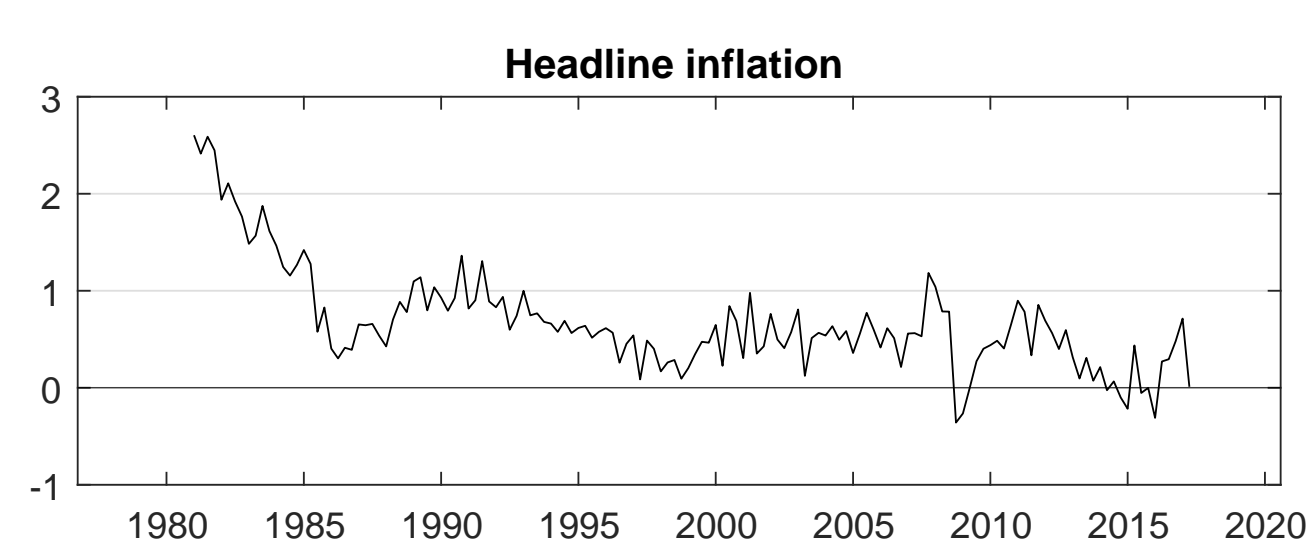
Our contribution

- ▶ We forecast HICP changes and core inflation in the Euro area and Japan using inflation forecast models from the recent literature
- ▶ We provide estimated probabilities of future deflation based on individual and pooled density forecasts
- ▶ We use scoring rules for binary outcomes to evaluate the probability forecasts

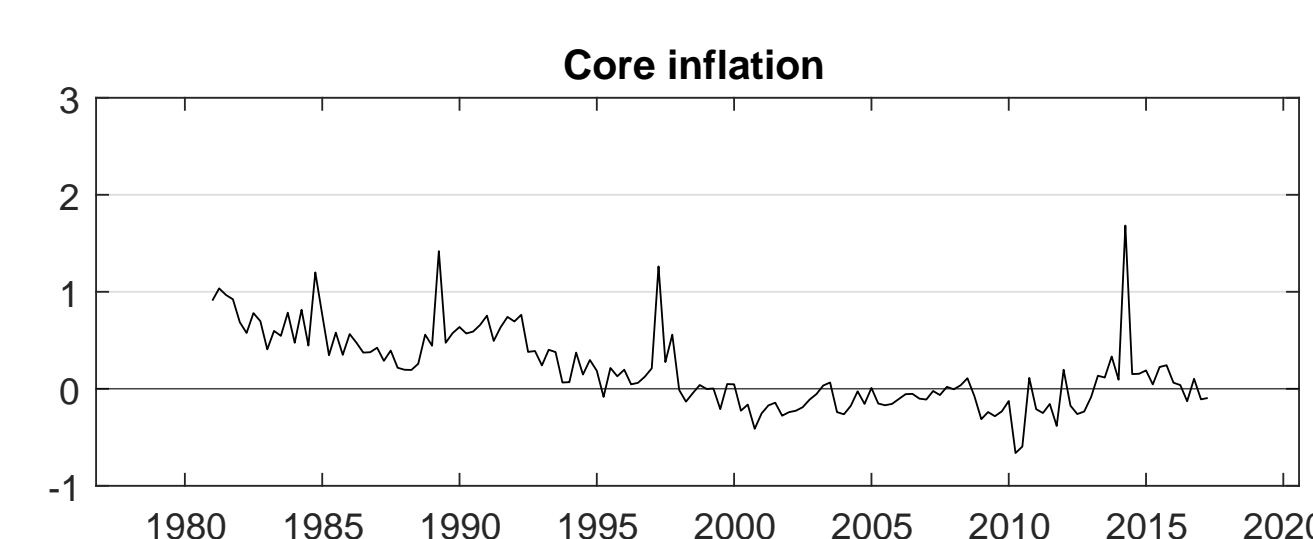
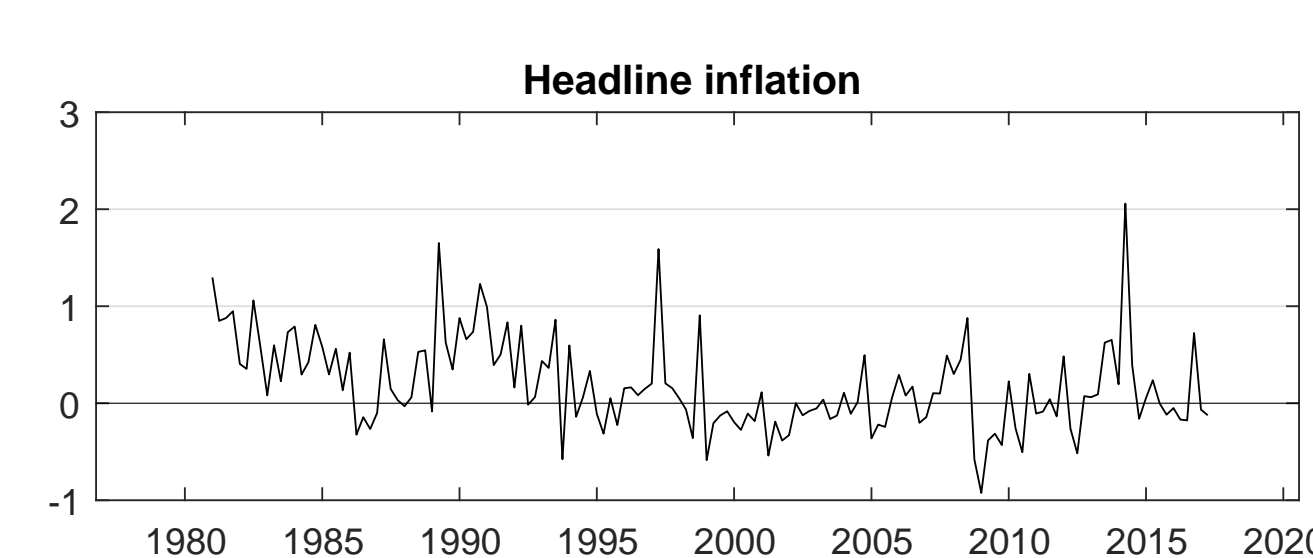
Data

- ▶ We use quarterly data for HICP rates of change, core inflation and real GDP for the Euro area and Japan from 1980Q1 until 2017Q2
- ▶ In the Euro area, there are very few short-lived deflation observations for HICP rates of change and no deflation observations at all for core inflation
- ▶ In Japan, several deflationary periods are visible in both measures
- ▶ Time series properties of headline and core inflation are quite different in the Euro area and in Japan

Euro area



Japan



Computation of deflation probabilities

- ▶ For "deflation" event \mathcal{D} , define binary variable $y_{\mathcal{D}} = 1$ if \mathcal{D} occurs and $y_{\mathcal{D}} = 0$ if \mathcal{D} does not occur
- ▶ In period T , we make prediction $P(\mathcal{D})$, i.e. we forecast the probability that \mathcal{D} occurs between period $T + 1$ and $T + h$ given observations $Y_{1:T}^o$: $P(\mathcal{D}) = P(\pi_{T+h}^h < 0 | Y_{1:T}^o)$
- ▶ We can use the sample-based estimator $P(\pi_{T+h}^h < 0 | Y_{1:T}^o) = \frac{1}{R} \sum_{r=1}^R \mathcal{I}(\pi_{T+h}^{h(r)} < 0 | Y_{1:T}^o)$ given samples from the predictive density $\pi_{T+h}^{h(r)} \sim p(\pi_{T+h}^h)$
- ▶ After period $T + h$, we have an observed value $y_{\mathcal{D}}^o$ that can be used to evaluate $P(\mathcal{D})$

Evaluation using scoring rules for binary outcomes

- ▶ Evaluation period starts in 2000Q1 and ends in 2017Q2
- ▶ To evaluate $P(\mathcal{D})$, we employ the logarithmic score (*LogS*), the Brier score (*BrS*) and an asymmetric score (*As*) from Elliott, Ghanem and Krüger (2016):

$$\text{LogS} = y_{\mathcal{D}}^o \log P(\mathcal{D}) + (1 - y_{\mathcal{D}}^o) \log(1 - P(\mathcal{D}))$$

$$\text{BrS} = y_{\mathcal{D}}^o [-(1 - P(\mathcal{D}))^2] + (1 - y_{\mathcal{D}}^o) [-P(\mathcal{D})^2] = -(y_{\mathcal{D}}^o - P(\mathcal{D}))^2$$

$$\text{As} = y_{\mathcal{D}}^o [\log P(\mathcal{D}) - P(\mathcal{D}) + 1] + (1 - y_{\mathcal{D}}^o) [-P(\mathcal{D})]$$

- ▶ While *BrS* and *LogS* are indifferent between false negatives and false positives, *As* implies stronger losses for deflation events that were not predicted

Forecasting models

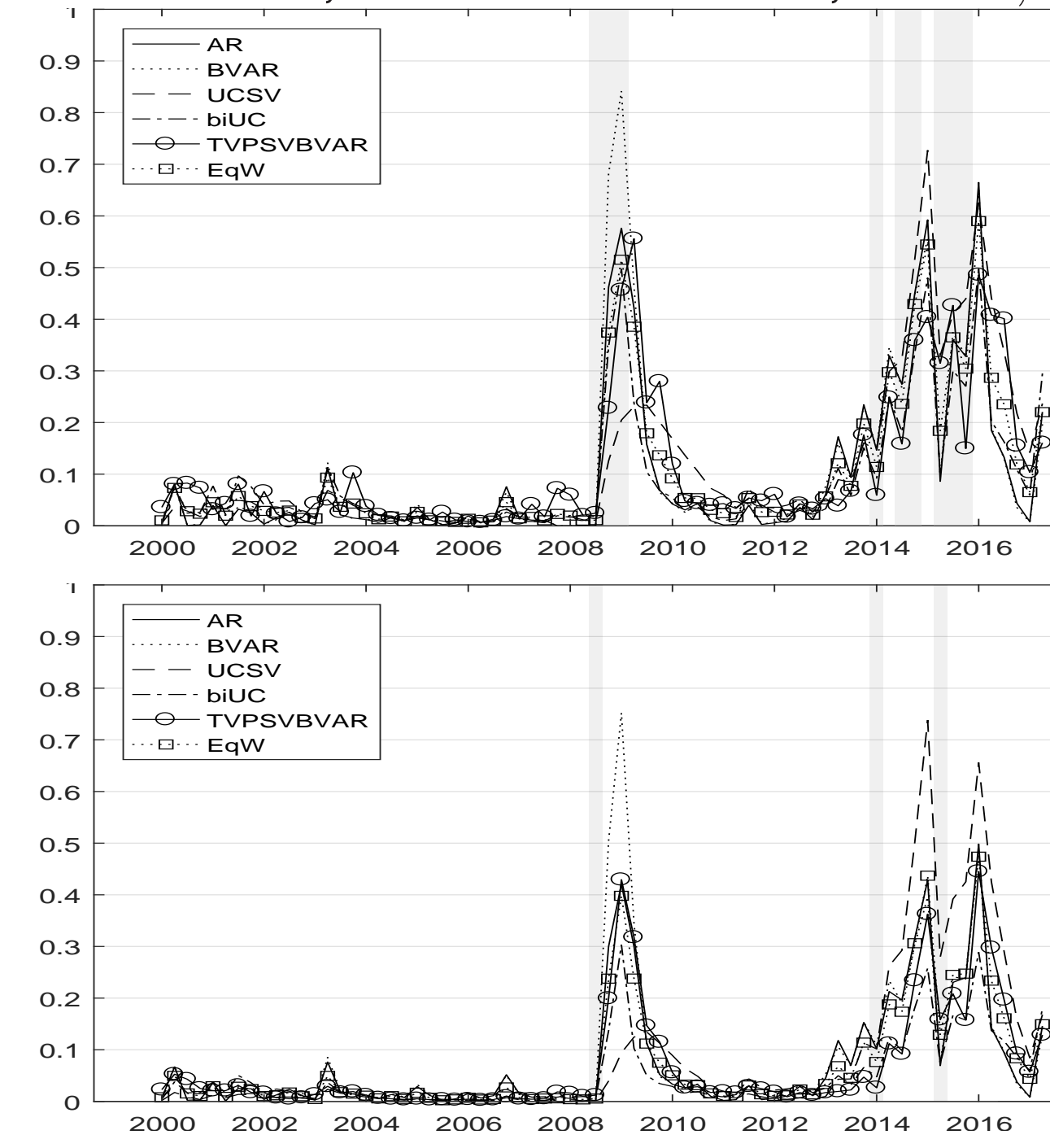
Identifier	Description
AR	Univariate, constant-parameter AR(4)
BVAR	Constant-parameter BVAR(4) in inflation and GDP growth
UCSV	Univariate unobserved component model with time-varying trend and stoch. volatility (SV), Chan (2013) variant of Stock and Watson (2007)
biUC	Bivariate Trend-Cycle Model in inflation and GDP growth with SV, Chan, Koop, Potter (2016)
TVPSVBVAR	Time-varying-parameter VAR(4) in inflation and GDP growth with SV, Primiceri (2005)
EqW	Pooled density from the above models with equal weights

- ▶ Models are estimated recursively with Bayesian MCMC methods
- ▶ Out-of sample path forecast densities are computed

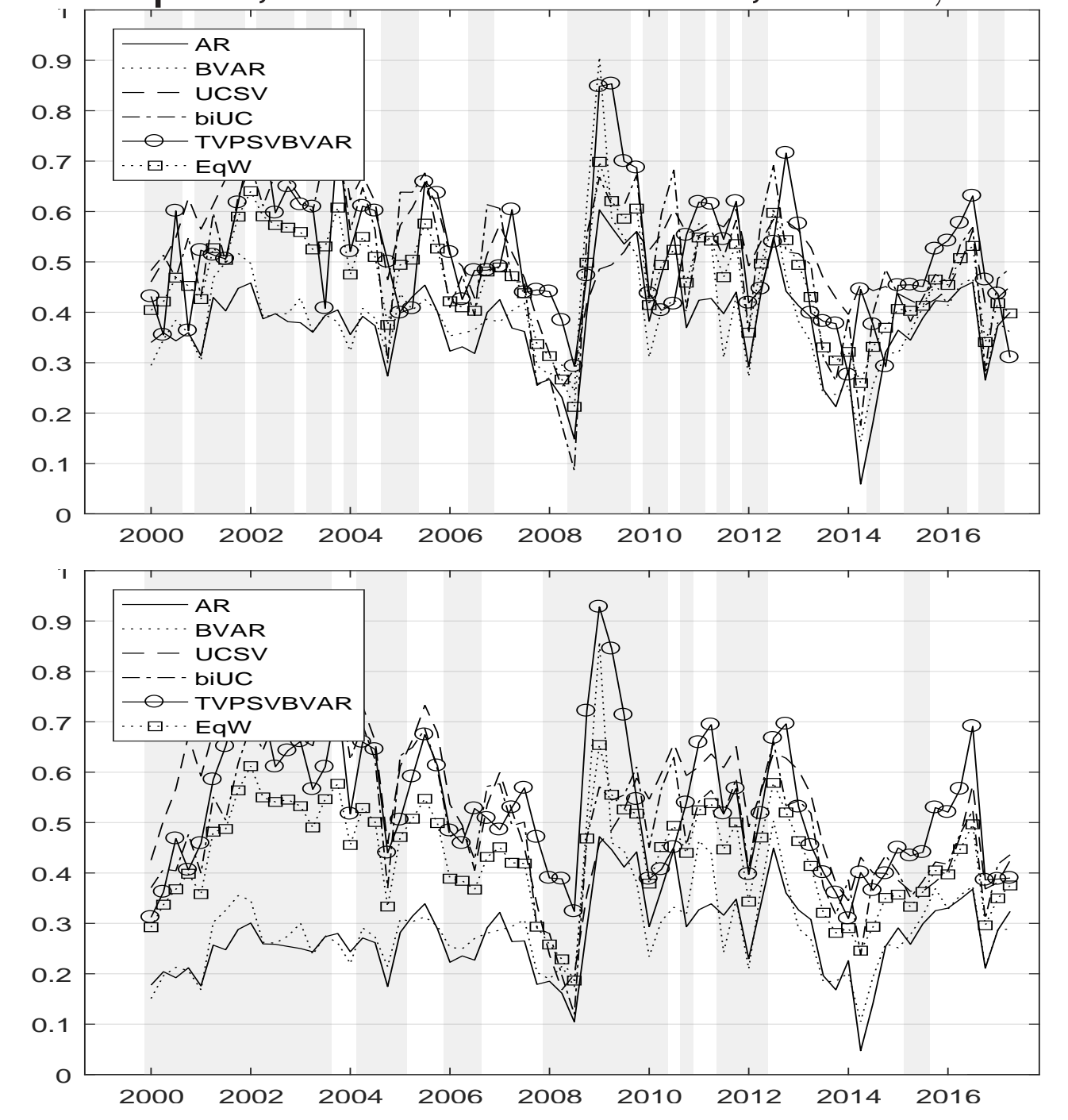
Main results

Estimated deflation probabilities

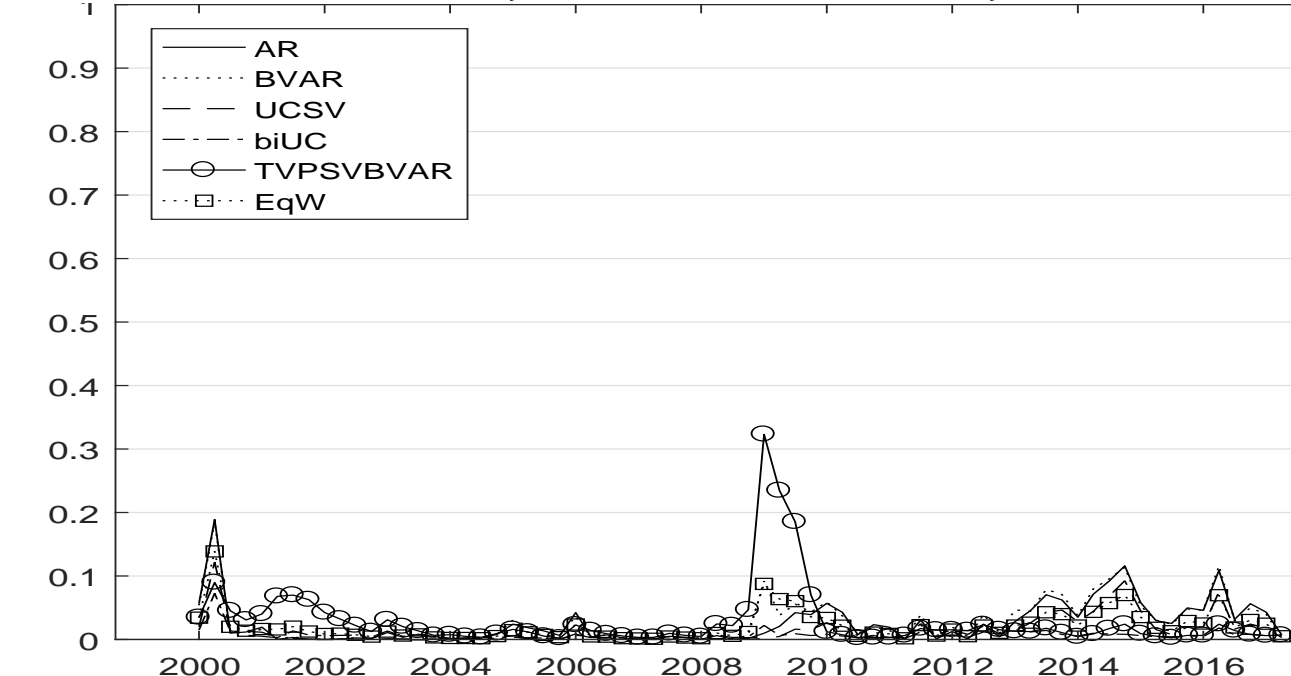
Euro area, headline inflation, $h = 1, 4$



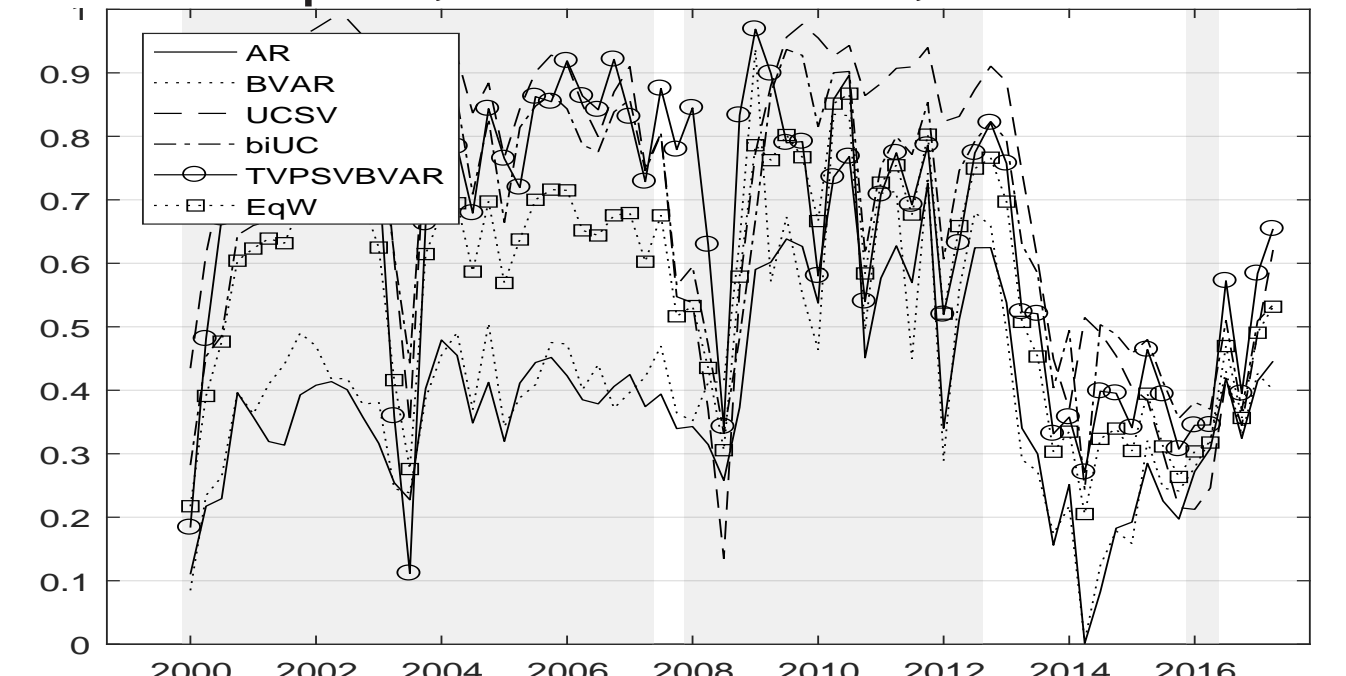
Japan, headline inflation, $h = 1, 4$



Euro area, core inflation, $h = 4$



Japan, core inflation, $h = 4$



- ▶ Note: shaded areas show observed values $y_{\mathcal{D}}^o$, i.e. a bar in period T indicates that $\sum_{i=1}^h \pi_{T+i}^o < 0$

Scores for binary outcomes

Headline inflation

horizon	Euro area		Japan		Euro area		Japan		Euro area		Japan	
	$h = 1$	$h = 4$	$h = 1$	$h = 4$	$h = 1$	$h = 4$	$h = 1$	$h = 4$	$h = 1$	$h = 4$	$h = 1$	$h = 4$
	Log Score				Brier Score				Asymmetric Score			
AR	-0.32	-0.26	-0.76	-0.95	-0.09	-0.06	-0.28	-0.36	-0.21	-0.21	-0.37	-0.49
BVAR	-0.30	-0.27	-0.73	-0.91	-0.08	-0.07	-0.27	-0.34	-0.20	-0.20	-0.36	-0.46
UCSV	-0.31	-0.26	-0.72	-0.70	-0.09	-0.06	-0.26	-0.25	-0.20	-0.20	-0.34	-0.33
biUC	-0.29	-0.21	-0.73	-0.74	-0.09	-0.05	-0.26	-0.27	-0.19	-0.16	-0.35	-0.35
TVPSVBVAR	-0.31	-0.22	-0.72	-0.69	-0.10	-0.05	-0.26	-0.25	-0.20	-0.17	-0.34	-0.32
EqW	-0.28	-0.22	-0.71	-0.74	-0.09	-0.05	-0.26	-0.27	-0.18	-0.17	-0.34	-0.35

Core inflation

horizon	Euro area		Japan		Euro area		Japan		Euro area		Japan	
	$h = 1$	$h = 4$	$h = 1$	$h = 4$	$h = 1$	$h = 4$	$h = 1$	$h = 4$	$h = 1$	$h = 4$	$h = 1$	$h = 4$
	Log Score				Brier Score				Asymmetric Score			
AR	-0.02	-0.03	-0.62	-0.80	0.00	0.00	-0.21	-0.30	-0.02	-0.03	-0.25	-0.34
BVAR	-0.02	-0.03	-0.60	-0.75	0.00	0.00	-0.21	-0.28	-0.02	-0.03	-0.25	-0.31
UCSV	-0.02	-0.01	-0.53	-0.45	0.00	0.00	-0.18	-0.15	-0.02	-0.01	-0.24	-0.19
biUC	-0.01	-0.01	-0.54	-0.48	0.00	0.00	-0.18	-0.16	-0.01	-0.01	-0.23	-0.19
TVPSVBVAR	-0.03	-0.03	-0.58	-0.49	0.00	0.00	-0.20	-0.16	-0.03	-0.03	-0.27	-0.20
EqW	-0.02	-0.02	-0.55	-0.54	0.00	0.00	-0.18	-0.18	-0.02	-0.02	-0.23	-0.21

Summary & conclusions

- ▶ Deflation is a rare event in the Euro area and difficult to forecast, deflation probabilities are lower and less volatile than in Japan
- ▶ In the Euro area, deflation probabilities for core inflation are negligible compared to headline inflation
- ▶ In Japan, deflation probabilities for core inflation are much higher than for headline inflation
- ▶ Time-varying-parameter models and density pool generally outperform constant-parameter models; gains are more pronounced for higher forecast horizons