What Do Professional Forecasters Actually Predict?

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- We examine what professional forecasters predict. Broad movements like trend or business cycle, or also an irregular component that is hard to predict by models and non-experts?
- Method: Use spectral analysis and state space modelling to decompose real-time economic time series into a trend, a business-cycle, and an irregular component. To examine which components are captured by forecasts of the Survey of Professional Forecasters (SPF), we regress the forecasts on the estimated components.
- Key finding: Forecasters can predict almost all variation

Spectral Analysis

We consider the model $y_t = \mu_t + c_t + \varepsilon_t$, where y_t is the observed time series, μ_t represents the trend, c_t the business cycle, and ε_t the irregular component. In other words, we have a slow-moving component, an intermediate component, and a high-frequency component. We isolate these different frequency bands by a low-pass filter derived by Baxter and King (1999).

due to the trend and the business-cycle, but forecasts contain little information about the irregular component.



State Space Model

The trend-cycle model of Harvey (1990): $y_t = \mu_t + c_t + \varepsilon_t, \qquad \varepsilon_t \sim N(0, \sigma_{\varepsilon}^2).$ The smooth trend component is specified as $\mu_{t+1} = \mu_t + \nu_t, \quad \nu_{t+1} = \nu_t + \zeta_t, \quad \zeta_t \sim \mathcal{N}(0, \sigma_{\zeta}^2),$ The business cycle component is represented by $c_{t+1} = \rho c_t \cos \lambda + \rho c_t^* \sin \lambda + \kappa_t, \qquad \kappa_t \sim N(0, \sigma_{\kappa}^2),$ $c_{t+1}^* = -\rho c_t \sin \lambda + \rho c_t^* \cos \lambda + \kappa_t^*, \quad \kappa_t^* \sim \mathcal{N}(0, \sigma_{\kappa}^2),$ where the unknown coefficients ρ , λ , and, σ_{κ}^2 represent the damping factor, the cyclical frequency, and the cycle error term variance, respectively.

Forecast Regression

The professional forecasts are related to the components of the historical time series by

 $f_t = \beta_0 + \beta_1 \hat{\mu}_t + \beta_2 \hat{c}_t + \beta_3 \hat{\varepsilon}_t + v_t,$ where f_t is the professional forecast for time period t. When the Survey of Professional Forecasters perfectly predicts the actual values, we have $\hat{\beta} = (0, 1, 1, 1)$.

	Based On Spectral Analysis						Based On State Space Model				• More results in paper! Including: Sensi-		
	Estimate			Wald		nate	ate		tivitu analucis stato snaco model identifica				
	Intercept	Trend	Cycle	Irreg.	Stat ^a	Intercept	Trend	Cycle	Irreg.	Stat ^a	tivity analysis state space model identifica-		
Perfect fcst	0	1	1	1		0	1	1	1		tion; Comparison to model-based forecast;		
NGDP	-1.178* (0.620)	1.001 (0.001)	0.954 (0.037)	0.249* (0.149)	34.897 {0.000}	-1.242* (0.589)	1.001 (0.001)	1.063 (0.045)	-0.596* (0.301)	42.717 {0.000}	Decomposition of forecasts		
PGDP	-0.197	1.000	0.990	-0.132*	43.620	-0.316	1.001	1.096*	-0.804^{*}	70.909	 In progress: Additional decompositions and 		
	(0.505)	(0.001)	(0.037)	(0.173)	{0.000}	(0.485)	(0.001)	(0.048)	(0.219)	{0.000}	link between them: Robustness to timing of		
CPROF	-1.552	1.001	0.849*	0.102*	50.164	-1.743	1.002	0.956	-0.621*	52.912	releases Sensitivity to Deal Time data		
	(2.548)	(0.005)	(0.041)	(0.155)	{0.000}	(2.538)	(0.004)	(0.034)	(0.283)	{0.000}	releases; Sensitivity to Real-Time data		
UNEMP	1.318 (1.960)	0.997 (0.011)	0.949* (0.016)	0.581* (0.104)	44.220 {0.000}	0.015* (1.919)	1.004 (0.011)	0.980* (0.010)	-0.024* (0.210)	53.677 {0.000}			
INDPROD	-3.491 (1.936)	1.006 (0.003)	0.938* (0.030)	0.441* (0.168)	18.540 {0.000}	-3.708* (1.874)	1.006 (0.003)			cion	S		
HOUSING	2.555* (0.880)	0.919* (0.022)	0.888* (0.038)	0.239* (0.119)	83.413 {0.000}	4.520* (1.331)	0.866* (0.037)	Co	nclu Trend a	nd cy	cle receive weights close to one:		
*: 5% level signif difference from perfect forecast; ()=Std. error									SPF can predict most of the value of significantly differ from zero:				
<i>^a</i> : Tests for <i>joint</i> difference from perfect forecast; {}= <i>p</i> -value								•	• Weights irregular components do may go we have a second				
									SPF does poorly predicting a SPF contains no new information				
									Compared to model-based forecasts, Stra				
									Compo				