



Simple Macroeconomic Forecast Distributions

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Overview



- 1. Setting
- 2. Methods
- 3. Results
- 4. Outlook

An economist's favorite pastime



- various institutions issue forecasts for annual macroeconomic targets
 - most prominent targets: (real) GDP growth and inflation
 - for Germany, sources are (among others) the Bundesbank, the ifo institute, the OECD
 - fixed-event forecasts: target date is fixed, forecast date is not
- forecasts are often disseminated widely
 - extensive media coverage, influence on political discussions
 - relevant for real-world outcomes (public budget planning, collective bargaining)

Can we really be that sure?



- usual practice: issue point forecasts only
 - uncertainty is at best acknowledged. rarely quantified
- forecasts of different horizons are often left uncontextualized
- distributional forecasts supposedly would require extra modeling effort

Prognose der EU

Deutschland vermeidet 2023 Rezession

Stand: 13.02.2023 12:56 Uhr

Nach Einschätzung der EU-Kommission dürfte sich die konjunkturelle Lage in Deutschland besser entwickeln als gedacht. Das Wachstum wird den Prognosen zufolge aber sehr gering ausfallen.

In ihrem aktuellen konjunkturellen Ausblick vertritt die EU-Kommission die Einschätzung, dass Deutschland im Jahr 2023 nicht in eine Rezession fallen wird. Die Brüsseler Behörde erwartet beim Bruttoinlandsprodukt (BIP) ein kleines Plus von 0,2 Prozent. Damit hat sich die Prognose deutlich gebessert, nachdem sie im November noch ein Minus von 0,6 Prozent veranschlagt hatte.

Figure: Source: tagesschau.de

Methods

Results

Contributions of this work



- show that attaching prediction intervals to an existing base of point forecasts can be
 - simple
 - cheap
 - transparent
- provide competitively performing distributional forecasts for
 - GDP growth and inflation
 - all G7 countries
 - current and next year targets

Data Source: IMF World Economic Outlook



- survey by the IMF staff, published bi-annually
 - contains forecasts with up to 6 years horizon and historic truth values
 - publication in April (horizon for current year \approx 8 months) and in October (\approx 2 months)
- publicly available¹ in an accessible format
- targets: real GDP growth, CPI inflation, current account balance
- time range: available since 1990, giving >30 years of forecast-truth pairs
- target locations: forecasts are issued for 196 countries in total

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Methods

Results

¹International Monetary Fund. 2024. World Economic Outlook Database, available at https://www.imf.org/en/Publications/WEO/weo-database/2024/April

Methods

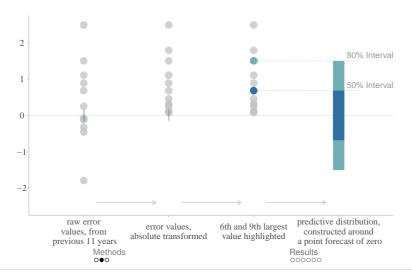


We apply an attractively simple and cheap method. For a given country and target:

- \blacksquare given forecasts $\hat{y}_{t,h}$ and the realized true values $y_{t,...}$
 - for target year t, horizon h
- ... construct sets $\mathcal{E}_{t,h} = \{\hat{e}_{t^*,h}^{abs} | t R \le t^* < t\}$, containing the last R = 11 forecast errors
 - based on absolute errors $\hat{e}_{t,h}^{abs} = |y_t \hat{y}_{t,h}|$
- for $\tau \in \{0.5, 0.8\}$, compute $q_{t,h}^{\tau} = Q(\mathcal{E}_{t,h}, \tau)$
 - with R = 11, quantiles can be taken directly from the order statistics
- and compute the upper and lower endpoints of a central prediction interval as
 - $u_{th}^{\tau} = \hat{y}_{th} + q_{th}^{\tau}$
 - $I_{th}^{\tau} = \hat{\mathbf{y}}_{t,h} \mathbf{q}_{th}^{\tau}$
- assess central interval coverage, score via the interval score

Methods - Visualized





Setting

Benchmarks



We compare with benchmarks:

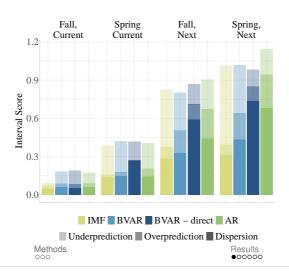
- same methodology, alternative point forecasts
 - autoregressive model ("AR")
 - Primiceri Bayesian vector autoregressive model² ("BVAR")
- directly generated distributional forecasts
 - obtained from the BVAR model ("BVAR direct")
- trained on quarterly data

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²Primiceri, G. 2005. *Time Varying Structural Vector Autoregressions and Monetary Policy*. Review of Economic Studies 72.





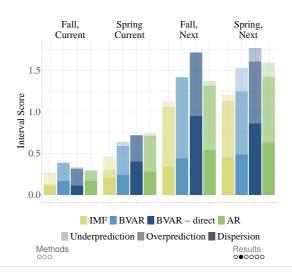


Setting

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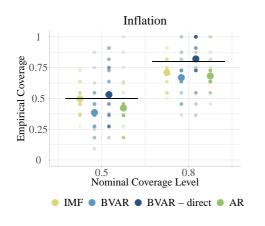




Setting

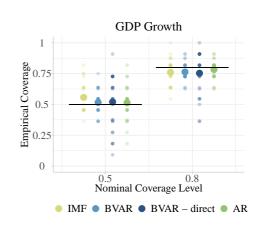
Calibration - on holdout data 2013-2023





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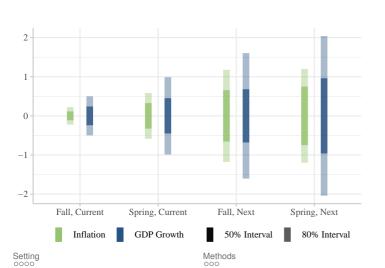


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Increasing Uncertainty





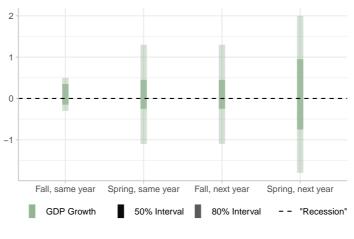
Average length of intervals

	50%	80%						
GDP Growth								
Fall, SY	0.5	1.1						
Spring, SY	1.0	2.1						
Fall, NY	1.7	3.3						
Spring, NY	2.1	4.0						
Inflation								
Fall, SY	0.3	0.5						
Spring, SY	0.6	1.2						
Fall, NY	1.2	2.0						
Spring, NY	1.4	2.2						

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Will Germany avoid recession?





Setting

Methods 000 Results oooo●o

Robustness Checks / Alternative Methods



- error extraction method: absolute vs. directional errors
 - similar scores, worse calibration link
- window method: rolling vs. expanding window
 - slightly improved coverage and scores, at the cost of interpretability
- alternative truth values from the OFCD
 - influences evaluation results at the shortest horizon (link)
- alternative BVAR specification
 - similar scores and calibration link



Summing up



- Attaching distributional forecasts via past forecast errors to an existing base of point forecasts is
 - cheap
 - transparent
 - competitive
- IMF forecasts are valuable source for distributional forecasts
- Uncertainty around point forecasts is often substantial, making its communication necessary
- Potential Extensions / Outlook
 - scale to more countries
 - make forecasts easily and publicly accessible via shiny app
 - preregistration of real-time forecasts

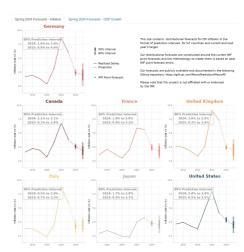


Making it public

Karlsruhe Institute of Technology

Shiny App: https://probability-forecasting. shinyapps.io/macropi/

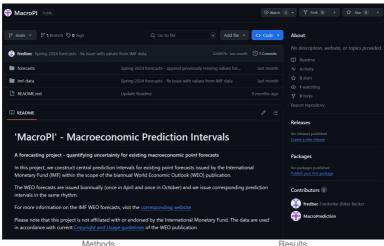
GitHub repo with our forecasts: https://github.com/ MacroPrediction/MacroPI



Setting Methods Results







Setting

Methods



Making it public - Timestamped Forecast Publication II





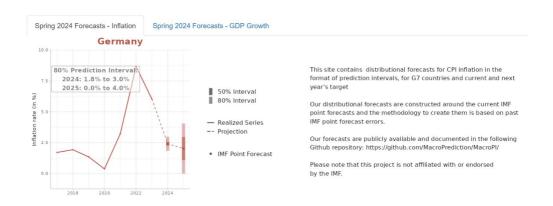
Setting

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Making it public - Visualization





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Methods

Results

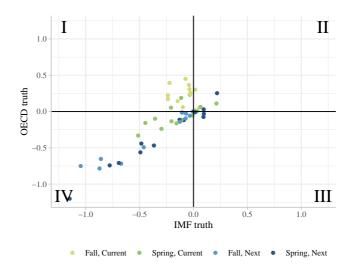




		Interval Score		50% Cvg.		80% Cvg.	
	Horizon	absolute	directional	absolute	directional	absolute	directional
GDP Growth	Fall, Current	0.23	0.24	0.49	0.43	0.76	0.65
	Spring, Current	0.41	0.41	0.56	0.54	0.76	0.67
	Fall, Next	0.91	0.88	0.49	0.42	0.73	0.70
	Spring, Next	1.14	1.15	0.50	0.40	0.64	0.55
Inflation	Fall, Current	0.12	0.12	0.52	0.44	0.76	0.64
	Spring, Current	0.26	0.25	0.43	0.39	0.75	0.65
	Fall, Next	0.47	0.50	0.40	0.31	0.67	0.54
	Spring, Next	0.52	0.55	0.42	0.38	0.67	0.54

Alternative truth value - Scores





Alternative BVAR specification - Scores



