

## Whose Inflation Expectations Forecast Best?

### Alternatives Based on Survey and Financial Data

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## **Motivation for this Paper**

- Increasing attention on using microdata to answer macro questions
  - Main focus of the literature: (1) Expectations formation using micro data, and (2) Modelling the behaviour (typically models incorporating heterogeneity)
  - Our Question: What does this mean for models that require a single measure of expected inflation, such the Quarterly Projection Model of the SARB?
- Principles of inflation forecasting (Faust and Wright, 2013)
  - 1. Subjective forecasts do best;
  - 2. Good forecasts must account for a slowly varying local mean;
  - 3. Good forecasts begin with high quality nowcasts; and
  - 4. Heavy shrinkage in the use of information improves inflation forecasts  $\langle I \rangle$

## **Our choices**

- Aggregation across groups:
  - 'BER aggregate' arithmetic mean
- Aggregation at the group level:
  - BER Financial analysts (FA)
  - BER Businesses (BUS)
  - BER Trade Unions (TU)
- Asset price data (Bloomberg)
- Some other form of aggregation
  - Factor models
- Q: Whose Inflation Forecast performs best?



#### Selected Aggregated Forecasts: 1 & 5 YR Horizons



SOUTH AFRICAN RESERVE BAN

## Alternative (possibly better) forms of aggregation

Demeaned inflation  
Forecast – could also  
be standardized  
Expected inflation (forecast)  
@ time t for horizon h  
Average inflation forecast  
NOTE: aggregate used BUT  
other benchmarks are possible  

$$\tilde{\pi}_{t,h}^e = \pi_{t,h}^e - \bar{\pi}_{t,h}^e = \sum_{i=1}^m \lambda_{i,h} f_{i,h} + \varepsilon_{i,h}$$
  
Factors: limited to 1 – often  
supported by the data ...but  
extensions are possible



## **Testing Forecasts: Which metrics?**

#### RMSE

$$RMSE_{t,h} = \sqrt{\sum_{j=1}^{k} (\pi_{t,h} - \pi_{t,h}^{e})^2}$$

Problems: (1) Scale-dependent. Arguably heterogeneity of forecasts across BUS, FIN, and LAB imply different distribution shapes not controlled for in RMSE; (2) sensitive to 'outliers'; (3) If FE = 0 then RMSE is undefined (or infinite) – a real problem only if this occurs frequently; (4) can be artificially skewed

• MASE

$$q_{t,h} = \frac{FE_{t,h}}{\frac{1}{|k-1\sum_{j=2}^{k}|\Delta\pi_{t,h}|}}$$

 $MASE_{t,h} = mean(|q_{t,h}|)$ 

$$FE_{t,h} = \pi_{t,h} - \pi^e_{t,h}$$

Hyndman & Koehler (2006) propose a scale invariant measure that is also easier to interpret. For example, an MASE < 1 means that it outperforms than a naïve one-step ahead forecast. If MASE > 1 then a naïve forecast is superior.

**Reminder: Superiority Relative to naïve forecast** 

## Reminder: Superiority Relative to naïve forecast (if MASE = 1 the two forecasts are equally accurate) **MASE Results**

Туре	СРІ ТО	CPI T1	CPI T2	CPI 5a
BER	0.515	0.402 <sup>3</sup>	0.333 <sup>3</sup>	0.451
BUS	0.618	0.633	0.583	0.990
FIN	0.324 <sup>2</sup>	0.026 <sup>2</sup>	0.076 <sup>2</sup>	0.743
LAB	0.494 <sup>3</sup>	0.512	0.491	0.339 <sup>3</sup>
Factor Model	0.003/0.088* 1	0.003/0.070* 1	0.016 <sup>1</sup> /0.057*	0.423/0.290* 2
Bloomberg	NA	NA	NA	0.315 <sup>1</sup>

#### "Calm" sub-sample

Volatile sub-sample

Туре	СРІ ТО	CPI T1	CPI T2	CPI 5a	Туре	СРІТО	CPIT1	CPIT2	CPI5a
BER	0.847	0.796 <sup>3</sup>	0.763 <sup>3</sup>	0.287 <sup>2</sup>	BER	0.411	<b>0.248</b> <sup>3</sup>	0.161 <sup>2</sup>	0.732
BUS	0.847	0.918	0.916	0.984	BUS	0.525	0.505	0.437	1
FIN	0.533 <sup>2</sup>	0.432 <sup>2</sup>	0.314 <sup>2</sup>	0.872	FIN	0.258 <sup>2</sup>	0.059 <sup>2</sup>	0.176 <sup>3</sup>	0.330 <sup>3</sup>
LAB	0.760 <sup>3</sup>	0.832	0.826	0.364 <sup>3</sup>	LAB	0.394 <sup>3</sup>	0.384	0.350	0.299 <sup>2</sup>
Factor Model	0.072 1	0.164 <sup>1</sup>	<b>0.246</b> <sup>1</sup>	0.154 <sup>1</sup>	Factor Model	0.026 1	0.055 <sup>1</sup>	0.060 <sup>1</sup>	0.749
Bloomberg	NA	NA	NA	0.491	Bloomberg	NA	NA	NA	0.100 <sup>1</sup>

## Summary

- Simple arithmetic averaging of BUS, FIN, and LAB inflation expectations (i.e., the BER forecast) rarely yields the best forecast
- A simple factor model seems to consistently perform quite (confirming principles 1 and 4 from Faust and Wright)
- Which forecasts do best in volatile periods ('non-normal times, when models too are performing poorly)?
  - We are less confident about these results
  - MASE suggests factor models, BER average and FAs do best in both calm and volatile periods, with TUs also offering good insight for T0 and T5
  - RMSE suggests financial analysts do best in 'normal times' and labour in volatile times (business not too far behind)



# THANK YOU



## **Selected Other Results**

Added to Appendix



## **RMSE** Results

RMSE ResultsFull Sample							
Туре	СРІТО	CPIT1	CPIT2	CPI5a			
BER	1.766 <sup>3</sup>	2.033 <sup>2</sup>	<b>2.214</b> <sup>3</sup>	<b>0.411</b> <sup>1</sup>			
BUS	2.085	2.210	2.362	0.688			
FIN	1.449 <sup>2</sup>	2.162	2.426	0.414 <sup>2</sup>			
LAB	1.955	2.063 <sup>3</sup>	2.184 <sup>2</sup>	0.541 <sup>3</sup>			
Factor Model	0.864 <sup>1</sup>	1.085 <sup>1</sup>	1.304 <sup>1</sup>	1.060			
Bloomberg	NA	NA	NA	0.917			
			"Volatile"	Sub-Sample			

#### "Calm" Sub-Sample

Туре	СРІТО	CPIT1	CPIT2	CPI5a	Туре	СРІТО	CPIT1	CPIT2	CPI5a
BER	0.959 <sup>3</sup>	1.117 <sup>3</sup>	1.233 <sup>3</sup>	0.827 <sup>2</sup>	BER	2.196 <sup>3</sup>	2.253 <sup>2</sup>	<b>2.741</b> <sup>3</sup>	1.389
BUS	1.291	1.447	1.562	1.112	BUS	2.526	2.646	2.821	1.397
FIN	0.555 <sup>1</sup>	0.854 <sup>1</sup>	0.967 <sup>1</sup>	0.720 <sup>1</sup>	FIN	1.860 <sup>2</sup>	2.769	3.106	1.614
LAB	1.162	1.269	1.355	0.868	LAB	2.387	2.503 <sup>3</sup>	2.645 <sup>2</sup>	1.246 <sup>2</sup>
Factor Model	0.833 <sup>2</sup>	0.966 <sup>2</sup>	1.134 <sup>2</sup>	0.851 <sup>3</sup>	Factor Model	<b>0.934</b> <sup>1</sup>	1.317 <sup>1</sup>	1.626 <sup>1</sup>	1.378 <sup>3</sup>
Bloomberg	NA	NA	NA	0.893	Bloomberg	NA	NA	NA	0.890 <sup>1</sup>

## **BER data**

Va	ariable	Trimmed CPI Inflation Forecast (excludes			Not Trimmed Inflation Forecast				
		+/-20%	Mean (S.D.)	luonj	Mean (S.D)				
Survey S	Source	Business	Fin. Analysts	Trade Union	Business	Fin. Analysts	Trade Union		
Current	Year:	6.18	5.60	5.96	6.19	5.96	5.96		
ТО		(1.53)	(1.76)	(1.56)	(1.53)	(1.56)	(1.56)		
One Yea	r Ahead: T1	6.20	5.37	6.02	6.29	5.37	6.05		
		(1.21)	(0.82)	(1.44)	(1.25)	(0.82)	(1.42)		
Two Yea	rs Ahead:	6.18	5.19	6.04	6.31	5.18	6.08		
T2		(1.06)	(0.47)	(1.33)	(1.09)	(0.48)	(1.32)		
Average	Over 5	5.88	5.18	5.49	5.95	5.15	5.54		
Years Ar	nead: 5Y	(0.55)	(0.48)	(0.71)	(0.55)	(0.44)	(0.72)		
Observe	ed inflation	5.32							
		(2.56)							
SA	AMPLE: 2000Q2	2-2023Q4							
<u>O</u> 1	<u>ther Summary</u>	<u>statistics</u>							

BUS: Up to 30000+ Observations By firm SIZE By Position of Respondent By SIC FIN: 1600+ Observations By firm SIZE By Position of Respondent By Sector

LAB: 1400+ Observations By firm SIZE

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Probability	BUST0	BUST1	BUDT2	BUS5A
BUS T1	0.94	1.00		
	18.67			
	0.00			
BUS T2	0.90	0.97	1.00	
	13.96	28.08		
12	0.00	0.00		
BUS 5A	0.03	0.02	0.05	1.00
	0.20	0.14	0.35	
	0.84	0.89	0.73	
FIN TO	0.33	0.26	0.22	0.10
	2.44	1.83	1.58	0.68
	0.02	0.07	0.12	0.50
	v 100 - 100-100		2.50 mm.a.	
FIN T1	0.48	0.46	0.49	0.21
6	3.81	3.57	3.90	1.48
	0.00	0.00	0.00	0.15

Simple correlation between pairs of forecasts



## **Encompassing tests**

• Evidence of 'superiority' in combining forecast data

Encompassing Tests: Full and Sub-samples 2009Q2-2023Q4

Forecast horizon	BER vs BUS		BER vs FIN		BER vs TU		BER vs SARB	
	Full	Sub	Full	Sub	Full	Sub	Full	Sub
T0	.00	.00	.00	.00	.00	.00	.00	.00
T1	.00	.00	.00	.01	.00	.00	.00	.00
T2	.00	.72	.00	.00	.03	.00	.00	.00
5a	.69	.69	.00	.00	.00	.00	.00	.00



#### Factor loadings

#### (b) Extended Model– Full Sample

Forecast	Business	Financial	Trade unions	SARB
horizon		Analysts		
	Full	Full	Full	Full
T0	0.965	0.893	0.966	0.898
T1	0.934	0.683	0.927	0.595
T2	0.952	0.785	0.924	0.485
5a	0.959	0.928	0.953	0.856



#### BER Survey: No. Observations By Year/Quarter



#### FE vs FH Forecasts: The Case of SARB Forecasts



#### SARB Credibility Over the Years: FIN Perspective





#### Outliers: The One Year Ahead Case from BUS

The Case of One-Year Ahead Forecasts

