

Has models' forecasting performance for US output growth and inflation changed over time, and when?

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Bank of Chile, Forecasting Workshop

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 - Existing tests only focus on average performance
- Main findings:
 - Forecasting inflation is harder than forecasting output
 - Models' forecasting ability for inflation worsened around the time of the Great Moderation, whereas that of output worsened in mid-1970s
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- Simple example
- General results
- Are the parameters in the Taylor rule identified?
- Conclusion

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Forecast comparisons over time: Fluctuation test

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- Giacomini and Rossi (2008) show that

$$F_{t,m}^{OOS} = m^{1/2} \sum_{j=t-m/2}^{t+m/2-1} rMSFE_t \implies [\mathcal{B}(\tau + \mu/2) - \mathcal{B}(\tau - \mu/2)] / \sqrt{\mu},$$

where $t = [\tau P]$, $m = [\mu P]$, $\mathcal{B}(\cdot)$ is a std univariate BM

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- They find critical values s.t.

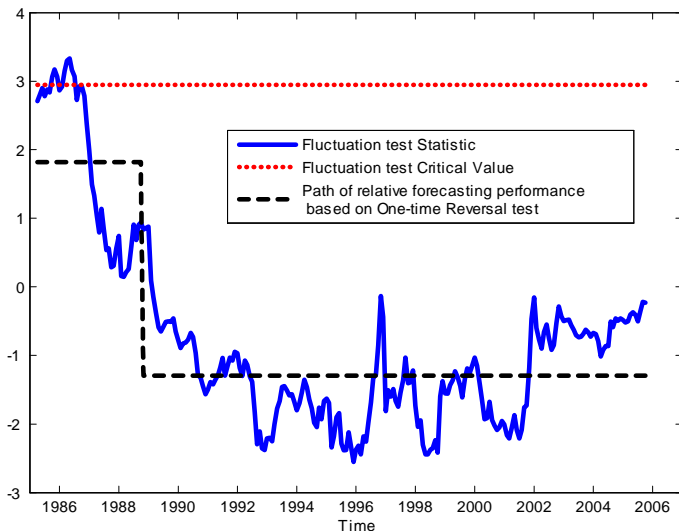
$$\Pr \left\{ \sup_{\tau} \left| F_{t,m}^{OOS} \right| > k_{\alpha} \right\} = \alpha$$

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Forecast comparisons over time: example



Empirical results: One-time reversal test I

Table 4. Forecasting Inflation: Tests of Average Equal Predictive Ability

Variable	rMSFE	p-value	Variable	rMSFE	p-value
rovnght level	0.10	0.92	ip $\Delta \ln$	2.74	0.01
rtbill level	0.15	0.88	ip gap	1.39	0.16
rbnds level	-0.13	0.90	emp $\Delta \ln$	-0.95	0.34
rbndm level	-0.75	0.45	emp gap	1.17	0.24
rbndl level	-0.89	0.37	unemp level	1.06	0.29
rovnght Δ	0.19	0.85	unemp $\Delta \ln$	-0.60	0.55
rtbill Δ	0.17	0.87	unemp gap	0.66	0.51
rbnds Δ	0.13	0.90	ppi $\Delta \ln$	0.56	0.58
rbndm Δ	0.53	0.59	ppi $\Delta^2 \ln$	0.75	0.45
rbndl Δ	0.69	0.49	earn $\Delta \ln$	-1.47	0.14
rrovnght level	0.10	0.92	earn $\Delta^2 \ln$	2.72	0.01
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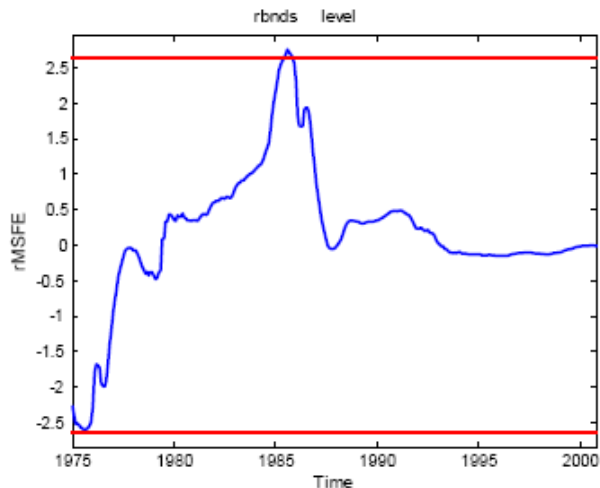
Empirical results: One-time reversal test II

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rrtbill	Δ	-0.71	0.48	m0	$\Delta^2 \ln$	2.13	0.03
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rrbndm	Δ	-1.20	0.23	m1	$\Delta^2 \ln$	3.09	0.00
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rspread	level	0.40	0.69	m2	$\Delta^2 \ln$	2.79	0.01
exrate	$\Delta \ln$	0.65	0.52	m3	$\Delta \ln$	-1.94	0.05
stockp	$\Delta \ln$	2.03	0.04	m3	$\Delta^2 \ln$	4.03	0.00
rstockp	$\Delta \ln$	2.03	0.04	rm0	$\Delta \ln$	-0.26	0.79
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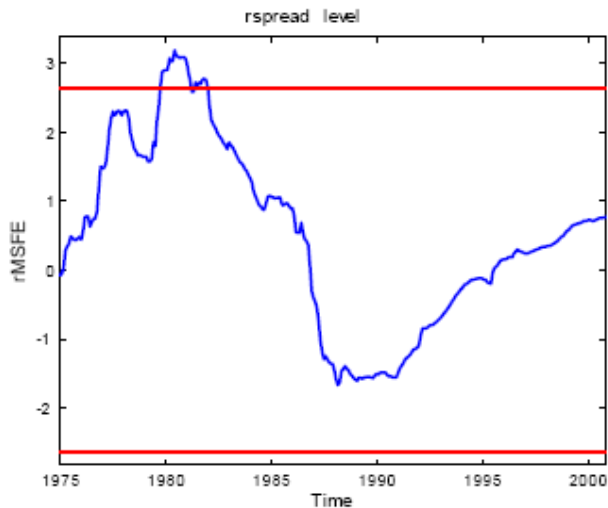
Empirical results: Fluctuation test

One-year Treasury Bond rate



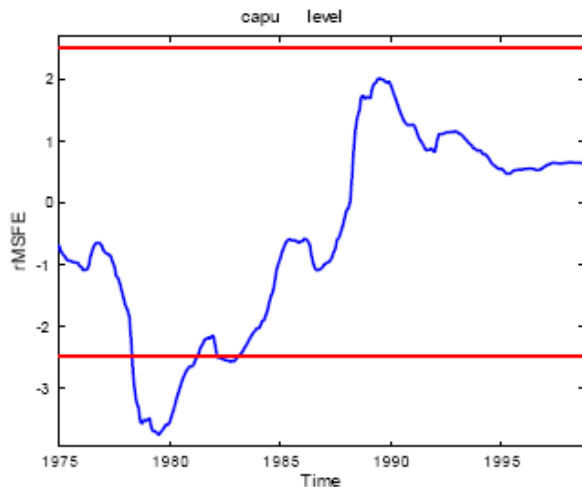
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Interest rate spread



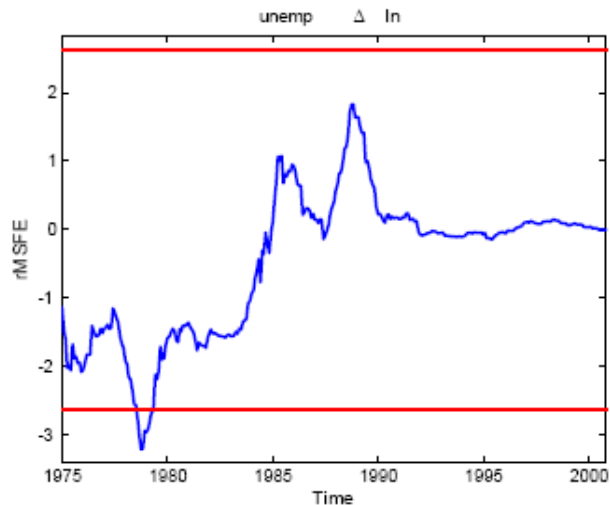
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Capacity Utilization

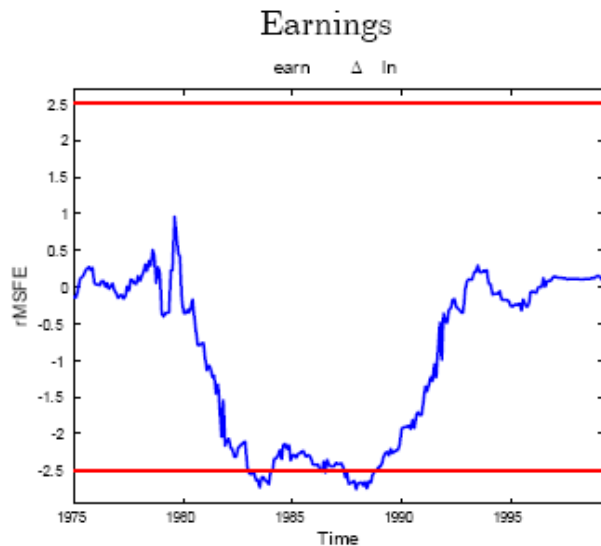


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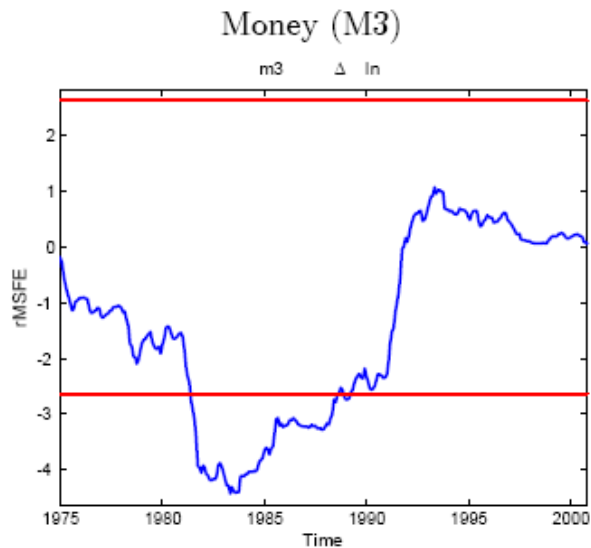
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Empirical results: Existing tests (Full sample I)

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- There is more evidence in favor of predictive ability for output than for inflation
- The time of the reversal in the predictive ability is around 1983-1984 for inflation (and mid-1970 for output).