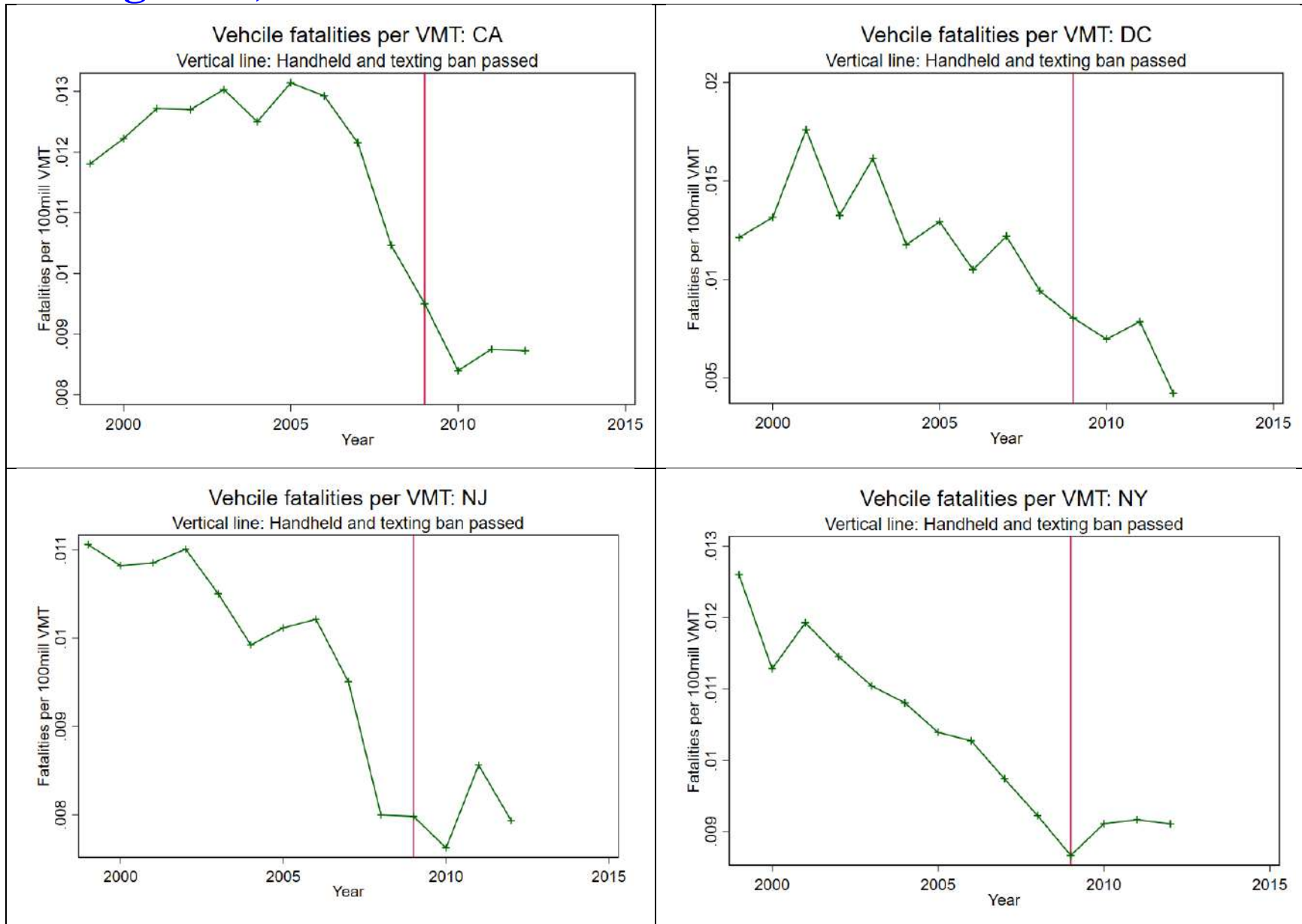


## Forecasting II

### Outline (Forecasting I and II)

1. Intro: Economic time series data & econometric issues
2. Lags, differences, autocorrelation, and stationarity
3. Autoregressive (AR) models
4. Autoregressive distributed lag (ADL) models & Granger Causality tests
5. Lag length selection using information criteria
6. Stationarity, nonstationarity, breaks, and model stability
7. Forecast intervals and fan charts

# PS 8 – selected state time series plots (states that passed HH & texting bans)



# PS 8 – selected regressions

```
. xtreg lvfrtot hhall textall hh_text $cvars1 i.year, fe vce(cluster state_id)
```

```
Fixed-effects (within) regression
Group variable: state_id
Number of obs      =      686
Number of groups   =      49
```

```
R-sq:
  within  = 0.7546
  between = 0.1243
  overall  = 0.1479
Obs per group:
  min     =      14
  avg     =     14.0
  max     =      14
```

```
corr(u_i, Xb) = -0.8530
F(21,48)      =      59.14
Prob > F      =      0.0000
```

(Std. Err. adjusted for 49 clusters in state\_id)

lvfrtot	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]
hhall	-.0428101	.0436995	-0.98	0.332	-.1306738 .0450535
textall	.0070935	.0251751	0.28	0.779	-.0435244 .0577113
hh_text	-.0235807	.0472056	-0.50	0.620	-.118494 .0713326
lvmttotpc	.5220621	.3972668	1.31	0.195	-.2766962 1.32082
lvmturbpc	.0002042	.1103617	0.00	0.999	-.2216928 .2221012
unrate	-.0234727	.0085182	-2.76	0.008	-.0405997 -.0063457
lpop	-.6976288	.2114066	-3.30	0.002	-1.12269 -.2725673
precipm	-.0494832	.0245379	-2.02	0.049	-.09882 -.0001463

year							
2000		-.0178226	.015817	-1.13	0.265	-.0496248	.0139797
2001		.0061997	.0158434	0.39	0.697	-.0256556	.0380549
2002		.0358903	.0176816	2.03	0.048	.0003391	.0714415
2003		.0444039	.0202771	2.19	0.033	.0036341	.0851737
2004		.0201023	.0263416	0.76	0.449	-.0328611	.0730657
2005		.0183229	.0236518	0.77	0.442	-.0292323	.0658782
2006		-.0096639	.0294972	-0.33	0.745	-.0689721	.0496442
2007		-.0646618	.0248846	-2.60	0.012	-.1146957	-.014628
2008		-.1158773	.0301706	-3.84	0.000	-.1765394	-.0552153
2009		-.1095308	.0527152	-2.08	0.043	-.2155218	-.0035399
2010		-.1386848	.0543131	-2.55	0.014	-.2478886	-.029481
2011		-.1775412	.0552717	-3.21	0.002	-.2886724	-.06641
2012		-.1773381	.0572856	-3.10	0.003	-.2925185	-.0621576
_cons		13.50535	2.577607	5.24	0.000	8.322719	18.68797
-----							
sigma_u		.72681311					
sigma_e		.09024689					
rho		.98481642	(fraction of variance due to u_i)				
-----							

. test hhall textall

- ( 1) hhall = 0
- ( 2) textall = 0

F( 2, 48) = 0.48  
 Prob > F = 0.6217

```
. test hhall textall hh_text
```

```
( 1) hhall = 0  
( 2) textall = 0  
( 3) hh_text = 0
```

```
F( 3, 48) = 0.71  
Prob > F = 0.5527
```

```
. lincom 1*hhall + 1*textall + 1*hh_text
```

```
( 1) hhall + textall + hh_text = 0
```

lvfrrtot	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
(1)	-.0592973	.0414473	-1.43	0.159	-.1426328	.0240381

```
. xtreg lvfrtot L(0/1).hhall L(0/1).textall L(0/1).hh_text L(0/1).alljr
$cvvars1 i.year,
> fe vce(cluster state_id)
```

```
Fixed-effects (within) regression
Group variable: state_id
```

```
Number of obs      =          637
Number of groups   =           49
```

```
R-sq:
  within  = 0.7602
  between = 0.1346
  overall  = 0.1524
```

```
Obs per group:
  min =          13
  avg  =         13.0
  max  =          13
```

```
corr(u_i, Xb) = -0.8870
F(25,48)      =          74.57
Prob > F      =          0.0000
```

(Std. Err. adjusted for 49 clusters in state\_id)

	lvfrtot	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
	hhall						
	--.	-.0632169	.0315942	-2.00	0.051	-.1267413	.0003075
	L1.	.012926	.056411	0.23	0.820	-.100496	.126348
	textall						
	--.	.0353531	.0246586	1.43	0.158	-.0142262	.0849325
	L1.	-.0491441	.0347472	-1.41	0.164	-.119008	.0207199
	hh_text						
	--.	.0133872	.0438997	0.30	0.762	-.074879	.1016534

L1.	-.0405331	.1024634	-0.40	0.694	-.2465496	.1654834
alljr						
--.	.013376	.0217748	0.61	0.542	-.0304052	.0571573
L1.	-.0170445	.0215304	-0.79	0.432	-.0603342	.0262452
lvmttotpc	.6677962	.414748	1.61	0.114	-.1661107	1.501703
lvmturbpc	-.0352352	.1183281	-0.30	0.767	-.2731497	.2026793
unrate	-.0242243	.0089776	-2.70	0.010	-.0422749	-.0061737
lpop	-.7895803	.2129782	-3.71	0.001	-1.217802	-.361359
precipm	-.0453472	.0265883	-1.71	0.095	-.0988066	.0081123
year						
2001	.0235622	.0131591	1.79	0.080	-.0028961	.0500204
2002	.0540475	.0184452	2.93	0.005	.016961	.091134
2003	.0634384	.020561	3.09	0.003	.0220978	.104779
2004	.0389619	.030236	1.29	0.204	-.0218316	.0997555
2005	.0398468	.0276706	1.44	0.156	-.0157886	.0954822
2006	.0127978	.0312404	0.41	0.684	-.0500152	.0756107
2007	-.0396245	.0246974	-1.60	0.115	-.089282	.0100329
2008	-.0889544	.0362502	-2.45	0.018	-.1618402	-.0160686
2009	-.0795439	.0635476	-1.25	0.217	-.207315	.0482271
2010	-.1082896	.0648636	-1.67	0.102	-.2387066	.0221274
2011	-.1326788	.0666262	-1.99	0.052	-.2666398	.0012822
2012	-.1279961	.0679146	-1.88	0.066	-.2645475	.0085552
_cons	15.3545	2.8091	5.47	0.000	9.706422	21.00257
sigma_u	.82072289					
sigma_e	.08935331					

rho | .98828585 (fraction of variance due to u\_i)

---

. test hhall textall

( 1) hhall = 0

( 2) textall = 0

F( 2, 48) = 2.46  
Prob > F = 0.0958

. test hhall textall hh\_text

( 1) hhall = 0

( 2) textall = 0

( 3) hh\_text = 0

F( 3, 48) = 1.68  
Prob > F = 0.1848



```
. xtreg lvfrtot L.hhall L.textall L.hh_text L.alljr $cvars1 i.year, fe
vce(cluster state
> _id)
```

```
Fixed-effects (within) regression
Group variable: state_id
```

```
Number of obs      =      637
Number of groups   =       49
```

```
R-sq:
  within  = 0.7588
  between = 0.1334
  overall  = 0.1496
```

```
Obs per group:
      min =      13
      avg =     13.0
      max =      13
```

```
corr(u_i, Xb) = -0.8959
F(21,48)      =      59.69
Prob > F      =      0.0000
```

(Std. Err. adjusted for 49 clusters in state\_id)

	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
lvfrtot						
hhall						
L1.	-.0362862	.0532782	-0.68	0.499	-.1434091	.0708368
textall						
L1.	-.0220725	.0280311	-0.79	0.435	-.0784328	.0342878
hh_text						
L1.	-.0372138	.0736555	-0.51	0.616	-.1853081	.1108804
alljr						

L1.	-.0034252	.0181765	-0.19	0.851	-.0399715	.0331211
lvmttotpc	.6841629	.4261656	1.61	0.115	-.1727005	1.541026
lvmturbpc	-.0429463	.12033	-0.36	0.723	-.284886	.1989934
unrate	-.0250812	.0092374	-2.72	0.009	-.0436542	-.0065083
lpop	-.8232446	.2077243	-3.96	0.000	-1.240902	-.405587
precipm	-.0466779	.0253272	-1.84	0.072	-.0976016	.0042458
year						
2001	.0241009	.0132384	1.82	0.075	-.0025167	.0507186
2002	.0545414	.0186616	2.92	0.005	.0170197	.0920632
2003	.0656634	.0209171	3.14	0.003	.0236068	.1077201
2004	.0408722	.0306598	1.33	0.189	-.0207735	.1025179
2005	.0415767	.0281596	1.48	0.146	-.0150419	.0981953
2006	.0150995	.0316307	0.48	0.635	-.0484983	.0786974
2007	-.0368568	.0248307	-1.48	0.144	-.0867822	.0130687
2008	-.0828747	.0364328	-2.27	0.027	-.1561278	-.0096217
2009	-.0682057	.0649987	-1.05	0.299	-.1988942	.0624829
2010	-.0876013	.0666542	-1.31	0.195	-.2216186	.0464159
2011	-.1157507	.0687188	-1.68	0.099	-.2539191	.0224177
2012	-.1132939	.0693709	-1.63	0.109	-.2527733	.0261856
_cons	15.90094	2.743076	5.80	0.000	10.38562	21.41627
sigma_u	.8546306					
sigma_e	.08931341					
rho	.98919663	(fraction of variance due to u_i)				

. test L.hhall L.textall

( 1) L.hhall = 0  
( 2) L.textall = 0

F( 2, 48) = 0.64  
Prob > F = 0.5341

. test L.hhall L.textall L.hh\_text

( 1) L.hhall = 0  
( 2) L.textall = 0  
( 3) L.hh\_text = 0

F( 3, 48) = 1.27  
Prob > F = 0.2940

. lincom L.hhall + L.textall + L.hh\_text + L.alljr

( 1) L.hhall + L.textall + L.hh\_text + L.alljr = 0

lvfirtot	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
(1)	-.0989977	.0601748	-1.65	0.106	-.2199872 .0219917

```
. xtreg lvfrbelow08 L.hhall L.textall L.hh_text $cvars1 i.year, fe vce(cluster state_id)
```

```
Fixed-effects (within) regression
Group variable: state_id
```

```
Number of obs      =          637
Number of groups   =           49
```

```
R-sq:
  within  = 0.7321
  between = 0.0962
  overall  = 0.1076
```

```
Obs per group:
      min =          13
      avg =         13.0
      max =          13
```

```
corr(u_i, Xb) = -0.9229
F(20,48)      =          99.52
Prob > F      =          0.0000
```

(Std. Err. adjusted for 49 clusters in state\_id)

	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
lvfrbelow08						
hhall						
L1.	-.0662835	.0594883	-1.11	0.271	-.1858926	.0533257
textall						
L1.	-.0390293	.0276785	-1.41	0.165	-.0946808	.0166221
hh_text						
L1.	-.006826	.066891	-0.10	0.919	-.1413194	.1276675
lvmttotpc	.6569602	.3999299	1.64	0.107	-.1471526	1.461073
lvmturbpc	-.0672488	.112906	-0.60	0.554	-.2942616	.1597641

unrate		-.0216938	.0086544	-2.51	0.016	-.0390947	-.0042929
lpop		-.9659531	.2175838	-4.44	0.000	-1.403435	-.5284716
precipm		-.0309324	.0242571	-1.28	0.208	-.0797046	.0178398
year							
2001		.0202515	.0126538	1.60	0.116	-.0051908	.0456937
2002		.0551645	.0204093	2.70	0.009	.0141288	.0962002
2003		.0726429	.0197473	3.68	0.001	.0329383	.1123475
2004		.064471	.0305532	2.11	0.040	.0030396	.1259024
2005		.058691	.0263509	2.23	0.031	.005709	.111673
2006		.0351487	.0307565	1.14	0.259	-.0266914	.0969888
2007		-.0228493	.0247515	-0.92	0.361	-.0726156	.0269169
2008		-.0585638	.0351558	-1.67	0.102	-.1292492	.0121217
2009		-.0673004	.0631247	-1.07	0.292	-.1942212	.0596204
2010		-.0673523	.0601177	-1.12	0.268	-.1882271	.0535225
2011		-.0844999	.0627228	-1.35	0.184	-.2106126	.0416127
2012		-.0753889	.064984	-1.16	0.252	-.2060479	.0552702
_cons		17.38018	2.998853	5.80	0.000	11.35058	23.40978
-----							
sigma_u		1.0105546					
sigma_e		.0922151					
rho		.99174184	(fraction of variance due to u_i)				
-----							

. test L.hhall L.textall

( 1) L.hhall = 0

( 2) L.textall = 0

F( 2, 48) = 1.75  
Prob > F = 0.1846

. test L.hhall L.textall L.hh\_text

- ( 1) L.hhall = 0
- ( 2) L.textall = 0
- ( 3) L.hh\_text = 0

F( 3, 48) = 3.10  
Prob > F = 0.0352

. lincom 1\*L.hhall + 1\*L.textall + 1\*L.hh\_text

- ( 1) L.hhall + L.textall + L.hh\_text = 0

lvfrbelow08	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
(1)	-.1121388	.045268	-2.48	0.017	-.2031561	-.0211214

Summary statistics (selected)

Variable	Obs	Mean	Std. Dev.	Min	Max
year	686	2005.5	4.03407	1999	2012
state_id	686	26.77551	14.4727	1	51
fatalacc	686	723.8907	711.3	14	3849
fatalsabo~08	686	249.0554	255.3927	4	1499
fatalsbel~08	686	549.691	541.3672	11	3068
fatalstot	686	798.7464	788.4438	15	4327
vfrtot	686	1.511268	.6151643	.2368071	3.919917
vfrabove08	686	.4801786	.22839	.0631486	1.41993
vfrbelow08	686	1.03109	.4120414	.1736585	2.648029

## Confidence intervals for combined effects, first lag only specifications:

### (a) Total VFR

lvfrtot	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
(1)	-.0989977	.0601748	-1.65	0.106	-.2199872 .0219917

In lives per state: (-160, 14)

### (b) VFR with BAC < .08

lvfrbelow08	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
(1)	-.1121388	.045268	-2.48	0.017	-.2031561 -.0211214

In lives per state: (-117, -12)



# BIC in AR( $p$ )

## Simple approach in STATA

```
. varsoc dlemp if tin(1962m1,$tnm1), maxlag(6)
```

Selection-order criteria

Sample: 1962m1 - 2016m9

Number of obs

=

657

lag	LL	LR	df	p	FPE	AIC	HQIC	SBIC
0	85.8531				.045222	-.258305	-.255656	-.251474
1	228.439	285.17	1	0.000	.029387	-.689313	-.684017	-.675652
2	290.685	124.49	1	0.000	.024389	-.875754	-.867809	-.855262
3	306.297	31.223	1	0.000	.023328	-.920233	-.909641	-.892911*
4	309.291	5.9889*	1	0.014	.023187*	-.926305*	-.913064*	-.892152
5	309.644	.70495	1	0.401	.023232	-.924334	-.908445	-.88335
6	309.68	.07222	1	0.788	.023301	-.9214	-.902863	-.873586

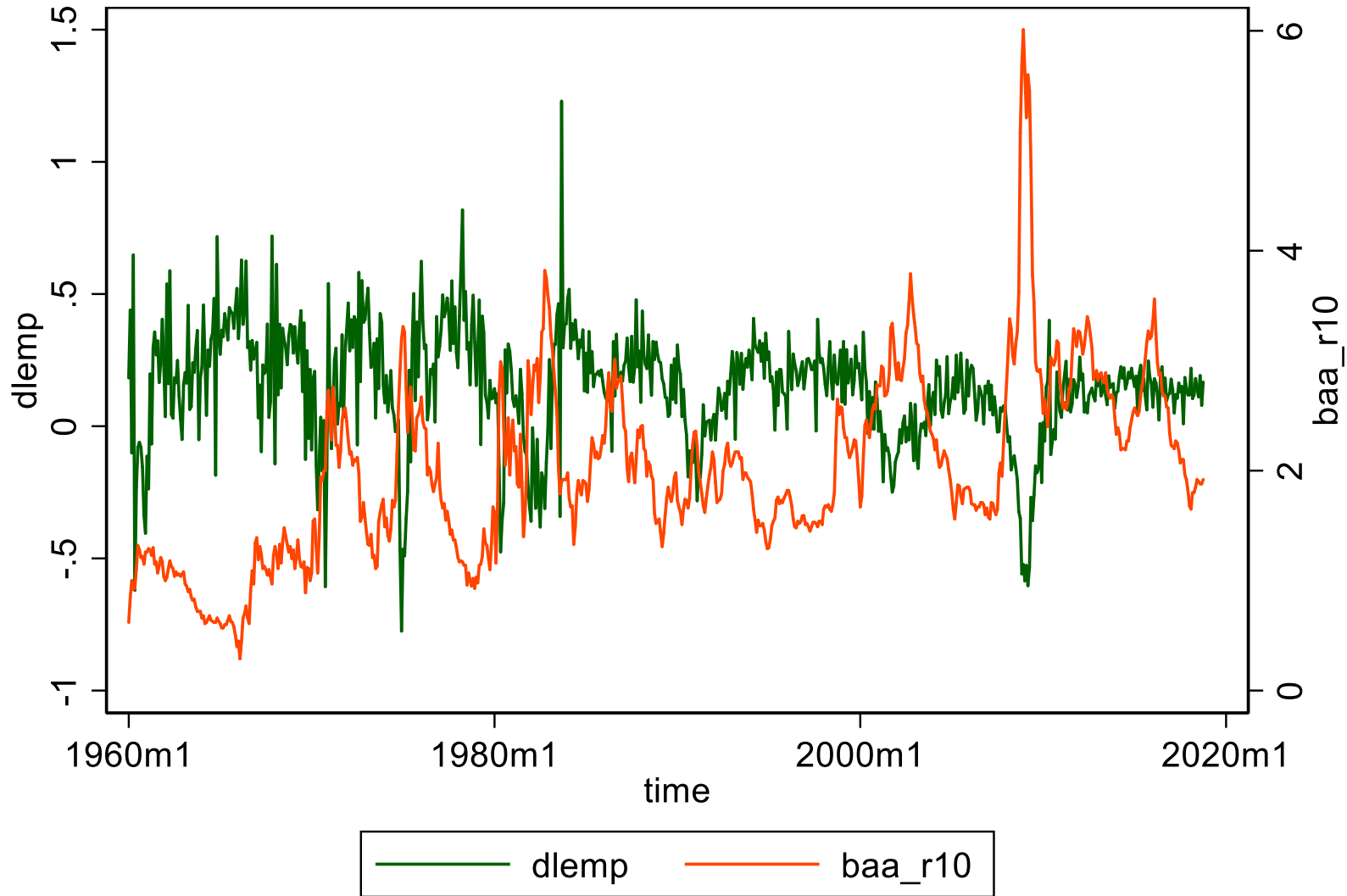
## BIC in multivariate models: choice of $p, q$ in ADL( $p, q$ )

```
* BIC calculation for ADL, max lags of 6,6
local bicmin = 1e+9
forvalues p = 1/6 {
  forvalues q = 1/6 {
    reg dlemp L(1/\`p') .dlemp L(1/\`q') .baa_r10 if tin(1962m1, $tn), r
    sca bic = ln(e(rss)/e(N)) + e(df_m)*ln(e(N))/e(N)
    dis "BIC = " bic
    dis "Adjusted Rsquared = " e(r2_a)
    dis "`q'," bic ", " e(r2_a)
    cap drop yhat_adl\`p'\`q'_62tn demp_adl\`p'\`q'_62tn
    predict yhat_adl\`p'\`q'_62tn
    qui gen demp_adl\`p'\`q'_62tn = payems[_n-1]*(yhat_adl\`p'\`q'_62tn/100)
    if bic < `bicmin' {
      local bicmin = bic
    }
  }
}
dis "BIC lengths: p = " `phat' "    q = " `qhat'

BIC lengths: p = 4    q = 3
```

# Stability/break analysis: Break at a known date (Chow test)

## Employment growth and BAA-Tbond spread



# Break test (“Chow test”): Break in 1986 (middle of sample)?

```

gen d85 = 0
  replace d85 = 1 if tin(1985m1,$tn)
gen d85_dlemp = d85*dlemp
gen d85_baa_r10 = d85*baa_r10
reg dlemp L(1/\`phat') .dlemp L(1/\`qhat') .baa_r10 ///
  d85 L(1/\`phat') .d85_dlemp L(1/\`qhat') .d85_baa_r10 if tin(1962m1,$tn), r
  testparm d85 L(1/\`phat') .d85_dlemp L(1/\`qhat') .d85_baa_r10

```

```

Linear regression                               Number of obs      =           682
                                                F(15, 666)         =           57.98
                                                Prob > F            =           0.0000
                                                R-squared           =           0.5273
                                                Root MSE            =           .14513

```

-----							
		Robust					
dlemp	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]		
-----							
dlemp							
L1.	.1693184	.0854578	1.98	0.048	.0015193	.3371175	
L2.	.2617734	.0688935	3.80	0.000	.1264987	.3970481	
L3.	.1688301	.0590402	2.86	0.004	.0529027	.2847575	
L4.	.0949334	.050982	1.86	0.063	-.0051714	.1950382	
baa_r10							
L1.	-.1298321	.0625494	-2.08	0.038	-.2526499	-.0070144	
L2.	-.0435966	.0785873	-0.55	0.579	-.1979054	.1107122	
L3.	.1612217	.0589227	2.74	0.006	.045525	.2769184	

d85		-.0334429	.0436531	-0.77	0.444	-.1191572	.0522713
d85_dlemp							
L1.		.0724783	.1016227	0.71	0.476	-.1270612	.2720179
L2.		.0477644	.0869543	0.55	0.583	-.1229731	.218502
L3.		-.0014344	.0782289	-0.02	0.985	-.1550394	.1521706
L4.		.0354407	.0694926	0.51	0.610	-.1010102	.1718916
d85_baa_r10							
L1.		.0161228	.0578198	0.28	0.780	-.0974084	.1296539
L2.		.0734735	.0659781	1.11	0.266	-.0560766	.2030236
L3.		-.0915837	.0463591	-1.98	0.049	-.1826112	-.0005562
_cons		.0821034	.0413868	1.98	0.048	.000839	.1633678

-----

. testparm d85 L(1/`phat').d85\_dlemp L(1/`qhat').d85\_baa\_r10

- ( 1) d85 = 0
- ( 2) L.d85\_dlemp = 0
- ( 3) L2.d85\_dlemp = 0
- ( 4) L3.d85\_dlemp = 0
- ( 5) L4.d85\_dlemp = 0
- ( 6) L.d85\_baa\_r10 = 0
- ( 7) L2.d85\_baa\_r10 = 0
- ( 8) L3.d85\_baa\_r10 = 0

F( 8, 666) = 1.01  
 Prob > F = 0.4256

## 2b. QLR test: Unknown break date (15% trimming):

Compute all possible Chow test stats for breaks in central 70% of data and report the largest Chow F-stat.

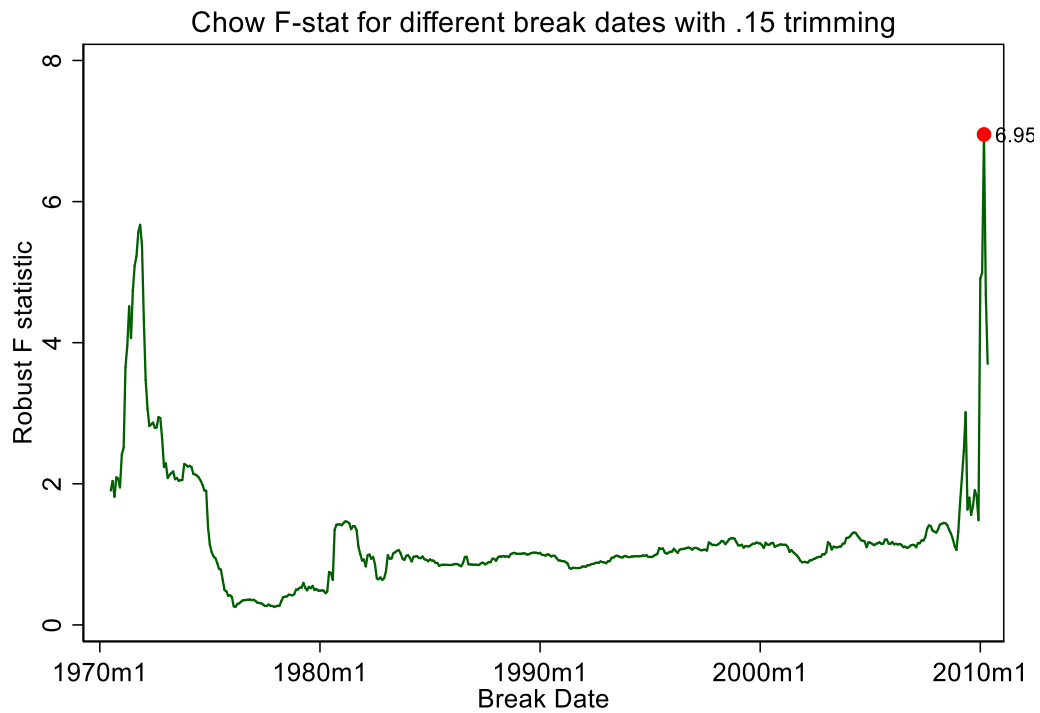
**TABLE 14.6** Critical Values of the QLR Statistic with 15% Trimming

Number of Restrictions ( $q$ )	10%	5%	1%
1	7.12	8.68	12.16
2	5.00	5.86	7.78
3	4.09	4.71	6.02
4	3.59	4.09	5.12
5	3.26	3.66	4.53
6	3.02	3.37	4.12
7	2.84	3.15	3.82
8	2.69	2.98	3.57
9	2.58	2.84	3.38
10	2.48	2.71	3.23

These critical values are larger than the  $F_{q,\infty}$  critical values – for example,  $F_{1,\infty}$  5% critical value is 3.84.

```
. qlr dlemp L(1/3) .dlemp L(1/2) .baa_r10 if tin(1962m1,$tn), graph
```

	date	F	restrict	trimming
1.	2010m3	6.95146	8	.15
2.	1971m11	5.671485	8	.15
3.	1971m10	5.566554	8	.15
4.	1971m12	5.38583	8	.15
5.	1971m9	5.227156	8	.15



## End of sample stability checks:

Estimate through 2014m12, check performance 2015m1 – end

```
. global tp "2012m12"
. reg dlemp L(1/\`phat') .dlemp L(1/\`qhat') .baa_r10 if tin(1972m1,$tp), r
```

```
Linear regression                               Number of obs   =           516
                                                F(7, 508)      =           87.57
                                                Prob > F       =           0.0000
                                                R-squared      =           0.6061
                                                Root MSE      =           .1337
```

---

	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
<hr/>						
dlemp						
L1.	.2368979	.0952696	2.49	0.013	.049727	.4240687
L2.	.349425	.0557016	6.27	0.000	.2399911	.4588589
L3.	.1584489	.0589444	2.69	0.007	.0426442	.2742537
L4.	.0399633	.0457087	0.87	0.382	-.049838	.1297646
<hr/>						
baa_r10						
L1.	-.1438489	.0401305	-3.58	0.000	-.2226911	-.0650066
L2.	.0031071	.0612409	0.05	0.960	-.1172096	.1234238
L3.	.1250056	.0478669	2.61	0.009	.0309641	.2190471
<hr/>						
_cons	.062125	.030093	2.06	0.039	.003003	.1212471

---

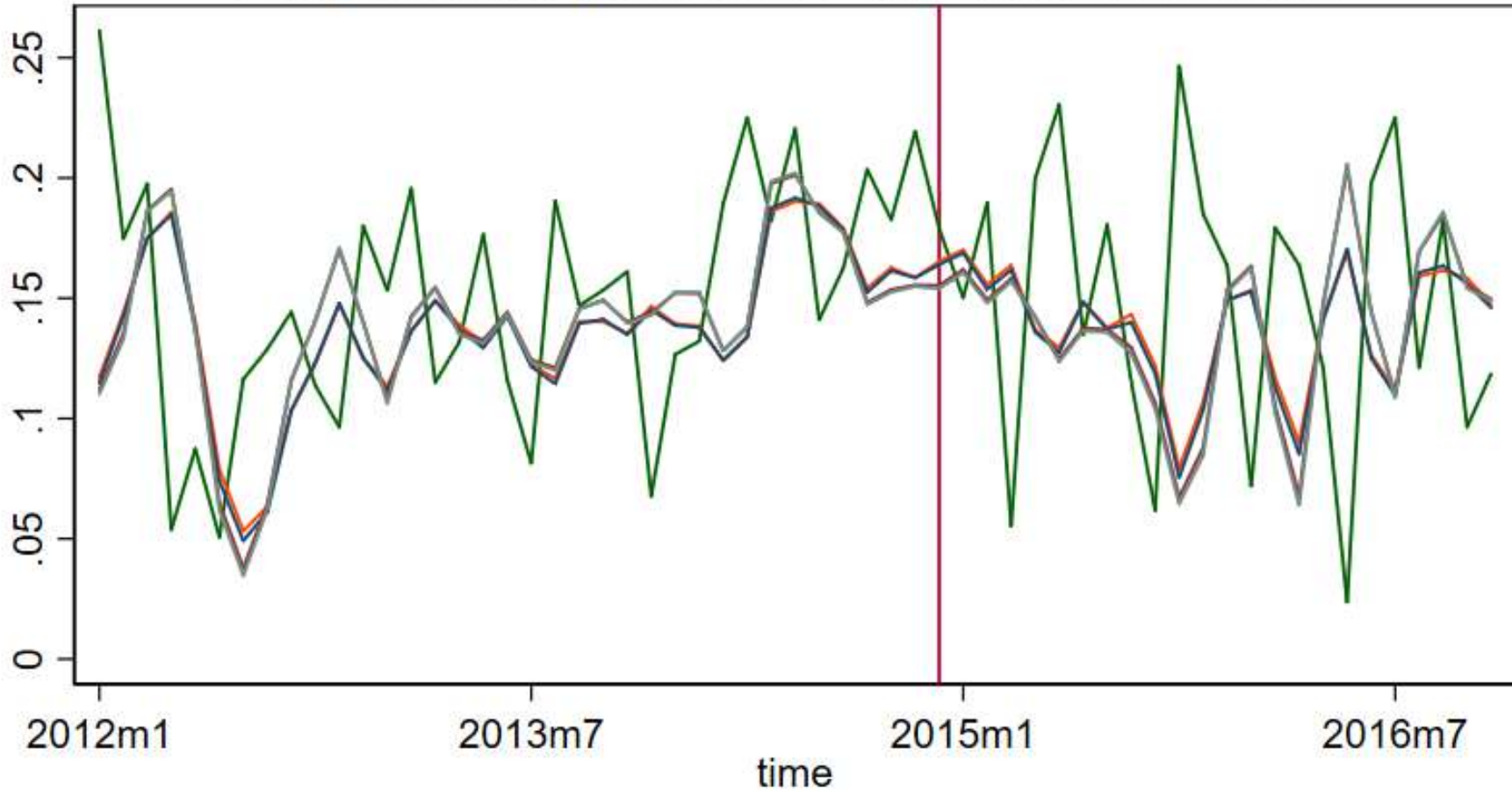


```

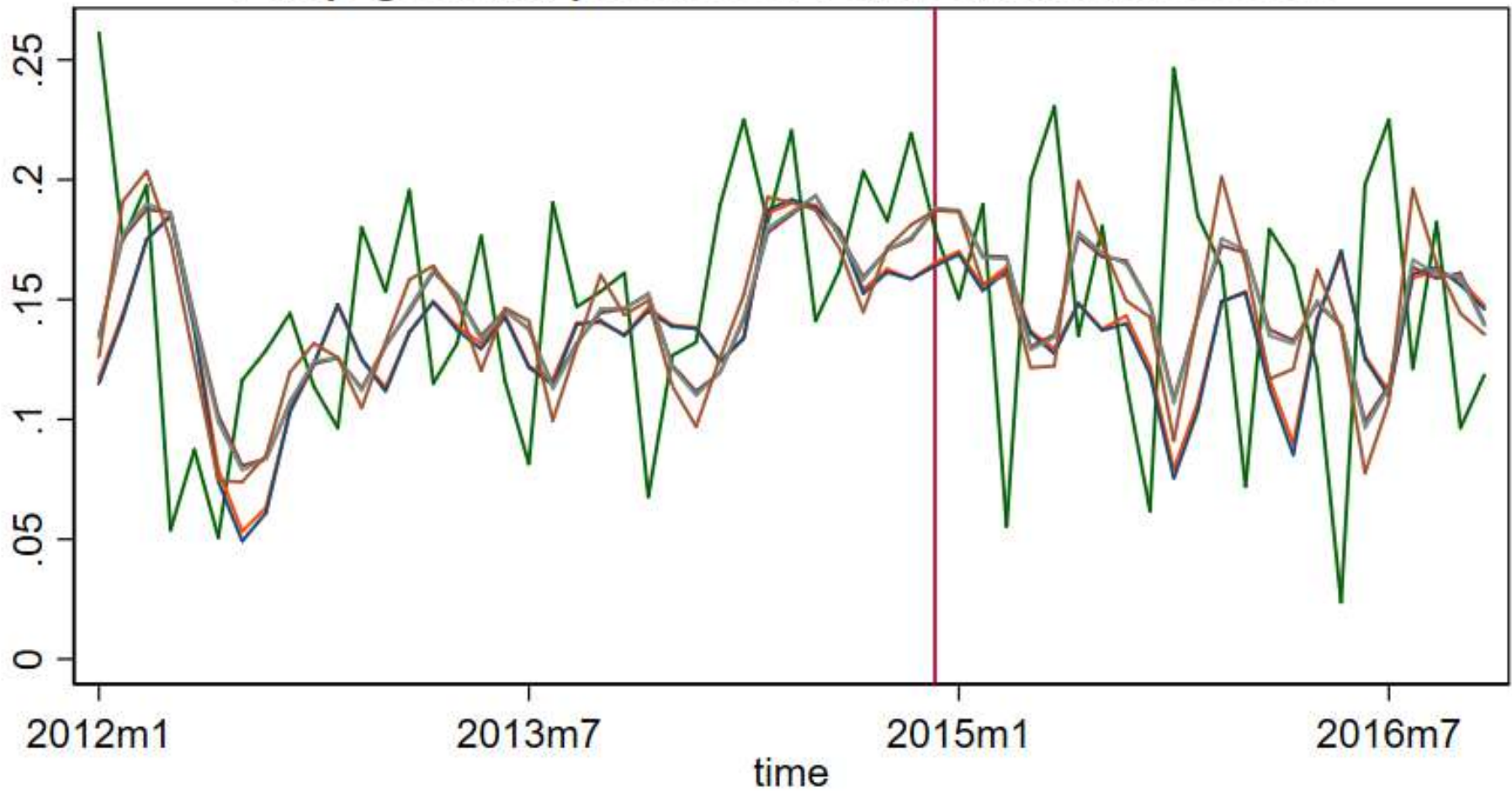
* 72 - tn
reg dlemp L(1/\`phat').dlemp L(1/\`qhat').baa_r10 if tin(1972m1,$tn), r
predict yhat_adl\`phat'\`qhat'\_72tn
gen demp_adl\`phat'\`qhat'\_72tn = payems[_n-1]*(yhat_adl\`phat'\`qhat'\_72tn/100)
* 72 - tp
reg dlemp L(1/\`phat').dlemp L(1/\`qhat').baa_r10 if tin(1972m1,$tp), r
predict yhat_adl\`phat'\`qhat'\_72tp
gen demp_adl\`phat'\`qhat'\_72tp = payems[_n-1]*(yhat_adl\`phat'\`qhat'\_72tp/100)
* 86 - tn
reg dlemp L(1/\`phat').dlemp L(1/\`qhat').baa_r10 if tin(1986m1,$tn), r
predict yhat_adl\`phat'\`qhat'\_86tn
gen demp_adl\`phat'\`qhat'\_86tn = payems[_n-1]*(yhat_adl\`phat'\`qhat'\_86tn/100)
* 86 - tp
reg dlemp L(1/\`phat').dlemp L(1/\`qhat').baa_r10 if tin(1986m1,$tp), r
predict yhat_adl\`phat'\`qhat'\_86tp
gen demp_adl\`phat'\`qhat'\_86tp = payems[_n-1]*(yhat_adl\`phat'\`qhat'\_86tp/100)
*
* graph the different in-sample fitted values and POOS forecasts
twoway tsline dlemp yhat_adl\`phat'\`qhat'\_72tn yhat_adl\`phat'\`qhat'\_72tp ///
yhat_adl\`phat'\`qhat'\_86tn yhat_adl\`phat'\`qhat'\_86tp if tin(2002m1,2016m11), ///
title("Emp growth, predicted values, and forecasts")
graph export fig\dlemp_poos.emf, replace

```

# Emp growth, predicted values, and forecasts



# Emp growth, predicted values, and forecasts



Check RMSEs: do the in-sample models fit a lot better than the pseudo out-of-sample models? Look at the forecast errors:

Forecast error = actual – predicted

```
gen e_adl`phat'`qhat'`_72tn = demp - demp_adl`phat'`qhat'`_72tn
gen e_adl`phat'`qhat'`_72tn_2 = e_adl`phat'`qhat'`_72tn*e_adl`phat'`qhat'`_72tn
su e_adl43_72tn_2 e_adl43_72tp_2 e_adl43_86tn_2 e_adl43_86tp_2 if
tin($tp,$tnp1)
```

Summary statistics for squared forecast error over pseudo out of sample period:

Variable	Obs	Mean	Std. Dev.	Min	Max
e_ar2_13tp_2	47	9591.721	11127.17	.063319	49019.5
e_ar4_72tp_2	47	8377.997	10360.5	2.291895	37608.59
e_adl4~2tp_2	47	10037.8	15410.83	4.214157	68598.75
e_ar4_86tp_2	47	8287.914	10125.43	.0841138	39613.93
e_adl4~6tp_2	47	8806.949	12499.92	.0828002	59371.02
e_ar2_13tn_2	47	9539.629	11077.59	.0032533	48675.44
e_ar4_72tn_2	47	8317.895	10308.77	4.492891	37317.93
e_adl4~2tn_2	47	9847.371	15071.32	11.99753	67430.69
e_ar4_86tn_2	47	8137.385	9991.039	2.592544	38623.08
e_adl4~6tn_2	47	8589.311	12148.67	1.569336	56790.43

Forecast RMSE for 8612 model is  $\sqrt{8138} = 90.2\text{k jobs (!)}$

## How do the forecasts compare?

AR(4) and ADL(4,3) forecasts of the change in employment for Oct and November, 86-18m10, and AR(2), 2013m1-2018m10

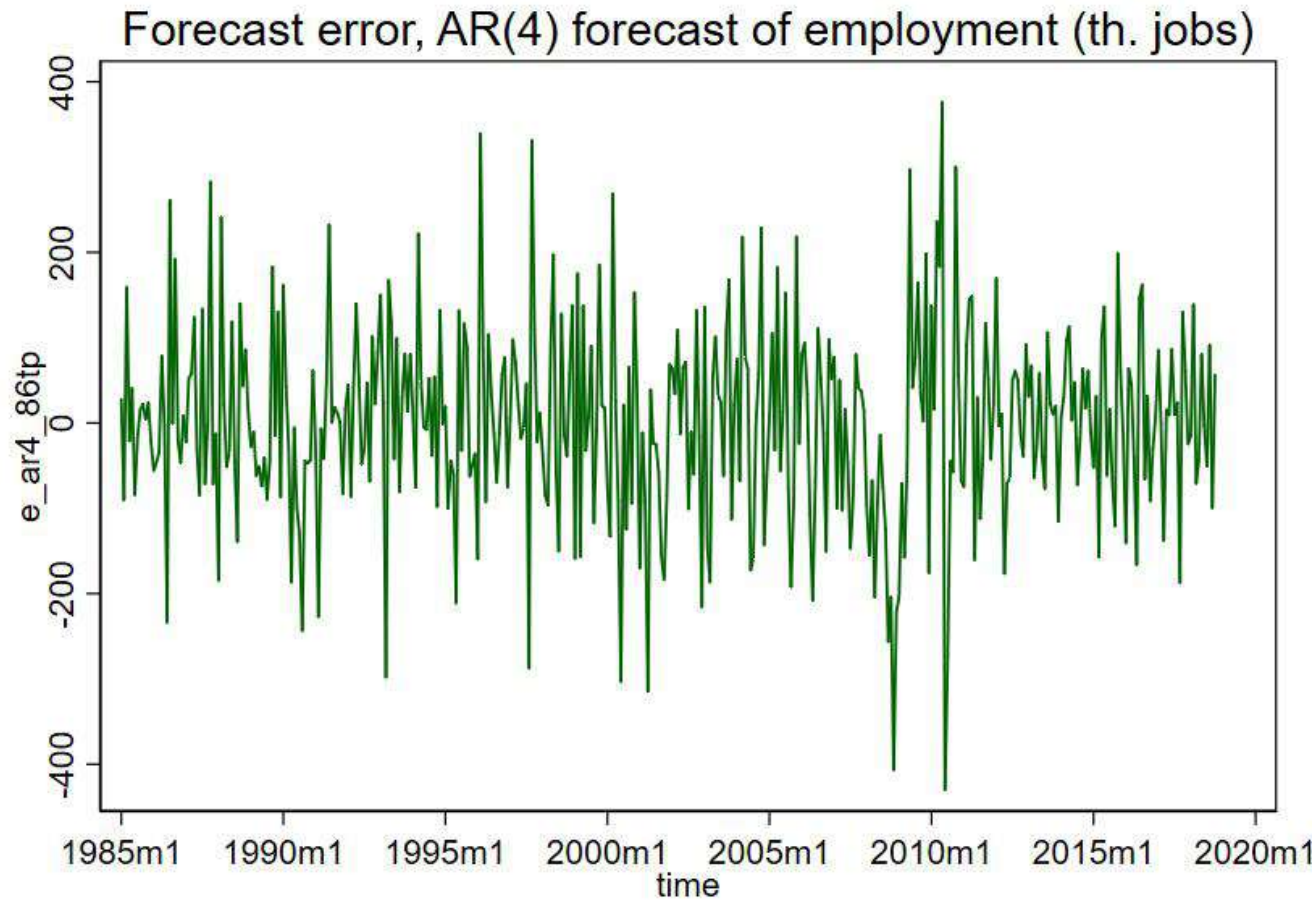
time	baa_r10	demp	d~4_86tn	d~3_86tn	demp~3tn
2018m9	1.88	118	218.02	228.5924	214.4359
2018m10	1.92	250	193.3913	205.5838	198.0245
2018m11	.	.	194.6607	195.5899	178.6618

Units of the forecasts are thousands of jobs, monthly change  
*November Employment Situation is released 8:30am Dec. 7, 2018*

# Forecast intervals

Use root mean squared (forecast) error

RMSFE = 90.2k over 2015-2018 (POOS estimate)

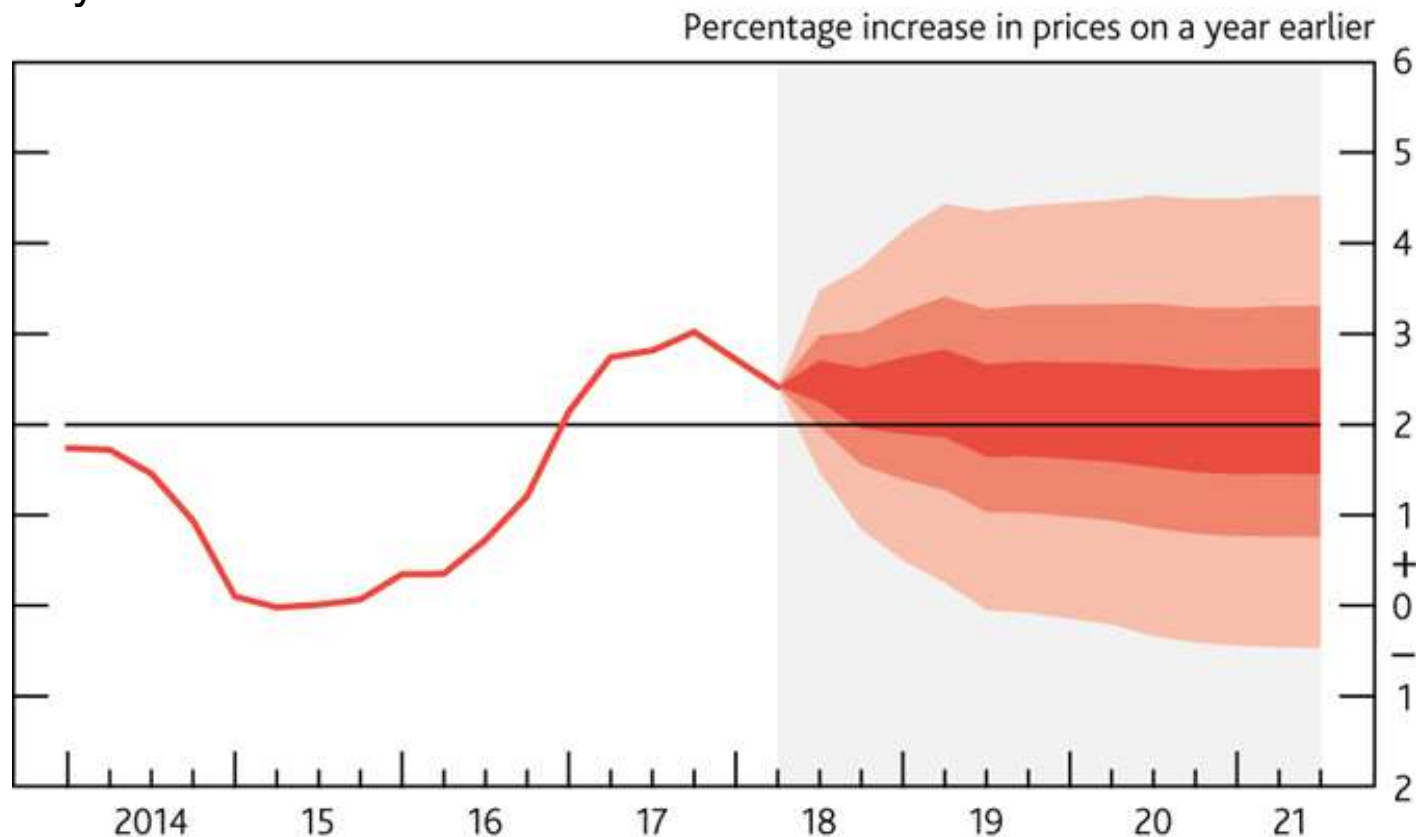


- 67% forecast interval for Nov 2016 = 195k  $\pm$  90k (AR(3) model estimated 1986m1-2018m10)

## Forecast distributions:

### Bank of England Inflation “Fan Chart”, November 2016

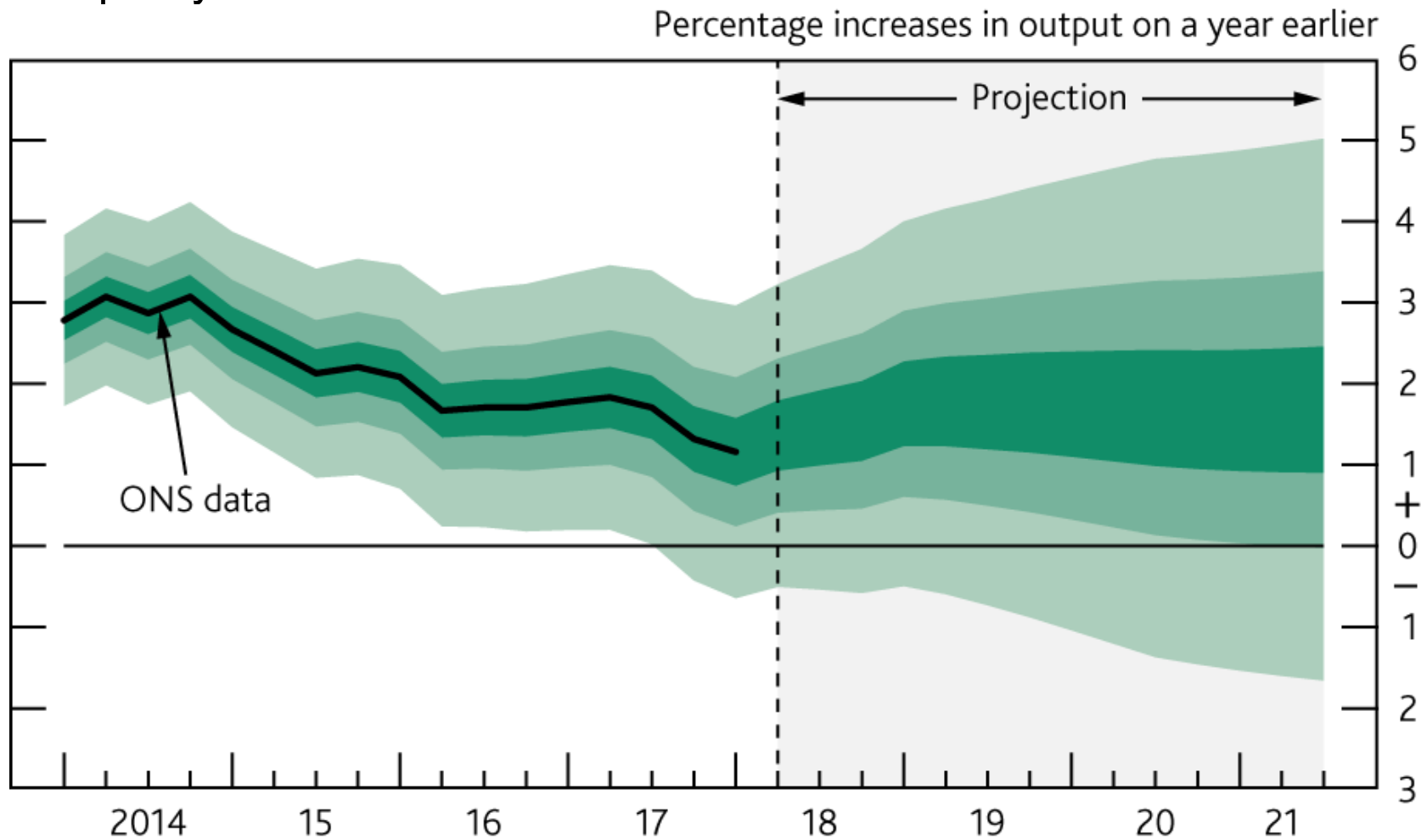
**Chart 5.3** CPI inflation projection based on market interest rate expectations, other policy measures as announced



<https://www.bankofengland.co.uk/inflation-report/2018/august-2018>

**Forecast distributions:** Bank of England GDP fan chart, 8/2018

**Chart 5.1** GDP projection based on market interest rate expectations, other policy measures as announced





# STATA code for lectures 21-22

```
chdir c:\courses\e1123\stata\macro
clear
set scheme slcolor
cap log close
cap program drop _all
log using e1123_f18_lect_21_22.txt, text replace
set more 1
*****
* e1123_f18_lect_21_22.do
*   Ec1123 TS lectures
*****
*aaa          float   %9.0g          Moody's Seasoned Aaa Corporate Bond Yield
*ambsl        float   %9.0g          St. Louis Adjusted Monetary Base
*awhi         float   %9.0g          Aggregate Weekly Hours Index: Total Private
Industries
*baa          float   %9.0g          Moody's Seasoned Baa Corporate Bond Yield
*bogambsl     float   %9.0g          Board of Governors Monetary Base, Adjusted
for Changes in Reserve
*cpiaucns     float   %9.0g          Consumer Price Index for All Urban
Consumers: All Items - NSA
*cpiaucsl     float   %9.0g          Consumer Price Index for All Urban
Consumers: All Items - SA
*cpilfens     float   %9.0g          Consumer Price Index for All Urban
Consumers: All Items Less Food & Energy - NSA
*cpilfesl     float   %9.0g          Consumer Price Index for All Urban
Consumers: All Items Less Food & Energy - SA
*cpn1m        float   %9.0g          1-Month AA Nonfinancial Commercial Paper
Rate
*cpn3m        str4     %9s           3-Month AA Nonfinancial Commercial Paper
Rate
*dspi         float   %9.0g          Disposable Personal Income - current dollar
```

*dspic96	float	%9.0g	Real Disposable Personal Income
*gs10	float	%9.0g	10-Year Treasury Constant Maturity Rate
*houst	int	%8.0g	Housing Starts: Total: New Privately Owned
Housing Units Started			
*hsn1f	int	%8.0g	New One Family Houses Sold: United States
*isratio	float	%9.0g	Inventory to Sales Ratio: Total Business
*m1sl	float	%9.0g	M1 Money Stock
*m2sl	float	%9.0g	M2 Money Stock
*mich	float	%9.0g	University of Michigan Inflation Expectation
- 1 year			
*oilprice	float	%9.0g	Spot Oil Price: West Texas Intermediate
*payems	long	%12.0g	Total Nonfarm Payrolls: All Employees
*pce	float	%9.0g	Personal Consumption Expenditures - current
dollar			
*pcec96	float	%9.0g	Real Personal Consumption Expenditures -
chained 20065 dollars			
*pcedgc96	float	%9.0g	Real Personal Consumption Expenditures:
Durable Goods			
*pcendc96	float	%9.0g	Real Personal Consumption Expenditures:
Nondurable Goods			
*pcepi	float	%9.0g	Personal Consumption Expenditures: Chain-
type Price Index			
*pcepilfe	float	%9.0g	Personal Consumption Expenditures: Chain-
Type Price Index Less Food & Energy			
*pcesc96	float	%9.0g	Real Personal Consumption Expenditures:
Services			
*pi	float	%9.0g	Personal Income - current dollars
*psavert	float	%9.0g	Personal Saving Rate
*tb3ms	float	%9.0g	3-Month Treasury Bill: Secondary Market Rate
*umcsent	float	%9.0g	University of Michigan: Consumer Sentiment
*unrate	float	%9.0g	Civilian Unemployment Rate
*time	float	%tm	

\*\*\*\*\*

use \macrodata\data\macro\_update\_m.dta

```

tsset time
drop if tin(,1958m12)
desc
global tnm1 "2016m9"
global tn   "2018m10"
global tnp1 "2018m11"
global tp   "2014m12"
*****
* additional data constructions
*****
gen ldspi = ln(dspi)
gen dldspi = 1200*D.ldspi
gen lcpi = ln(cpiaucsl)
gen infcpi = 1200*(D.lcpi)
gen infcpi12 = 100*(lcpi - L12.lcpi)
gen r90 = tb3ms
gen r10 = gs10
gen r10_r90 = r10-r90
  replace baa = 4.34 if tin(2016m10,2016m10)
gen baa_r10 = baa-r10
replace houst = houst/1000
gen dhoust = D.houst
gen lindpro = ln(indpro)
gen dlindpro = 100*D.lindpro
gen lemp = ln(payems)
gen dlemp = 100*D.lemp
gen demp = D.payems
* Annual employment growth
gen d12lemp = 100*(lemp-L12.lemp)
*****
* Time series plots
*****
*
twoway tsline infcpi, title("Monthly inflation rate, CPI")

```

```

graph export fig\infcp12.emf, replace
twoway tsline infcp12, title("12-month inflation rate, CPI")
graph export fig\infcp12.emf, replace
twoway tsline payems, title("Payroll employment in U.S. (thousands of workers)")
graph export fig\empl.emf, replace
twoway tsline lemp, title("Payroll employment, logs")
graph export fig\lemp.emf, replace
twoway tsline dlemp, title("Payroll employment, monthly % change")
graph export fig\dlemp.emf, replace
twoway tsline d12lemp, title("Payroll employment: growth over previous 12 months")
graph export fig\d12lemp.emf, replace
twoway tsline unrate, title("Civilian unemployment rate, U.S.")
graph export fig\unrate.emf, replace
twoway tsline tb3ms, title("Interest rate on 90-day Treasury bills")
graph export fig\tb3ms.emf, replace
twoway tsline gs10, title("Yield on 10-year Treasury bonds")
graph export fig\gs10.emf, replace
twoway tsline baa, title("Moody's Seasoned Baa Corporate Bond Yield")
graph export fig\baa.emf, replace
twoway tsline baa r10 baa_r10, title("Baa bonds, 10-yr Tbond rates, & spread")
graph export fig\baa_r10.emf,replace
twoway tsline dlemp baa_r10, title("Employment growth and BAA-Tbond spread")
graph export fig\dlemp_baar10.emf,replace
twoway (tsline dlemp) (tsline baa_r10, yaxis(2)) if tin(1960m1,$tnp1), ///
title("Employment growth and BAA-Tbond spread")
graph export fig\dlemp_baar10_6015.emf,replace
twoway (tsline dlemp) (tsline baa_r10, yaxis(2)) if tin(2005m1, $tnp1), ///
title("Employment growth and BAA-Tbond spread")
graph export fig\dlemp_baar10_0515.emf,replace
twoway (tsline dlemp) (tsline houst, yaxis(2)) if tin(1960m1,$tnp1), ///
title("Employment growth and Housing Starts")
graph export fig\dlemp_houst_6015.emf,replace
twoway (tsline dlemp) (tsline houst, yaxis(2)) if tin(2005m1, $tnp1), ///
title("Employment growth and Housing Starts")

```

```

*
*****
* autocorrelations
*****
*
corrgram unrate
corrgram dlemp
list time payems demp lemp dlemp if tin(2018m1,$tn)
su demp if tin(2011q1,$tn)
gen demp_mn = r(mean)
tsline demp demp_mn if tin(2011q1,$tn), ///
  title("Monthly change in employment (th. jobs)") legend(off) tttitle("")
*
*****
* AR(1) - AR(6) models of employment growth
*****
*
reg dlemp if tin(1962m1,$tnm1), r
  sca bic = ln(e(rss)/e(N)) + e(df_m)*ln(e(N))/e(N)
  dis "BIC = " bic
  dis "Adjusted Rsquared = " e(r2_a)
  dis 0 ", " bic ", " e(r2_a)
predict yhat_ar0_62tn
forvalues p = 1/6 {
  reg dlemp L(1/`p').dlemp if tin(1962m1,$tnm1), r
    sca bic = ln(e(rss)/e(N)) + `p'*e(df_m)*ln(e(N))/e(N)
    dis "BIC = " bic
    dis "Adjusted Rsquared = " e(r2_a)
    dis `p' ", " bic ", " e(r2_a)
  predict yhat_ar`p'_62tn
  qui gen demp_ar`p'_62tn = payems[_n-1]*(yhat_ar`p'_62tn/100)
  testparm L(1/`p').dlemp
  if `p'>1 {
    testparm L(2/`p').dlemp
  }
}

```

```

}
}
*
* simpler command to produce BIC
varsoc dlemp if tin(1962m1,$tnm1), maxlag(6)*
list time payems dlemp dlemp yhat_ar1_62tn if tin(2015m1,$tnp1), noobs
*
*****
* ADL(1,1) and ADL(4,4) models using Baa-10 year T-bond spread
* including Granger Causality test (4 lags)
*****
*
reg dlemp L1.dlemp L1.baa_r10 if tin(1962m1,$tnm1), r
dis "Adjusted Rsquared = " e(r2_a)
reg dlemp L(1/4).dlemp L(1/4).baa_r10 if tin(1962m1,$tn), r
dis "Adjusted Rsquared = " e(r2_a)
cap drop yhat_adl44_62tn
predict yhat_adl44_62tn
testparm L(1/4).baa_r10
testparm L(2/4).baa_r10
testparm L(2/4).dlemp
*
gen dlemp_adl44_62tn = payems[_n-1]*(yhat_adl44_62tn/100)
list time dlemp dlemp_ar1_62tn dlemp_ar2_62tn dlemp_adl44_62tn if tin(2016m1,$tnp1), noobs
*
*****
* ADL(p,q) models using Baa-10 year T-bond spread, p = 1,..., 6, q = 1,..., 6
*****
*
reg dlemp L(1/3).dlemp if tin(1962m1,$tn), r
dis "BIC = " ln(e(rss)/e(N)) + e(df_m)*ln(e(N))/e(N)
dis "Adjusted Rsquared = " e(r2_a)
local bicmin = 1e+9
forvalues p = 1/6 {

```

```

forvalues q = 1/6 {
  reg dlemp L(1/\`p').dlemp L(1/\`q').baa_r10 if tin(1962m1,$tn), r
  sca bic = ln(e(rss)/e(N)) + e(df_m)*ln(e(N))/e(N)
  dis "BIC = " bic
  dis "Adjusted Rsquared = " e(r2_a)
  dis "`q'," bic "," e(r2_a)
  cap drop yhat_adl`p'`q' _62tn demp_adl`p'`q' _62tn
  predict yhat_adl`p'`q' _62tn
  qui gen demp_adl`p'`q' _62tn = payems[_n-1]*(yhat_adl`p'`q' _62tn/100)
  if bic < `bicmin' {
    local bicmin = bic
  }
}

local qhat = `q'

local phat = `p'
}
}
}
dis "BIC lengths: p = " `phat' "    q = " `qhat'
*
*****
* In-sample predicted values
*****
*
list time dlemp yhat_ar1_62tn yhat_ar4_62tn yhat_adl44_62tn if tin(2016m1,$tn), noobs
list time demp demp_ar1_62tn demp_ar2_62tn demp_adl44_62tn demp_adl`phat'`qhat' _62tn if
tin(2016m1,$tn), noobs
*
*****
* Stability analysis: Chow test for break mid sample (1986)
*****
reg dlemp L(1/\`phat').dlemp L(1/\`qhat').baa_r10 if tin(1962m1,1985m12), r
reg dlemp L(1/\`phat').dlemp L(1/\`qhat').baa_r10 if tin(1986m1,$tn), r
*

```

```

gen d85 = 0
  replace d85 = 1 if tin(1985m1,$tn)
gen d85_dlemp = d85*dlemp
gen d85_baa_r10 = d85*baa_r10
reg dlemp L(1/\`phat').dlemp L(1/\`qhat').baa_r10 ///
  d85 L(1/\`phat').d85_dlemp L(1/\`qhat').d85_baa_r10 if tin(1962m1,$tn), r
  testparm d85 L(1/\`phat').d85_dlemp L(1/\`qhat').d85_baa_r10
*
*****
* Stability analysis: QLR test (note: the QLR test is slow)
*****
*
qlr dlemp L(1/\`phat').dlemp L(1/\`qhat').baa_r10 if tin(1962m1,$tn), graph
  graph export fig\qlr_adl\`phat'\`qhat'\_sprd_62_2015.png,replace
*
*****
* end-of-sample Pseudo out-of-sample performance, 2013-end
*****
global tp "2014m12"
* 72 - tn
reg dlemp L(1/\`phat').dlemp if tin(1972m1,$tn), r
  predict yhat_ar\`phat'\_72tn
  gen demp_ar\`phat'\_72tn = payems[_n-1]*(yhat_ar\`phat'\_72tn/100)
reg dlemp L(1/\`phat').dlemp L(1/\`qhat').baa_r10 if tin(1972m1,$tn), r
  predict yhat_adl\`phat'\`qhat'\_72tn
  gen demp_adl\`phat'\`qhat'\_72tn = payems[_n-1]*(yhat_adl\`phat'\`qhat'\_72tn/100)
* 72 - tp
reg dlemp L(1/\`phat').dlemp if tin(1972m1,$tp), r
  predict yhat_ar\`phat'\_72tp
  gen demp_ar\`phat'\_72tp = payems[_n-1]*(yhat_ar\`phat'\_72tp/100)
reg dlemp L(1/\`phat').dlemp L(1/\`qhat').baa_r10 if tin(1972m1,$tp), r
  predict yhat_adl\`phat'\`qhat'\_72tp
  gen demp_adl\`phat'\`qhat'\_72tp = payems[_n-1]*(yhat_adl\`phat'\`qhat'\_72tp/100)
* 86 - tn

```



```

reg dlemp L(1/\`phat').dlemp if tin(1986m1,$tn), r
  predict yhat_ar`phat'_86tn
  gen demp_ar`phat'_86tn = payems[_n-1]*(yhat_ar`phat'_86tn/100)
reg dlemp L(1/\`phat').dlemp L(1/\`qhat').baa_r10 if tin(1986m1,$tn), r
  predict yhat_adl`phat'\`qhat'_86tn
  gen demp_adl`phat'\`qhat'_86tn = payems[_n-1]*(yhat_adl`phat'\`qhat'_86tn/100)
* 86 - tp
reg dlemp L(1/\`phat').dlemp if tin(1986m1,$tp), r
  predict yhat_ar`phat'_86tp
  gen demp_ar`phat'_86tp = payems[_n-1]*(yhat_ar`phat'_86tp/100)
reg dlemp L(1/\`phat').dlemp L(1/\`qhat').baa_r10 if tin(1986m1,$tp), r
  predict yhat_adl`phat'\`qhat'_86tp
  gen demp_adl`phat'\`qhat'_86tp = payems[_n-1]*(yhat_adl`phat'\`qhat'_86tp/100)
*
* 2013 - AR(2)
reg dlemp L(1/2).dlemp if tin(1986m1,$tp), r
  predict yhat_ar2_13tp
  gen demp_ar2_13tp = payems[_n-1]*(yhat_ar2_13tp/100)
reg dlemp L(1/2).dlemp if tin(1986m1,$tn), r
  predict yhat_ar2_13tn
  gen demp_ar2_13tn = payems[_n-1]*(yhat_ar2_13tn/100)

* graph the different in-sample fitted values and POOS forecasts
label var yhat_ar`phat'_72tn "yhat_ar`phat'_72tn"
label var yhat_ar`phat'_72tp "yhat_ar`phat'_72tp"
label var yhat_ar`phat'_86tn "yhat_ar`phat'_86tn"
label var yhat_ar`phat'_86tp "yhat_ar`phat'_86tp"
label var yhat_ar2_13tn "yhat_ar`phat'_13tn"
label var yhat_ar2_13tp "yhat_ar`phat'_13tp"
label var yhat_adl`phat'\`qhat'_72tn "yhat_adl`phat'\`qhat'_72tn"
label var yhat_adl`phat'\`qhat'_72tp "yhat_adl`phat'\`qhat'_72tp"
label var yhat_adl`phat'\`qhat'_86tn "yhat_adl`phat'\`qhat'_86tn"
label var yhat_adl`phat'\`qhat'_86tp "yhat_adl`phat'\`qhat'_86tp"
twoway tsline dlemp yhat_adl`phat'\`qhat'_86tn yhat_adl`phat'\`qhat'_86tp ///

```

```

yhat_adl`phat'`qhat'_72tn yhat_adl`phat'`qhat'_72tp if tin(2012m1,2016m11), ///
title("Emp growth, predicted values, and forecasts") tline($tp)
graph export fig\dlemp_poos_adl.png, replace
twoway tline dlemp yhat_adl`phat'`qhat'_86tn yhat_adl`phat'`qhat'_86tp ///
          yhat_ar`phat'_86tn yhat_ar`phat'_86tp yhat_ar2_13tp if
tin(2012m1,2016m11), ///
title("Emp growth, predicted values, and forecasts") tline($tp)
graph export fig\dlemp_poos_ar_adl.png, replace
*
* compare forecast errors over the POOS period
*
global vlist ""
foreach nn in "tp" "tn" {
  gen e_ar2_13`nn' = demp - demp_ar2_13`nn'
  gen e_ar2_13`nn'_2 = e_ar2_13`nn'*e_ar2_13`nn'
  global vlist "$vlist e_ar2_13`nn'_2"
  foreach tt in "72" "86" {
    gen e_ar`phat'`_`tt'`nn' = demp - demp_ar`phat'`_`tt'`nn'
    gen e_ar`phat'`_`tt'`nn'_2 = e_ar`phat'`_`tt'`nn'*e_ar`phat'`_`tt'`nn'
    gen e_adl`phat'`qhat'`_`tt'`nn' = demp - demp_adl`phat'`qhat'`_`tt'`nn'
    gen e_adl`phat'`qhat'`_`tt'`nn'_2 = e_adl`phat'`qhat'`_`tt'`nn'*e_adl`phat'`qhat'`_`tt'`nn'
    global vlist "$vlist e_ar`phat'`_`tt'`nn'_2 e_adl`phat'`qhat'`_`tt'`nn'_2"
  }
}
su $vlist if tin($tp,$tnp1)
*
*****
* Forecasts and forecast interval
*****
su e_ar`phat'_86tp_2 if tin($tp,$tn)
dis "POOS RMSE = " sqrt(r(mean))
*
tline e_ar`phat'_86tp if tin(1985m1,2018m12), ///
title("Forecast error, AR(4) forecast of employment (th. jobs)")

```

```
graph export fig\e_ar`phat'_86tp.png, replace
*
* forecasts for November
list time baa_r10 demp demp_ar`phat'_86tn demp_adl`phat`qhat'_86tn demp_ar2_13tn if
tin(2018m1,2018m11), noobs
*
*****
log close
```