- 1) Information Criterion and Cross Validation
- c) Compare model selection using the four criteria. Did they agree in both cases (n=30 and n=300)? For each criterion, are the differences between the "best" model and the next best more or less pronounced with more data?

	N=30				N=300			
	5-Fold*	LOOCV			10-Fold	LOOCV		
Lags	RMSE**	RMSE	AIC	BIC	RMSE**	RMSE	AIC	BIC
1	1.200	1.181	84.905	87.496	0.971	0.973	826.860	834.248
2	1.106	1.161	83.709	86.301	1.030	1.031	860.809	868.196
3	1.179	1.178	83.284	85.876	1.006	1.003	845.514	852.901
1,2	1.208	1.165	83.683	87.571	0.977	0.977	828.387	839.468
1,3	1.102	1.117	81.155	85.043	0.952	0.951	813.588	824.669
2,3	1.007	1.129	81.986	85.874	1.005	1.006	847.514	858.595
1,2,3	0.987	1.002	76.828	82.011	0.948	0.946	811.069	825.844

^{*}I used K=5 because there are only 27 observations, if you leave out the first three to make sure all samples used are based on the same set (30 if you did not).

Some other things to be aware of:

- You will have different numbers and a different pattern than in my table.
- In both cases, K-fold, LOOCV, and AIC agree on the first choice for my results.
- The BIC selection is "wrong" for n=300. But only slightly.
- Some of you may have gotten more disagreement. Part of the point of this is just to drive home that that can happen and there is no perfect selection criteria.
- 2) From the table below, we see performance does not differ much across these models by these selection criteria. Model 3 is best if I had to pick one. It is a good bit simpler than 1 or 2, since it drops one variable and has fewer lags than 1.

Model	1	2	3	4
Lagg	1/12	1/12	1/12	1/12,24
Lags	0/12	0/2	0/2, 12	0/2, 12, 24
FL Bld Permits	Yes	Yes	Yes	No
US EPR	Yes	Yes	No	Yes
10-Fold	0.0039	0.0039	0.0037	0.0039
LOOCV	0.0039	0.0038	0.0038	0.0037
AIC	-2655.729	-2774.178	-2781.418	-2679.016
BIC	-2902.086	-2905.055	-2908.446	-2804.882

^{**}To get the K-Fold RMSE, I squared each folds RMSE to get the MSE, averaged them, then took the square root, to get the RMSE as it would be for all out validation set observations if calculated all together. If you just averaged, OK, it will be very close.

```
Appendix A: Do File
*Problem Set 3 Solution
clear
set more off
cd "C:\Users\jdewey\Documents\A S20 Time Series\Problem Sets\"
log using "Problem Set 3 Work", replace
*Question 1a
clear
set obs 30
gen t=[_n]
tsset t
gen r=rnormal()
gen y=r if t<4
replace y=0.5+0.5*1.y-0.1*12.y+0.25*13.y+r if t>=4
drop r
*Note, I restrict the sample to t>3 so the same observations
** are compared for each model
crossfold reg y l(1).y if t>3 , k(5)
loocv reg y l(1).y if t>3
reg y l(1).y if t>3
estat ic
crossfold reg y 1(2).y if t>3, k(5)
loocv reg y 1(2).y if t>3
reg y 1(2).y if t>3
estat ic
crossfold reg y 1(3).y if t>3, k(5)
loocv reg y 1(3).y if t>3
reg y 1(3).y if t>3
estat ic
crossfold reg y 1(1,2).y if t>3 , k(5)
loocv reg y l(1,2).y if t>3
reg y 1(1,2).y if t>3
estat ic
crossfold reg y l(1,3).y if t>3, k(5)
loocv reg y 1(1,3).y if t>3
reg y 1(1,3).y if t>3
estat ic
crossfold reg y 1(2,3).y if t>3, k(5)
loocv reg y 1(2,3).y if t>3
reg y 1(2,3).y if t>3
estat ic
crossfold reg y 1(1/3).y if t>3, k(5)
loocv reg y 1(1/3).y if t>3
reg y 1(1/3).y if t>3
estat ic
*Question 1b
clear
set obs 300
gen t=[n]
tsset t
```

```
gen r=rnormal()
gen y=r if t<4
replace y=0.5+0.5*1.y-0.1*12.y+0.25*13.y+r if t>=4
drop r
crossfold reg y l(1).y if t>3 , k(10)
loocv reg y l(1).y if t>3
reg y l(1).y if t>3
estat ic
crossfold reg y 1(2).y if t>3, k(10)
loocv reg y 1(2).y if t>3
reg y 1(2).y if t>3
estat ic
crossfold reg y 1(3).y if t>3, k(10)
loocv reg y 1(3).y if t>3
reg y 1(3).y if t>3
estat ic
crossfold reg y 1(1,2).y if t>3 , k(10)
loocv reg y 1(1,2).y if t>3
reg y 1(1,2).y if t>3
estat ic
crossfold reg y 1(1,3).y if t>3, k(10)
loocv reg y 1(1,3).y if t>3
reg y 1(1,3).y if t>3
estat ic
crossfold reg y 1(2,3).y if t>3, k(10)
loocv reg y 1(2,3).y if t>3
reg y 1(2,3).y if t>3
estat ic
crossfold reg y 1(1/3).y if t>3 , k(10)
loocv reg y 1(1/3).y if t>3
reg y 1(1/3).y if t>3
estat ic
*Ouestion 2
clear
set more off
** data prep
import delimited using "us and florida economic time series.txt"
rename observation date datestring
gen dateday=date(datestring,"YMD")
gen date=mofd(dateday)
format date %tm
tsset date
generate month=month(dateday)
keep if tin(1990m1,2019m12)
rename flbppriv fl bp
rename fllfn fl lf
rename flnan fl nonfarm
rename lnu02300000 20200110 us epr
gen lnflnonfarm=ln( fl nonfarm)
```

```
gen lnfllf=ln( fl lf)
gen lnusepr = ln(us epr)
gen lnflbp=ln( fl bp)
*fit and evaluate models
*Note I restrict estimation to year>1989 so the same observations
*Are compared for all models
crossfold reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/12)d.lnfllf ///
      1(0/12)d.lnusepr 1(0/12)d.lnflbp i.month date, k(10)
loocv reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/12)d.lnfllf ///
      1(0/12) d.lnusepr 1(0/12) d.lnflbp i.month date
reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/12)d.lnfllf ///
      1(0/12) d.lnusepr 1(0/12) d.lnflbp i.month date
estat ic
crossfold reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/2)d.lnfllf ///
      1(0/2)d.lnusepr 1(0/2)d.lnflbp i.month date, k(10)
loocv reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/2)d.lnfllf ///
      1(0/2)d.lnusepr 1(0/2)d.lnflbp i.month date
reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/2)d.lnfllf 1(0/2)d.lnusepr ///
      1(0/2) d.lnflbp i.month date
estat ic
crossfold reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/2,12)d.lnfllf ///
      1(0/2,12) d.lnflbp i.month date, k(10)
loocv reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/2,12)d.lnfllf ///
      1(0/2,12) d.lnflbp i.month date
reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/2,12)d.lnfllf //
      1(0/2,12) d.lnflbp i.month date
crossfold reg d.lnflnonfarm 1(1/12,24)d.lnflnonfarm 1(1/2,12,24)d.lnfllf ///
      1(1/2,12,24) d.lnusepr i.month if tin(1990m1,2019m12), k(10)
loocv reg d.lnflnonfarm 1(1/12,24) d.lnflnonfarm 1(1/2,12,24) d.lnfllf ///
      1(1/2,12,24) d.lnusepr i.month
reg d.lnflnonfarm 1(1/12,24)d.lnflnonfarm 1(1/2,12,24)d.lnfllf ///
      1(1/2,12,24) d.lnusepr i.month
estat ic
```

log close

```
Appendix B: Log File
. *Question 1a
. clear
. set obs 30
number of observations ( N) was 0, now 30
. gen t=[n]
. tsset t
      time variable: t, 1 to 30
             delta: 1 unit
. gen r=rnormal()
. gen y=r if t<4
(27 missing values generated)
. replace y=0.5+0.5*1.y-0.1*12.y+0.25*13.y+r if t>=4
(27 real changes made)
. drop r
. *Note, I restrict the sample to t>3 so the same observations
. ** are compared for each model
. crossfold reg y l(1).y if t>3, k(5)
          | RMSE
-----
      est1 | .4477374
      est2 | .9144028
       est3 | .8166789
est4 | 1.028801
       est5 | 1.644439
. loocv reg y l(1).y if t>3
Leave-One-Out Cross-Validation Results
_____
Method | Value
Root Mean Squared Errors | 1.0506173
Mean Absolute Errors | .71878381
```

.20815721

					_
reg	У	1(1)	• y	if	t>3

Pseudo-R2

Source	SS	df	MS	Number of obs		27
26 1 3				1 (1/ 20)	=	9.41
	9.93272065		9.93272065		=	0.0051
Residual	26.4001087	25	1.05600435	- 1	=	0.2734
				Adj R-squared	l =	0.2443
Total	36.3328294	26	1.39741651	Root MSE	=	1.0276
у	Coef.	Std. Err.	t 	P> t [95% C	Conf.	Interval]

- 1	C 4 C 0 C 7 4	0100500	2 07	0.005	0105061	1 00140
I	.6469674					
_cons	.3563493	.2464536	1.45 	0.161	1512315 	.8639
estat ic						
kaike's infor	mation criter	ion and Ba	yesian inf	formation	criterion	
 Model	Obs	 ll(null)	 ll(model)	df	AIC	BIC
	 27 -					82.6077
					[R] BIC note	
	1 (2)		`			
crossioia re	eg y 1(2).y if RMSE	L/3 , K(3)			
'						
'	1.03154 1.343264					
	.9272938					
	1.330561					
est5	1.88108					
	od +-					
	red Errors Errors					
reg y 1(2).y	 , if +>3					
		df	MC	Numb	er of obs =	2
+	SS 			- F(1,	25) =	0.0
Model	.073995561	1	.07399556	1 Prob	> F =	0.823
Residual	36.2588338	25	1.4503533	5 R-sq	uared =	0.002
Total	36.3328294	26	1.3974165	1 Root	MSE =	1.204
у	Coef.				 [95% Conf.	 Interval
y						
	.0571522	.253027	0.23	0.823	4639667	.578271
_cons	.7695221	.2860344	2.69	0.013	.1804233	1.35862
estat ic						
kaike's infor	mation criter	ion and Ba	yesian inf	formation	criterion	

Model | Obs ll(null) ll(model) df

AIC BIC

27 -42.31929 -42.29176 2 88.58353 91.1752 . | Note: N=Obs used in calculating BIC; see [R] BIC note. . crossfold reg y 1(3).y if t>3, k(5)| RMSE est1 | 1.009151 est2 | 2.343802 est3 | 1.13863 est4 | .5679428 est5 | 1.529219 . loocv reg y 1(3).y if t>3Leave-One-Out Cross-Validation Results Method Value Root Mean Squared Errors | 1.4733996 Mean Absolute Errors | 1.0713477 .61797207 Pseudo-R2 . reg y 1(3).y if t>3 ----- Adj R-squared = -0.0351 Total | 36.3328294 26 1.39741651 Root MSE =

У	Coef.	Std. Err.		P> t	[95% Conf.	Interval]
у 13.		.2571441	-0.34	0.735	6176241	.4415722
_cons	.8671521	.2899284	2.99	0.006	.2700335	1.464271

. estat ic

Akaike's information criterion and Bayesian information criterion

Model	Obs	,	ll(model)		AIC	BIC
.	27	-42.31929	-42.25615	2	88.51231	91.10398

Note: N=Obs used in calculating BIC; see [R] BIC note.

. crossfold reg y 1(1,2).y if t>3 , k(5)

| RMSE ----est1 | .6746962 est2 | 1.563159 est3 | .7311557 est4 | 1.007034 est5 | 1.523913

. loocv reg y 1(1,2).y if t>3

Leave-One-Out Cross-Validation Results

Method		Value
Root Mean Squared Errors Mean Absolute Errors Pseudo-R2		1.0986212 .83345356 .19030643

. reg y 1(1,2).y if t>3

Source	SS	df	MS	Number of ob	s =	27 6.10
Model Residual	12.2396339 24.0931955	2 24	6.11981694 1.00388315	F(2, 24) Prob > F R-squared	=	0.0072 0.3369
Total	36.3328294	26	1.39741651	Adj R-square Root MSE	a = =	0.2816
у	Coef.	Std. Err.	t 1	P> t [95%	 Conf.	Interval]
y L1. L2.	.8283623 3691902	.2379548		0.002 .3372 0.1438718		1.319477
cons	.4744731	.2526132	1.88 (0.0730468	949	.995841

. estat ic

Akaike's information criterion and Bayesian information criterion

Model	Obs ll(null)	ll(model)	df	AIC	BIC
.	27 -42.31929	-36.77359	3	79.54718	83.43469

Note: N=Obs used in calculating BIC; see [R] BIC note.

. crossfold reg y 1(1,3).y if t>3, k(5)

	RMSE
- 1 -	
	.9226874
	1.217376
	1.108357
	1.757104
	.8038165
	 -+-

. loocv reg y 1(1,3).y if t>3

Leave-One-Out Cross-Validation Results

Method		Value
Root Mean Squared Errors Mean Absolute Errors Pseudo-R2		1.2125174 .86848364 .08552092

. reg y 1(1,3).y if t>3

Source	SS	df	MS	Numb	er of obs	=	27
+				- F(2,	24)	=	5.32
Model	11.1555742	2	5.577787	1 Prob	> F	=	0.0123
Residual	25.1772552	24	1.0490523	3 R-sq	uared	=	0.3070
+				- Adj	R-squared	=	0.2493
Total	36.3328294	26	1.3974165	1 Root	MSE	=	1.0242
у		Std. Err.		P> t	-	nf.	Interval]
v							
L1.	.6962714	.2151573	3.24	0.004	.252208	6	1.140334
L3.	2419406	.2240888	-1.08	0.291	704437	2	.2205559

. estat ic

Akaike's information criterion and Bayesian information criterion

Model	Obs ll(null)	ll(model)	df	AIC	BIC
.	27 -42.31929	-37.36775	3	80.73549	84.623

Note: N=Obs used in calculating BIC; see [R] BIC note.

_cons | .4862484 .2735234 1.78 0.088 -.0782762 1.050773

. crossfold reg y 1(2,3).y if t>3, k(5)

		RMSE
	-+-	
est1		1.19416
est2		1.132158
est3		.788579
est4		1.710587
est5		2.293407

. loocv reg y 1(2,3).y if t>3

Leave-One-Out Cross-Validation Results

Method	Ţ	Value
	+-	
Root Mean Squared Errors		1.5206717
Mean Absolute Errors		1.1282346
Pseudo-R2		.50131148

. reg y 1(2,3).y if t>3

Model Residual	.552294111 35.7805353	2 24	.276147055 1.49085564	Pro R-s	, 24) bb > F quared R-squared	= = = =	0.19 0.8321 0.0152 -0.0669
Total	36.3328294	26	1.39741651	Roc	t MSE	=	1.221
у			t 		[95% Cc	onf.	Interval]
y L2. L3.	.1584684 1802594	.3127402		0.617 0.576	486995 837091		.8039325 .4765729
_cons	.8247929	.3059778	2.70	0.013	.193285	57	1.4563

Akaike's information criterion and Bayesian information criterion

Model	Obs	ll(null)	11 (model)	df	AIC	BIC
.	27	-42.31929	-42.1125	3	90.22499	94.11251

Note: N=Obs used in calculating BIC; see [R] BIC note.

. crossfold reg y 1(1/3).y if t>3 , k(5)

. loocv reg y 1(1/3).y if t>3

Leave-One-Out Cross-Validation Results

Method	1	Value
	+-	
Root Mean Squared Errors		1.2397812
Mean Absolute Errors		.93246793
Pseudo-R2		.08044492

. reg y 1(1/3).y if t>3

Source	SS	df	MS	Number of obs $F(3, 23)$; = =	27 3.94
Model	12.3437179 23.9891114	3	4.11457264 1.04300485	Prob > F	=	
	36.3328294		1.39741651	Adj R-squared Root MSE	l = =	0.2536 1.0213
у	Coef.	Std. Err.		 P> t [95% C	onf.	Interval]

L1.	.8201704	.2439294	3.36	0.003	.315564	1.324777
L2.	3174417	.2974217	-1.07	0.297	9327052	.2978219
L3.	0845685	.2677069	-0.32	0.755	6383624	.4692255
_cons	.5033211	.2732026	1.84	0.078	0618415	1.068484

Akaike's information criterion and Bayesian information criterion

Model	Obs ll(null)	ll(model)	df	AIC	BIC
.	27 -42.31929	-36.71514	4	81.43029	86.61363

Note: N=Obs used in calculating BIC; see [R] BIC note.

•

. *Question 1b

. clear

. set obs 300 number of observations (_N) was 0, now 300 $\,$

. gen t=[n]

. tsset t

time variable: t, 1 to 300 delta: 1 unit

. gen r=rnormal()

. gen y=r if t<4
(297 missing values generated)</pre>

. replace y=0.5+0.5*1.y-0.1*12.y+0.25*13.y+r if t>=4 (297 real changes made)

. drop r

•

. crossfold reg y l(1).y if t>3, k(10)

. loocv reg y l(1).y if t>3

Leave-One-Out Cross-Validation Results

Method		Value
Root Mean Squared Errors Mean Absolute Errors Pseudo-R2	 	.99349549 .81452099 .1658777

. reg y 1(1).y if t>3

Source	SS	df	MS	Number o		251
Model Residual	60.5416388	1 295	60.5416388 .983173043		d =	0.0000
Total		296	1.18438408			
у	Coef.	Std. Err.	t	P> t [95% Conf.	Interval]
y L1.	.4140138	.0527597			3101806	.517847
_cons	.8642433	.0963091	8.97	0.000 .	6747034	1.053783

. estat ic

Akaike's information criterion and Bayesian information criterion

Model	Obs ll(null)	ll(model)	df	AIC	BIC
.	297 -446.0535	-417.9013	2 8	339.8026	847.1901

Note: N=Obs used in calculating BIC; see [R] BIC note.

. crossfold reg y 1(2).y if t>3 , k(10)

	1	RMSE
	-+-	
est1		.9484897
est2		1.064733
est3		.9910389
est4		1.091403
est5		1.027012
est6		1.024085
est7		1.062075
est8		1.078493
est9		1.027734
est10		1.422612

. loocv reg y 1(2).y if t>3

Leave-One-Out Cross-Validation Results

Method	Value
Root Mean Squared Errors	•
Mean Absolute Errors	.88628806

. reg y 1(2).y if t>3

Source	SS	df	MS	Number c		297 7.88
Model Residual	9.12416691 341.45352	1 295	9.12416691 1.15746956	Prob > F	ed =	0.0053 0.0260
Total	350.577686	296	1.18438408	-	=	
у	Coef.	Std. Err.	t	-	95% Conf.	Interval]
y L2.	.1603387	.057108	2.81	0.005 .	0479481	.2727294
_cons	1.235926	.1042415	11.00			1.4410//

. estat ic

Akaike's information criterion and Bayesian information criterion

Model	Obs ll(null)	ll(model)	df	AIC	BIC
.	297 -446.0535	-442.1374	2	888.2749	895.6623

Note: N=Obs used in calculating BIC; see [R] BIC note.

. crossfold reg y l(3).y if t>3, k(10)

	-	RMSE
est.1	1	.9204688
esti	- 1	. 9204000
est2		1.229033
est3		1.139943
est4		1.081345
est5		.8443094
est6		.9478467
est7		.9490367
est8		1.187149
est9		1.090669
est10		1.070583

. loocv reg y 1(3).y if t>3

Leave-One-Out Cross-Validation Results

Method	Value
Root Mean Squared Errors	1.0596216
Mean Absolute Errors	.84556019
Pseudo-R2	.05520804

. reg y 1(3).y if t>3

Model Residual	23.604574 326.973112	1 295	23.604574	F(1, 295) Prob > F R-squared Adj R-square	= = = d =	21.30 0.0000 0.0673 0.0642
Total	350.577686	296	1.18438408	Root MSE	=	1.0528
у	Coef.	Std. Err.	t F	°> t [95%	Conf.	Interval]
y L3.	.2577714	.0558575	4.61 0	.000 .1478	417	.3677011
_cons	1.093973	.1018939	10.74 0	.000 .8934	415	1.294504

Akaike's information criterion and Bayesian information criterion

Model	Obs ll(null)	ll(model)	df	AIC	BIC
.	297 -446.0535	-435.7024	2	875.4048	882.7923

Note: N=Obs used in calculating BIC; see [R] BIC note.

. crossfold reg y l(1,2).y if t>3 , k(10)

		RMSE
	- -	0.00000
est1		.8753155
est2		1.079413
est3		.9750601
est4		.9606736
est5		.9878336
est6		.9170026
est7		.9686407
est8		.9939085
est9		1.064011
est10		1.142216

. loocv reg y 1(1,2).y if t>3

Leave-One-Out Cross-Validation Results

Method	!	Value
Root Mean Squared Errors Mean Absolute Errors Pseudo-R2	 	.9981137 .81971413 .16107318

. reg y 1(1,2).y if t>3

	Source	SS	df	MS	Number of obs	=	297
_	+-				F(2, 294)	=	30.74
	Model	60.6241574	2	30.3120787	Prob > F	=	0.0000
	Residual	289.953529	294	.986236494	R-squared	=	0.1729
_	+-				Adj R-squared	=	0.1673
	Total	350.577686	296	1.18438408	Root MSE	=	.99309

у	Coef.	Std. Err.	t	P> t	[95% Conf.	. Interval]
y L1. L2.			7.23 -0.29	0.000 0.773	.3064298 131233	.5358146 .0976001
_cons	.8784199	.1081959	8.12	0.000	.6654833	1.091357

Akaike's information criterion and Bayesian information criterion

Model	Obs ll(null)	ll(model)	df	AIC	BIC
.	297 -446.0535	-417.859	3	841.7181	852.7993

Note: N=Obs used in calculating BIC; see [R] BIC note.

. crossfold reg y l(1,3).y if t>3 , k(10)

		RMSE
est1 est2 est3 est4 est5 est6 est7 est8 est9	-+· 	.8379843 1.119864 .9830091 1.027417 .9641554 .8945947 1.130843 .8870858 .9412586

. loocv reg y 1(1,3).y if t>3

Leave-One-Out Cross-Validation Results

Method		Value
Root Mean Squared Errors Mean Absolute Errors Pseudo-R2	 	.97580523 .78232156 .1985535

. reg y 1(1,3).y if t>3

Source	SS	df	MS	Number of obs $F(2, 294)$	=	297 39.17
Model Residual 	73.7635366 276.81415	2 294 	36.8817683 .941544728 1.18438408	Prob > F R-squared Adj R-squared Root MSE	= =	0.0000 0.2104 0.2050 .97033
у	Coef.	Std. Err.		P> t [95% Co	onf.	Interval]
y L1.	.3819759	.0523338		0.000 .27897	95	.4849722

L3.	.1955498	.0521832	3.75	0.000	.0928498	.2982499
_cons	.6256466	.1137394	5.50	0.000	.4018001	.8494931

Akaike's information criterion and Bayesian information criterion

Model	Obs ll(null)	ll(model)	df	AIC	BIC
.	297 -446.0535	-410.9725	3 	827.9449	839.0261

Note: N=Obs used in calculating BIC; see [R] BIC note.

. crossfold reg y 1(2,3).y if t>3 , k(10)

	1	RMSE
	-+-	
est1		.9761117
est2		1.134196
est3		1.220372
est4		1.025265
est5		.9111706
est6		.950201
est7		1.096652
est8		1.072383
est9		1.130017
est10		1.036515

. loocv reg y 1(2,3).y if t>3

Leave-One-Out Cross-Validation Results

Method		Value
	+-	
Root Mean Squared Errors		1.0602008
Mean Absolute Errors		.8499411
Pseudo-R2		.05470602

. reg y 1(2,3).y if t>3

Source	SS	df	MS	Number of obs F(2, 294)	=	297 11.16
Model Residual	24.7384365 325.83925	2 294	12.3692183 1.10829677	Prob > F R-squared	=	0.0000
Total	350.577686	296	1.18438408	Adj R-squared Root MSE	=	1.0528
У	•	Std. Err.	t F		 onf.	Interval]
У L2.	1 .0623826	.0616753	1.01 0	.31305899	85	.1837637
L3.		.0616462	3.75 0	.11006	33	.3527109

Akaike's information criterion and Bayesian information criterion

Model	Obs ll(null)	ll(model)	df	AIC	BIC
.	297 -446.0535	-435.1865	3	876.3731 	887.4543

Note: N=Obs used in calculating BIC; see [R] BIC note.

. crossfold reg y 1(1/3).y if t>3 , k(10)

	1	RMSE
	-+-	
est1		.8134784
est2		.9355836
est3		.9581936
est4		1.063685
est5		.8773045
est6		.8814801
est7		1.296819
est8		1.046323
est9		.9691692
est10		.7505918

. loocv reg y 1(1/3).y if t>3

Leave-One-Out Cross-Validation Results

·	
Root Mean Squared Errors .973312 Mean Absolute Errors .777557 Pseudo-R2 .202824	733

. reg y 1(1/3).y if t>3

Source	SS	df	MS	Number of obs	; = =	297 27.63
Model Residual	77.3017268 273.27596	3 293	25.7672423 .932682456	Prob > F R-squared Adj R-squared	=	0.0000 0.2205 0.2125
Total	350.577686	296	1.18438408	Root MSE	=	.96575
У	Coef.	Std. Err.	t P	?> t [95% C	conf.	Interval]
у L1. L2.	.425519	.0566819 .0615696 .0565612	-1.95 0	.000 .31396 .05224109 .000 .12785	43	.5370743 .0012552 .3504937
L3.	.2391761 .6735114	.1158396		.000 .44552		.9014945

. estat ic

Akaike's information criterion and Bayesian information criterion

```
Obs ll(null) ll(model)
                                            df
     Model |
                                                       AIC
_______
                  297 -446.0535 -409.0621
                                             4
                                                  826.1242 840.8992
______
            Note: N=Obs used in calculating BIC; see [R] BIC note.
. *Question 2
. clear
. set more off
. ** data prep
. import delimited using "us and florida economic time series.txt"
(5 vars, 972 obs)
. rename observation date datestring
. gen dateday=date(datestring,"YMD")
. gen date=mofd(dateday)
. format date %tm
. tsset date
      time variable: date, 1939m1 to 2019m12
             delta: 1 month
. generate month=month(dateday)
. keep if tin(1990m1, 2019m12)
(612 observations deleted)
. rename flbppriv fl bp
. rename fllfn fl_lf
. rename flnan fl_nonfarm
. rename lnu02300000 20200110 us epr
. gen lnflnonfarm=ln( fl nonfarm)
. gen lnfllf=ln( fl lf)
. gen lnusepr = ln(us epr)
. gen lnflbp=ln( fl bp)
. *fit and evaluate models
. *Note I restrict estimation to year>1989 so the same observations
. *Are compared for all models
. crossfold reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/12)d.lnfllf ///
        1(0/12) d.lnusepr 1(0/12) d.lnflbp i.month date if tin(1991m1,2019m12) , k(10)
```

RMSE

```
est1 | .004854

est2 | .0028465

est3 | .0048683

est4 | .0034981

est5 | .0030781

est6 | .0040035

est7 | .0037033

est8 | .0036501

est9 | .0039859

est10 | .0033776
```

. loocv reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/12)d.lnfllf /// 1(0/12)d.lnusepr 1(0/12)d.lnflbp i.month date

Leave-One-Out Cross-Validation Results

Method		Value
Root Mean Squared Errors Mean Absolute Errors Pseudo-R2		.00389888 .00288576 .8474397

. reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/12)d.lnfllf /// > 1(0/12)d.lnusepr 1(0/12)d.lnflbp i.month date

Source	SS	df	MS		per of obs 3, 283)	=	347 42.26
Model	.030831254	63	.00048938		b > F	=	0.0000
Residual	.003277128	283	.0000115		quared	=	0.9039
	+				R-squared	=	0.8825
Total	.034108382	346	.00009857	9 Roo	t MSE	=	.0034
D.	 						
lnflnonfarm	Coef.	Std. Err.	t	P> t	[95% Con	f.	<pre>Interval]</pre>
lnflnonfarm	 						
LD.	1574149	.0580953	-2.71	0.007	2717687		0430611
L2D.	1574789	.0594298	-2.65	0.009	2744594		0404983
L3D.	.1395365	.0599708	2.33	0.021	.021491		.2575819
L4D.	.1302252	.0601073	2.17	0.031	.0119111		.2485394
L5D.	.036204	.0614863	0.59	0.556	0848245		.1572325
L6D.	.0953011	.0634205	1.50	0.134	0295346		.2201369
L7D.	.0117236	.0621461	0.19	0.851	1106038		.1340509
L8D.	0508451	.0615834	-0.83	0.410	1720649		.0703746
L9D.	.0859865	.0595028	1.45	0.150	0311377		.2031106
L10D.	207794	.0584651	-3.55	0.000	3228756		0927124
L11D.	0434531	.0578108	-0.75	0.453	1572468		.0703406
L12D.	.3097061	.056476	5.48	0.000	.1985398		.4208723
lnfllf	 						
D1.	.1062022	.1000859	1.06	0.290	090805		.3032094
LD.	1178716	.1004531	-1.17	0.242	3156017		.0798584
L2D.	1274227	.1028799	-1.24	0.217	3299297		.0750842
L3D.	1037616	.1043474	-0.99	0.321	3091571		.1016339
L4D.	0764095	.1050053	-0.73	0.467	2831		.1302811
L5D.	0037077	.1037939	-0.04	0.972	2080138		.2005984
L6D.	05965	.1041819	-0.57	0.567	2647198		.1454198
L7D.	0006012	.1042636	-0.01	0.995	2058318		.2046294
L8D.	.0146311	.1043507	0.14	0.889	1907709		.2200331

L9D.		.1037189	1.71	0.088	0263885	.3819283
L10D.		.1068988	2.03	0.043	.0069375	.4277728
L11D.		.1069936	0.18	0.859	1915499	.2296585
L12D.		.1057575	-1.56	0.120	372972	.0433703
lnusepr D1. LD. L2D. L3D. L4D. L5D. L6D. L7D. L8D. L9D. L10D. L11D. L12D.	203726	.1354639	-1.50	0.134	4703707	.0629187
	.2001192	.1371366	1.46	0.146	069818	.4700563
	.0790793	.1387443	0.57	0.569	1940226	.3521811
	.1610994	.1376737	1.17	0.243	109895	.4320939
	.2234233	.1391924	1.61	0.110	0505605	.497407
	02279	.1364623	-0.17	0.867	2914	.24582
	.2954135	.1352384	2.18	0.030	.0292126	.5616144
	.1286871	.1367504	0.94	0.347	1404899	.3978641
	.0747586	.1381187	-0.54	0.589	346629	.1971119
	1316723	.1391333	-0.95	0.345	4055396	.1421951
	2945532	.1419814	-2.07	0.039	5740268	0150796
	.1266267	.1410913	0.90	0.370	1510948	.4043483
	.224092	.1373989	1.63	0.104	0463615	.4945455
lnflbp	 .0044437 .0039814 .0064503 .0068853 .0051008 .0046621 .004366 .0038368 .003177 .0028721 .0033299 .0035077 .0031535	.0016422 .0019214 .0020021 .0020155 .0020277 .0020346 .0020655 .0020881 .0021186 .0021085 .0020848 .0019783 .0016353	2.71 2.07 3.22 3.42 2.52 2.29 2.11 1.84 1.50 1.36 1.60 1.77 1.93	0.007 0.039 0.001 0.001 0.012 0.023 0.035 0.067 0.135 0.174 0.111 0.077 0.055	.0012112 .0001994 .0025094 .002918 .0011094 .0006572 .0003003 0002733 0002733 0012782 0007738 0003864 0000653	.0076761 .0077635 .0103913 .0108525 .0090921 .008667 .0084316 .0079469 .0073473 .0070224 .0074337
month 2 3 4 5 6 7 8 9 10 11	 .0112254 .009569 .0103909 .0053637 0019222 .0044899 .013014 .0109015 .0179031 .0109558 .0157008	.0039079 .0040687 .0044854 .0034821 .0041834 .0040453 .0040663 .0034322 .004553 .0040924 .0037763	2.87 2.35 2.32 1.54 -0.46 1.11 3.20 3.18 3.93 2.68 4.16	0.004 0.019 0.021 0.125 0.646 0.268 0.002 0.002 0.002 0.000 0.008	.0035332 .0015602 .001562 0014904 0101568 0034728 .0050099 .0041457 .0089411 .0029005 .0082677	.0189175 .0175777 .0192198 .0122179 .0063124 .0124526 .021018 .0176574 .0268652 .0190112 .0231339
date	-3.71e-06	2.01e-06	-1.85	0.066	-7.67e-06	2.43e-07
_cons	0054646	.003224	-1.69	0.091	0118107	.0008814

Akaike's information criterion and Bayesian information criterion

Model	Obs	ll(null)	11 (model)	df	AIC	BIC
.	347	1108.606	1515.043	64	-2902.086	-2655.729

Note: N=Obs used in calculating BIC; see [R] BIC note.

. crossfold reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/2)d.lnfllf ///
> 1(0/2)d.lnusepr 1(0/2)d.lnflbp i.month date , k(10)

		RMSE
	-+	
est1		.0033667
est2		.0040457
est3		.0053336
est4		.0033043
est5		.0034043
est6		.0032772
est7		.0033094
est8		.0035109
est9		.0048244
est10		.0029423

. loocv reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/2)d.lnfllf /// > 1(0/2)d.lnusepr 1(0/2)d.lnflbp i.month date

Leave-One-Out Cross-Validation Results

. reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/2)d.lnfllf 1(0/2)d.lnusepr /// > 1(0/2)d.lnflbp i.month date

Source	SS	df	MS	Number of obs	=	347
+-				F(33, 313)	=	74.27
Model	.030245856	33	.000916541	Prob > F	=	0.0000
Residual	.003862526	313	.00001234	R-squared	=	0.8868
+-				Adj R-squared	=	0.8748
Total	.034108382	346	.000098579	Root MSE	=	.00351

D.		Q1 1 F		D> 1.1.1	505° a 6	T
lnflnonfarm	Coei.	Sta. Err.	t 	P> t 	[95% Conf.	Interval
lnflnonfarm						
LD.	0926596	.0539475	-1.72	0.087	1988052	.0134861
L2D.	1013494	.0545464	-1.86	0.064	2086733	.0059745
L3D.	.2257017	.0538977	4.19	0.000	.1196541	.3317493
L4D.	.1584207	.0515924	3.07	0.002	.0569089	.2599325
L5D.	.1456743	.0524817	2.78	0.006	.0424128	.2489358
L6D.	.140864	.0550022	2.56	0.011	.0326431	.2490849
L7D.	.0853223	.0547739	1.56	0.120	0224494	.1930939
L8D.	.0145033	.0536544	0.27	0.787	0910657	.1200722
L9D.	.1028963	.0513696	2.00	0.046	.0018229	.2039698
L10D.	1763997	.0503805	-3.50	0.001	2755269	0772724
L11D.	0750815	.0518778	-1.45	0.149	1771547	.0269918
L12D.	.3597356	.0514983	6.99	0.000	.2584091	.4610622
lnfllf						
D1.	.0215757	.0964236	0.22	0.823	1681446	.2112961
LD.	1651507	.096038	-1.72	0.086	3541123	.023811
L2D.	141473	.0978523	-1.45	0.149	3340044	.0510585

Inusepr						
D1.	1612876	.1291331	-1.25	0.213	4153664	.0927912
LD.	.1960407	.1287356	1.52	0.129	0572559	.4493373
L2D.	.0613202	.1305917	0.47	0.639	1956284	.3182687
i						
lnflbp						
D1.	.00441	.0015488	2.85	0.005	.0013625	.0074574
LD.	.0029326	.0017603	1.67	0.097	000531	.0063961
L2D.	.0034816	.0015737	2.21	0.028	.0003853	.0065779
month						
2	.0125423	.0028953	4.33	0.000	.0068455	.018239
3	.0092578	.0031904	2.90	0.004	.0029805	.015535
4	.0110159	.0032953	3.34	0.001	.0045322	.0174996
5	.0044171	.0029959	1.47	0.141	0014776	.0103118
6	.0027434	.003072	0.89	0.373	0033008	.0087877
7	.0039961	.0027956	1.43	0.154	0015045	.0094967
8	.0134171	.002811	4.77	0.000	.0078862	.0189479
9	.016115	.0027803	5.80	0.000	.0106446	.0215854
10	.0235632	.0032462	7.26	0.000	.017176	.0299504
11	.0153235	.0029086	5.27	0.000	.0096006	.0210463
12	.0142819	.0023467	6.09	0.000	.0096647	.0188992
Ì						
date	-1.94e-06	1.93e-06	-1.01	0.315	-5.74e-06	1.85e-06
_cons	0087229	.0025667	-3.40	0.001	0137731	0036728

Akaike's information criterion and Bayesian information criterion

Model	Obs	ll(null)	ll(model)	df	AIC	BIC
.	347	1108.606	1486.527	34 	-2905.055	-2774.178

Note: N=Obs used in calculating BIC; see [R] BIC note.

. crossfold reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/2,12)d.lnfllf /// > 1(0/2,12)d.lnflbp i.month date , k(10)

	1	RMSE
	-+	
est1		.002985
est2		.002963
est3		.0028823
est4		.003077
est5		.0046443
est6		.0056406
est7		.0035519
est8		.0027221
est9		.0033546
est10	1	.0048672

. loocv reg d.lnflnonfarm 1(1/12) d.lnflnonfarm 1(0/2,12) d.lnfllf /// > 1(0/2,12) d.lnflbp i.month date

Leave-One-Out	Cross-Validation Results
Method	l Value

. reg d.lnflnonfarm 1(1/12)d.lnflnonfarm 1(0/2,12)d.lnfllf /// > 1(0/2,12)d.lnflbp i.month date

Source	SS	df	MS			= 347 = 77.19
Model Residual	.030261308	32 314	.00094566	66 Prob 52 R-sc	> F quared	= 0.0000 = 0.8872
Total	.034108382	346	.00009857	_		= 0.8757 = .0035
D.	 I					
lnflnonfarm	Coef.	Std. Err.	t 	P> t	[95% Conf	. Interval]
lnflnonfarm						
LD.		.0518028	-2.04	0.042	2075438	0036948
L2D.	0921418	.052223	-1.76	0.079	1948931	.0106095
L3D.	.2294649	.0513697	4.47	0.000	.1283925	.3305372
L4D.	.1576544	.0512838	3.07	0.002	.0567511	.2585577
L5D.	.1465611	.0519056	2.82	0.005	.0444345	.2486878
L6D.	.1448653	.0541136	2.68	0.008	.0383941	.2513364
L7D.	.0950336	.0533402	1.78	0.076	0099158	.1999831
L8D.	.0395854	.05266	0.75	0.453	0640258	.1431965
L9D.	.1072418	.0508503	2.11	0.036	.0071915	.2072922
L10D.	1768537	.0498853	-3.55	0.000	2750055	0787019
L11D.	0876855	.0512101	-1.71 6.91	0.088	1884439 .2527984	.0130729
L12D.	.3534041	.0511326	6.91	0.000	.232/984	.4540099
lnfllf	 					
D1.	 056149	.0556547	-1.01	0.314	1656523	.0533543
LD.	0363912	.0539611	-0.67	0.501	1425623	.0535343
L2D.	1069649	.0537053	-1.99	0.047	2126327	0012972
L12D.	1056342	.0539506	-1.96	0.051	2117847	.0005162
1120.		.000000	1.50	0.001	.211/01/	.0000102
lnflbp						
D1.	.0040897	.0015363	2.66	0.008	.0010669	.0071125
LD.	.003202	.0017507	1.83	0.068	0002426	.0066466
L2D.	.0032357	.0015535	2.08	0.038	.000179	.0062923
L12D.	.0014876	.0013168	1.13	0.259	0011032	.0040784
month	 					
2	.0098277	.0021732	4.52	0.000	.0055519	.0141036
3	.0085637	.0025121	3.41	0.001	.003621	.0135064
4	.0096017	.0028128	3.41	0.001	.0040674	.015136
5	.004879	.0027898	1.75	0.081	0006101	.010368
6	.0009795	.0025176	0.39	0.697	003974	.0059329
7	.0038837	.0022606	1.72	0.087	0005642	.0083316
8	.0127237	.002469	5.15	0.000	.0078659	.0175815
9	.0160257	.0026769	5.99	0.000	.0107587	.0212926
10	.0209295	.0026518	7.89	0.000	.015712	.0261471
11	.0147903	.0024106	6.14	0.000	.0100473	.0195333
12	.0137729	.0019983	6.89	0.000	.0098412	.0177046
date	 -1.98e-06	1.89e-06	-1.05	0.295	-5.71e-06	1.74e-06
cons	0078223	.0019928	-3.93	0.000	0117433	0039013

Akaike's information criterion and Bayesian information criterion

Model	Obs	ll(null)	ll(model)	df	AIC	BIC
.	347 	1108.606	1487.223	33 	-2908.446	-2781.418

Note: N=Obs used in calculating BIC; see [R] BIC note.

. crossfold reg d.lnflnonfarm 1(1/12,24)d.lnflnonfarm 1(1/2,12,24)d.lnfllf ///1(1/2,12,24) d.lnusepr i.month , k(10)

	1	RMSE
est1 est2 est3 est4 est5 est6 est7 est8 est9	-+-	.0050724 .0042645 .0033856 .0030582 .0019545 .0035341 .0053942 .0037919 .0033725
00010	- 1	

. loocv reg d.lnflnonfarm 1(1/12,24)d.lnflnonfarm 1(1/2,12,24)d.lnfllf // 1(1/2,12,24) d.lnusepr i.month

Leave-One-Out Cross-Validation Results

Method		Value
Root Mean Squared Errors Mean Absolute Errors Pseudo-R2	 	.00370499 .00265047 .86035869

. reg d.lnflnonfarm 1(1/12,24) d.lnflnonfarm 1(1/2,12,24) d.lnfllf ///1(1/2,12,24) d.lnusepr i.month

Source	SS	df	MS	Number of obs	=	335
Model Residual	.029334712	32 302	.00091671	F(32, 302) Prob > F R-squared	=	74.38 0.0000 0.8874
Total	.033056529	334	.000098972	Adj R-squared Root MSE	=	0.8755
D. lnflnonfarm	Coef.	Std. Err.	t P>	> t [95% C	 onf.	Interval]
<pre>Inflnonfarm </pre>	0893296 0665814 .2241205 .1514706 .1580576 .1598189 .0713947	.0527122 .0558162 .0541419 .0521838 .0532447 .055306 .0554199	-1.19 0. 4.14 0. 2.90 0. 2.97 0. 2.89 0. 1.29 0.	.09119305 .23417641 .000 .11757 .004 .04878 .003 .05328 .004 .0509 .19903766 .71608788	94 75 06 01 85 35	.0144 .0432566 .3306636 .2541605 .2628351 .2686528 .1804528

L9D. L10D. L11D. L12D. L24D.	.076253 1636484 0908334 .3044588 .1201683	.0525006 .0515785 .0531678 .0598043 .05152	1.45 -3.17 -1.71 5.09 2.33	0.147 0.002 0.089 0.000 0.020	0270605 2651471 1954596 .1867729 .0187846	.1795664 0621497 .0137928 .4221446 .2215519
lnfllf	 1225263	.0962344	-1.27	0.204	3119012	.0668486
L2D. L12D. L24D.	1611748 1592969 .1558841	.0968689 .1014979 .0998312	-1.66 -1.57 1.56	0.097 0.118 0.119	3517982 3590296 0405688	.0294486 .0404358 .352337
lnusepr	.1330011	.0330312	1.00	0.113	.0103000	.332337
LD.	.159602	.1313701	1.21	0.225	0989147	.4181187
L2D.	.0849591	.1315279	0.65	0.519	1738682	.3437864
L12D.	.1076075	.1327893	0.81	0.418	1537019	.368917
L24D.	330308	.1254379	-2.63	0.009	577151	0834651
month						
2	.012446	.0031029	4.01	0.000	.00634	.018552
3	.0108327	.0033014	3.28	0.001	.004336	.0173294
4	.0111764	.0035758	3.13	0.002	.0041397	.018213
5	.0053687	.0030705	1.75	0.081	0006736	.0114109
6	.0041449	.0033227	1.25	0.213	0023936	.0106835
7	.006134	.0027733	2.21	0.028	.0006765	.0115915
8	.0129795	.0027654	4.69	0.000	.0075375	.0184214
9	.0159311	.0027644	5.76	0.000	.0104912	.0213709
10	.0227959	.0035244	6.47	0.000	.0158605	.0297313
11	.0145436	.003114	4.67	0.000	.0084157	.0206715
12	.0140586	.002605	5.40	0.000	.0089323	.019185
_cons	 0102486 	.002402	-4.27	0.000	0149754	0055219

Akaike's information criterion and Bayesian information criterion

Model	Obs	 ll(null)	11 (model)	df	AIC	BIC
.	335	1069.62	1435.441	33	-2804.882	-2679.016

Note: N=Obs used in calculating BIC; see [R] BIC note.

. log close

name: <unnamed>

log: C:\Users\jdewey\Documents\A S20 Time Series\Problem Sets\Problem Set 3

Work.smcl

log type: smcl

closed on: 17 Feb 2020, 10:09:33
