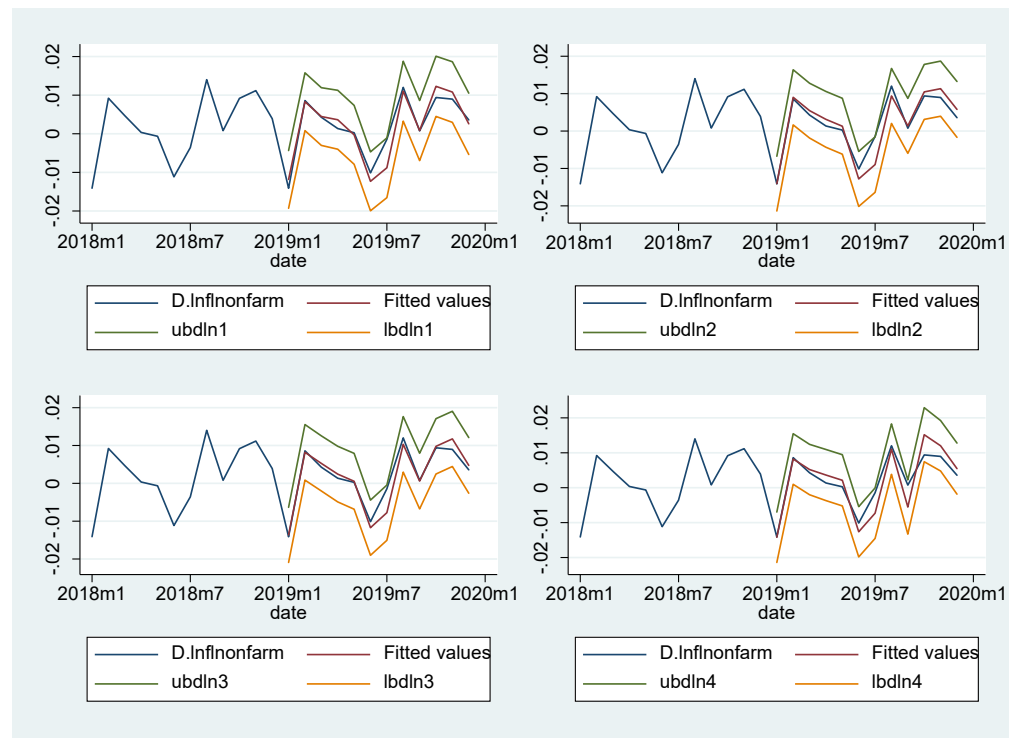
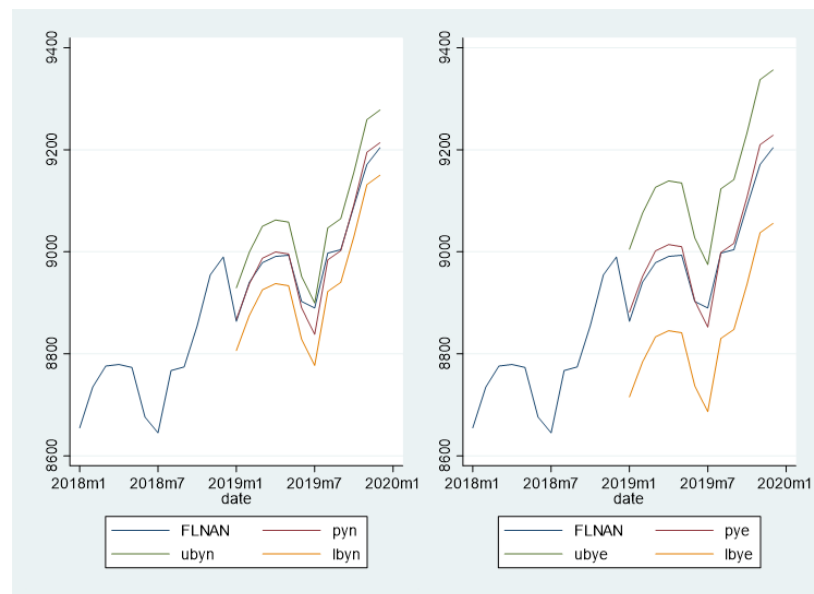


The table shows the fit statistics for the models. Model 3 is first on three criteria and second on two. Looking at the visuals below, model 3 is at least as good as the others at matching the overall pattern over the past two years.

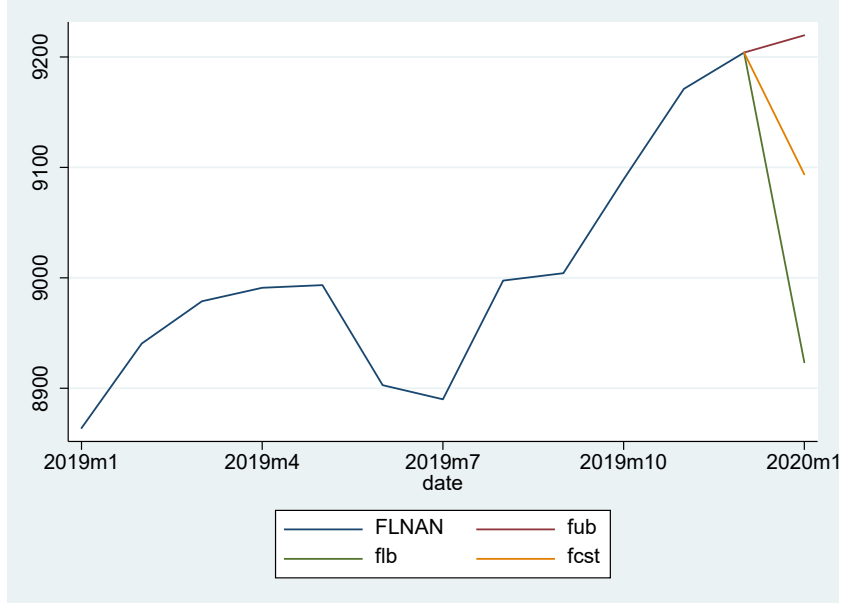
Model	# Vars	AIC	BIC	10-Fold RMSE	N-Fold RMSE	Out Smpl RMSE
1	60	-2897	-2662	0.00380	0.00390	0.00195
2	30	-2899	-2779	0.00376	0.00382	0.00126
3	30	-2903	-2784	0.00375	0.00381	0.00089
4	32	-2805	-2679	0.00374	0.00370	0.00216



The figure shows the point forecast and the normal and empirical forecast intervals.



The point forecast is 9093.688, and the empirical prediction interval is 8923.5 to 9219.6. This is shown in the figure below.



Appendix A: Do file for Problem Set 4

*Problem Set 4 Solution

```

clear
set more off
cd "C:\Users\jdewey\Documents\A S20 Time Series\Problem Sets\"
log using "Problem Set 4 Work", replace

** 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 flllfn fl_lf
rename flnan fl_nonfarm
rename lnu02300000_20200110 us_epr
gen lnflnonfarm=ln( fl_nonfarm)
gen lnflllf=ln( fl_lf)
gen lnusepr = ln(us_epr)
gen lnflbp=ln( fl_bp)

tsappend, add(1)
replace month=month(dofm(date)) if month==.

*fit and evaluate models
*Note I restricted estimation to year>1989 so the same observations are
* compared for all models by dropping earlier years above

*Model 1
set seed 22045 // to make sure the same folds are used for each model
crossfold reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/12)d.lnflllf ///
            l(1/12)d.lnusepr l(1/12)d.lnflbp i.month date , k(10)
scalar define k=10
matrix kSSE=r(est)'*r(est)
scalar krmse1=(e1(kSSE,1,1)/k)^.5
matrix drop kSSE
scalar drop k
loocv reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/12)d.lnflllf ///
        l(1/12)d.lnusepr l(1/12)d.lnflbp i.month date
scalar loormse1=r(rmse)
reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/12)d.lnflllf ///
    l(1/12)d.lnusepr l(1/12)d.lnflbp i.month date
estat ic
scalar aic1=(e1(r(S),1,5))
scalar bic1=(e1(r(S),1,6))
reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/12)d.lnflllf ///
    l(1/12)d.lnusepr l(1/12)d.lnflbp i.month date if tin(1991m1,2018m12)
scalar NVar1=e(df_m)
predict res, residual
predict pdln1
predict stdf1, stdf
gen ressq=res^2

```

```

summ ressq if tin(2018q1,2018q4)
scalar osrmse1=r(mean)^0.5
drop res ressq
gen ubdln1=pdln1+1.96*stdf1
gen lbdln1=pdln1-1.96*stdf1
twoway (tsline d.lnflnonfarm if tin(2018m1,2019m12)) ///
       (tsline pdln1 ubdln1 lbdln1 if tin(2019m1,2019m12) ) , ///
       saving(m1tslines, replace)

*Model 2
set seed 22045 // to make sure the same folds are used for each model
crossfold reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/2)d.lnfl1lf ///
          l(1/2)d.lnusepr l(1/2)d.lnflbp i.month date , k(10)
scalar define k=10
matrix kSSE=r(est)'*r(est)
scalar krmse2=(el(kSSE,1,1)/k)^.5
matrix drop kSSE
scalar drop k
loocv reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/2)d.lnfl1lf ///
        l(1/2)d.lnusepr l(1/2)d.lnflbp i.month date
scalar loormse2=r(rmse)
reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/2)d.lnfl1lf l(1/2)d.lnusepr ///
    l(1/2)d.lnflbp i.month date
estat ic
scalar aic2=(el(r(S),1,5))
scalar bic2=(el(r(S),1,6))
reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/2)d.lnfl1lf l(1/2)d.lnusepr ///
    l(1/2)d.lnflbp i.month date if tin(1991m1,2018m12)
scalar NVar2=e(df_m)
predict res, residual
predict pdln2
predict stdf2, stdf
gen ressq=res^2
summ ressq if tin(2018q1,2018q4)
scalar osrmse2=r(mean)^0.5
drop res ressq
gen ubdln2=pdln2+1.96*stdf2
gen lbdln2=pdln2-1.96*stdf2
twoway (tsline d.lnflnonfarm if tin(2018m1,2019m12)) ///
       (tsline pdln2 ubdln2 lbdln2 if tin(2019m1,2019m12) ) , ///
       saving(m2tslines, replace)

*Model 3
set seed 22045 // to make sure the same folds are used for each model
crossfold reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/2,12)d.lnfl1lf ///
          l(1/2,12)d.lnflbp i.month date , k(10)
scalar define k=10
matrix kSSE=r(est)'*r(est)
scalar krmse3=(el(kSSE,1,1)/k)^.5
matrix drop kSSE
scalar drop k
loocv reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/2,12)d.lnfl1lf ///
        l(1/2,12)d.lnflbp i.month date
scalar loormse3=r(rmse)
reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/2,12)d.lnfl1lf ///
    l(1/2,12)d.lnflbp i.month date
estat ic

```

```

scalar aic3=(el(r(S),1,5))
scalar bic3=(el(r(S),1,6))
reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/2,12)d.lnfl1f ///
    l(1/2,12)d.lnflbp i.month date if tin(1991m1,2018m12)
scalar NVar3=e(df_m)
predict res, residual
predict pdln3
predict stdf3, stdf
gen ressq=res^2
summ ressq if tin(2018q1,2018q4)
scalar osrmse3=r(mean)^0.5
drop res ressq
gen ubdln3=pdln3+1.96*stdf3
gen lbdln3=pdln3-1.96*stdf3
twoway (tsline d.lnflnonfarm if tin(2018m1,2019m12)) ///
    (tsline pdln3 ubdln3 lbdln3 if tin(2019m1,2019m12) ) , ///
    saving(m3tslines, replace)

*Model 4
set seed 22045 // to make sure the same folds are used for each model
crossfold reg d.lnflnonfarm l(1/12,24)d.lnflnonfarm l(1/2,12,24)d.lnfl1f ///
    l(1/2,12,24)d.lnusepr i.month , k(10)
scalar define k=10
matrix kSSE=r(est) '*r(est)
scalar krmse4=(el(kSSE,1,1)/k)^.5
matrix drop kSSE
scalar drop k
loocv reg d.lnflnonfarm l(1/12,24)d.lnflnonfarm l(1/2,12,24)d.lnfl1f ///
    l(1/2,12,24)d.lnusepr i.month
scalar loormse4=r(rmse)
reg d.lnflnonfarm l(1/12,24)d.lnflnonfarm l(1/2,12,24)d.lnfl1f ///
    l(1/2,12,24)d.lnusepr i.month
estat ic
scalar aic4=(el(r(S),1,5))
scalar bic4=(el(r(S),1,6))
reg d.lnflnonfarm l(1/12,24)d.lnflnonfarm l(1/2,12,24)d.lnfl1f ///
    l(1/2,12,24)d.lnusepr i.month if tin(1991m1,2018m12)
scalar NVar4=e(df_m)
predict res, residual
predict pdln4
predict stdf4, stdf
gen ressq=res^2
summ ressq if tin(2018q1,2018q4)
scalar osrmse4=r(mean)^0.5
drop res ressq
gen ubdln4=pdln4+1.96*stdf4
gen lbdln4=pdln4-1.96*stdf4
twoway (tsline d.lnflnonfarm if tin(2018m1,2019m12)) ///
    (tsline pdln4 ubdln4 lbdln4 if tin(2019m1,2019m12) ) ///
    , saving(m4tslines, replace)

matrix M1=(NVar1,aic1,bic1,krmse1,loormse1,osrmse1)
matrix M2=(NVar2,aic2,bic2,krmse2,loormse2,osrmse2)
matrix M3=(NVar3,aic3,bic3,krmse3,loormse3,osrmse3)
matrix M4=(NVar4,aic4,bic4,krmse4,loormse4,osrmse4)
matrix MStats=(M1\M2\M3\M4)

```

```

matrix colnames MStats=NVar AIC BIC RMSE10F RMSENF RMSEOS
matrix rownames MStats=Model1 Model2 Model3 Model4
matrix list MStats

graph combine m1tslines.gph m2tslines.gph m3tslines.gph m4tslines.gph , ///
    saving(mtslines, replace)

STOP

*Going to go with model 3

drop pdl* ub* lb* stdf*
scalar drop _all

reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/2,12)d.lnfl1f ///
    l(1/2,12)d.lnflbp i.month date
predict pdln
predict stdf
scalar rmse=e(rmse)
gen corrnorm=exp((rmse^2)/2)
predict res
gen expres=exp(res)
summ expres
gen corrempr=r(mean)

gen pyn=corrnorm*exp(l.lnflnon+pdln)
gen ubyn=corrnorm*exp(l.lnflnon+pdln+1.96*rmse)
gen lbyn=corrnorm*exp(l.lnflnon+pdln-1.96*rmse)
twoway (tsline fl_nonfarm if tin(2018m1,2019m12)) ///
    (tsline pyn ubyn lbyn if tin(2019m1,2019m12) ) ///
    , saving(m3ynorm, replace)

_pctile res, percentiles(2.5,97.5)
gen pye=corrempr*exp(l.lnflnon+pdln)
gen ubye=corrempr*exp(l.lnflnon+pdln+r(r2))
gen lbye=corrempr*exp(l.lnflnon+pdln+r(r1))
twoway (tsline fl_nonfarm if tin(2018m1,2019m12)) ///
    (tsline pye ubye lbye if tin(2019m1,2019m12) ) ///
    , saving(m3yemp, replace)

graph combine m3ynorm.gph m3yemp.gph , ///
    saving(m3yen, replace)

gen fub=ubye if tin(2020m1,)
gen flb=lbye if tin(2020m1,)
gen fcst=pye if tin(2020m1,)
replace fcst=fl_non if tin(2019m12,2019m12)
replace fub=fl_non if tin(2019m12,2019m12)
replace flb=fl_non if tin(2019m12,2019m12)

tsline fl_nonfarm fub flb fcst if tin(2019m1,2020m1) , saving(fcst, replace)

list fcst fup flb if date=tm(2020m1)

log close

```

Appendix B: Log File

```

-----
      name:  <unnamed>
      log:   C:\Users\jdewey\Documents\A S20 Time Series\Problem Sets\Problem Set 4
Work.smcl
      log type:  smcl
      opened on:  5 Apr 2020, 21:37:41

.
.
. ** 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 flllfn fl_lf

. rename flnan fl_nonfarm

. rename lnu02300000_20200110 us_epr

. gen lnflnonfarm=ln( fl_nonfarm)

. gen lnflllf=ln( fl_lf)

. gen lnusepr = ln(us_epr)

. gen lnflbp=ln( fl_bp)

.
. tsappend, add(1)

. replace month=month(dofm(date)) if month==.
(1 real change made)

.
.
.
.
. *fit and evaluate models
. *Note I restricted estimation to year>1989 so the same observations are
. * compared for all models by dropping earlier years above
.
.
. *Model 1

```

```
. set seed 22045 // to make sure the same folds are used for each model

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

	RMSE
est1	.0042477
est2	.0046809
est3	.0036137
est4	.0049964
est5	.0031522
est6	.0040797
est7	.0026135
est8	.0035943
est9	.0029642
est10	.0033241

```
. scalar define k=10

. matrix kSSE=r(est)'\*r(est)

. scalar krmse1=(e1(kSSE,1,1)/k)^.5

. matrix drop kSSE

. scalar drop k

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

Leave-One-Out Cross-Validation Results

Method	Value
Root Mean Squared Errors	.0039009
Mean Absolute Errors	.00282589
Pseudo-R2	.84712569

```
. scalar loormse1=r(rmse)

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

Source	SS	df	MS	Number of obs	=	347

Model	.030722026	60	.000512034	F(60, 286)	=	43.24
Residual	.003386356	286	.00001184	Prob > F	=	0.0000

Total	.034108382	346	.000098579	R-squared	=	0.9007

				Adj R-squared	=	0.8799
				Root MSE	=	.00344

D.	lnflnonfarm	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]

lnflnonfarm						
LD.		-.186234	.056416	-3.30	0.001	-.2972773 - .0751908
L2D.		-.1547999	.0587146	-2.64	0.009	-.2703675 - .0392324
L3D.		.1401796	.0598191	2.34	0.020	.0224379 .2579212
L4D.		.1286005	.0601072	2.14	0.033	.010292 .246909
L5D.		.0462484	.0611613	0.76	0.450	-.074135 .1666317

Time Series Modeling and Forecasting

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L6D.		.1077401	.0613787	1.76	0.080	-.0130711	.2285513
L7D.		.0218677	.061864	0.35	0.724	-.0998988	.1436343
L8D.		-.0431733	.0618821	-0.70	0.486	-.1649755	.0786288
L9D.		.0889303	.0600003	1.48	0.139	-.0291679	.2070284
L10D.		-.194563	.0588503	-3.31	0.001	-.3103977	-.0787283
L11D.		-.0458498	.0583995	-0.79	0.433	-.1607971	.0690976
L12D.		.3152677	.0570512	5.53	0.000	.2029741	.4275613
lnfllf							
LD.		-.1375696	.1011412	-1.36	0.175	-.3366451	.061506
L2D.		-.1430712	.1026083	-1.39	0.164	-.3450345	.058892
L3D.		-.1307211	.1038122	-1.26	0.209	-.3350539	.0736117
L4D.		-.0463644	.1055798	-0.44	0.661	-.2541764	.1614476
L5D.		-.0220199	.1045149	-0.21	0.833	-.2277358	.1836961
L6D.		-.073281	.1038964	-0.71	0.481	-.2777796	.1312176
L7D.		-.0004624	.104082	-0.00	0.996	-.2053263	.2044015
L8D.		.0108387	.1048545	0.10	0.918	-.1955457	.2172232
L9D.		.1967346	.1046425	1.88	0.061	-.0092326	.4027017
L10D.		.2476003	.1075338	2.30	0.022	.0359424	.4592583
L11D.		.0133503	.1079256	0.12	0.902	-.1990789	.2257794
L12D.		-.1570608	.1068642	-1.47	0.143	-.3674008	.0532793
lnusepr							
LD.		.2110302	.1362474	1.55	0.123	-.0571446	.479205
L2D.		.0944642	.1395361	0.68	0.499	-.1801838	.3691122
L3D.		.1740808	.1387099	1.25	0.211	-.0989409	.4471026
L4D.		.1557738	.1381626	1.13	0.260	-.1161707	.4277182
L5D.		-.0132056	.1370974	-0.10	0.923	-.2830536	.2566424
L6D.		.2790443	.1353574	2.06	0.040	.0126214	.5454673
L7D.		.1298337	.1376584	0.94	0.346	-.1411183	.4007857
L8D.		-.0757072	.1389392	-0.54	0.586	-.3491802	.1977658
L9D.		-.1653567	.1396094	-1.18	0.237	-.4401489	.1094354
L10D.		-.3470377	.1422668	-2.44	0.015	-.6270605	-.0670149
L11D.		.1304803	.142392	0.92	0.360	-.149789	.4107496
L12D.		.1802605	.137825	1.31	0.192	-.0910195	.4515404
lnflbp							
LD.		.0014532	.0016699	0.87	0.385	-.0018337	.0047401
L2D.		.0048252	.0019409	2.49	0.013	.0010049	.0086456
L3D.		.0062142	.0020251	3.07	0.002	.0022282	.0102001
L4D.		.0045958	.0020418	2.25	0.025	.0005769	.0086146
L5D.		.0045223	.0020493	2.21	0.028	.0004886	.0085559
L6D.		.0049516	.0020707	2.39	0.017	.0008759	.0090272
L7D.		.0041298	.0020715	1.99	0.047	.0000525	.0082071
L8D.		.0038086	.0021258	1.79	0.074	-.0003756	.0079928
L9D.		.0034564	.0021233	1.63	0.105	-.0007229	.0076357
L10D.		.0041043	.0020888	1.96	0.050	-7.01e-06	.0082155
L11D.		.0034516	.0019993	1.73	0.085	-.0004836	.0073869
L12D.		.003117	.001649	1.89	0.060	-.0001287	.0063626
month							
2		.0089558	.0036845	2.43	0.016	.0017036	.016208
3		.0082561	.003799	2.17	0.031	.0007785	.0157338
4		.0081944	.0042342	1.94	0.054	-.0001398	.0165285
5		.0028597	.0033823	0.85	0.399	-.0037976	.009517
6		-.004196	.00394	-1.06	0.288	-.0119511	.003559
7		.0023721	.0037789	0.63	0.531	-.0050659	.0098101
8		.0120413	.0039413	3.06	0.002	.0042835	.019799
9		.0092995	.0033932	2.74	0.007	.0026206	.0159784
10		.0152326	.0042659	3.57	0.000	.0068361	.0236291
11		.008236	.0038779	2.12	0.035	.0006031	.0158689
12		.0144426	.0036928	3.91	0.000	.007174	.0217111

```

      date | -3.79e-06  2.03e-06  -1.87  0.062  -7.78e-06  1.99e-07
      _cons | -.003468  .0029584  -1.17  0.242  -.0092911  .0023551
-----

```

```
. estat ic
```

Akaike's information criterion and Bayesian information criterion

```

-----
      Model |      Obs  ll(null)  ll(model)      df      AIC      BIC
-----+-----
      . |      347  1108.606  1509.354      61  -2896.709  -2661.9
-----

```

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

```
. scalar aic1=(e1(r(S),1,5))
```

```
. scalar bic1=(e1(r(S),1,6))
```

```
. reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/12)d.lnfllf ///
>      l(1/12)d.lnusepr l(1/12)d.lnflbp i.month date if tin(1991m1,2018m12)
```

```

-----
      Source |      SS      df      MS      Number of obs      =      335
-----+-----
      Model | .030103405      60 .000501723      F(60, 274)      =      41.35
      Residual | .003324854     274 .000012135      Prob > F      =      0.0000
-----+-----
      Total | .03342826     334 .000100085      R-squared      =      0.9005
                                           Adj R-squared  =      0.8788
                                           Root MSE      =      .00348
-----

```

```

-----
D.
lnflnonfarm |      Coef.   Std. Err.      t    P>|t|     [95% Conf. Interval]
-----+-----
lnflnonfarm |
  LD. | -.1930854   .0579173    -3.33  0.001   - .3071049   -.0790659
  L2D. | -.1553992   .0601116    -2.59  0.010   - .2737385   -.0370599
  L3D. | .1418981   .0609142     2.33  0.021   .0219788   .2618173
  L4D. | .1296244   .0613088     2.11  0.035   .0089282   .2503206
  L5D. | .0437587   .0624723     0.70  0.484   -.0792279   .1667454
  L6D. | .1075893   .0626099     1.72  0.087   -.0156682   .2308468
  L7D. | .0162405   .0632799     0.26  0.798   -.108336    .140817
  L8D. | -.0519209   .0634909    -0.82  0.414   -.1769129   .0730711
  L9D. | .0872121   .0618319     1.41  0.160   -.0345138   .208938
  L10D. | -.1969749   .0603978    -3.26  0.001   -.3158776   -.0780721
  L11D. | -.0514218   .0598301    -0.86  0.391   -.1692069   .0663633
  L12D. | .300731    .0586177     5.13  0.000   .1853327   .4161292
lnfllf |
  LD. | -.1626785   .1087132    -1.50  0.136   -.3766978   .0513407
  L2D. | -.1595182   .1091353    -1.46  0.145   -.3743684   .055332
  L3D. | -.0926912   .1109798    -0.84  0.404   -.3111727   .1257903
  L4D. | .0237447   .1140586     0.21  0.835   -.200798    .2482873
  L5D. | .008565     .1131701     0.08  0.940   -.2142284   .2313583
  L6D. | -.0801172   .1125639    -0.71  0.477   -.3017172   .1414828
  L7D. | -.0344974   .1119908    -0.31  0.758   -.2549691   .1859744
  L8D. | -.018564     .1120342    -0.17  0.869   -.2391211   .2019932
  L9D. | .1765756    .1111867     1.59  0.113   -.0423131   .3954643
  L10D. | .245674     .1110628     2.21  0.028   .0270291   .4643189
  L11D. | .0221368    .1111832     0.20  0.842   -.1967451   .2410188
  L12D. | -.1376978   .1096828    -1.26  0.210   -.3536259   .0782304
lnusepr |
  LD. | .245242     .1441601     1.70  0.090   -.0385602   .5290441
-----

```

L2D.		.1329345	.1469418	0.90	0.366	-.156344	.422213
L3D.		.1485017	.146062	1.02	0.310	-.1390447	.4360481
L4D.		.0878049	.1462348	0.60	0.549	-.2000817	.3756916
L5D.		-.0549679	.1450908	-0.38	0.705	-.3406023	.2306666
L6D.		.2834977	.1449964	1.96	0.052	-.0019509	.5689463
L7D.		.1775538	.1473984	1.20	0.229	-.1126235	.467731
L8D.		-.035584	.1478749	-0.24	0.810	-.3266993	.2555312
L9D.		-.125733	.1480825	-0.85	0.397	-.417257	.165791
L10D.		-.3463431	.1472148	-2.35	0.019	-.6361589	-.0565273
L11D.		.1366832	.1466964	0.93	0.352	-.152112	.4254784
L12D.		.1575691	.1415662	1.11	0.267	-.1211265	.4362647
lnflbp							
LD.		.0018831	.0017334	1.09	0.278	-.0015294	.0052957
L2D.		.0053573	.0020279	2.64	0.009	.001365	.0093496
L3D.		.006421	.002118	3.03	0.003	.0022513	.0105907
L4D.		.0044499	.002123	2.10	0.037	.0002705	.0086293
L5D.		.0044712	.0021038	2.13	0.034	.0003296	.0086128
L6D.		.0052013	.0021137	2.46	0.014	.00104	.0093625
L7D.		.0042838	.0021139	2.03	0.044	.0001223	.0084454
L8D.		.0036872	.0021623	1.71	0.089	-.0005696	.0079441
L9D.		.0034861	.0021654	1.61	0.109	-.0007769	.0077491
L10D.		.0042454	.0021347	1.99	0.048	.0000428	.0084479
L11D.		.0035563	.0020478	1.74	0.084	-.0004752	.0075877
L12D.		.0033161	.0016977	1.95	0.052	-.000026	.0066583
month							
2		.0100688	.0039138	2.57	0.011	.0023639	.0177737
3		.0096305	.0040135	2.40	0.017	.0017292	.0175317
4		.0090093	.0044667	2.02	0.045	.000216	.0178027
5		.0025839	.0035085	0.74	0.462	-.0043232	.0094909
6		-.0042804	.0041245	-1.04	0.300	-.0124002	.0038393
7		.0027876	.0038998	0.71	0.475	-.0048898	.0104651
8		.0133054	.0041323	3.22	0.001	.0051703	.0214406
9		.0103615	.0035622	2.91	0.004	.0033488	.0173742
10		.0169306	.0045707	3.70	0.000	.0079325	.0259287
11		.0089775	.004059	2.21	0.028	.0009867	.0169683
12		.0146969	.0038918	3.78	0.000	.0070352	.0223585
date		-4.02e-06	2.15e-06	-1.87	0.063	-8.25e-06	2.20e-07
_cons		-.0040271	.0030982	-1.30	0.195	-.0101264	.0020723

```

. scalar NVar1=e(df_m)

. predict res, residual
(14 missing values generated)

. predict pdln1
(option xb assumed; fitted values)
(13 missing values generated)

. predict stdf1, stdf
(13 missing values generated)

. gen ressq=res^2
(14 missing values generated)

. summ ressq if tin(2018q1,2018q4)

```

Variable		Obs	Mean	Std. Dev.	Min	Max
resqq		4	3.81e-06	4.28e-06	7.82e-08	8.88e-06

```

. scalar osrmse1=r(mean)^0.5

. drop res ressq

. gen ubdln1=pdln1+1.96*stdf1
(13 missing values generated)

. gen lbdln1=pdln1-1.96*stdf1
(13 missing values generated)

. twoway (tsline d.lnflnonfarm if tin(2018m1,2019m12)) ///
>      (tsline pdln1 ubdln1 lbdln1 if tin(2019m1,2019m12) ) , ///
>      saving(mltslines, replace)
(file mltslines.gph saved)

.
.
.
.
.
. *Model 2
. set seed 22045 // to make sure the same folds are used for each model

. crossfold reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/2)d.lnfl1f ///
>      l(1/2)d.lnusepr l(1/2)d.lnflbp i.month date , k(10)

-----+-----
      |          RMSE
-----+-----
est1 | .0044081
est2 | .0048725
est3 | .0037067
est4 | .004862
est5 | .0030362
est6 | .0038931
est7 | .002654
est8 | .0028316
est9 | .0027686
est10 | .0036666

. scalar define k=10

. matrix kSSE=r(est)'*r(est)

. scalar krmse2=(e1(kSSE,1,1)/k)^.5

. matrix drop kSSE

. scalar drop k

. loocv reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/2)d.lnfl1f ///
>      l(1/2)d.lnusepr l(1/2)d.lnflbp i.month date

```

Leave-One-Out Cross-Validation Results

Method	Value
Root Mean Squared Errors	.00382463
Mean Absolute Errors	.00271565
Pseudo-R2	.85235776

```
. scalar loormse2=r(rmse)
```

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

Source	SS	df	MS	Number of obs	=	347
Model	.030103875	30	.001003463	F(30, 316)	=	79.18
Residual	.004004507	316	.000012672	Prob > F	=	0.0000
				R-squared	=	0.8826
				Adj R-squared	=	0.8714
Total	.034108382	346	.000098579	Root MSE	=	.00356

D.		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
lnflnonfarm						
lnflnonfarm						
LD.		-.1292626	.0520146	-2.49	0.013	-.2316014 -.0269238
L2D.		-.0978374	.0545818	-1.79	0.074	-.2052271 .0095523
L3D.		.223866	.0542971	4.12	0.000	.1170365 .3306956
L4D.		.1581467	.0521274	3.03	0.003	.0555862 .2607073
L5D.		.1393954	.0526783	2.65	0.009	.0357509 .24304
L6D.		.1399744	.0534127	2.62	0.009	.0348849 .2450639
L7D.		.0839072	.0551957	1.52	0.129	-.0246903 .1925047
L8D.		.0119944	.0541148	0.22	0.825	-.0944765 .1184652
L9D.		.0924972	.0517874	1.79	0.075	-.0093944 .1943889
L10D.		-.1661393	.0501878	-3.31	0.001	-.2648838 -.0673948
L11D.		-.0841851	.0518044	-1.63	0.105	-.1861104 .0177401
L12D.		.3717099	.0518522	7.17	0.000	.2696906 .4737291
lnfllf						
LD.		-.1826015	.097081	-1.88	0.061	-.3736084 .0084054
L2D.		-.1482414	.0984729	-1.51	0.133	-.3419868 .0455041
lnusepr						
LD.		.2129745	.1296114	1.64	0.101	-.0420359 .4679849
L2D.		.0721516	.1313385	0.55	0.583	-.186257 .3305601
lnflbp						
LD.		.0002202	.0015249	0.14	0.885	-.0027801 .0032205
L2D.		.0021079	.0015343	1.37	0.170	-.0009108 .0051266
month						
2		.0091515	.0023933	3.82	0.000	.0044427 .0138604
3		.0070618	.0028439	2.48	0.014	.0014665 .0126571
4		.0083358	.0028878	2.89	0.004	.002654 .0140176
5		.0017523	.002821	0.62	0.535	-.003798 .0073025
6		-.0003743	.002558	-0.15	0.884	-.0054071 .0046586
7		.0011175	.0022308	0.50	0.617	-.0032717 .0055066
8		.0113968	.0025791	4.42	0.000	.0063225 .0164712
9		.0140913	.0026692	5.28	0.000	.0088397 .0193429
10		.0204449	.0026992	7.57	0.000	.0151342 .0257556
11		.0126423	.0025707	4.92	0.000	.0075843 .0177002
12		.0122226	.0021036	5.81	0.000	.0080837 .0163615
date		-2.03e-06	1.94e-06	-1.04	0.297	-5.85e-06 1.80e-06
_cons		-.0061237	.0020764	-2.95	0.003	-.010209 -.0020384

```
. estat ic
```

Akaike's information criterion and Bayesian information criterion

Model	Obs	ll(null)	ll(model)	df	AIC	BIC
.	347	1108.606	1480.264	31	-2898.529	-2779.2

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

```
. scalar aic2=(e1(r(S),1,5))
```

```
. scalar bic2=(e1(r(S),1,6))
```

```
. reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/2)d.lnfl1f l(1/2)d.lnusepr ///
> l(1/2)d.lnflbp i.month date if tin(1991m1,2018m12)
```

Source	SS	df	MS	Number of obs	=	335
Model	.029502637	30	.000983421	F(30, 304)	=	76.16
Residual	.003925622	304	.000012913	Prob > F	=	0.0000
				R-squared	=	0.8826
				Adj R-squared	=	0.8710
Total	.03342826	334	.000100085	Root MSE	=	.00359

D.		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
lnflnonfarm						
LD.		-.1432869	.0532355	-2.69	0.008	-.2480436 -.0385301
L2D.		-.1039357	.0556954	-1.87	0.063	-.2135331 .0056617
L3D.		.2222094	.0552671	4.02	0.000	.1134548 .330964
L4D.		.1665216	.0531033	3.14	0.002	.0620252 .2710181
L5D.		.1482078	.0539376	2.75	0.006	.0420695 .254346
L6D.		.143469	.0546519	2.63	0.009	.0359251 .251013
L7D.		.0849044	.0564479	1.50	0.134	-.0261738 .1959825
L8D.		.0118469	.0551023	0.21	0.830	-.0965832 .1202771
L9D.		.0950046	.0528395	1.80	0.073	-.0089728 .198982
L10D.		-.1667371	.0513332	-3.25	0.001	-.2677504 -.0657237
L11D.		-.0973744	.0527623	-1.85	0.066	-.2012 .0064512
L12D.		.3558615	.0529947	6.72	0.000	.2515787 .4601442
lnfl1f						
LD.		-.2207878	.1039964	-2.12	0.035	-.4254316 -.0161439
L2D.		-.1729889	.1039895	-1.66	0.097	-.3776193 .0316415
lnusepr						
LD.		.2605882	.1366604	1.91	0.057	-.0083319 .5295083
L2D.		.1144682	.1376348	0.83	0.406	-.1563694 .3853057
lnflbp						
LD.		.0004482	.0015817	0.28	0.777	-.0026643 .0035607
L2D.		.0023902	.0015791	1.51	0.131	-.0007172 .0054976
month						
2		.0095659	.002479	3.86	0.000	.0046877 .0144441
3		.0074317	.0029681	2.50	0.013	.0015911 .0132722
4		.0082161	.0029643	2.77	0.006	.0023829 .0140493
5		.0014169	.0029168	0.49	0.627	-.0043227 .0071565
6		-.0007554	.0026257	-0.29	0.774	-.0059222 .0044113
7		.0004755	.002295	0.21	0.836	-.0040405 .0049916
8		.0111367	.002636	4.22	0.000	.0059495 .0163238
9		.0144071	.0027386	5.26	0.000	.0090182 .019796
10		.0212425	.0027835	7.63	0.000	.015765 .0267199
11		.0131857	.0026669	4.94	0.000	.0079378 .0184336
12		.0122006	.0021503	5.67	0.000	.0079691 .016432

date		-2.31e-06	2.07e-06	-1.11	0.266	-6.40e-06	1.77e-06
_cons		-.0058985	.0021464	-2.75	0.006	-.0101222	-.0016749

```
. scalar NVar2=e(df_m)
```

```
. predict res, residual
(14 missing values generated)
```

```
. predict pdln2
(option xb assumed; fitted values)
(13 missing values generated)
```

```
. predict stdf2, stdf
(13 missing values generated)
```

```
. gen ressq=res^2
(14 missing values generated)
```

```
. summ ressq if tin(2018q1,2018q4)
```

Variable		Obs	Mean	Std. Dev.	Min	Max
ressq		4	1.59e-06	1.08e-06	4.81e-07	3.02e-06

```
. scalar osrmse2=r(mean)^0.5
```

```
. drop res ressq
```

```
. gen ubdln2=pdln2+1.96*stdf2
(13 missing values generated)
```

```
. gen lbdln2=pdln2-1.96*stdf2
(13 missing values generated)
```

```
. twoway (tsline d.lnflnonfarm if tin(2018m1,2019m12)) ///
> (tsline pdln2 ubdln2 lbdln2 if tin(2019m1,2019m12) ) , ///
> saving(m2tslines, replace)
(file m2tslines.gph saved)
```

```
.
.
```

```
. *Model 3
. set seed 22045 // to make sure the same folds are used for each model
```

```
. crossfold reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/2,12)d.lnflf ///
> l(1/2,12)d.lnflbp i.month date , k(10)
```

		RMSE
est1		.0044445
est2		.0049586
est3		.0035232
est4		.0046603
est5		.0033702
est6		.0039719
est7		.0026611
est8		.0029074
est9		.0025942
est10		.0035816

```

. scalar define k=10

. matrix kSSE=r(est) '*r(est)

. scalar krmse3=(e1(kSSE,1,1)/k)^.5

. matrix drop kSSE

. scalar drop k

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

```

Leave-One-Out Cross-Validation Results

Method	Value
Root Mean Squared Errors	.00381403
Mean Absolute Errors	.00267812
Pseudo-R2	.85320103

```

. scalar loormse3=r(rmse)

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

```

Source	SS	df	MS	Number of obs	=	347
Model	.030160659	30	.001005355	F(30, 316)	=	80.47
Residual	.003947723	316	.000012493	Prob > F	=	0.0000
				R-squared	=	0.8843
				Adj R-squared	=	0.8733
Total	.034108382	346	.000098579	Root MSE	=	.00353

D.		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnflnonfarm							
LD.		-.1179473	.0517143	-2.28	0.023	-.2196951	-.0161995
L2D.		-.0834614	.052317	-1.60	0.112	-.1863951	.0194723
L3D.		.2351664	.0515731	4.56	0.000	.1336964	.3366363
L4D.		.1596382	.0516095	3.09	0.002	.0580966	.2611797
L5D.		.1494304	.0522109	2.86	0.004	.0467055	.2521552
L6D.		.1590254	.0533389	2.98	0.003	.054081	.2639697
L7D.		.1012275	.0538178	1.88	0.061	-.0046589	.207114
L8D.		.0386844	.0531118	0.73	0.467	-.065813	.1431819
L9D.		.098232	.0511176	1.92	0.056	-.0023419	.1988058
L10D.		-.1775155	.0498607	-3.56	0.000	-.2756165	-.0794146
L11D.		-.0970244	.0510761	-1.90	0.058	-.1975166	.0034678
L12D.		.3567019	.0514544	6.93	0.000	.2554654	.4579385
lnfllf							
LD.		-.044002	.0540869	-0.81	0.417	-.150418	.0624139
L2D.		-.1081494	.053964	-2.00	0.046	-.2143235	-.0019754
L12D.		-.122442	.0525707	-2.33	0.020	-.2258747	-.0190092
lnflbp							
LD.		.0006484	.0015122	0.43	0.668	-.0023269	.0036237
L2D.		.0020716	.0015132	1.37	0.172	-.0009056	.0050488
L12D.		.0016323	.0013283	1.23	0.220	-.0009811	.0042457

month							
2		.0087985	.0021525	4.09	0.000	.0045635	.0130335
3		.0082388	.0025179	3.27	0.001	.0032848	.0131928
4		.0091183	.0028339	3.22	0.001	.0035427	.014694
5		.0038005	.0027832	1.37	0.173	-.0016755	.0092764
6		.0003569	.0025185	0.14	0.887	-.0045984	.0053121
7		.0033529	.0021834	1.54	0.126	-.0009428	.0076487
8		.0123898	.0024816	4.99	0.000	.0075074	.0172723
9		.0152734	.0026658	5.73	0.000	.0100285	.0205184
10		.0202373	.0026631	7.60	0.000	.0149976	.025477
11		.0138288	.0024088	5.74	0.000	.0090896	.0185681
12		.0133178	.0020107	6.62	0.000	.0093618	.0172738
date		-1.85e-06	1.91e-06	-0.97	0.333	-5.61e-06	1.90e-06
_cons		-.0073374	.0020026	-3.66	0.000	-.0112776	-.0033972

```
. estat ic
```

Akaike's information criterion and Bayesian information criterion

Model	Obs	ll(null)	ll(model)	df	AIC	BIC
.	347	1108.606	1482.742	31	-2903.484	-2784.155

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

```
. scalar aic3=(e1(r(S),1,5))
```

```
. scalar bic3=(e1(r(S),1,6))
```

```
. reg d.lnflnonfarm l(1/12)d.lnflnonfarm l(1/2,12)d.lnfllf ///
> l(1/2,12)d.lnflbp i.month date if tin(1991m1,2018m12)
```

Source	SS	df	MS	Number of obs	=	335
Model	.029533325	30	.000984444	F(30, 304)	=	76.84
Residual	.003894935	304	.000012812	Prob > F	=	0.0000
Total	.03342826	334	.000100085	R-squared	=	0.8835
				Adj R-squared	=	0.8720
				Root MSE	=	.00358

D.		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
lnflnonfarm						
LD.		-.1293841	.0531129	-2.44	0.015	-.2338996
L2D.		-.0858645	.0535234	-1.60	0.110	-.1911877
L3D.		.2384584	.0526225	4.53	0.000	.1349079
L4D.		.1690107	.0528191	3.20	0.002	.0650734
L5D.		.1578785	.0536036	2.95	0.003	.0523975
L6D.		.1622324	.0546873	2.97	0.003	.0546189
L7D.		.1049544	.0553378	1.90	0.059	-.0039392
L8D.		.0405085	.0545189	0.74	0.458	-.0667738
L9D.		.0997632	.0524099	1.90	0.058	-.0033688
L10D.		-.1781345	.0510472	-3.49	0.001	-.2785852
L11D.		-.1082445	.052337	-2.07	0.039	-.2112331
L12D.		.3445404	.0527947	6.53	0.000	.2406511
lnfllf						
LD.		-.0509275	.0564462	-0.90	0.368	-.1620023
L2D.		-.1055925	.0559263	-1.89	0.060	-.2156442

L12D.		-.1157475	.0547446	-2.11	0.035	-.2234739	-.0080211
lnflbp							
LD.		.0008645	.0015703	0.55	0.582	-.0022255	.0039545
L2D.		.0023376	.0015671	1.49	0.137	-.0007462	.0054215
L12D.		.0017084	.0013632	1.25	0.211	-.0009741	.0043909
month							
2		.0086588	.0022125	3.91	0.000	.004305	.0130125
3		.0082427	.0026136	3.15	0.002	.0030996	.0133858
4		.0091522	.0029301	3.12	0.002	.0033864	.0149179
5		.0036547	.0028793	1.27	0.205	-.0020112	.0093205
6		.0001017	.0025957	0.04	0.969	-.0050061	.0052094
7		.002862	.0022433	1.28	0.203	-.0015523	.0072764
8		.0124005	.0025404	4.88	0.000	.0074015	.0173995
9		.0155017	.0027433	5.65	0.000	.0101034	.0209
10		.0207081	.002745	7.54	0.000	.0153065	.0261098
11		.0143846	.0024842	5.79	0.000	.0094963	.0192729
12		.0134458	.0020583	6.53	0.000	.0093955	.017496
date		-2.02e-06	2.04e-06	-0.99	0.325	-6.04e-06	2.01e-06
_cons		-.0072834	.0020777	-3.51	0.001	-.0113719	-.0031948

```

. scalar NVar3=e(df_m)

. predict res, residual
(14 missing values generated)

. predict pdln3
(option xb assumed; fitted values)
(13 missing values generated)

. predict stdf3, stdf
(13 missing values generated)

. gen ressq=res^2
(14 missing values generated)

. summ ressq if tin(2018q1,2018q4)

```

Variable		Obs	Mean	Std. Dev.	Min	Max
ressq		4	7.86e-07	1.38e-06	1.69e-08	2.85e-06

```

. scalar osrmse3=r(mean)^0.5

. drop res ressq

. gen ubdln3=pdln3+1.96*stdf3
(13 missing values generated)

. gen lbdln3=pdln3-1.96*stdf3
(13 missing values generated)

. twoway (tsline d.lnflnonfarm if tin(2018m1,2019m12)) ///
>       (tsline pdln3 ubdln3 lbdln3 if tin(2019m1,2019m12) ) , ///
>       saving(m3tslines, replace)
(file m3tslines.gph saved)

.
.
. *Model 4

```

```
. set seed 22045 // to make sure the same folds are used for each model

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

	RMSE
est1	.0042473
est2	.004756
est3	.0035679
est4	.0049788
est5	.0033895
est6	.0038525
est7	.0026151
est8	.0030183
est9	.0025266
est10	.0036381

```
. scalar define k=10

. matrix kSSE=r(est)'\r(est)

. scalar krmse4=(e1(kSSE,1,1)/k)^.5

. matrix drop kSSE

. scalar drop k

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

Leave-One-Out Cross-Validation Results

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

```
. scalar loormse4=r(rmse)

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

Source	SS	df	MS	Number of obs	=	335
Model	.029334712	32	.00091671	F(32, 302)	=	74.38
Residual	.003721818	302	.000012324	Prob > F	=	0.0000
Total	.033056529	334	.000098972	R-squared	=	0.8874
				Adj R-squared	=	0.8755
				Root MSE	=	.00351

D.	lnflnonfarm	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
lnflnonfarm						
LD.		-.0893296	.0527122	-1.69	0.091	-.1930593 .0144
L2D.		-.0665814	.0558162	-1.19	0.234	-.1764194 .0432566
L3D.		.2241205	.0541419	4.14	0.000	.1175775 .3306636
L4D.		.1514706	.0521838	2.90	0.004	.0487806 .2541605
L5D.		.1580576	.0532447	2.97	0.003	.0532801 .2628351

L6D.		.1598189	.055306	2.89	0.004	.050985	.2686528
L7D.		.0713947	.0554199	1.29	0.199	-.0376635	.1804528
L8D.		.0199339	.0547887	0.36	0.716	-.0878821	.12775
L9D.		.076253	.0525006	1.45	0.147	-.0270605	.1795664
L10D.		-.1636484	.0515785	-3.17	0.002	-.2651471	-.0621497
L11D.		-.0908334	.0531678	-1.71	0.089	-.1954596	.0137928
L12D.		.3044588	.0598043	5.09	0.000	.1867729	.4221446
L24D.		.1201683	.05152	2.33	0.020	.0187846	.2215519
lnfl1f							
LD.		-.1225263	.0962344	-1.27	0.204	-.3119012	.0668486
L2D.		-.1611748	.0968689	-1.66	0.097	-.3517982	.0294486
L12D.		-.1592969	.1014979	-1.57	0.118	-.3590296	.0404358
L24D.		.1558841	.0998312	1.56	0.119	-.0405688	.352337
lnusepr							
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

```
. estat ic
```

Akaike's information criterion and Bayesian information criterion

Model	Obs	ll(null)	ll(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.

```
. scalar aic4=(e1(r(S),1,5))
```

```
. scalar bic4=(e1(r(S),1,6))
```

```
. reg d.lnflnonfarm l(1/12,24)d.lnflnonfarm l(1/2,12,24)d.lnfl1f ///
> l(1/2,12,24)d.lnusepr i.month if tin(1991m1,2018m12)
```

Source	SS	df	MS	Number of obs	=	323
Model	.028763234	32	.000898851	F(32, 290)	=	72.13
Residual	.003613751	290	.000012461	Prob > F	=	0.0000
				R-squared	=	0.8884
				Adj R-squared	=	0.8761
Total	.032376985	322	.00010055	Root MSE	=	.00353

D.		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnflnonfarm							
lnflnonfarm							
	LD.	-.0948899	.0538787	-1.76	0.079	-.2009328	.011153
	L2D.	-.0699134	.0566996	-1.23	0.219	-.1815084	.0416815
	L3D.	.2254863	.0548686	4.11	0.000	.1174951	.3334775
	L4D.	.1605308	.0530143	3.03	0.003	.0561892	.2648724
	L5D.	.1707254	.0543612	3.14	0.002	.0637329	.2777179
	L6D.	.1686264	.0562834	3.00	0.003	.0578507	.279402
	L7D.	.0701475	.0563324	1.25	0.214	-.0407247	.1810198
	L8D.	.01683	.0556574	0.30	0.763	-.0927136	.1263736
	L9D.	.0730854	.0535314	1.37	0.173	-.032274	.1784448
	L10D.	-.1617793	.0523278	-3.09	0.002	-.2647697	-.0587889
	L11D.	-.1031841	.0538404	-1.92	0.056	-.2091516	.0027834
	L12D.	.2633745	.0622896	4.23	0.000	.1407774	.3859716
	L24D.	.1726133	.0585465	2.95	0.003	.0573833	.2878433
	lnfl1lf						
	LD.	-.1634339	.1031209	-1.58	0.114	-.3663941	.0395263
	L2D.	-.1805075	.102243	-1.77	0.079	-.38174	.0207249
	L12D.	-.1331463	.1039788	-1.28	0.201	-.3377951	.0715024
	L24D.	.1745973	.1017924	1.72	0.087	-.0257484	.3749429
	lnusepr						
	LD.	.2178432	.138245	1.58	0.116	-.0542475	.4899338
	L2D.	.1240292	.1373462	0.90	0.367	-.1462925	.3943509
	L12D.	.0737005	.1362179	0.54	0.589	-.1944007	.3418016
	L24D.	-.3825953	.1298244	-2.95	0.003	-.6381128	-.1270778
	month						
	2	.0133682	.0032032	4.17	0.000	.0070638	.0196727
	3	.0114276	.0034255	3.34	0.001	.0046855	.0181696
	4	.0118775	.0036738	3.23	0.001	.0046468	.0191082
	5	.0054937	.0031766	1.73	0.085	-.0007585	.0117459
	6	.0048945	.0034447	1.42	0.156	-.0018852	.0116742
	7	.0065019	.0028725	2.26	0.024	.0008484	.0121554
	8	.0128559	.0028172	4.56	0.000	.0073111	.0184006
	9	.0159705	.0028304	5.64	0.000	.0103998	.0215413
	10	.0243941	.0036365	6.71	0.000	.0172369	.0315513
	11	.0153289	.0032299	4.75	0.000	.0089718	.021686
	12	.0141916	.0026701	5.32	0.000	.0089364	.0194469
	_cons	-.0107474	.0024715	-4.35	0.000	-.0156118	-.0058831

```

. scalar NVar4=e(df_m)

. predict res, residual
(26 missing values generated)

. predict pdln4
(option xb assumed; fitted values)
(25 missing values generated)

. predict stdf4, stdf
(25 missing values generated)

. gen ressq=res^2
(26 missing values generated)

. summ ressq if tin(2018q1,2018q4)

```

Variable	Obs	Mean	Std. Dev.	Min	Max
-----+-----					
ressq	4	4.65e-06	7.94e-06	3.50e-08	.0000165

```

. scalar osrmse4=r(mean)^0.5

. drop res ressq

. gen ubdln4=pdln4+1.96*stdf4
(25 missing values generated)

. gen lbdln4=pdln4-1.96*stdf4
(25 missing values generated)

. twoway (tsline d.lnflnonfarm if tin(2018m1,2019m12)) ///
>      (tsline pdln4 ubdln4 lbdln4 if tin(2019m1,2019m12) ) ///
>      , saving(m4tslines, replace)
(file m4tslines.gph saved)

.
.
. matrix M1=(NVar1,aic1,bic1,krmse1,loormse1,osrmse1)

. matrix M2=(NVar2,aic2,bic2,krmse2,loormse2,osrmse2)

. matrix M3=(NVar3,aic3,bic3,krmse3,loormse3,osrmse3)

. matrix M4=(NVar4,aic4,bic4,krmse4,loormse4,osrmse4)

. matrix MStats=(M1\M2\M3\M4)

. matrix colnames MStats=NVar AIC BIC RMSE10F RMSENF RMSEOS

. matrix rownames MStats=Model1 Model2 Model3 Model4

. matrix list MStats

MStats[4,6]
      NVar      AIC      BIC      RMSE10F      RMSENF      RMSEOS
Model1      60 -2896.7087 -2661.8999   .00379661   .0039009   .00195263
Model2      30 -2898.5286 -2779.1995   .00375628   .00382463   .00126116
Model3      30 -2903.4842 -2784.1552   .00375098   .00381403   .00088639
Model4      32 -2804.8818 -2679.0155   .00374267   .00370499   .00215691

.
. graph combine m1tslines.gph m2tslines.gph m3tslines.gph m4tslines.gph , ///
>      saving(mtslines, replace)
(file mtslines.gph saved)

.
. STOP
command STOP is unrecognized
r(199);

end of do-file

r(199);

. do "C:\Users\jdewey\AppData\Local\Temp\STD3380_000000.tmp"

. *Going to go with model 3
.
. drop pdl* ub* lb* stdf*

```

```
. scalar drop _all
```

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

Source	SS	df	MS	Number of obs	=	347
Model	.030160659	30	.001005355	F(30, 316)	=	80.47
Residual	.003947723	316	.000012493	Prob > F	=	0.0000
				R-squared	=	0.8843
				Adj R-squared	=	0.8733
Total	.034108382	346	.000098579	Root MSE	=	.00353

D.		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
lnflnonfarm						
LD.		-.1179473	.0517143	-2.28	0.023	-.2196951 -.0161995
L2D.		-.0834614	.052317	-1.60	0.112	-.1863951 .0194723
L3D.		.2351664	.0515731	4.56	0.000	.1336964 .3366363
L4D.		.1596382	.0516095	3.09	0.002	.0580966 .2611797
L5D.		.1494304	.0522109	2.86	0.004	.0467055 .2521552
L6D.		.1590254	.0533389	2.98	0.003	.054081 .2639697
L7D.		.1012275	.0538178	1.88	0.061	-.0046589 .207114
L8D.		.0386844	.0531118	0.73	0.467	-.065813 .1431819
L9D.		.098232	.0511176	1.92	0.056	-.0023419 .1988058
L10D.		-.1775155	.0498607	-3.56	0.000	-.2756165 -.0794146
L11D.		-.0970244	.0510761	-1.90	0.058	-.1975166 .0034678
L12D.		.3567019	.0514544	6.93	0.000	.2554654 .4579385
lnfllf						
LD.		-.044002	.0540869	-0.81	0.417	-.150418 .0624139
L2D.		-.1081494	.053964	-2.00	0.046	-.2143235 -.0019754
L12D.		-.122442	.0525707	-2.33	0.020	-.2258747 -.0190092
lnflbp						
LD.		.0006484	.0015122	0.43	0.668	-.0023269 .0036237
L2D.		.0020716	.0015132	1.37	0.172	-.0009056 .0050488
L12D.		.0016323	.0013283	1.23	0.220	-.0009811 .0042457
month						
2		.0087985	.0021525	4.09	0.000	.0045635 .0130335
3		.0082388	.0025179	3.27	0.001	.0032848 .0131928
4		.0091183	.0028339	3.22	0.001	.0035427 .014694
5		.0038005	.0027832	1.37	0.173	-.0016755 .0092764
6		.0003569	.0025185	0.14	0.887	-.0045984 .0053121
7		.0033529	.0021834	1.54	0.126	-.0009428 .0076487
8		.0123898	.0024816	4.99	0.000	.0075074 .0172723
9		.0152734	.0026658	5.73	0.000	.0100285 .0205184
10		.0202373	.0026631	7.60	0.000	.0149976 .025477
11		.0138288	.0024088	5.74	0.000	.0090896 .0185681
12		.0133178	.0020107	6.62	0.000	.0093618 .0172738
date		-1.85e-06	1.91e-06	-0.97	0.333	-5.61e-06 1.90e-06
_cons		-.0073374	.0020026	-3.66	0.000	-.0112776 -.0033972

```
. predict pdln
(option xb assumed; fitted values)
(13 missing values generated)
```

```

. predict stdf
(option xb assumed; fitted values)
(13 missing values generated)

. scalar rmse=e(rmse)

. gen corrnorm=exp((rmse^2)/2)

. predict res
(option xb assumed; fitted values)
(13 missing values generated)

. gen expres=exp(res)
(13 missing values generated)

. summ expres

      Variable |           Obs       Mean   Std. Dev.       Min       Max
-----+-----
      expres |           348    1.001602    .0093405    .978271    1.017811

. gen corremp=r(mean)

.
.
. gen pyn=corrnorm*exp(1.lnflnon+pdln)
(13 missing values generated)

. gen ubyn=corrnorm*exp(1.lnflnon+pdln+1.96*rmse)
(13 missing values generated)

. gen lbyn=corrnorm*exp(1.lnflnon+pdln-1.96*rmse)
(13 missing values generated)

. twoway (tsline fl_nonfarm if tin(2018m1,2019m12)) ///
>      (tsline pyn ubyn lbyn if tin(2019m1,2019m12) ) ///
>      , saving(m3ynorm, replace)
(file m3ynorm.gph saved)

.
.
. _pctile res, percentiles(2.5,97.5)

. gen pye=corremp*exp(1.lnflnon+pdln)
(13 missing values generated)

. gen ubye=corremp*exp(1.lnflnon+pdln+r(r2))
(13 missing values generated)

. gen lbye=corremp*exp(1.lnflnon+pdln+r(r1))
(13 missing values generated)

. twoway (tsline fl_nonfarm if tin(2018m1,2019m12)) ///
>      (tsline pye ubye lbye if tin(2019m1,2019m12) ) ///
>      , saving(m3yemp, replace)
(file m3yemp.gph saved)

.
. graph combine m3ynorm.gph m3yemp.gph , ///
>      saving(m3yen, replace)
(file m3yen.gph saved)

.

```



```

.
. gen fub=ubye if tin(2020m1,)
(360 missing values generated)

. gen flb=lbye if tin(2020m1,)
(360 missing values generated)

. gen fcst=pye if tin(2020m1,)
(360 missing values generated)

. replace fcst=fl_non if tin(2019m12,2019m12)
(1 real change made)

. replace fub=fl_non if tin(2019m12,2019m12)
(1 real change made)

. replace flb=fl_non if tin(2019m12,2019m12)
(1 real change made)

.
. tsline fl_nonfarm fub flb fcst if tin(2019m1,2020m1) , saving(fcst, replace)
(file fcst.gph saved)

.
. list fcst fup flb if date=tm(2020m1)
variable fup not found
r(111);

end of do-file

r(111);

. do "C:\Users\jdewey\AppData\Local\Temp\STD3380_000000.tmp"

. list fcst fub flb if date=tm(2020m1)
=exp not allowed
r(101);

end of do-file

r(101);

. do "C:\Users\jdewey\AppData\Local\Temp\STD3380_000000.tmp"

. list fcst fub flb if date==tm(2020m1)

      +-----+
      |      fcst      fub      flb |
      |-----|
361. | 9093.688   9219.636   8923.498 |
      +-----+

.
. log close
      name: <unnamed>
      log:  C:\Users\jdewey\Documents\A S20 Time Series\Problem Sets\Problem Set 4
Work.smcl
      log type:  smcl
      closed on:   5 Apr 2020, 21:42:04
-----

```