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Economic Forecasting (Econ 460)
May 8, 2022
Project Report

Project Description and Historical Data

This project aims to predict point and interval forecasts of the consumer price index for all urban consumers: all items in U.S. city average. Data for this variable is released through the U.S. Bureau of Labor Statistics (BLS) every month. BLS also has historical data on this variable since January 1st, 1947. I accessed the data using Stata's Fred API. The variable code is CPIAUCSL. The variable is seasonally adjusted and has the index 1982-1984 equals 100.

Description of the Forecasting Method

The first thing I did was graph the consumer price index against time. From this graph, I was able to see that there was an increasing trend. At this point, I had to test if the graph was increasing linearly or exponentially. I performed a natural log transformation of the CPI data and then graphed the results. I thought the graph for the natural log of CPI was more linear so I decided to use the natural log of CPI. From this point on I will refer to the natural log of CPI as CPI.

The next thing I did was determine if the data had any cyclical components. In order to determine this, I graphed the autocorrelation function. The results of the graph showed that the current CPI was highly correlated with past lags of itself. This implied that I should add cyclical components to my model.

The next decision I made was determining if I should use an autoregressive, mean-adjusted model, or a combination of both. I chose to use a purely autoregressive model to measure the cyclical components because the data was highly correlated to past lags and the graph did not revolve around a constant value.

At this point, my equation looked like this. $y_t = T_t + C_t + E_t$

In order to find the best autoregressive model, I regressed CPI against time and lags of itself. I ran tests until I included twelve lags. Then I used the AIC algorithm to find the best model. This algorithm compiled the data in a table. I selected the economic model with the lowest AIC value. The model I chose was an AR(10) model because it was the most negative value in the table. I also used AIC instead of BIC because AIC provides the best model.

My final equation was the following. $y_t = \alpha + \beta_0 \times time_t + AR(10) + E_t$

Now that I had a complete model, one that accounted for trends and cyclical components, I was ready to predict point and interval estimates by using the direct method to calculate point and 90% interval forecasts.

In order to predict point and interval forecasts for April, I executed the following code in stata.

```

regress y time L(1/10).y
predict y1
predict sf1, stdf
generate y1L=y1-1.645*sf1
generate y1U=y1+1.645*sf1

```

I ran nearly identical code to predict the next eleven intervals. After running the regressions, I was able to create the data in the following section. Do not forget the values are in natural log form.

Point and Interval Forecasts

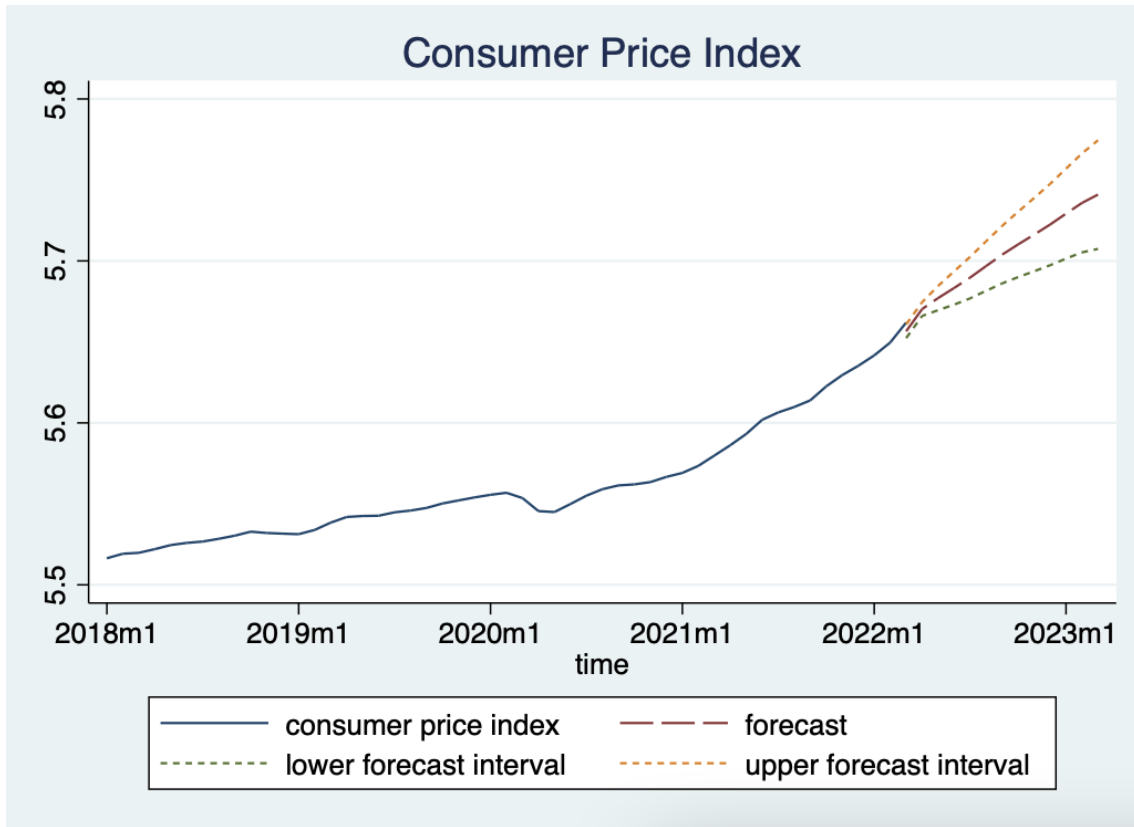
Figure 1

time	pL	p	pU
2022m4	5.666039	5.670297	5.674556
2022m5	5.669603	5.677082	5.68456
2022m6	5.673034	5.6833	5.693565
2022m7	5.676917	5.689733	5.702548
2022m8	5.681571	5.696838	5.712106
2022m9	5.686102	5.703846	5.72159
2022m10	5.690009	5.710173	5.730338
2022m11	5.6936	5.716213	5.738826
2022m12	5.69732	5.722432	5.747544
2023m1	5.701456	5.729199	5.756943
2023m2	5.705278	5.735812	5.766345
2023m3	5.70744	5.740982	5.774524

After converting the estimates out of log form, the estimates are as follows. The point estimate for April is 290.12. There is a ninety percent probability that the real value lies in the following interval [288.89, 291.36].

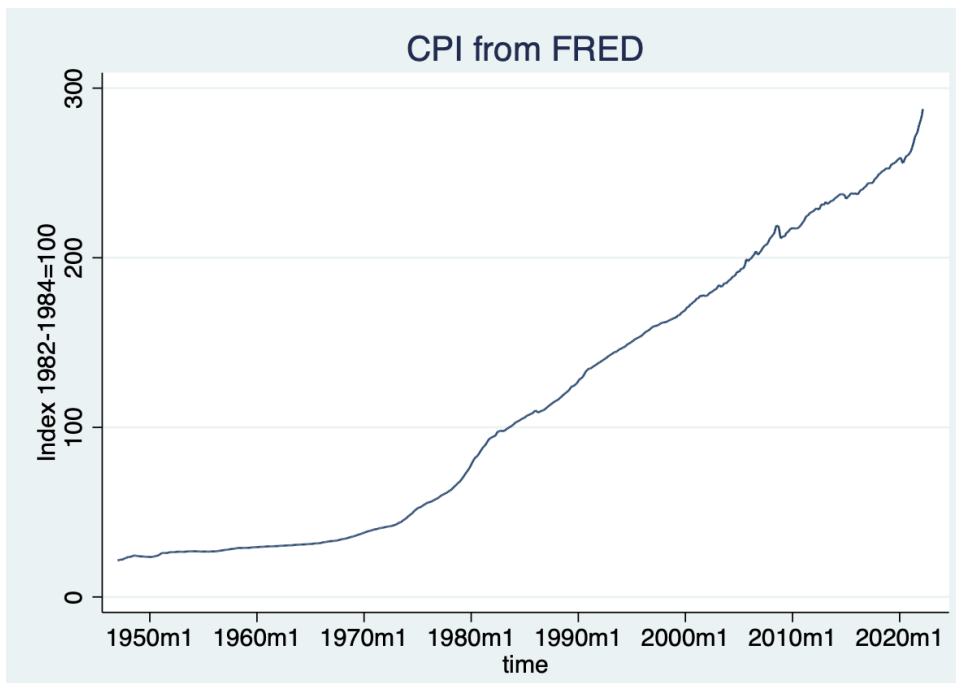
¹ Figure 1 is a table containing point and 90% interval estimates for the CPI in the next twelve intervals. The column pL contains estimates for the lower bound. The column p contains point estimates for the true value of CPI. The column pU contains estimates for the upper bound.

Figure 2



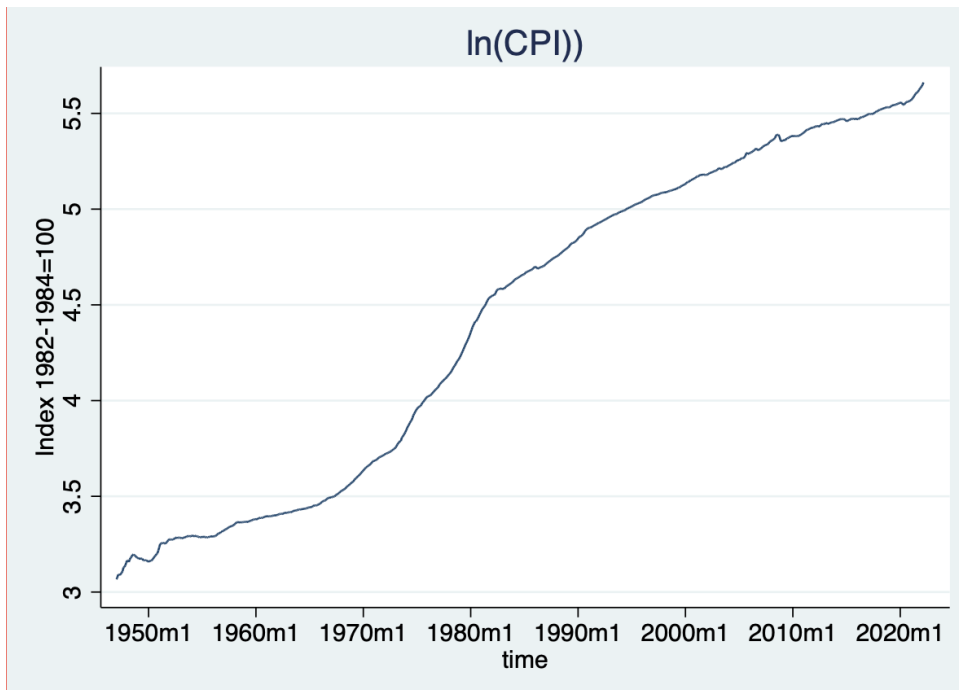
2

Appendix 1: Graph of CPI

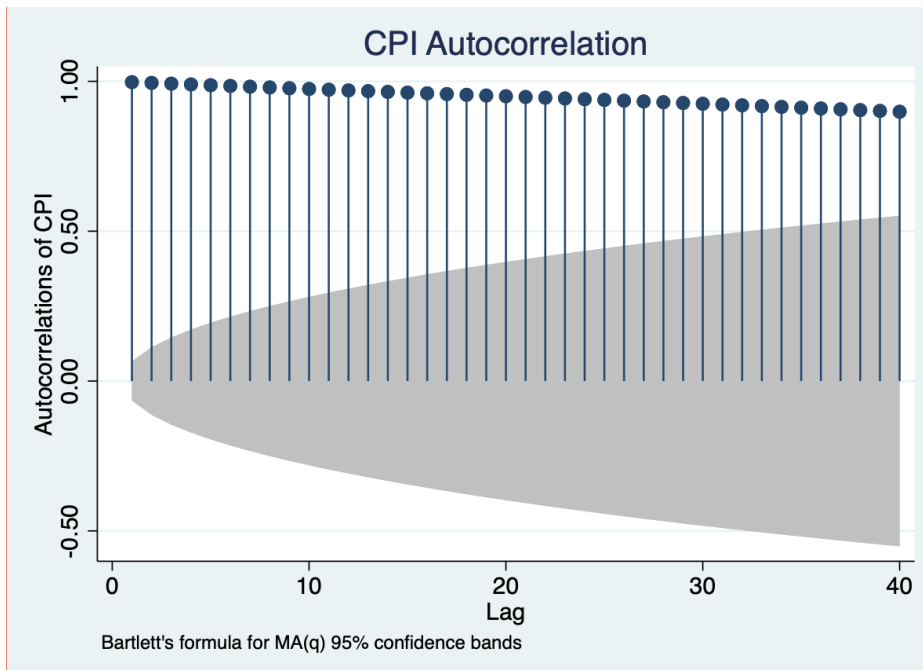


² Figure 2 is a graphical representation of the data in Figure 1.

Appendix 2: Graph of $\ln(\text{CPI})$



Appendix 3: CPI Autocorrelation Function



Appendix 4: Stata .do file and Output

```

-----
> -----
      name: <unnamed>
      log: /Users/andrewpeterson/Desktop/Stata-Workspace/econ460/project2.lo
> g
  log type: text
  opened on: 6 May 2022, 18:53:06

.
. freduse CPIAUCSL
(903 observations read)

. rename CPIAUCSL cpi

.
. tsmktime time, start(1947m1)

Time variable: time, 1947m1 to 2022m3
      Delta: 1 month

. tsset time

Time variable: time, 1947m1 to 2022m3
      Delta: 1 month

.
. describe

Contains data
  Observations:          903
  Variables:              4
-----
> -----
Variable      Storage   Display   Value
  name         type      format   label   Variable label
-----
> -----
date          str10    %10s          fed string date
cpi           double  %10.0g      Consumer Price Index for All Urb
> an Consumers: All

              Items in U.S. City Average
daten        float   %td          numeric (daily) date
time         float   %tm
-----
> -----
Sorted by: time
      Note: Dataset has changed since last saved.

```

```

. // browse
. // summarize
.
. tsappend, add(12)

.
. generate x = cpi
(12 missing values generated)

.
. generate y = ln(cpi)
(12 missing values generated)

.
. tsline x, title("CPI from FRED")

. graph rename cpi

.
. tsline y, title("ln(CPI)")

. graph rename ln_cpi

.
. ac y, title("CPI Autocorrelation")

. graph rename cpi_autocorrelation

.
.
. // test aic and bic to find the best model
. regress y if time>=tm(1947m1), r

```

```

Linear regression           Number of obs   =           903
                          F(0, 902)       =           0.00
                          Prob > F         =           .
                          R-squared        =           0.0000
                          Root MSE     =           .85378

```

		Robust				
	y	Coefficient	std. err.	t	P> t	[95% conf. interval]
	_cons	4.422797	.0284122	155.67	0.000	4.367035 4.478558

```
. estimates store ar0
```

```
.  
. regress y time L.y if time>=tm(1947m1), r
```

```
Linear regression                Number of obs   =           902  
                                F(2, 899)      >          99999.00  
                                Prob > F          =           0.0000  
                                R-squared         =           1.0000  
                                Root MSE       =           .00343
```

		Robust				
y	Coefficient	std. err.	t	P> t	[95% conf. interval]	
time	-2.63e-06	2.19e-06	-1.20	0.229	-6.93e-06	1.66e-06
y						
L1.	1.000551	.0006152	1626.51	0.000	.9993437	1.001758
_cons	.0012185	.0020815	0.59	0.558	-.0028666	.0053036

```
. estimates store ar1
```

```
.  
. regress y time L(1/2).y if time>=tm(1947m1), r
```

```
Linear regression                Number of obs   =           901  
                                F(3, 897)      >          99999.00  
                                Prob > F          =           0.0000  
                                R-squared         =           1.0000  
                                Root MSE       =           .0028
```

		Robust				
y	Coefficient	std. err.	t	P> t	[95% conf. interval]	
time	5.30e-08	1.83e-06	0.03	0.977	-3.53e-06	3.64e-06
y						
L1.	1.582806	.0473105	33.46	0.000	1.489954	1.675658
L2.	-.5829132	.0473185	-12.32	0.000	-.675781	-.4900454
_cons	.001661	.0017461	0.95	0.342	-.0017659	.0050879

```
. estimates store ar2
```

```
.  
. regress y time L(1/3).y if time>=tm(1947m1), r
```

```
Linear regression                Number of obs    =           900  
                                F(4, 895)       >    99999.00  
                                Prob > F           =           0.0000  
                                R-squared           =           1.0000  
                                Root MSE        =           .00275
```

		Robust				
y	Coefficient	std. err.	t	P> t	[95% conf. interval]	
time	6.55e-07	1.80e-06	0.36	0.716	-2.88e-06	4.19e-06
y						
L1.	1.511166	.0575492	26.26	0.000	1.398218	1.624113
L2.	-.3936845	.0951505	-4.14	0.000	-.5804285	-.2069405
L3.	-.1177317	.0488717	-2.41	0.016	-.2136482	-.0218152
_cons	.0019705	.0017219	1.14	0.253	-.0014089	.00535

```
. estimates store ar3
```

```
.  
. regress y time L(1/4).y if time>=tm(1947m1), r
```

```
Linear regression                Number of obs    =           899  
                                F(5, 893)       >    99999.00  
                                Prob > F           =           0.0000  
                                R-squared           =           1.0000  
                                Root MSE        =           .0027
```

		Robust				
y	Coefficient	std. err.	t	P> t	[95% conf. interval]	
time	8.92e-07	1.74e-06	0.51	0.609	-2.53e-06	4.31e-06
y						
L1.	1.513002	.0566027	26.73	0.000	1.401912	1.624092
L2.	-.479042	.0941492	-5.09	0.000	-.6638214	-.2942625
L3.	.1144011	.0853891	1.34	0.181	-.0531856	.2819878
L4.	-.1486882	.0473101	-3.14	0.002	-.2415402	-.0558363
_cons	.0020581	.0017014	1.21	0.227	-.0012812	.0053974

. estimates store ar4

.
. regress y time L(1/5).y if time>=tm(1947m1), r

Linear regression

Number of obs	=	898
F(6, 891)	>	99999.00
Prob > F	=	0.0000
R-squared	=	1.0000
Root MSE	=	.00268

		Robust				
y	Coefficient	std. err.	t	P> t	[95% conf. interval]	
time	1.14e-06	1.76e-06	0.65	0.519	-2.32e-06	4.59e-06
y						
L1.	1.488532	.0598443	24.87	0.000	1.37108	1.605984
L2.	-.4458292	.0960157	-4.64	0.000	-.6342726	-.2573859
L3.	.050001	.0871346	0.57	0.566	-.121012	.221014
L4.	.0147856	.0878471	0.17	0.866	-.1576258	.1871971
L5.	-.1078893	.0522904	-2.06	0.039	-.2105161	-.0052625
_cons	.0022086	.0016985	1.30	0.194	-.001125	.0055422

. estimates store ar5

.
. regress y time L(1/6).y if time>=tm(1947m1),r

Linear regression

Number of obs	=	897
F(7, 889)	>	99999.00
Prob > F	=	0.0000
R-squared	=	1.0000
Root MSE	=	.00267

y	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
time	1.55e-06	1.77e-06	0.88	0.381	-1.92e-06	5.02e-06
y						
L1.	1.480202	.0612142	24.18	0.000	1.360061	1.600343
L2.	-.4425083	.0957369	-4.62	0.000	-.6304049	-.2546116
L3.	.0410465	.0885411	0.46	0.643	-.1327276	.2148205
L4.	-.0184738	.0872152	-0.21	0.832	-.1896455	.1526978
L5.	.0423117	.0782278	0.54	0.589	-.111221	.1958445
L6.	-.1030847	.046761	-2.20	0.028	-.1948596	-.0113098
_cons	.0024731	.0016953	1.46	0.145	-.0008542	.0058003

. estimates store ar6

.
 . regress y time L(1/7).y if time>=tm(1947m1),r

Linear regression	Number of obs	=	896
	F(8, 887)	>	99999.00
	Prob > F	=	0.0000
	R-squared	=	1.0000
	Root MSE	=	.00266

y	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
time	1.89e-06	1.74e-06	1.09	0.278	-1.53e-06	5.31e-06
y						
L1.	1.469883	.0621121	23.67	0.000	1.347979	1.591787
L2.	-.4355676	.0960885	-4.53	0.000	-.6241549	-.2469804
L3.	.038369	.0875908	0.44	0.661	-.1335404	.2102783
L4.	-.0188853	.0875836	-0.22	0.829	-.1907806	.15301
L5.	.008419	.0822318	0.10	0.918	-.1529726	.1698107
L6.	.0261994	.096067	0.27	0.785	-.1623456	.2147445
L7.	-.0890175	.0548467	-1.62	0.105	-.1966619	.0186269
_cons	.0027142	.0016779	1.62	0.106	-.0005789	.0060073

y	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
time	2.81e-06	1.72e-06	1.63	0.103	-5.68e-07	6.20e-06
y						
L1.	1.448421	.0620725	23.33	0.000	1.326594	1.570248
L2.	-.4256651	.0957317	-4.45	0.000	-.6135532	-.2377769
L3.	.0364757	.0889797	0.41	0.682	-.1381607	.2111121
L4.	.0015135	.0860848	0.02	0.986	-.1674412	.1704681
L5.	.0067462	.0842967	0.08	0.936	-.1586992	.1721915
L6.	-.0574047	.0892548	-0.64	0.520	-.2325809	.1177716
L7.	.0594525	.0953075	0.62	0.533	-.1276033	.2465082
L8.	.0005011	.0831061	0.01	0.995	-.1626076	.1636097
L9.	-.0708726	.0388677	-1.82	0.069	-.1471565	.0054114
_cons	.0033672	.0016667	2.02	0.044	.0000961	.0066384

. estimates store ar9

. regress y time L(1/10).y if time>=tm(1947m1),r

Linear regression	Number of obs	=	893
	F(11, 881)	>	99999.00
	Prob > F	=	0.0000
	R-squared	=	1.0000
	Root MSE	=	.00256

y	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
time	3.08e-06	1.71e-06	1.80	0.072	-2.74e-07	6.44e-06
y						
L1.	1.45889	.0604415	24.14	0.000	1.340264	1.577516
L2.	-.4504831	.09426	-4.78	0.000	-.6354836	-.2654827
L3.	.0451048	.0862481	0.52	0.601	-.1241709	.2143805
L4.	.0034788	.0831631	0.04	0.967	-.1597422	.1666997
L5.	-.0057763	.0810915	-0.07	0.943	-.1649315	.1533788
L6.	-.0635437	.0880846	-0.72	0.471	-.2364238	.1093365
L7.	.1024191	.0904915	1.13	0.258	-.075185	.2800231
L8.	-.0834838	.0808066	-1.03	0.302	-.2420797	.075112
L9.	.1215453	.0649197	1.87	0.062	-.00587	.2489606
L10.	-.1290718	.0388334	-3.32	0.001	-.2052885	-.052855

_cons		.0035903	.0016517	2.17	0.030	.0003485	.0068321
-------	--	----------	----------	------	-------	----------	----------

. estimates store ar10

. regress y time L(1/11).y if time>=tm(1947m1),r

Linear regression		Number of obs	=	892
		F(12, 879)	>	99999.00
		Prob > F	=	0.0000
		R-squared	=	1.0000
		Root MSE	=	.00255

		Robust				
y	Coefficient	std. err.	t	P> t	[95% conf. interval]	
time	3.57e-06	1.73e-06	2.07	0.039	1.79e-07	6.96e-06
y						
L1.	1.449651	.0614513	23.59	0.000	1.329043	1.57026
L2.	-.4502919	.0961906	-4.68	0.000	-.6390819	-.2615018
L3.	.0498737	.0888662	0.56	0.575	-.124541	.2242884
L4.	.0036028	.0813187	0.04	0.965	-.1559986	.1632042
L5.	-.0069576	.0782414	-0.09	0.929	-.1605195	.1466042
L6.	-.0582275	.0851387	-0.68	0.494	-.2253263	.1088713
L7.	.0990836	.0904994	1.09	0.274	-.0785367	.2767038
L8.	-.0874038	.0843609	-1.04	0.300	-.2529762	.0781686
L9.	.0888051	.0674843	1.32	0.189	-.0436441	.2212544
L10.	.0085365	.065488	0.13	0.896	-.1199945	.1370676
L11.	-.0977262	.038526	-2.54	0.011	-.1733399	-.0221126
_cons	.0039853	.0016628	2.40	0.017	.0007217	.0072489

. estimates store ar11

```
.
. regress y time L(1/12).y if time>=tm(1947m1),r
```

```
Linear regression                Number of obs    =          891
                                F(13, 877)      >    99999.00
                                Prob > F             =         0.0000
                                R-squared            =         1.0000
                                Root MSE         =         .00254
```

		Robust				
y	Coefficient	std. err.	t	P> t	[95% conf. interval]	
time	3.65e-06	1.74e-06	2.09	0.036	2.30e-07	7.07e-06
y						
L1.	1.448317	.0610664	23.72	0.000	1.328464	1.56817
L2.	-.4316041	.0947499	-4.56	0.000	-.6175672	-.245641
L3.	.0068347	.0822067	0.08	0.934	-.1545101	.1681795
L4.	.0366986	.0768994	0.48	0.633	-.1142296	.1876269
L5.	-.01439	.0767257	-0.19	0.851	-.1649775	.1361975
L6.	-.06074	.0846034	-0.72	0.473	-.2267888	.1053088
L7.	.1124253	.089442	1.26	0.209	-.0631201	.2879707
L8.	-.083409	.0839433	-0.99	0.321	-.2481621	.0813442
L9.	.0509965	.0611875	0.83	0.405	-.0690945	.1710876
L10.	.0489537	.0647	0.76	0.449	-.0780312	.1759385
L11.	-.141336	.0784219	-1.80	0.072	-.2952525	.0125805
L12.	.0261893	.0471709	0.56	0.579	-.0663918	.1187704
_cons	.0040258	.0016712	2.41	0.016	.0007459	.0073058

```
. estimates store ar12
```

```
.
. estimates stats ar0 ar1 ar2 ar3 ar4 ar5 ar6 ar7 ar8 ar9 ar10 ar11 ar12
```

Akaike's information criterion and Bayesian information criterion

Model	N	ll(null)	ll(model)	df	AIC	BIC
ar0	903	-1138.058	-1138.058	1	2278.117	2282.922
ar1	902	-1136.034	3839.704	3	-7673.409	-7658.995
ar2	901	-1134.018	4020.062	4	-8032.124	-8012.91
ar3	900	-1132.03	4031.329	5	-8052.658	-8028.646
ar4	899	-1130.038	4044.684	6	-8077.368	-8048.56
ar5	898	-1128.037	4046.753	7	-8079.505	-8045.904
ar6	897	-1126.044	4047.826	8	-8079.653	-8041.26
ar7	896	-1124.059	4046.7	9	-8075.399	-8032.218
ar8	895	-1122.085	4046.134	10	-8072.268	-8024.3
ar9	894	-1120.142	4057.652	11	-8093.304	-8040.551
ar10	893	-1118.201	4066.661	12	-8109.322	-8051.787
ar11	892	-1116.269	4066.799	13	-8107.598	-8045.283
ar12	891	-1114.36	4068.081	14	-8108.163	-8041.07

Note: BIC uses N = number of observations. See [R] BIC note.

```

.
. // use an AR(10) model because AIC is lowest for AR(10)
.
. regress y time L(1/10).y

```

Source	SS	df	MS	Number of obs	=	893
Model	639.745822	11	58.1587111	F(11, 881)	>	99999.00
Residual	.005792715	881	6.5752e-06	Prob > F	=	0.0000
				R-squared	=	1.0000
				Adj R-squared	=	1.0000
Total	639.751615	892	.71721033	Root MSE	=	.00256

y	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
time	3.08e-06	1.79e-06	1.72	0.086	-4.37e-07	6.60e-06
Y						
L1.	1.45889	.0332628	43.86	0.000	1.393606	1.524173
L2.	-.4504831	.0581463	-7.75	0.000	-.5646045	-.3363617
L3.	.0451048	.0593502	0.76	0.447	-.0713794	.161589
L4.	.0034788	.059332	0.06	0.953	-.1129698	.1199274
L5.	-.0057763	.0593284	-0.10	0.922	-.1222178	.1106651
L6.	-.0635437	.0592094	-1.07	0.283	-.1797516	.0526643
L7.	.1024191	.0592723	1.73	0.084	-.0139123	.2187505
L8.	-.0834838	.0583922	-1.43	0.153	-.198088	.0311203
L9.	.1215453	.0561556	2.16	0.031	.011331	.2317597
L10.	-.1290718	.0320957	-4.02	0.000	-.1920647	-.0660788
_cons	.0035903	.0018799	1.91	0.056	-.0000994	.00728


```

-----
. predict y1
(option xb assumed; fitted values)
(21 missing values generated)

```

```

. predict sf1,stdf
(21 missing values generated)

```

```

. generate y1L=y1-1.645*sf1
(21 missing values generated)

```

```

. generate y1U=y1+1.645*sf1
(21 missing values generated)

```

```

. regress y time L(2/11).y

```

Source	SS	df	MS	Number of obs	=	892
				F(11, 880)	>	99999.00
Model	638.04309	11	58.0039172	Prob > F	=	0.0000
Residual	.017837883	880	.00002027	R-squared	=	1.0000
				Adj R-squared	=	1.0000
Total	638.060927	891	.716117764	Root MSE	=	.0045

y	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
time	7.67e-06	3.15e-06	2.43	0.015	1.48e-06	.0000139
Y						
L2.	1.661922	.0584233	28.45	0.000	1.547257	1.776587
L3.	-.6004638	.1021056	-5.88	0.000	-.8008628	-.4000648
L4.	.0696008	.1042081	0.67	0.504	-.1349247	.2741263
L5.	-.0031081	.104178	-0.03	0.976	-.2075743	.2013582
L6.	-.0709581	.1041995	-0.68	0.496	-.2754666	.1335504
L7.	.0145226	.1040514	0.14	0.889	-.1896952	.2187404
L8.	.0599704	.1040727	0.58	0.565	-.1442893	.2642302
L9.	-.0318404	.1025258	-0.31	0.756	-.2330641	.1693833
L10.	.1776195	.0986836	1.80	0.072	-.0160631	.3713022
L11.	-.2795629	.0564356	-4.95	0.000	-.390327	-.1687987
_cons	.008901	.0033018	2.70	0.007	.0024206	.0153814

```
. predict y2
(option xb assumed; fitted values)
(21 missing values generated)
```

```
. predict sf2,stdf
(21 missing values generated)
```

```
. generate y2L=y2-1.645*sf2
(21 missing values generated)
```

```
. generate y2U=y2+1.645*sf2
(21 missing values generated)
```

```
.
. regress y time L(3/12).y
```

Source	SS	df	MS	Number of obs	=	891
-----+-----				F(11, 879)	>	99999.00
Model	636.371885	11	57.8519896	Prob > F	=	0.0000
Residual	.033572205	879	.000038194	R-squared	=	0.9999
-----+-----				Adj R-squared	=	0.9999
Total	636.405458	890	.715062312	Root MSE	=	.00618

y	Coefficient	Std. err.	t	P> t	[95% conf. interval]
time	.000013	4.34e-06	3.00	0.003	4.51e-06 .0000215
y					
L3.	1.783432	.0802004	22.24	0.000	1.626025 1.940838
L4.	-.6368718	.1401641	-4.54	0.000	-.9119671 -.3617765
L5.	.0649271	.1430469	0.45	0.650	-.2158262 .3456804
L6.	-.0724314	.1430431	-0.51	0.613	-.3531773 .2083145
L7.	.0239252	.1431558	0.17	0.867	-.2570419 .3048923
L8.	-.0384648	.1428297	-0.27	0.788	-.3187918 .2418623
L9.	.0940868	.1428572	0.66	0.510	-.1862943 .3744678
L10.	.0296856	.1408288	0.21	0.833	-.2467143 .3060855
L11.	.0706972	.1354912	0.52	0.602	-.1952268 .3366211
L12.	-.3228762	.0774733	-4.17	0.000	-.4749305 -.1708219
_cons	.0150327	.0045342	3.32	0.001	.0061336 .0239318

```
. predict y3
(option xb assumed; fitted values)
(21 missing values generated)
```

```
. predict sf3,stdf
(21 missing values generated)
```

```
. generate y3L=y3-1.645*sf3
(21 missing values generated)
```

```
. generate y3U=y3+1.645*sf3
(21 missing values generated)
```

```
.
. regress y time L(4/13).y
```

Source	SS	df	MS	Number of obs	=	890
-----+-----				F(11, 878)	>	99999.00
Model	634.723438	11	57.7021308	Prob > F	=	0.0000
Residual	.052257392	878	.000059519	R-squared	=	0.9999
-----+-----				Adj R-squared	=	0.9999
Total	634.775696	889	.714033403	Root MSE	=	.00771

y	Coefficient	Std. err.	t	P> t	[95% conf. interval]
time	.0000181	5.43e-06	3.34	0.001	7.46e-06 .0000288
y					
L4.	1.924334	.1001171	19.22	0.000	1.727837 2.120831
L5.	-.6825371	.1749729	-3.90	0.000	-1.025951 -.3391231
L6.	-.0060952	.178625	-0.03	0.973	-.3566771 .3444867
L7.	.0230004	.1787206	0.13	0.898	-.3277691 .37377
L8.	-.0239283	.1787082	-0.13	0.894	-.3746735 .3268169
L9.	-.0045311	.1782994	-0.03	0.980	-.3544738 .3454117
L10.	.1367909	.1784669	0.77	0.444	-.2134805 .4870624
L11.	-.0017856	.1758588	-0.01	0.992	-.3469382 .3433671
L12.	-.1943267	.1691403	-1.15	0.251	-.5262933 .1376399
L13.	-.1763351	.0967231	-1.82	0.069	-.3661706 .0135004
_cons	.0210863	.0056629	3.72	0.000	.0099719 .0322006

```
. predict y4
(option xb assumed; fitted values)
(21 missing values generated)
```

```
. predict sf4,stdf
(21 missing values generated)
```

```
. generate y4L=y4-1.645*sf4
(21 missing values generated)
```

```
. generate y4U=y4+1.645*sf4
(21 missing values generated)
```

```
.
. regress y time L(5/14).y
```

Source	SS	df	MS	Number of obs	=	889
-----+-----				F(11, 877)	>	99999.00
Model	633.067109	11	57.5515554	Prob > F	=	0.0000
Residual	.074077489	877	.000084467	R-squared	=	0.9999
-----+-----				Adj R-squared	=	0.9999
Total	633.141187	888	.712996832	Root MSE	=	.00919

y	Coefficient	Std. err.	t	P> t	[95% conf. interval]
time	.0000233	6.49e-06	3.59	0.000	.0000106 .000036
y					
L5.	2.10068	.119269	17.61	0.000	1.866594 2.334766
L6.	-.8343481	.2085121	-4.00	0.000	-1.243589 -.4251072
L7.	.098777	.2129776	0.46	0.643	-.3192283 .5167823
L8.	-.0226085	.2129098	-0.11	0.915	-.4404809 .3952638
L9.	-.0006862	.212893	-0.00	0.997	-.4185256 .4171531
L10.	.0304672	.2125628	0.14	0.886	-.3867239 .4476582
L11.	.1399795	.2126711	0.66	0.511	-.2774242 .5573832
L12.	-.2789834	.2094986	-1.33	0.183	-.6901606 .1321939
L13.	-.0914518	.201496	-0.45	0.650	-.4869225 .3040189
L14.	-.1487976	.115237	-1.29	0.197	-.3749701 .0773749
_cons	.0273509	.0067495	4.05	0.000	.0141037 .040598

```
. predict y5
(option xb assumed; fitted values)
(21 missing values generated)
```

```
. predict sf5,stdf
(21 missing values generated)
```

```
. generate y5L=y5-1.645*sf5
(21 missing values generated)
```

```
. generate y5U=y5+1.645*sf5
(21 missing values generated)
```

```
.
. regress y time L(6/15).y
```

Source	SS	df	MS	Number of obs	=	888
-----+-----				F(11, 876)	>	99999.00
Model	631.384563	11	57.3985967	Prob > F	=	0.0000
Residual	.099918765	876	.000114063	R-squared	=	0.9998
-----+-----				Adj R-squared	=	0.9998
Total	631.484482	887	.7119329	Root MSE	=	.01068

y	Coefficient	Std. err.	t	P> t	[95% conf. interval]
time	.0000285	7.55e-06	3.78	0.000	.0000137 .0000434
y					
L6.	2.227759	.138778	16.05	0.000	1.955383 2.500136
L7.	-.8341028	.2425337	-3.44	0.001	-1.310118 -.3580879
L8.	.0667579	.247495	0.27	0.787	-.4189946 .5525104
L9.	.0036936	.2474139	0.01	0.988	-.4818996 .4892869
L10.	.0184948	.2475742	0.07	0.940	-.467413 .5044027
L11.	.0228434	.2470865	0.09	0.926	-.4621073 .507794
L12.	-.0866995	.2471366	-0.35	0.726	-.5717487 .3983496
L13.	-.2056596	.2434529	-0.84	0.398	-.6834787 .2721595
L14.	-.0844847	.2341796	-0.36	0.718	-.5441032 .3751339
L15.	-.137182	.1339178	-1.02	0.306	-.4000193 .1256552
_cons	.0338874	.0078472	4.32	0.000	.018486 .0492888

```
. predict y6
(option xb assumed; fitted values)
(21 missing values generated)
```

```
. predict sf6,stdf
(21 missing values generated)
```

```
. generate y6L=y6-1.645*sf6
(21 missing values generated)
```

```
. generate y6U=y6+1.645*sf6
(21 missing values generated)
```

```
.
. regress y time L(7/16).y
```

Source	SS	df	MS	Number of obs	=	887
-----+-----				F(11, 875)	>	99999.00
Model	629.729857	11	57.2481688	Prob > F	=	0.0000
Residual	.128878636	875	.00014729	R-squared	=	0.9998
-----+-----				Adj R-squared	=	0.9998
Total	629.858735	886	.710901507	Root MSE	=	.01214

y	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
time	.0000342	8.60e-06	3.98	0.000	.0000173	.0000511
y						
L7.	2.387637	.1577016	15.14	0.000	2.078119	2.697154
L8.	-.8953414	.275613	-3.25	0.001	-1.436281	-.3544016
L9.	.0939886	.2812428	0.33	0.738	-.4580007	.645978
L10.	.0122549	.2813551	0.04	0.965	-.5399547	.5644645
L11.	.0293067	.2814179	0.10	0.917	-.5230263	.5816397
L12.	-.2121946	.2807786	-0.76	0.450	-.7632729	.3388836
L13.	-.0304454	.2808402	-0.11	0.914	-.5816446	.5207538
L14.	-.2148541	.2766647	-0.78	0.438	-.7578579	.3281498
L15.	.059004	.2661346	0.22	0.825	-.4633328	.5813407
L16.	-.2396783	.1522083	-1.57	0.116	-.5384143	.0590577
_cons	.040907	.0089222	4.58	0.000	.0233955	.0584185

```
. predict y7
(option xb assumed; fitted values)
(21 missing values generated)
```

```
. predict sf7,stdf
(21 missing values generated)
```

```
. generate y7L=y7-1.645*sf7
(21 missing values generated)
```

```
. generate y7U=y7+1.645*sf7
(21 missing values generated)
```

```
.
. regress y time L(8/17).y
```

Source	SS	df	MS	Number of obs	=	886
-----+-----				F(11, 874)	>	99999.00
Model	628.087673	11	57.0988794	Prob > F	=	0.0000
Residual	.16187688	874	.000185214	R-squared	=	0.9997
-----+-----				Adj R-squared	=	0.9997
Total	628.24955	885	.709886497	Root MSE	=	.01361

y	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
time	.0000407	9.67e-06	4.21	0.000	.0000217	.0000597
y						
L8.	2.549556	.1768663	14.42	0.000	2.202424	2.896689
L9.	-.9331907	.3090652	-3.02	0.003	-1.539788	-.3265939
L10.	.1096883	.3156044	0.35	0.728	-.5097427	.7291194
L11.	.0217538	.3156002	0.07	0.945	-.5976691	.6411767
L12.	-.2027355	.3155748	-0.64	0.521	-.8221084	.4166374
L13.	-.1672518	.3148631	-0.53	0.595	-.785228	.4507244
L14.	-.0266918	.314948	-0.08	0.932	-.6448345	.5914509
L15.	-.1093604	.310265	-0.35	0.725	-.7183119	.4995912
L16.	.0787541	.2984542	0.26	0.792	-.5070166	.6645247
L17.	-.3328085	.1706823	-1.95	0.052	-.6678037	.0021866
_cons	.0486964	.0100109	4.86	0.000	.0290482	.0683445

```
. predict y8
(option xb assumed; fitted values)
(21 missing values generated)
```

```
. predict sf8,stdf
(21 missing values generated)
```

```
. generate y8L=y8-1.645*sf8
(21 missing values generated)
```

```
. generate y8U=y8+1.645*sf8
(21 missing values generated)
```

```
.
. regress y time L(9/18).y
```

Source	SS	df	MS	Number of obs	=	885
-----+-----				F(11, 873)	>	99999.00
Model	626.452087	11	56.9501897	Prob > F	=	0.0000
Residual	.199382457	873	.000228388	R-squared	=	0.9997
-----+-----				Adj R-squared	=	0.9997
Total	626.65147	884	.708881753	Root MSE	=	.01511

y	Coefficient	Std. err.	t	P> t	[95% conf. interval]
time	.0000477	.0000108	4.43	0.000	.0000266 .0000688
y					
L9.	2.74927	.1964475	13.99	0.000	2.363705 3.134834
L10.	-.9920734	.3435312	-2.89	0.004	-1.666317 -.3178299
L11.	.1264704	.3505782	0.36	0.718	-.5616043 .814545
L12.	-.210121	.350459	-0.60	0.549	-.8979616 .4777196
L13.	-.1605136	.3504375	-0.46	0.647	-.848312 .5272848
L14.	-.1744144	.3496596	-0.50	0.618	-.860686 .5118572
L15.	.1015672	.3497717	0.29	0.772	-.5849244 .7880589
L16.	-.1144246	.3445939	-0.33	0.740	-.790754 .5619048
L17.	.0290145	.3315051	0.09	0.930	-.6216256 .6796546
L18.	-.369188	.1896103	-1.95	0.052	-.7413332 .0029573
_cons	.0570948	.0111216	5.13	0.000	.0352666 .0789231


```
. predict y9
(option xb assumed; fitted values)
(21 missing values generated)
```

```
. predict sf9,stdf
(21 missing values generated)
```

```
. generate y9L=y9-1.645*sf9
(21 missing values generated)
```

```
. generate y9U=y9+1.645*sf9
(21 missing values generated)
```

```
.
. regress y time L(10/19).y
```

Source	SS	df	MS	Number of obs	=	884
-----+-----				F(11, 872)	>	99999.00
Model	624.832712	11	56.8029738	Prob > F	=	0.0000
Residual	.24303835	872	.000278714	R-squared	=	0.9996
-----+-----				Adj R-squared	=	0.9996
Total	625.07575	883	.707900057	Root MSE	=	.01669

y	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
time	.0000555	.0000119	4.66	0.000	.0000321	.0000789
y						
L10.	2.974485	.2172741	13.69	0.000	2.548044	3.400927
L11.	-1.054734	.3795916	-2.78	0.006	-1.799754	-.3097138
L12.	-.099288	.3872824	-0.26	0.798	-.8594026	.6608267
L13.	-.1687557	.387159	-0.44	0.663	-.928628	.5911167
L14.	-.1635436	.3871519	-0.42	0.673	-.9234021	.596315
L15.	-.0582244	.386303	-0.15	0.880	-.8164166	.6999679
L16.	.1028883	.3864561	0.27	0.790	-.6556046	.8613812
L17.	-.1710897	.3807247	-0.45	0.653	-.9183337	.5761543
L18.	.0184805	.3662184	0.05	0.960	-.7002921	.737253
L19.	-.3969865	.2095449	-1.89	0.058	-.8082578	.0142848
_cons	.0663402	.0122898	5.40	0.000	.0422192	.0904612

```
. predict y10
(option xb assumed; fitted values)
(21 missing values generated)
```

```
. predict sf10,stdf
(21 missing values generated)
```

```
. generate y10L=y10-1.645*sf10
(21 missing values generated)
```

```
. generate y10U=y10+1.645*sf10
(21 missing values generated)
```

```
.
. regress y time L(11/20).y
```

Source	SS	df	MS	Number of obs	=	883
-----+-----				F(11, 871)	>	99999.00
Model	623.205567	11	56.6550515	Prob > F	=	0.0000
Residual	.293983617	871	.000337524	R-squared	=	0.9995
-----+-----				Adj R-squared	=	0.9995
Total	623.499551	882	.70691559	Root MSE	=	.01837

y	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
time	.0000632	.0000131	4.81	0.000	.0000374	.000089
y						
L11.	3.244455	.2392235	13.56	0.000	2.774933	3.713977
L12.	-1.389035	.417752	-3.33	0.001	-2.208953	-.569117
L13.	-.0475338	.4262003	-0.11	0.911	-.8840334	.7889657
L14.	-.168014	.4260798	-0.39	0.693	-1.004277	.6682492
L15.	-.0508016	.4260828	-0.12	0.905	-.8870707	.7854675
L16.	-.0749539	.4251699	-0.18	0.860	-.9094313	.7595234
L17.	.0862453	.4253227	0.20	0.839	-.7485318	.9210224
L18.	-.2060742	.4189797	-0.49	0.623	-1.028402	.6162537
L19.	.0004979	.4030565	0.00	0.999	-.7905775	.7915733
L20.	-.4139226	.2305956	-1.80	0.073	-.8665106	.0386654
_cons	.0757665	.0135302	5.60	0.000	.0492109	.1023222

```
. predict y11
(option xb assumed; fitted values)
(21 missing values generated)
```

```
. predict sf11,stdf
(21 missing values generated)
```

```
. generate y11L=y11-1.645*sf11
(21 missing values generated)
```

```
. generate y11U=y11+1.645*sf11
(21 missing values generated)
```

```
.
. regress y time L(12/21).y
```

Source	SS	df	MS	Number of obs	=	882
-----+-----				F(11, 870)	>	99999.00
Model	621.55824	11	56.5052945	Prob > F	=	0.0000
Residual	.354311136	870	.000407254	R-squared	=	0.9994
-----+-----				Adj R-squared	=	0.9994
Total	621.912551	881	.70591663	Root MSE	=	.02018

y	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
time	.0000712	.0000145	4.92	0.000	.0000428	.0000996
-----+-----						
y						
L12.	3.311054	.262855	12.60	0.000	2.79515	3.826958
L13.	-1.471639	.4588797	-3.21	0.001	-2.37228	-.5709983
L14.	-.0296985	.4681862	-0.06	0.949	-.948605	.8892081
L15.	-.051703	.4680704	-0.11	0.912	-.9703821	.8669762
L16.	-.0788038	.4680991	-0.17	0.866	-.9975393	.8399317
L17.	-.1065286	.4670852	-0.23	0.820	-1.023274	.8102169
L18.	.0987374	.467201	0.21	0.833	-.8182355	1.01571
L19.	-.2667252	.4602564	-0.58	0.562	-1.170068	.6366174
L20.	.0265587	.4427697	0.06	0.952	-.8424629	.8955803
L21.	-.4528509	.2533367	-1.79	0.074	-.9500733	.0443715
-----+-----						
_cons	.0856399	.0148707	5.76	0.000	.0564533	.1148265

```

. predict y12
(option xb assumed; fitted values)
(21 missing values generated)

. predict sf12,stdf
(21 missing values generated)

. generate y12L=y12-1.645*sf12
(21 missing values generated)

. generate y12U=y12+1.645*sf12
(21 missing values generated)

.
.
. egen p=rowfirst(y1 y2 y3 y4 y5 y6 y7 y8 y9 y10 y11 y12) if time>=tm(2022m3)
(902 missing values generated)

. egen pL=rowfirst(y1L y2L y3L y4L y5L y6L y7L y8L y9L y10L y11L y12L) if time
> >=tm(2022m3)
(902 missing values generated)

. egen pU=rowfirst(y1U y2U y3U y4U y5U y6U y7U y8U y9U y10U y11U y12U) if time
> >=tm(2022m3)
(902 missing values generated)

.
. label variable y "consumer price index"

. label variable p "forecast"

. label variable pL "lower forecast interval"

. label variable pU "upper forecast interval"

.
. tsline y p pL pU if time>=tm(2018m1), title(Consumer Price Index) lpattern (
> solid longdash shor
> tdash shortdash)

```

```
. graph rename cpi_prediction
.
. list time pL p pU if time>=tm(2022m4)
```

	time	pL	p	pU
904.	2022m4	5.666039	5.670297	5.674556
905.	2022m5	5.669603	5.677082	5.68456
906.	2022m6	5.673034	5.6833	5.693565
907.	2022m7	5.676917	5.689733	5.702548
908.	2022m8	5.681571	5.696838	5.712106
909.	2022m9	5.686102	5.703846	5.72159
910.	2022m10	5.690009	5.710173	5.730338
911.	2022m11	5.6936	5.716213	5.738826
912.	2022m12	5.69732	5.722432	5.747544
913.	2023m1	5.701456	5.729199	5.756943
914.	2023m2	5.705278	5.735812	5.766345
915.	2023m3	5.70744	5.740982	5.774524

```
.
. log close
  name: <unnamed>
  log: /Users/andrewpeterson/Desktop/Stata-Workspace/econ460/project2.lo
> g
  log type: text
  closed on: 6 May 2022, 18:53:11
-----
> -----
```