

Time Series Analysis and Forecasting

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Workshop 4: Stationarity and ARIMA

This workshop consists of two main sections:

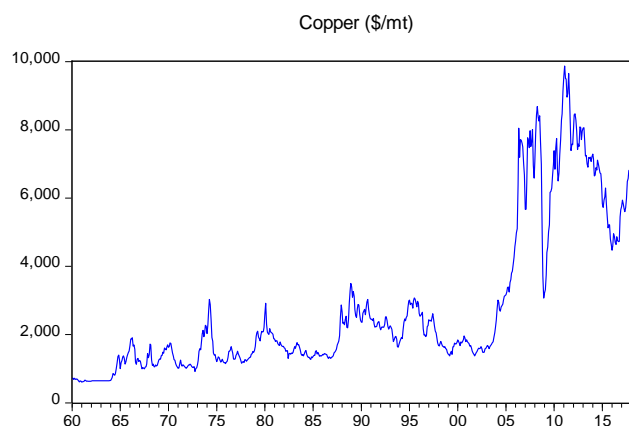
- (a) Dickey-Fuller Tests
- (b) Arima Modelling

(a) Dickey-Fuller Tests

In this section, we shall test for stationarity in a time series copper prices from 1960-Jan to 2018-Dec. The series for this first part is stored in the EViews file **session 4 workshop.wfl** and we want to analyse the series of **copper** prices.

- (i) First plot the series by selecting from the main menu

Quick Graph



Plot the correlogram by selecting

Quick Series Statistics Correlogram

Specify **copper**, then click **OK**

Does the series look stationary?

- (ii) Carry out a basic Dickey-Fuller test with no constant or trend term:

From the main EViews menu bar select

Quick Series Statistics Unit Root Test

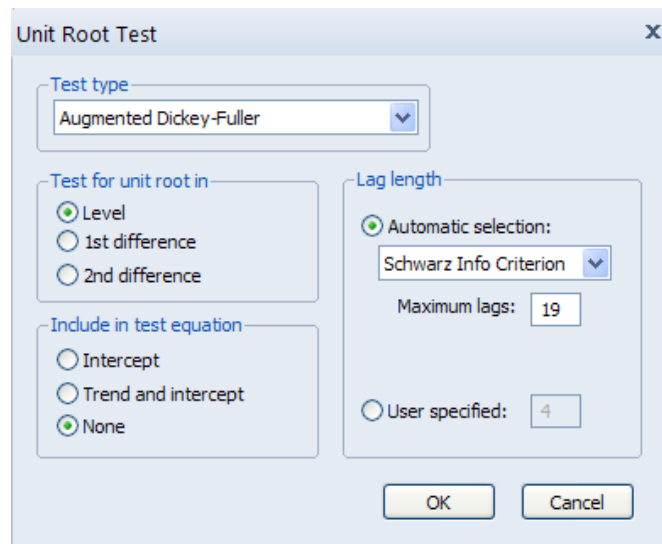
Specify **copper** then click **OK**

In the ensuing dialog box, specify

Test Type:	Augmented Dickey-Fuller
Test for unit root in:	Level
Include in test equation:	None

This basic Dickey-Fuller test involves estimation of the regression

$$\Delta Y_t = \gamma Y_{t-1} + \varepsilon_t$$



The results window shows the regression output at the bottom. The t -statistic for γ is also shown at the top left of the window and the corresponding critical values are shown in the top right.

What is the test's null hypothesis? Can it be rejected for the exchange rate series? Is the series stationary?

(iii) Include a constant and a trend in the basic Dickey-Fuller test:

From the menu bar of the Dickey-Fuller results window, select

View Unit Root Test

The Unit Root Test specification dialog box should reappear. Specify

Include in test equation: Trend and intercept

Does the test indicate stationarity?

(iv) Carry out Augmented Dickey-Fuller tests, including lagged values of the dependent variable.

$$\Delta Y_t = a_0 + a_2 t + \gamma Y_{t-1} + \sum_{i=1}^m \beta_i \Delta Y_{t-i} + \varepsilon_t$$

If the extra terms are not significant, throw them out of the regression.

As in regression analysis, when you have reasonable t -values you are not finished; you must also check for autocorrelation in the residuals.

You can do this by selecting from the main menu bar:

Quick Series Statistics Correlogram

Then specify **resid**

Do the Augmented Dickey-Fuller tests indicate that the series is stationary?

(v) If the series is non-stationary, examine whether the first differences are stationary. Graph the differenced series, plot its correlogram and then perform unit root tests. (The difference operator is D(copper) in EViews.)

(b) Arima Modelling

In the file *session 4 workshop.wf1*, which you can download from the intranet, you have the data on the **Gold** prices from 1960 to 2018 (Dec).

The Box-Jenkins methodology first requires that we establish stationarity, before embarking on the three main steps: Identification, Estimation and Testing.

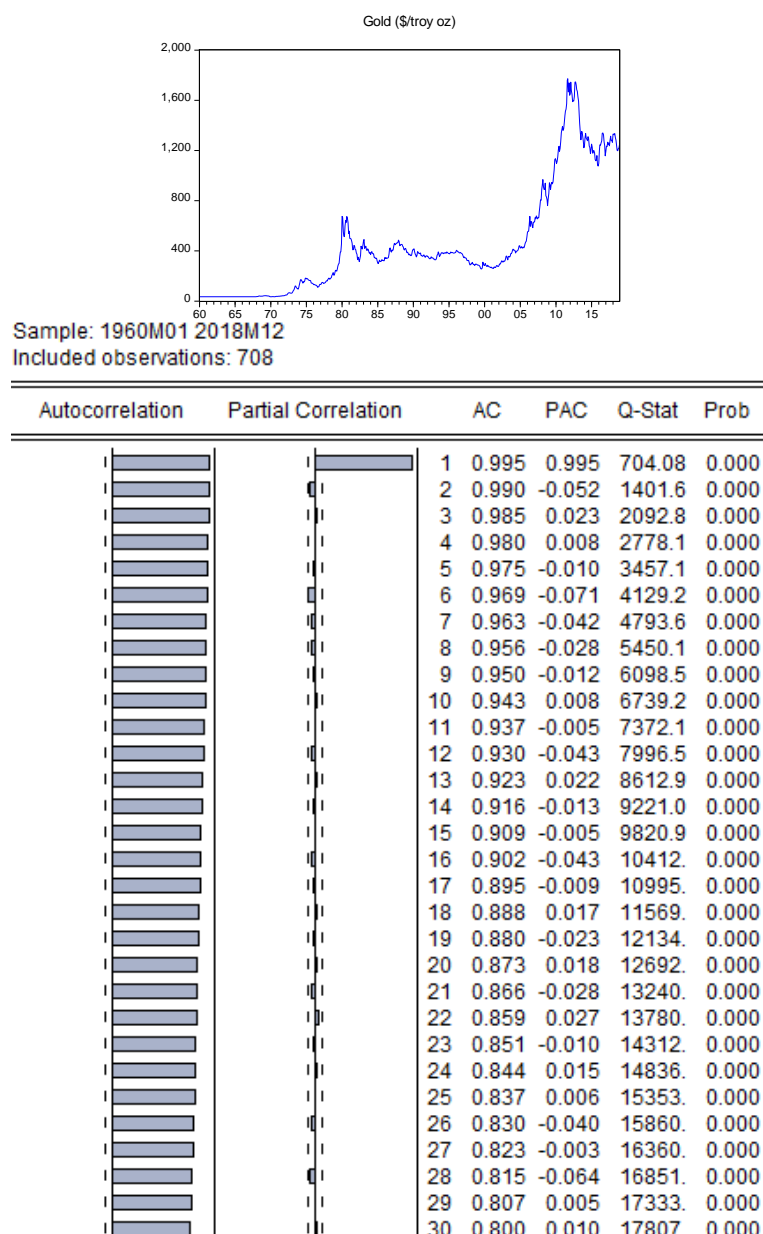
(i) Establishing Stationarity

Graph the Gold time series by selecting **Quick** and then **Graph**. **Plot the autocorrelation function for the Gold** time series. The persistence of autocorrelations at long lags is typical of a non-stationary series. Another name for the autocorrelation function is the correlogram.

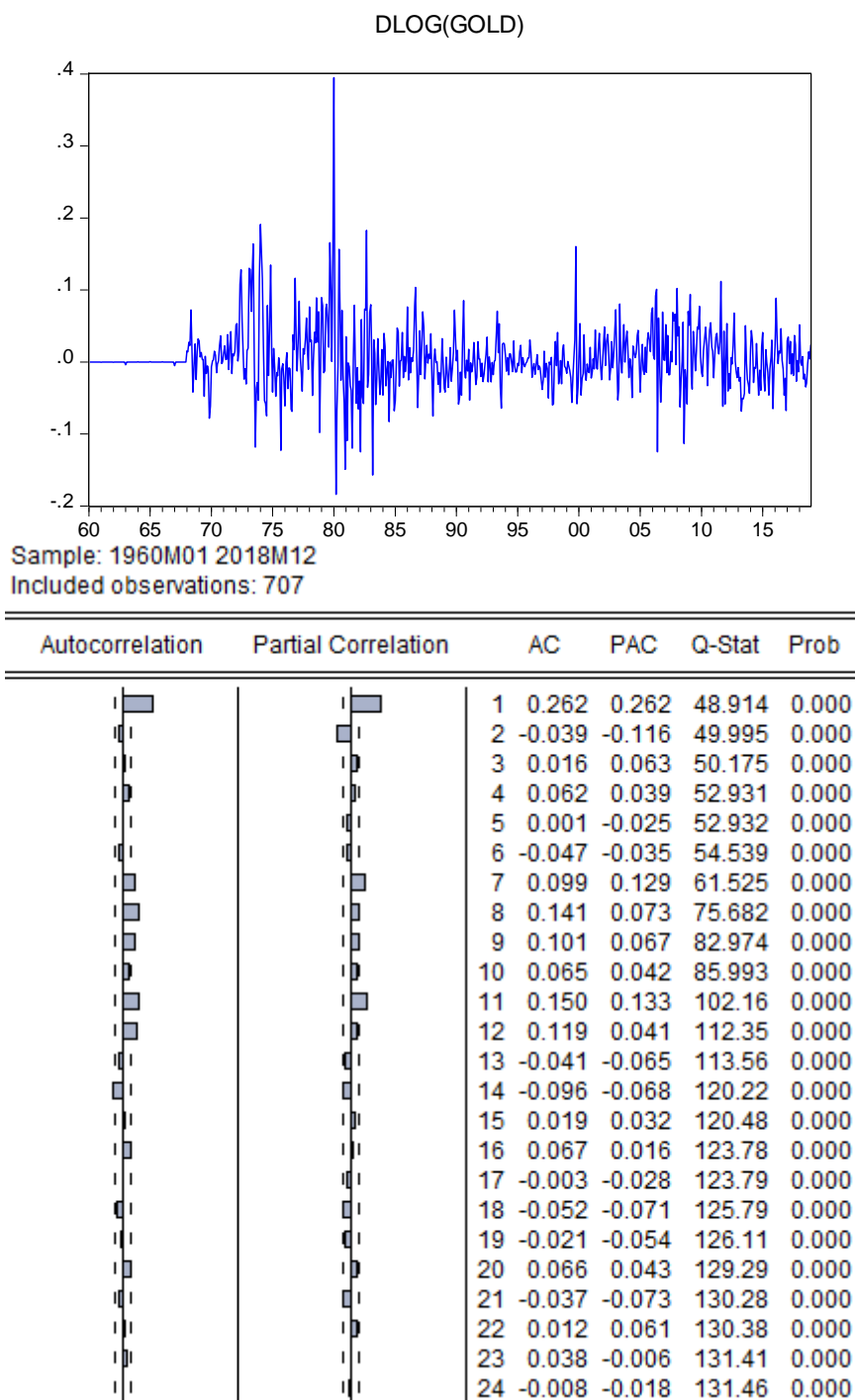
From the main EViews menu, select

Quick Series Statistics Correlogram

Specify the series as **Gold** and click **OK**.

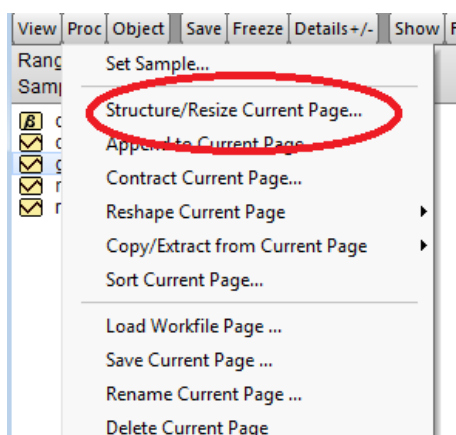


Graph the log returns of the series, DLOG(gold). The series looks stationary. **Plot the autocorrelation function of DLOG(gold).** The autocorrelations are not very persistent which suggests stationarity so let's build a model for DLOG(gold). (If Dickey-Fuller tests were performed, they would support stationarity in the mean.)

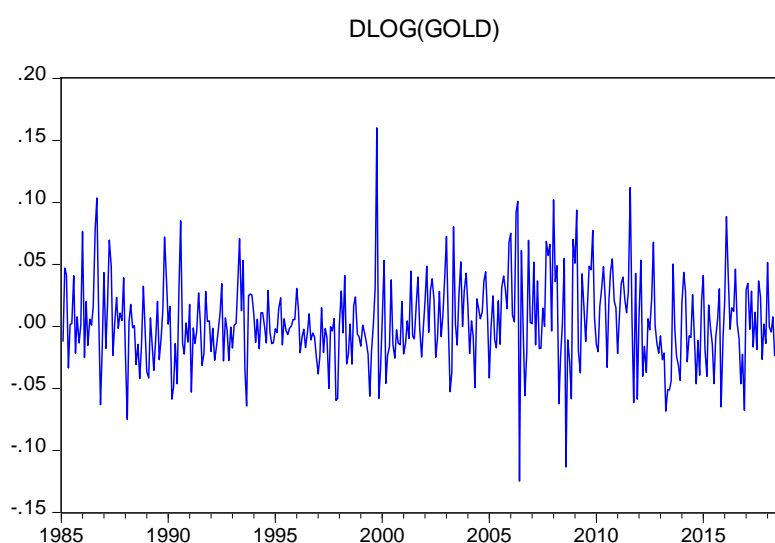
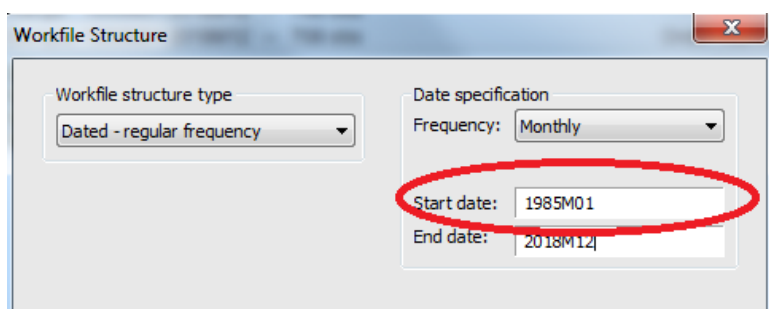


You may want to remove the initial data, as it is not representative.

From the Workfile Window menu, select
Proc Structure/Resize Current Page



Set the new workfile structure type to start from 1985M01











































































(ii) Identification

Plot the autocorrelation function (ACF) and partial autocorrelation function (PACF) for the series DLOG(gold).

Both of these can be found by selecting from the main EViews menu

Quick Series Statistics Correlogram

Sample: 1985M01 2018M12
Included observations: 407

Autocorrelation	Partial Correlation		AC	PAC	Q-Stat	Prob
		1	0.136	0.136	7.6014	0.006
		2	-0.085	-0.106	10.580	0.005
		3	-0.000	0.028	10.580	0.014
		4	0.047	0.035	11.506	0.021
		5	0.018	0.009	11.647	0.040
		6	0.005	0.009	11.657	0.070
		7	0.072	0.074	13.829	0.054
		8	0.024	0.001	14.060	0.080
		9	-0.013	-0.004	14.130	0.118
		10	0.009	0.013	14.163	0.166
		11	0.154	0.148	24.152	0.012
		12	0.030	-0.017	24.534	0.017
		13	-0.001	0.027	24.534	0.027
		14	0.035	0.027	25.056	0.034
		15	0.045	0.027	25.923	0.039
		16	0.042	0.034	26.664	0.045
		17	0.019	0.015	26.820	0.061
		18	0.057	0.038	28.209	0.059
		19	0.021	0.002	28.390	0.076
		20	0.081	0.089	31.181	0.053
		21	0.027	-0.002	31.496	0.066
		22	0.036	0.020	32.060	0.076
		23	0.037	0.025	32.656	0.087
		24	0.047	0.037	33.608	0.092
		25	0.009	-0.016	33.647	0.116
		26	0.062	0.064	35.321	0.105
		27	-0.010	-0.056	35.367	0.130
		28	0.001	0.013	35.368	0.159
		29	-0.001	-0.030	35.368	0.193
		30	-0.024	-0.031	35.615	0.221
		31	0.052	0.025	36.817	0.218
		32	-0.005	-0.033	36.829	0.255
		33	-0.043	-0.052	37.634	0.265
		34	-0.047	-0.056	38.641	0.268
		35	0.051	0.038	39.790	0.265
		36	-0.032	-0.073	40.258	0.287

The PACF and ACF function have two significant spikes and then cut-off (even though the ACF seems to fading away). However, we still have some spikes appearing later on, and also at lags 7, 11 and 20.

A number of ARMA processes could result in this pattern. Verify that reasonable candidate models are: ARMA(2,0), ARMA(0,2), ARMA(1,0), ARMA(0,1), and ARMA(1,1).

(iii) Estimation

Models can be estimated by selecting from the main EViews menu

Quick Estimate Equation

and in the Equation Specification dialog box, we might choose to fit an ARMA(0,1)

DLOG(gold) c MA(1)

(iv) Testing

Both of the coefficients are significant, so the model looks quite reasonable. However, as in regression analysis, we must consider the residuals.

The quickest way to graph the residuals is to select from the menu bar of the equation window

View Actual, Fitted, Residual Actual, Fitted, Residual Graph

Do the **residuals** look random?

We should check for **autocorrelation in the residuals**. From the menu bar of the equation window, select

View Residual Diagnostics Correlogram

Does the correlogram suggest random residuals?

The **Box-Pierce Q-statistic** is shown to the right of the ACF and PACF plots. This statistic considers autocorrelation in several lags together. The p-value of the Q-statistic is given for convenience. When testing at 5% significance, a p-value less than 5% indicates significance. Are any of the Q-statistics significant?

Sample: 1985M01 2018M12
Included observations: 407
Q-statistic probabilities adjusted for 1 ARMA term

Autocorrelation	Partial Correlation	AC	PAC	Q-Stat	Prob
		1	-0.014	-0.014	0.0774
		2	-0.083	-0.084	2.9417
		3	0.006	0.004	2.9572
		4	0.044	0.037	3.7588
		5	0.012	0.015	3.8229
		6	-0.009	-0.002	3.8558
		7	0.071	0.073	5.9516
		8	0.014	0.013	6.0305
		9	-0.013	-0.002	6.0984
		10	-0.014	-0.013	6.1854
		11	0.155	0.150	16.262
		12	0.006	0.005	16.277
		13	-0.007	0.020	16.296
		14	0.030	0.028	16.689
		15	0.034	0.025	17.181
		16	0.035	0.037	17.715
		17	0.004	0.014	17.723
		18	0.056	0.042	19.070
		19	-0.001	-0.006	19.071
		20	0.079	0.087	21.741
		21	0.009	0.011	21.778
		22	0.030	0.016	22.175
		23	0.025	0.021	22.446
		24	0.044	0.045	23.295
		25	-0.009	-0.021	23.332
		26	0.067	0.069	25.267
		27	-0.022	-0.047	25.478
		28	0.003	0.007	25.483
		29	0.004	-0.022	25.491
		30	-0.034	-0.040	25.989
		31	0.059	0.024	27.526
		32	-0.009	-0.021	27.565
		33	-0.032	-0.046	28.030
		34	-0.052	-0.071	29.253
		35	0.065	0.038	31.144
		36	-0.033	-0.054	31.638

Take note of the **RSS**, **SBC**, **AIC**, and **HQ**; these indicate the quality of the fit of the model: the lower the better. SBC, AIC and HQ can be negative.

Try to find an alternative model which has better or comparable fit, and which also has no residual autocorrelation. The quickest way to estimate another model is to select **Estimate** from the menu bar of the equation window; the Equation Specification dialog box should reappear.

ARMA(2, 0): **DLOG(gold) c AR(1) AR(2)**
ARMA(0, 2): **DLOG(gold) c MA(1) MA(2)**
ARMA(1, 0): **DLOG(gold) c AR(1)**
ARMA(1, 1): **DLOG(gold) c AR(1) MA(1)**

(v) Forecasting

Estimate an ARMA(0,1) model for the DLOG(gold) series using all the data **except the last 12 observations**. You can specify the shortened sample at the bottom of the equation specification dialog box.

Equation Estimation

Specification Options

Equation specification
Dependent variable followed by list of regressors including ARMA and PDL terms, OR an explicit equation like $Y=c(1)+c(2)*X$.

dlog(gold) c ma(1)

Estimation settings
Method: LS - Least Squares (NLS and ARMA)
Sample: 1985m01 2017m12

Equation Estimation

Specification Options

Coefficient covariance
Covariance method: Ordinary
Information matrix: OPG
☒ d.f. Adjustment

ARMA
Method: GLS
Starting ARMA coefficient values: Automatic

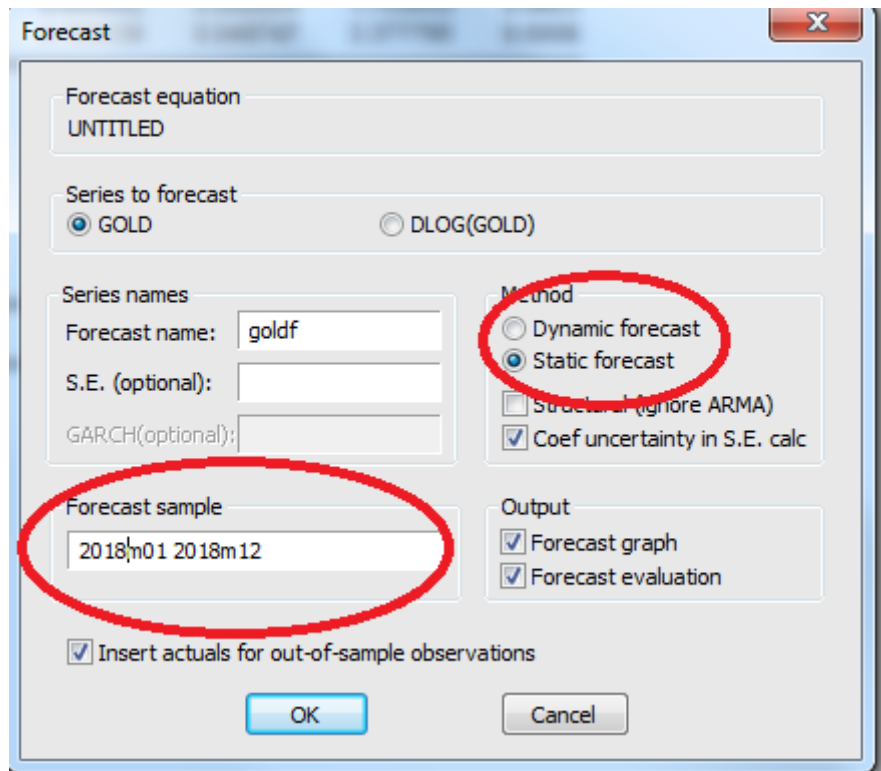
Optimization
Optimization method: BFGS
Step method: Marquardt
Maximum iterations: 500
Convergence tolerance: 0.0001
☐ Display settings in output

Coefficient name
c

OK Cancel

Click OK. Check that the model has reasonable diagnostics.

Click the **Forecast** button which can be found in the menu at the top of the current equation window.



In the top right of the dialog box, change the forecasting **Method** to **Static** and click OK.

Now try a **Dynamic** forecast for gold.

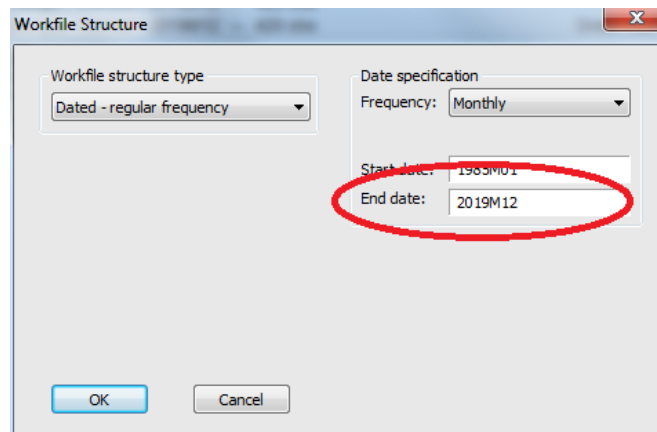
Both Dynamic and Static forecasting produce 12 forecasts but they are not the same. Dynamic produces a 1-step-ahead forecast, a 2-step-ahead forecast, a 3-step-ahead forecast,, and a 12-step-ahead forecast. By contrast, Static forecasting produces twelve 1-step-ahead forecasts.

Produce Static forecasts for Gold and then plot these forecasts, goldF, and gold on the same graph by selecting **Quick Graph**. How good is the fit?

Produce Static forecasts for DLOG(gold) by specifying this at the top left of the forecast dialog box. Plot the forecasts and DLOG(gold) on the same graph. How good is the fit?

Do you expect gold prices to go up or down in the next 12 months?
How about risk? What is the prediction interval?

Click: *workfile window ... Procs ... Structure/Resize Current Page*
 Date Specification. **End Date: 2019:M12**



Once again, estimate an ARMA(0,1) model for the DLOG(gold) series, but now using all the data (**2018M12**).

Click the **Forecast** button which can be found in the menu at the top of the current equation window. [FORECAST observations **2019m01 to 2019m12**]. In the top right of the dialog box, try a **Dynamic** forecasting **Method** and click OK.

Why can't you use the **Static** Forecasting method?