

# ***EViews tutorial: time series analysis***

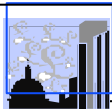
***Professor Roy Batchelor***

***City University Business School, London***

***& ESCP, Paris***

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## ***EViews***

☐ On the City University system, EViews 3.1 is in  
Start/ Programs/ Departmental Software/CUBS

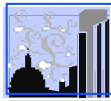
☐ Earnings time series in FT500M.wf1

☐ Box Jenkins analysis  
– ACF, PACF  
– comparing models

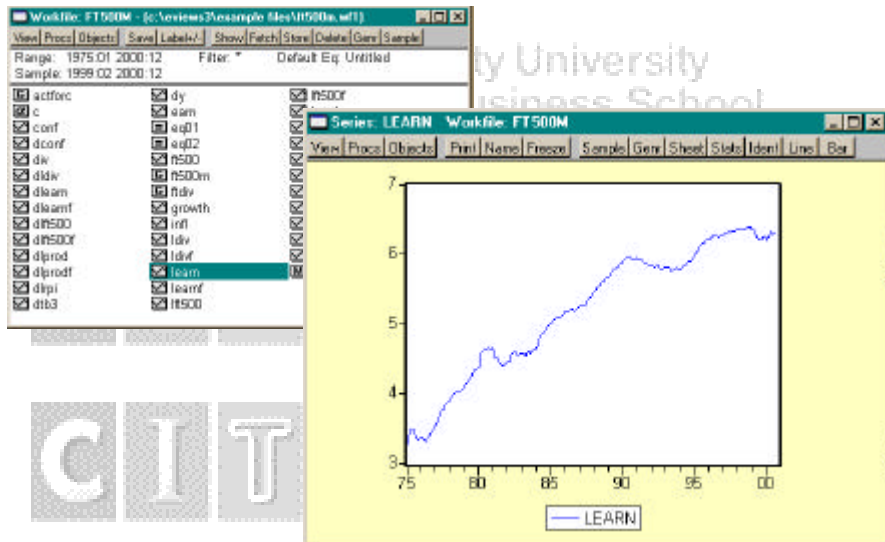
☐ Exponential smoothing models

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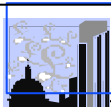


## Log-Earnings data



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## Box-Jenkins (ARIMA) modeling



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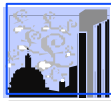
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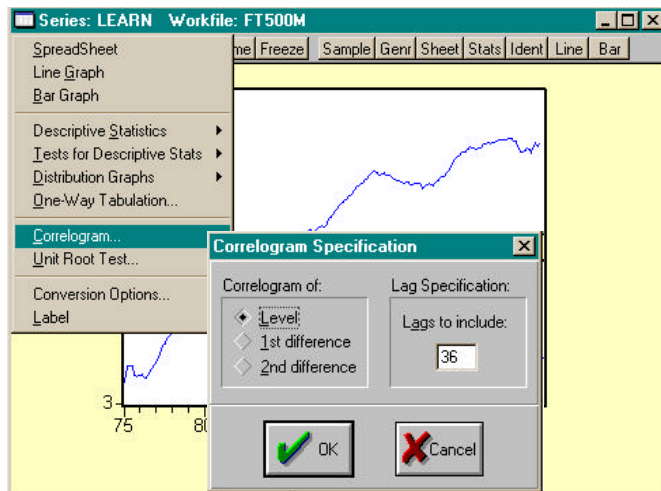
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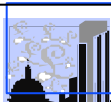


## Correlogram command

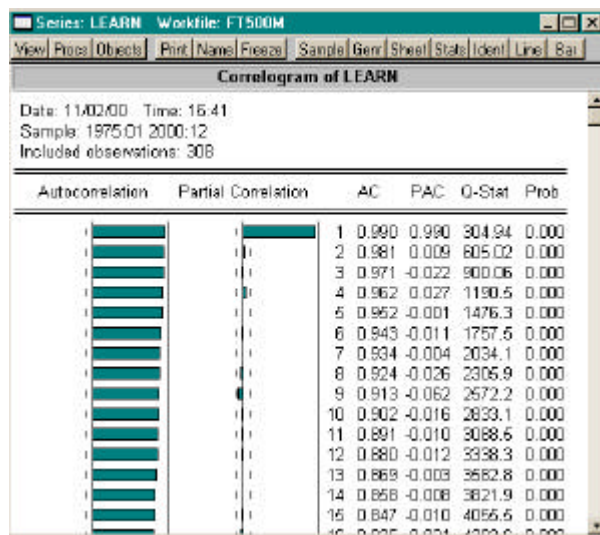


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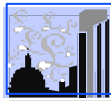


## ACF, PACF for level variable:LEARN

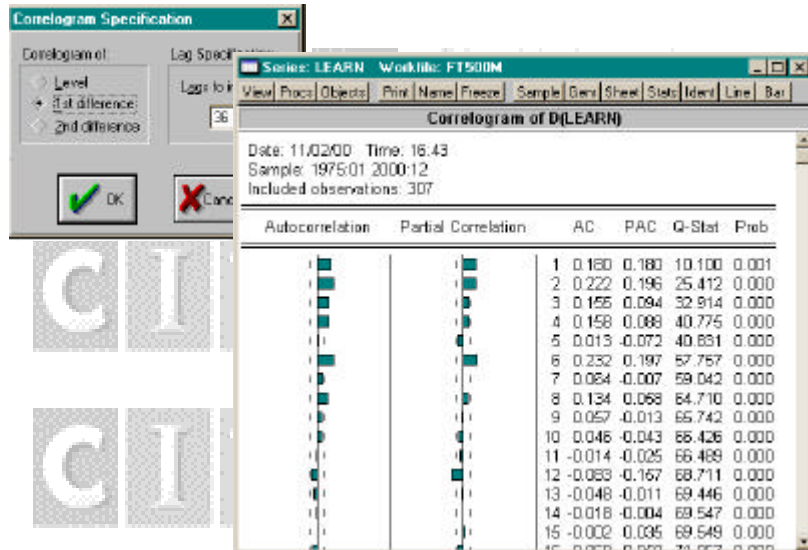


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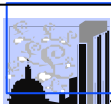


## ACF, PACF for 1st difference: DLEARN



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## ARIMA: the Eviews notation

### ARIMA Theory

ARIMA (autoregressive integrated moving average) models are generalizations of the simple AR model that use three tools for modeling the serial correlation in the disturbance:

- The first tool is the autoregressive, or AR, term. The AR(1) model introduced above uses only the first-order term but, in general, you may use additional, higher-order AR terms. Each AR term corresponds to the use of a lagged value of the residual in the forecasting equation for the unconditional residual. An autoregressive model of order  $p$ , AR( $p$ ), has the form

$$u_t = \rho_1 u_{t-1} + \rho_2 u_{t-2} + \dots + \rho_p u_{t-p} + \epsilon_t$$

- The second tool is the integration order term. Each integration order corresponds to differencing the series being forecast. A first-order integrated component means that the forecasting model is designed for the first difference of the original series. A second-order component corresponds to using second differences, and so on.

- The third tool is the MA, or moving average term. A moving average forecasting model uses lagged values of the forecast error to improve the current forecast. A first-order moving average term uses the most recent forecast error, a second-order term uses the forecast error from the two most recent periods, and so on. An MA( $q$ ) has the form:

$$u_t = \epsilon_t + \theta_1 \epsilon_{t-1} + \theta_2 \epsilon_{t-2} + \dots + \theta_q \epsilon_{t-q}$$

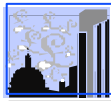
Some authors and software packages use the opposite sign convention for the  $\theta$  coefficients so that the interpretation of the signs of the MA coefficients may be reversed.

The autoregressive and moving average specifications can be combined to form an ARIMA( $p,q$ ) specification

$$u_t = \rho_1 u_{t-1} + \rho_2 u_{t-2} + \dots + \rho_p u_{t-p} + \epsilon_t + \theta_1 \epsilon_{t-1} + \theta_2 \epsilon_{t-2} + \dots + \theta_q \epsilon_{t-q}$$

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## AR(6) model for DLEARN

**Equation Specification**

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

DLEARN C ar(1) ar(2) ar(3) ar(4) ar(5) ar(6)

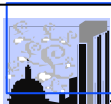
Estimation Settings:  
Method: LS - Least Squares (NLS and ARMA)  
Sample: 1975:01 2000:12

OK Cancel Options

This is ARIMA(6, 1, 0) for EARN  
or equivalently  
ARIMA(6, 0, 0) for DLEARN  
or equivalently  
AR(6) for DLEARN

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## AR(6) model output

**Equation: UNTITLED** Workfile: FT500M

View Proc Objects Print Name Release Estimate Forecast Stats Resids

Dependent Variable: DLEARN  
Method: Least Squares  
Date: 11/02/00 Time: 16:50  
Sample (adjusted): 1975:08 2000:08  
Included observations: 301 after adjusting endpoints  
Convergence achieved after 3 iterations

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	0.009416	0.003164	2.976260	0.0032
AR(1)	0.162254	0.057336	2.829696	0.0050
AR(2)	0.143496	0.057773	2.483600	0.0136
AR(3)	0.082823	0.058684	1.411346	0.1592
AR(4)	0.066240	0.058446	1.173503	0.2415
AR(5)	-0.103060	0.055840	-1.935162	0.0539
AR(6)	0.209015	0.055818	3.744591	0.0002

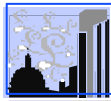
R-squared	0.143314	Mean dependent var	0.003450
Adjusted R-squared	0.125830	S.D. dependent var	0.026075
S.E. of regression	0.024379	Akaike info criterion	-4.567207
Sum squared resid	0.174735	Schwarz criterion	-4.480995
Log likelihood	694.3647	F-statistic	8.197136
Durbin-Watson stat	1.965974	Prob(F-statistic)	0.000000

Inverted AR Roots				
.84	.41 - .57i	.41 + .57i	-.34 - .70i	
-.34 + .70i	-.82			

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## Seasonal MA model

Equation: UNTITLED Workfile: FT500M

View/Proc/Objects Print Name Freeze Estimate Forecast Stats Reads

Dependent Variable: DLEARN  
 Method: Least Squares  
 Date: 11/02/00 Time: 16:58  
 Sample(adjusted): 1975:02 2000:08  
 Included observations: 307 after adjusting endpoints  
 Convergence achieved after 7 iterations  
 Backcast: 1974:02 1975:01

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	0.010081	0.002030	4.964507	0.0000
MA(1)	0.158360	0.048495	3.199504	0.0016
MA(6)	0.374806	0.057731	6.492319	0.0000
MA(12)	-0.122945	0.057704	-2.130600	0.0339

R-squared	0.134116	Mean dependent var	0.009915
Adjusted R-squared	0.125543	S.D. dependent var	0.027020
S.E. of regression	0.025267	Akaike info criterion	-4.606676
Sum squared resid	0.193445	Schwarz criterion	-4.457117
Log likelihood	895.6211	F-statistic	15.64379
Durbin-Watson stat	1.944395	Prob(F-statistic)	0.000000

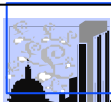
Inverted MA Roots				
.77 -.46i	.77+.46i	.76	.38+.67i	
.36 -.67i	-.02+.91i	-.02 -.91i	-.39+.67i	
-.39 -.67i	-.76	-.61 -.46i	-.81+.46i	

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This model is

$$\text{DLEARN}_t = r_0 + u_t + \theta_1 u_{t-1} + \theta_6 u_{t-6} + \theta_{12} u_{t-12}$$



## Mixed AR/MA model

Equation: UNTITLED Workfile: FT500M

View/Proc/Objects Print Name Freeze Estimate Forecast Stats Reads

Dependent Variable: DLEARN  
 Method: Least Squares  
 Date: 11/02/00 Time: 17:04  
 Sample(adjusted): 1975:03 2000:08  
 Included observations: 306 after adjusting endpoints  
 Convergence achieved after 8 iterations  
 Backcast: 1974:09 1975:02

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	0.009095	0.003292	2.762929	0.0061
AR(1)	0.636316	0.081055	7.860416	0.0000
MA(1)	-0.902642	0.081350	-5.175739	0.0000
MA(6)	0.354359	0.060216	7.066672	0.0000

R-squared	0.169043	Mean dependent var	0.009793
Adjusted R-squared	0.160789	S.D. dependent var	0.026980
S.E. of regression	0.024716	Akaike info criterion	-4.549742
Sum squared resid	0.164487	Schwarz criterion	-4.601067
Log likelihood	700.1105	F-statistic	20.47982
Durbin-Watson stat	2.026506	Prob(F-statistic)	0.000000

Inverted AR Roots				
.64				
.83 -.40i	.83+.40i	.08+.62i	.06 -.83i	

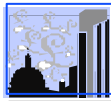
Inverted MA Roots				
-.66+.41i	-.66-.41i			

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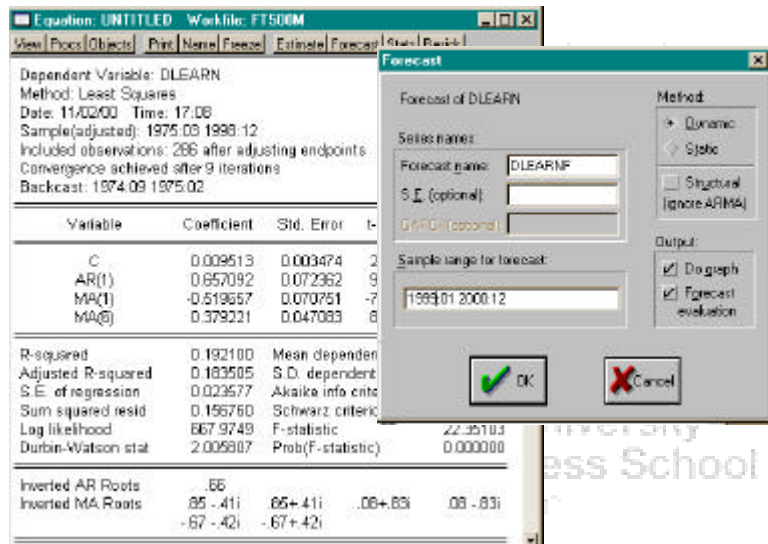
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This model is

$$\text{DLEARN}_t = r_0 + \rho_1 \text{DLEARN}_{t-1} + u_t + \theta_1 u_{t-1} + \theta_6 u_{t-6}$$



## Forecast evaluation

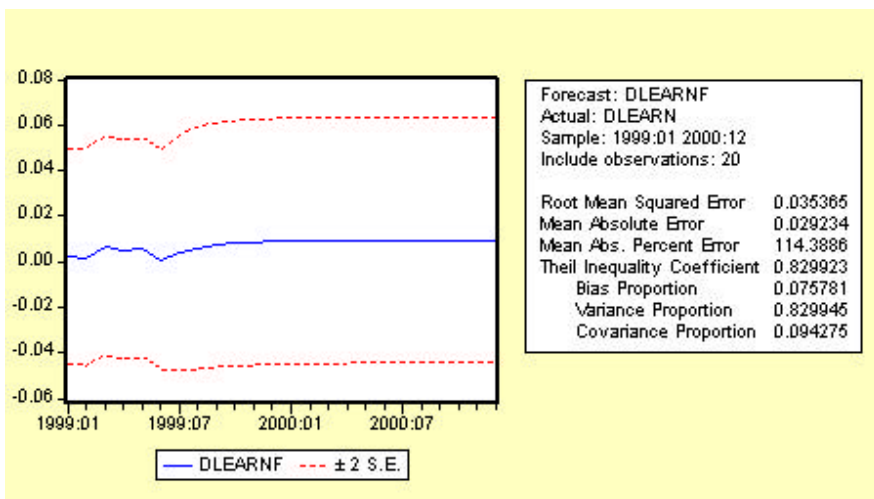


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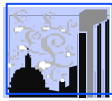


## Multiperiod out of sample forecasts: 1999-2000

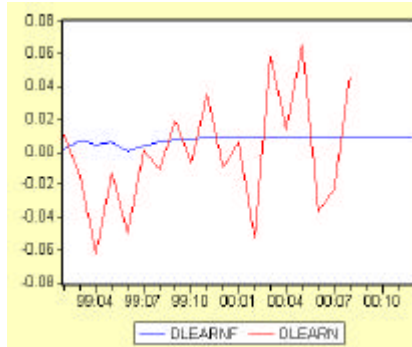


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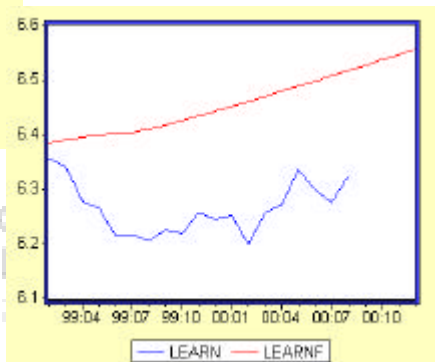


## Multiperiod forecasts v. actual



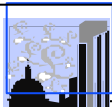
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## One-period ahead forecasts

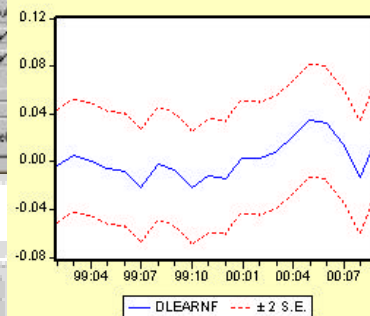
Forecast of DLEARN

Series names:  
Forecast name: DLEARNF  
S.E. (optional):  
BARCH (optional):

Sample range for forecast:  
1999:02 2000:12

Method:  
Dynamic  
+ Static  
Structural  
(ignore AFMA)

OK Cancel



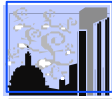
Forecast: DLEARNF  
Actual: DLEARN  
Sample: 1999:02 2000:09  
Include observations: 19

Root Mean Squared Error	0.037482
Mean Absolute Error	0.030890
Mean Abs. Percent Error	193.3657
Theil Inequality Coefficient	0.739447
Bias Proportion	0.002691
Variance Proportion	0.277181
Covariance Proportion	0.720128

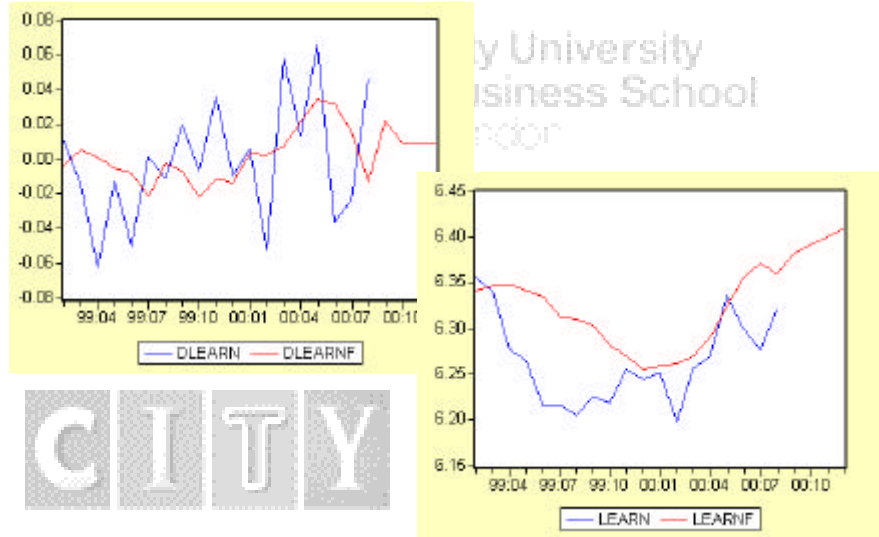
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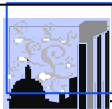


## *One-period ahead forecasts v. actual*



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## *Exponential smoothing models*



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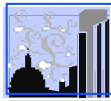
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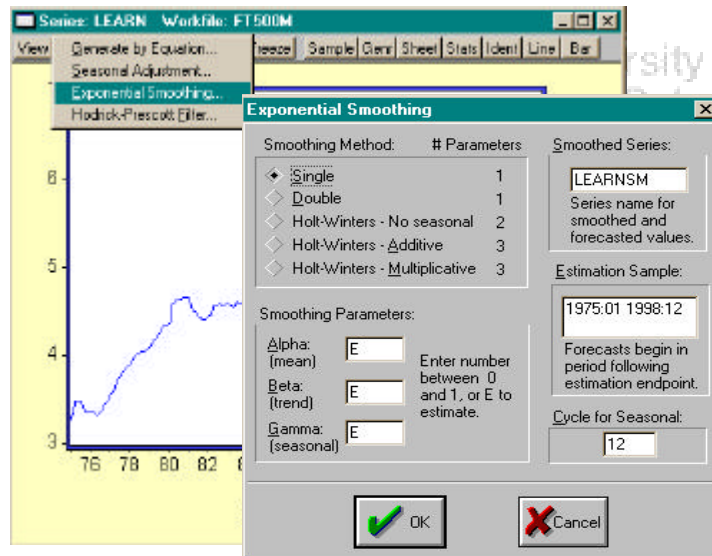
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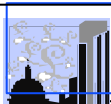


## PROCS/ Exponential smoothing menu

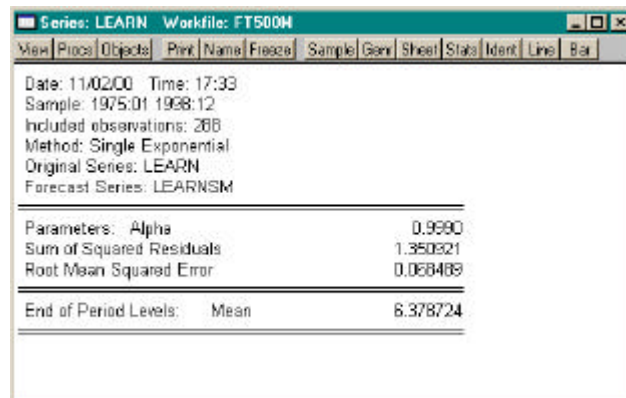


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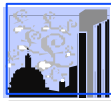


## Single exponential smoothing

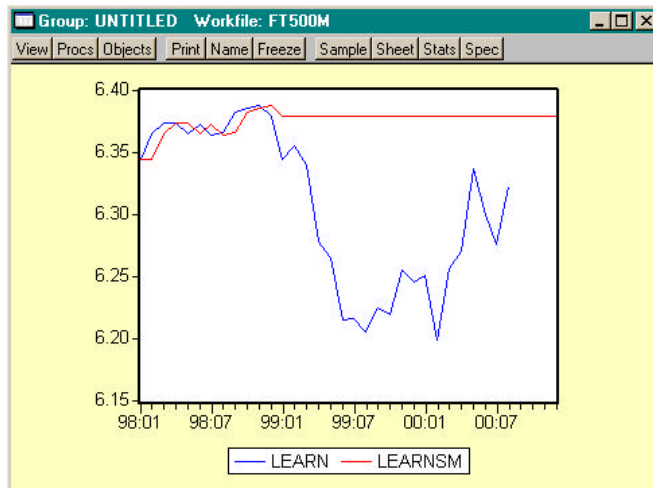


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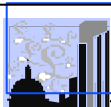


## SES forecast for 1999-2000



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## Holt linear trend model

**Exponential Smoothing**

Smoothing Method: # Parameters

- Single 1
- Double 1
- Holt-Winters - No seasonal 2**
- Holt-Winters - Additive 3
- Holt-Winters - Multiplicative 3

Smoothing Parameters:

Alpha (mean)  Enter number between 0 and 1, or E to estimate

Beta (trend)

Gamma (seasonal)

OK Cancel

**Series: LEARN Workfile: FT500M**

Date: 11/02/00 Time: 17:41

Sample: 1975:01 1998:12

Included observations: 288

Method: Holt-Winters No Seasonal

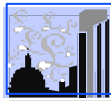
Original Series: LEARN

Forecast Series: LEARNSM

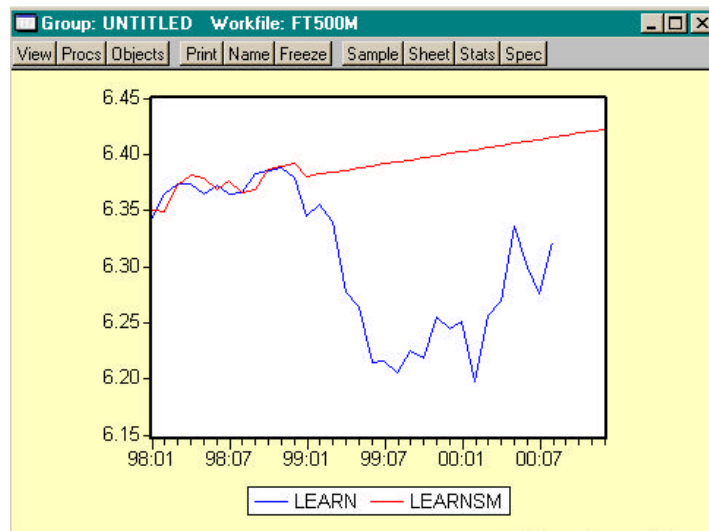
Parameters:	Alpha	1.0000
	Beta	0.1700
	Sum of Squared Residuals	0.187256
	Root Mean Squared Error	0.025499
End of Period Levels:	Mean	6.378715
	Trend	0.001829

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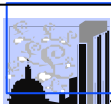


## Holt linear trend forecast for 1999-2000

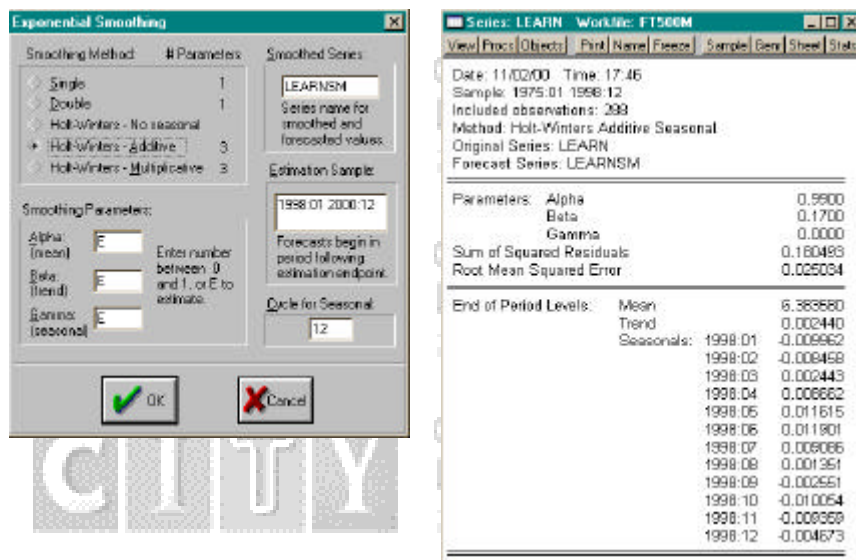


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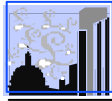


## Holt-Winters seasonal model

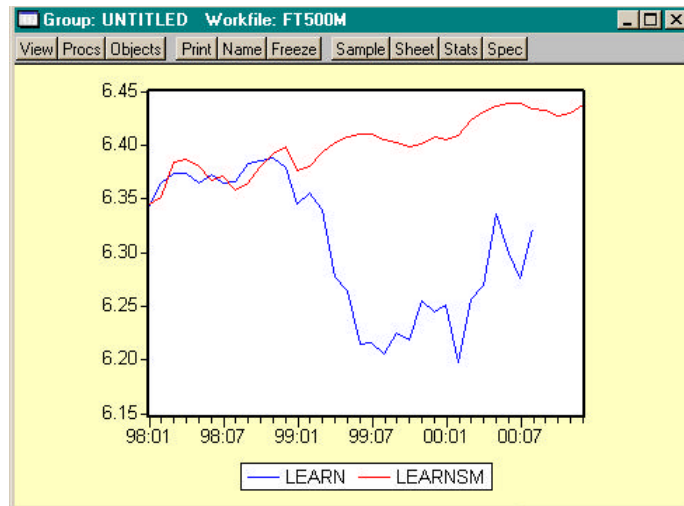


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## *Holt-Winters forecast for 1999-2000*



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