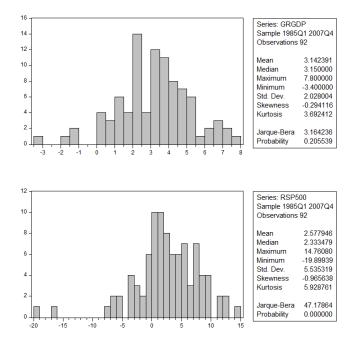
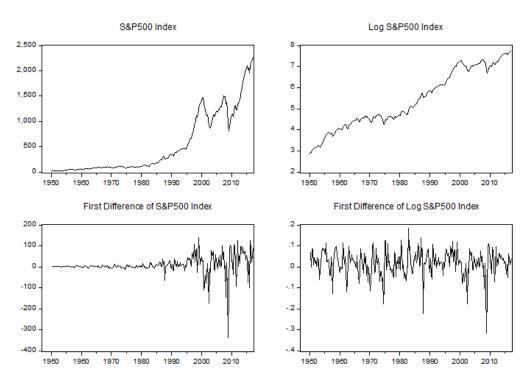
Practice Problems 1

- Question 1. Explain the concepts of point forecast, interval forecast, density forecast.
- Question 2. Define first order and second order weakly stationary processes.
- Question 3. Define white noise.
- Question 4. Explain what loss function is.
- Question 5. Give two examples of loss function, one symmetric, one asymmetric.
- Question 6. Consider Fed forecasting inflation. Is it likely to have (1) a symmetric loss function, or (2) an asymmetric loss function with larger losses for negative forecast errors, or (3) an asymmetric loss function with larger losses for positive forecast errors? Explain.
- Question 7. Consider Congressional Budget Office producing forecasts of future budget deficits. Is it likely to have a symmetric loss function or are the relative costs of over- and under-predicting public deficits different, and the loss function is thus asymmetric? Explain.
- Question 8. Explain how increasing ϕ_1 in an AR(1) model changes the behavior of time series Y_t .
- Question 9. Define an AR(2) model and describe how its AC and PAC functions look like.
- Question 10. Define an MA(4) model and describe how its AC and PAC functions look like.
- Question 11. Explain the role of the adjusted R^2 , AIC and SIC, in model selection.

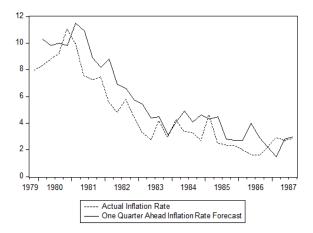
Question 14. Figure below shows the histograms for the real GDP growth rate and the quarterly return for S&P500 Index during the period 1985Q1-2007Q4. Is the GDP growth rate normally distributed in this sample? How about the returns for S&P500 Index?



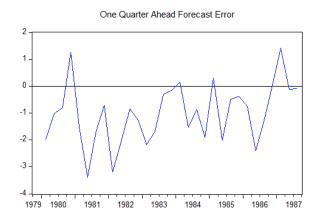
Question 15. Figure below shows the time series for the S&P500 Index, the log transformed S&P500 Index, and also their first differences. Explain which of the four series are nonstationary, first order weakly stationary, second order weakly stationary.



Question 16. Consider the Fed's one quarter ahead forecast for inflation during the 1979Q4-1987Q3 period.



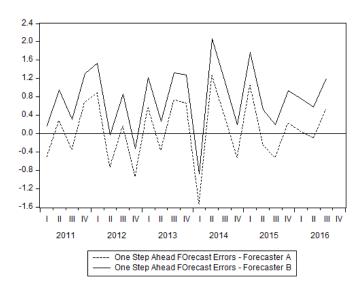
Suppose that we want to test whether the Fed's forecast are optimal under the symmetric quadratic loss function, which would imply that $E(y_{t+1}) = f_{t,1}$ and thus the forecast error $e_{t,1} = y_{t+1} - f_{t,1}$ would have to satisfy $E(e_{t,1}) = 0$. In other words, if the Fed's forecast are optimal under the symmetric quadratic loss function, the forecast error $e_{t,1}$ should fluctuate around zero, have zero mean, and in the regression $e_{t,1} = \beta_0 + e_t$ coefficient β_0 should be zero. Figure below shows that time series plot for the forecast errors, and the results of that regression. Interpret these results; what can we say about Fed's loss function during 1979Q4-1987Q3 based on them?



Dependent Variable: GPGDP_E1 Method: Least Squares Date: 02/24/17 Time: 19:34 Sample (adjusted): 1980Q1 1987Q3 Included observations: 31 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
С	-1.017073	0.202722	-5.017080	0.0000
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood Durbin-Watson stat	0.000000 0.000000 1.128708 38.21948 -47.23215 1.562466	Mean depend S.D. depende Akaike info cri Schwarz crite Hannan-Quin	ent var iterion rion	-1.017073 1.128708 3.111751 3.158009 3.126830

Question 17. Consider two forecasters, A and B, who use the same AR model to forecast the real GDP growth rate during 2011Q1-2016Q4, but produce different forecasts, $f_{t,1}^{(A)} = \mu_{t+1|t}$ and $f_{t,1}^{(B)} = \mu_{t+1|t} - \sigma_{t+1|t}^2$, where $\mu_{t+1|t} = E(y_{t+1}|I_t)$ is the conditional mean, $\sigma_{t+1|t}^2 = var(y_{t+1}|I_t)$ the conditional variance. The forecast errors are thus $e_{t,1}^{(A)} = y_{t+1} - \mu_{t+1|t}$ and $e_{t,1}^{(A)} = y_{t+1} - \mu_{t+1|t} + \sigma_{t+1|t}^2$ shown below. Based the forecasts they choose and their forecasting errors, what can we say about the loss functions of these two forecasters - are they symmetric or asymmetric?



Question 18. Figure below show the correlogram for the percentage change in the house price index in San Diego MSA during 1975Q1-2008Q3. Discuss which AR/MA/ARMA models would you consider as plausible candidates for this time series and explain why.

Date: 02/25/17 Time: 13:30 Sample: 1975Q1 2008Q3

Included observations: 134									
Autocorrelation	Partial Correlation	AC		PAC	Q-Stat	Prob			
		1	0.487	0.487	32.436	0.000			
; =		3	0.487 0.403	0.328 0.123	65.201 87.756	0.000			
		5		0.225 -0.142	118.26 127.65	0.000			
: 📙		6	0.277 0.265	-0.001 0.075	138.56 148.65	0.000			
·	'III ' '(('	8		-0.093 -0.041	153.61 155.51	0.000			
, j j.,		10 11		-0.115 -0.090	155.86 155.88	0.000			
. d .			-0.065 -0.073		156.51 157.32	0.000			
<u> </u>			-0.124 -0.157		159.65 163.41	0.000			
		16	-0.164 -0.123	0.006 0.077	167.55 169.90	0.000			
		18	-0.176 -0.227	-0.027	174.75 182.90	0.000			
급:	'5		-0.22 <i>1</i> -0.117	0.123	185.07	0.000			

Question 19. Figure below shows the correlogram for the residuals from AR(2) and AR(4) models for the percentage change in the house price index in San Diego MSA. For a good model, the residuals should be white noise with no time dependence. Do the residuals from AR(2) and AR(4) model satisfy this property?

Date: 02/25/17 Time: 13:32 Sample: 1975Q1 2008Q3 Included observations: 134 residuals from AR(2) model

Date: 02/25/17 Time: 13:31 Sample: 1975Q1 2008Q3 Included observations: 134 residuals from AR(4) model

Autocorrelation	Partial Correlation	AC	PAC	Q-Stat	Prob	Autocorrelation	Partial Correlation	AC	PAC	Q-Stat	Prob
Autocorrelation	Partial Correlation	1 -0.060 2 -0.157 3 0.081 4 0.272 5 -0.106 6 0.002 7 0.161 8 0.061 9 -0.004 10 -0.001 11 0.022 12 -0.054 13 0.005 14 -0.033 15 -0.074 16 -0.013	-0.060 -0.162 0.062 0.265 -0.053 0.065 0.116 0.030 -0.028 -0.047 -0.078 -0.036 -0.036 -0.036 -0.036 -0.034	0.4987 3.9233 4.8320 15.227 16.821 16.822 20.544 21.086 21.089 21.158 21.599 21.602 21.765 22.599 22.625 23.056	0.028 0.000 0.001 0.002 0.001 0.002 0.004 0.007 0.012 0.017 0.028 0.040 0.047	Autocorrelation	Partial Correlation	1 0.032 2 0.031 3 0.034 4 0.069 5 -0.077 6 0.036 7 0.149 8 0.043 9 0.046 10 -0.005 11 0.002 12 -0.078 13 -0.021 14 -0.038 15 -0.041 16 -0.023 17 0.034	0.032 0.030 0.032 0.066 -0.083 0.036 0.149 0.032 0.044 -0.031 -0.017 -0.063 -0.027 -0.024	0.1413 0.2749 0.4337 1.0957 1.9315 5.2995 5.5614 5.8654 5.8700 6.7815 6.8510 7.0708 7.3252	0.165 0.347 0.151 0.234 0.320 0.438 0.555 0.560 0.653 0.719 0.772 0.829 0.869
		19 -0.182	-0.146		0.026			19 -0.161		12.652 13.079	0.629