

Texas Tech University
Department of Economics
Spring 2017
Eco 4306: Economic and Business Forecasting
Final Exam

Name:

ID:

Short questions (40 points)

- Q1.** 4 points
- Q2.** 4 points
- Q3.** 4 points
- Q4.** 4 points
- Q5.** 4 points
- Q6.** 4 points
- Q7.** 4 points
- Q8.** 4 points
- Q9.** 4 points
- Q10.** 4 points

Applied problems (64 points)

- Q11.** 8 points
- Q12.** 8 points
- Q13.** 8 points
- Q14.** 8 points
- Q15.** 8 points
- Q16.** 8 points
- Q17.** 8 points
- Q18.** 8 points

Good luck!

Question 1. (4 points)

What is Granger causality and how do we test it?

Question 2. (4 points)

What are impulse-response functions?

Question 3. (4 points)

Explain what spurious regression problem is and give an example.

Question 4. (4 points)

Explain what is means if X_t and Y_t are cointegrated. Give an example.

Question 5. (4 points)

Explain the idea behind error correction model. Draw a diagram illustrating the error correction mechanism.

Question 6. (4 points)

Explain what volatility clustering means.

Question 7. (4 points)

Explain the difference between moving average (MA) and exponentially weighted moving average (EWMA) models of the conditional variance.

Question 8. (4 points)

Write the equation for the autoregressive conditional heteroscedasticity ARCH(1) model. Explain the intuition behind this model.

Question 9. (4 points)

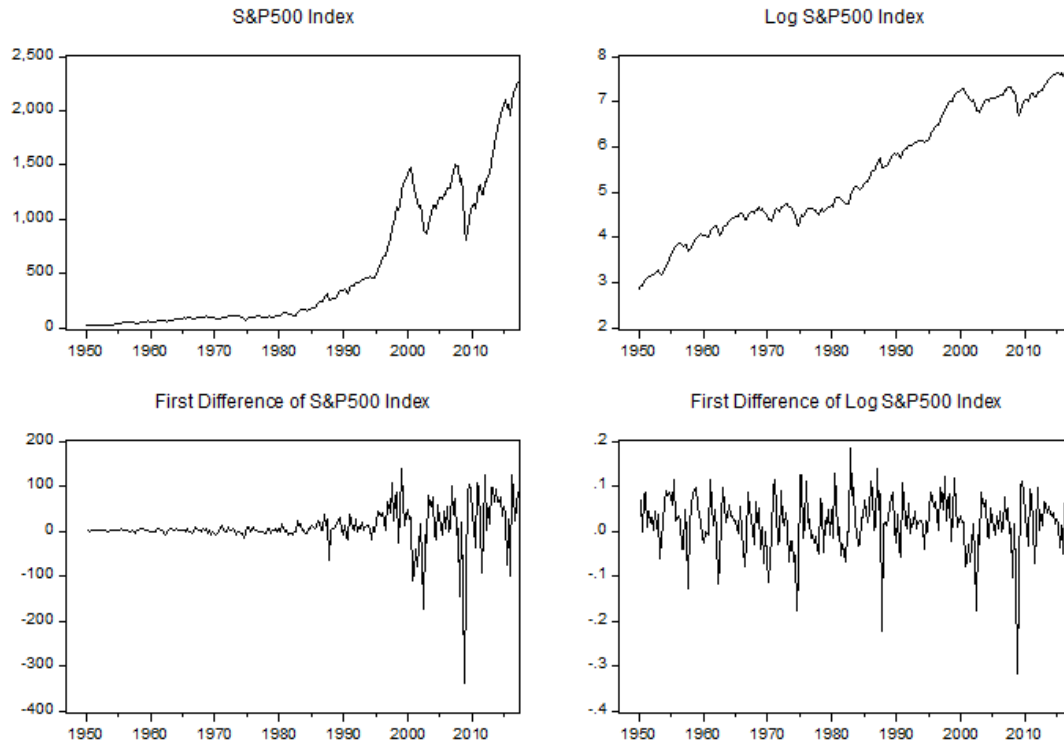
Explain what 1% VaR is and draw a diagram to illustrate this.

Question 10. (4 points)

Consider a GARCH(1,1) model for daily S&P 500 returns from 1/2/1998 to 7/25/2008 sample. With normal innovations, the number of violations $r_t < r_t^{VaR(0.01)}$ is 42 which represents 1.58% of observations. With innovations from Student-t distribution the number of violations $r_t < r_t^{VaR(0.01)}$ is 30 or 1.13% of the sample. Which of these two models would be more suitable for risk management purposes and why?

Question 11 (8 points)

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.



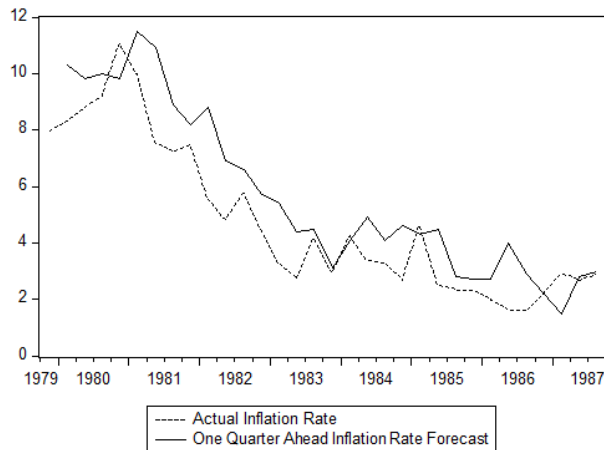
Question 12 (8 points)

Consider the one quarter ahead Fed's 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 we perform an F-test for the joint hypothesis $H_0 : \beta_0 = 0, \beta_1 = 1$ in a regression $y_{t+1} = \beta_0 + \beta_1 f_{t,1} + \varepsilon_{t+1}$ where y_{t+1} is the actual inflation and $f_{t,1}$ is the Fed's 1-quarter-ahead forecast.

Interpret the results of this test below.

What can we say about Fed's loss function during 1979Q4-1987Q3 based on this test?



Wald Test:
Equation: EQ_GPGDP_F1

Test Statistic	Value	df	Probability
F-statistic	17.79592	(2, 29)	0.0000
Chi-square	35.59184	2	0.0000

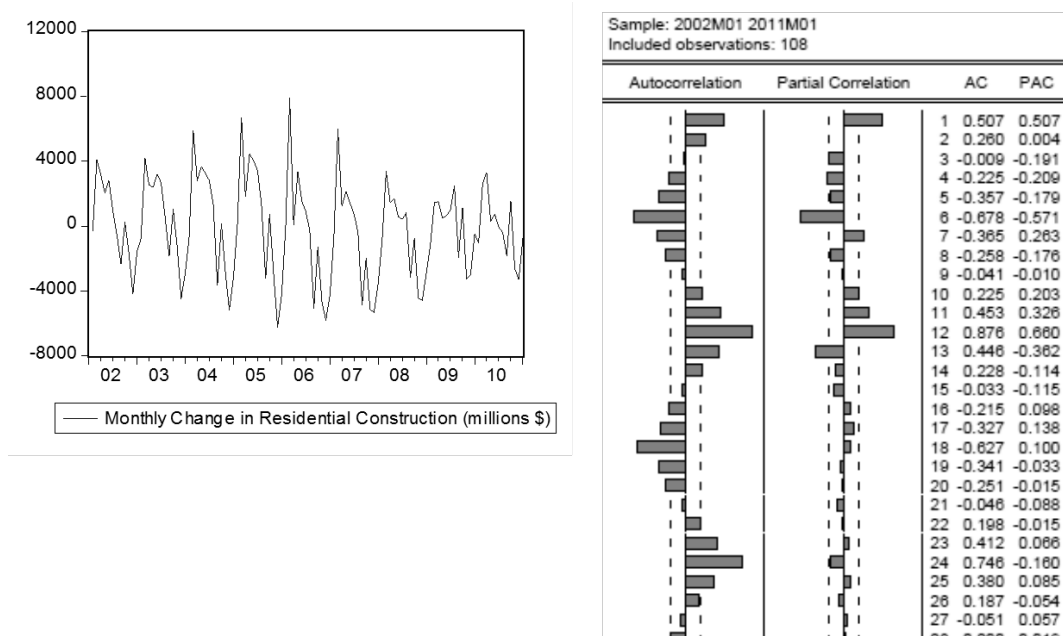
Null Hypothesis: C(1)=0, C(2)=1
Null Hypothesis Summary:

Normalized Restriction (= 0)	Value	Std. Err.
C(1)	-0.117978	0.408818
-1 + C(2)	-0.158453	0.064037

Restrictions are linear in coefficients.

Question 13 (8 points)

Consider the data for monthly changes in U.S. residential construction for the period January 2002-January 2011 shown below. Discuss what kind of model you would estimate for this time series, write its equation, and explain why you would choose this model.



Question 14 (8 points)

Consider two candidate models for change in private residential construction spending, AR(1)+SAR(1) and AR(2)+SAR(1), the results for which are below. Discuss which of these models would be preferred based on the correlograms of residuals, AIC and BIC, and statistical significance of coefficients.

Dependent Variable: DCONST
Method: ARMA Maximum Likelihood (BFGS)
Date: 04/08/17 Time: 16:14
Sample: 1993M02 2013M12
Included observations: 251
Convergence achieved after 5 iterations
Coefficient covariance computed using outer product of gradients

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	218.5690	1235.621	0.176890	0.8597
AR(1)	0.563509	0.044817	12.57349	0.0000
SAR(12)	0.944410	0.013547	69.71517	0.0000
SIGMASQ	379202.0	27048.41	14.01938	0.0000

R-squared	0.933686	Mean dependent var	49.44223
Adjusted R-squared	0.932881	S.D. dependent var	2396.078
S.E. of regression	620.7600	Akaike info criterion	15.82347
Sum squared resid	95179704	Schwarz criterion	15.87965
Log likelihood	-1981.845	Hannan-Quinn criter.	15.84608
F-statistic	1159.242	Durbin-Watson stat	2.130377
Prob(F-statistic)	0.000000		

Date: 04/08/17 Time: 16:14
Sample: 1993M01 2013M12
Included observations: 251
Q-statistic probabilities adjusted for 2 ARMA terms

Autocorrelation	Partial Correlation	AC	PAC	Q-Stat	Prob
1		-0.066	-0.066	1.1109	
2		0.136	0.132	5.8246	
3		-0.022	-0.005	5.9460	0.015
4		-0.010	-0.030	5.9704	0.051
5		0.058	0.061	6.8484	0.077
6		-0.075	-0.065	8.3212	0.080
7		-0.081	-0.109	10.030	0.074
8		0.116	0.132	13.549	0.035
9		-0.088	-0.054	15.602	0.029
10		-0.017	-0.073	15.675	0.047
11		-0.044	-0.013	16.196	0.063
12		-0.004	0.010	16.201	0.094

Dependent Variable: DCONST
Method: ARMA Maximum Likelihood (BFGS)
Date: 04/08/17 Time: 16:14
Sample: 1993M02 2013M12
Included observations: 251
Convergence achieved after 7 iterations
Coefficient covariance computed using outer product of gradients

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	214.4219	1374.109	0.156044	0.8761
AR(1)	0.497140	0.054595	9.105947	0.0000
AR(2)	0.116143	0.052147	2.227211	0.0268
SAR(12)	0.944592	0.013249	71.29646	0.0000
SIGMASQ	373960.6	26485.07	14.11968	0.0000

R-squared	0.934603	Mean dependent var	49.44223
Adjusted R-squared	0.933540	S.D. dependent var	2396.078
S.E. of regression	617.7066	Akaike info criterion	15.81783
Sum squared resid	93864109	Schwarz criterion	15.88805
Log likelihood	-1980.137	Hannan-Quinn criter.	15.84609
F-statistic	878.9103	Durbin-Watson stat	1.975678
Prob(F-statistic)	0.000000		

Date: 04/08/17 Time: 16:14
Sample: 1993M01 2013M12
Included observations: 251
Q-statistic probabilities adjusted for 3 ARMA terms

Autocorrelation	Partial Correlation	AC	PAC	Q-Stat	Prob
1		0.011	0.011	0.0322	
2		0.054	0.054	0.7743	
3		-0.054	-0.056	1.5282	
4		-0.032	-0.034	1.7972	0.180
5		0.042	0.049	2.2539	0.324
6		-0.091	-0.092	4.3953	0.222
7		-0.081	-0.089	6.1091	0.191
8		0.112	0.132	9.3965	0.094
9		-0.073	-0.078	10.799	0.095
10		-0.031	-0.065	11.048	0.137
11		-0.041	-0.009	11.501	0.175
12		-0.005	-0.001	11.507	0.243

Question 15 (8 points)

Consider two models for U.S. real GDP, used to construct forecast for the period 2010Q1-2016Q4:

- model A: deterministic trend model for which the sequence of 1-step ahead forecasts has RMSE=103.45 and the multistep forecast has RMSE=1649.06
- model B: stochastic trend model for which the sequence of 1-step ahead forecasts has RMSE=77.32 and the multistep forecast has RMSE=905.18.

The 1-step ahead forecasts are then used to perform the test of equal predictive ability by estimating

$$\Delta L_{t+j,1} = \beta_0 + u_{t+j} \quad \text{with } j = 0, 1, 2, \dots, T - t - 1$$

where $\Delta L_{t+j,1} = (e_{t+j,1}^A)^2 - (e_{t+j,1}^B)^2$, and $e_{t+j,1}^k$ is the one step ahead forecast error for forecast from model k in period $t + j$. Explain the idea behind this test and interpret its results below. Discuss how we would use it together with above RMSE values in model selection process.

Dependent Variable: DL_TREND
 Method: Least Squares
 Date: 04/09/17 Time: 18:34
 Sample (adjusted): 2010Q1 2016Q4
 Included observations: 28 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	5454.311	1293.939	4.215275	0.0002
R-squared	0.000000	Mean dependent var	5454.311	
Adjusted R-squared	0.000000	S.D. dependent var	6846.884	
S.E. of regression	6846.884	Akaike info criterion	20.53604	
Sum squared resid	1.27E+09	Schwarz criterion	20.58361	
Log likelihood	-286.5045	Hannan-Quinn criter.	20.55058	
Durbin-Watson stat	2.683486			

Question 16. (8 points)

Interpret the results of the Granger causality test for a VAR with three variables: $y_{1,t} = 400\Delta \log GDP_t$ is the growth rate of the U.S. real GDP and $y_{2,t} = 400(\Delta \log SP500_t - \Delta \log p_t^{GDP})$ is the inflation adjusted return of S&P 500, and $y_{3,t}$ is the Leading Index for the United States.

Explain what these Granger causality imply about the usefulness of each of the three variables when it comes to predicting the other ones. Is there any economic intuition behind these results?

Dependent variable: GRGDP

Excluded	Chi-sq	df	Prob.
RRSP500	3.689833	1	0.0547
LI	22.08652	1	0.0000
All	27.70217	2	0.0000

Dependent variable: RRSP500

Excluded	Chi-sq	df	Prob.
GRGDP	0.021518	1	0.8834
LI	0.206983	1	0.6491
All	0.673715	2	0.7140

Dependent variable: LI

Excluded	Chi-sq	df	Prob.
GRGDP	2.487304	1	0.1148
RRSP500	9.320140	1	0.0023
All	12.78463	2	0.0017

Question 17. (8 points)

Consider a bivariate VEC

$$\begin{aligned}\Delta \log p_t^{GAS} &= \gamma_1 z_{t-1} + \kappa_{11} \Delta \log p_{t-1}^{GAS} + \kappa_{12} \Delta \log p_{t-2}^{GAS} + \phi_{11} \Delta \log p_{t-1}^{OIL} + \phi_{12} \Delta \log p_{t-2}^{OIL} + \varepsilon_{1,t} \\ \Delta \log p_t^{OIL} &= \gamma_2 z_{t-1} + \kappa_{21} \Delta \log p_{t-1}^{GAS} + \kappa_{22} \Delta \log p_{t-2}^{GAS} + \phi_{21} \Delta \log p_{t-1}^{OIL} + \phi_{22} \Delta \log p_{t-2}^{OIL} + \varepsilon_{2,t}\end{aligned}$$

where $z_{t-1} = \log p_{t-1}^{GAS} - \beta_1 \log p_{t-1}^{OIL} - \beta_0$ is the error terms period $t - 1$.

Is the coefficient β_1 statistically significant? Interpret what the estimated value for β_1 means.

Are γ_1 and γ_2 statistically significant? Are the signs of γ_1 and γ_2 in the estimated VEC model consistent with error correction mechanism that moves the system back to the long run equilibrium, whenever there is a disruption and $z_{t-1} \neq 0$?

Sample: 1995M01 2010M12

Included observations: 192

Standard errors in () & t-statistics in []

Cointegrating Eq:		CointEq1	
LOG(PGAS(-1))		1.000000	
LOG(POIL(-1))		-0.597621 (0.01461) [-40.9023]	
C		1.587414 (0.05198) [30.5370]	
Error Correction:		D(LOG(PGAS))	D(LOG(POIL))
CointEq1		-0.327074 (0.07340) [-4.45601]	-0.108005 (0.12811) [-0.84307]
D(LOG(PGAS(-1)))		0.352504 (0.09747) [3.61644]	-0.115845 (0.17012) [-0.68095]
D(LOG(PGAS(-2)))		-0.127554 (0.09037) [-1.41151]	-0.032472 (0.15772) [-0.20588]
D(LOG(POIL(-1)))		0.103709 (0.06431) [1.61277]	0.201301 (0.11223) [1.79359]
D(LOG(POIL(-2)))		0.011155 (0.06267) [0.17799]	0.081658 (0.10939) [0.74652]

Question 18. (8 points)

Consider the GARCHJ(1,1) model for the S&P 500 daily returns. Write the equations for the estimated model, with estimated parameter values plugged into these equations.

Dependent Variable: R Method: ML - ARCH (BHHH) - Normal distribution Sample: 5815 8471 Included observations: 2657 Convergence achieved after 10 iterations Bollerslev-Wooldrige robust standard errors & covariance Variance backcast: ON GARCH = C(2) + C(3)*RESID(-1)^2 + C(4)*GARCH(-1)				
	Coefficient	Std. Error	z-Statistic	Prob.
C	0.036267	0.017439	2.079665	0.0376
Variance Equation				
C	0.010421	0.005245	1.987099	0.0469
RESID(-1)^2	0.065649	0.011338	5.790038	0.0000
GARCH(-1)	0.927400	0.011045	83.96233	0.0000
R-squared	-0.000534	Mean dependent var		0.009761
Adjusted R-squared	-0.001666	S.D. dependent var		1.146761
S.E. of regression	1.147716	Akaike info criterion		2.888638
Sum squared resid	3494.671	Schwarz criterion		2.897498
Log likelihood	-3833.556	Durbin-Watson stat		2.079139

On April 2, 2008, the 1-day-ahead conditional mean is $\mu_{t|t-1} = 0.036$, the 1-day-ahead conditional standard deviation $\sigma_{t|t-1} = 1.785$. Calculate the 1% VaR and 5% VaR, given that $\Phi^{-1}(0.05) = -1.645$ and $\Phi^{-1}(0.01) = -2.326$. Interpret these numbers, given a portfolio worth 1 million dollars.