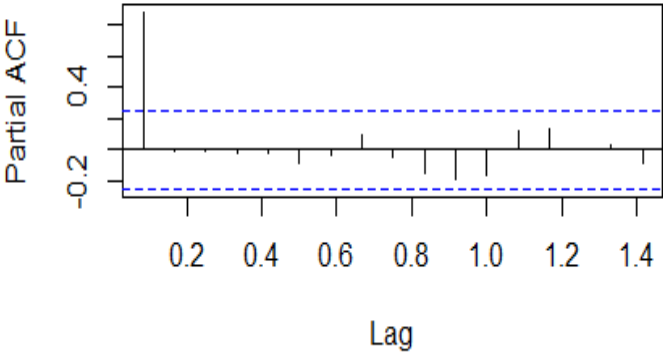
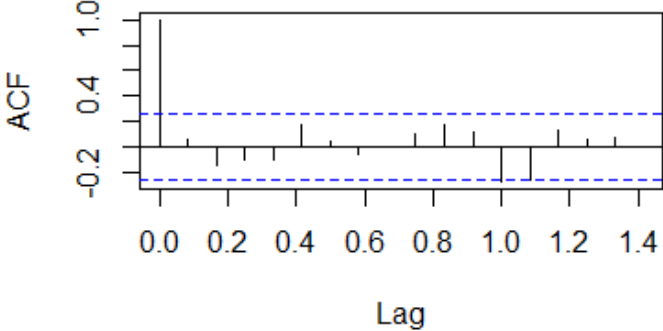




**SOMAIYA**  
VIDYAVIHAR UNIVERSITY

Trimester: Oct – Dec 24		
Maximum Marks: 50	Examination: ETE Exam	Date: 17/01/2025      Duration: 3 Hours
Programme code: 01 Programme: MBA (BADS Minor)	Class: SY	Semester/Trimester: V
College: K. J. Somaiya Institute of Management	Name of the department/Section/Center: Business Analytics	
Course Code: 217P01M532	Name of the Course: Predictive Analytics	
<b>Instructions:</b> <b>1.</b> All questions are compulsory. <b>2.</b> Make suitable assumptions if required and state them.		

Question No.		Max. Marks
1	<p>Here are the Autocorrelation Function (ACF) and Partial Autocorrelation Function (PACF) graphs for the inflation data:</p> <p style="text-align: center;"><b>Series data1_ts</b></p>  <p style="text-align: center;"><b>Inflation</b></p>  <p>a) Based on the ACF plot, do you observe any significant lags with high autocorrelation? How would this influence your model</p>	10

	<p>choice?</p> <p>b) Looking at the PACF plot, do you see a sharp cutoff after a specific lag? What does that suggest about the AR process of the data?</p> <p>c) If the ACF shows a slow decay, what type of model would you consider for this time series? Would you prefer an AR model or an MA model, and why?</p> <p>d) In the context of time series modelling, how do you use the ACF and PACF plots to determine the parameters for ARIMA or similar models?</p>	
2	<p>Following is the R code for the SARIMA model.</p> <pre> setwd("D:/Predictive Modelling/triple exponential smoothing") y&lt;-read.csv("sales goods company.csv") yt&lt;-ts(y, start=2012, frequency=4) plot(yt, col="black") trend&lt;-lm(yt ~ time(yt)) abline(trend, col="red")  adf.test(yt) ytd1&lt;-diff(yt, k=1) adf.test(ytd1) acf(ytd1) # q=0 pacf(ytd1) # p=1 arima(yt, order=c(1,1,0)) arima(yt, order=c(2,1,0)) arima(yt, order=c(1,2,1))  yt_seasonal&lt;-diff(yt, k=4) # adf.test(yt_seasonal) # D=0 acf(yt_seasonal) # Q=0 pacf(yt_seasonal) # P=2  arima(yt, order=c(2,2,1), seasonal = list(order=c(2,0,0), period=4))  model1&lt;-arima(yt, order=c(2,2,1), seasonal = list(order=c(2,1,0), period=4)) arima(yt, order=c(2,2,1), seasonal = list(order=c(2,1,1), period=4))  library("forecast") auto.arima(yt)  Box.test(residuals(model1), lag=10, type="Ljung-Box") </pre> <p>Give the answer to the following questions based on the above code</p> <p>a) What is the purpose of using ts() function in the code? How does it transform the data?</p> <p>b) What does the Augmented Dickey-Fuller test (adf.test()) assess in time series analysis? Frame the Null and Alternative hypotheses for this test. If the p-value for this test is 0.257, what conclusion would you draw, and what would be the next step to perform the analysis?</p> <p>c) How does the ARIMA model (1,1,0) differ from (2,1,0) and (1,2,1) in terms of model complexity and interpretation?</p> <p>d) What do the parameters order=c(p,d,q) represent in the ARIMA model?</p> <p>e) How does the Box.test(residuals(model1), lag=10, type="Ljung-Box") function help in evaluating the model's residuals? What would indicate a good model fit? Frame the hypotheses.</p>	20
3	<p>Following is the R code for exponential smoothing.</p> <pre> # Set working directory and load the dataset </pre>	20

	<pre>setwd("D:/time series forecasting/ expo smoothing")  # Load data demand_data &lt;- read.csv("weekly_demand.csv")  # Convert to time series demand_ts &lt;- ts(demand_data, start = c(2020, 1), frequency = 52)  # Plot the original time series plot(demand_ts, main = "Weekly Demand Time Series", col = "blue", pch = 19)  # Load the forecast package install.packages("forecast") library(forecast)  # Apply exponential smoothing ets_model &lt;- ets(demand_ts, model="ANN") print("Exponential Smoothing Model:") print(ets_model)  # Plot the forecast plot(ets_model, main = " Exponential Smoothing Forecast", col = "darkgreen")  # Calculate residuals and MSE residuals_ets &lt;- residuals(ets_model) mse_ets &lt;- mean(residuals_ets^2) print(paste("MSE for Exponential Smoothing:", mse_ets))</pre> <p>Give the answer to the following questions based on the above code</p> <ul style="list-style-type: none"><li>a) What steps would you take if there are missing values in the demand_data before converting it into a time series?</li><li>b) Explain the meaning of the "ANN" model components in the ets() function.</li><li>c) In what scenarios might the "ANN" model be insufficient for capturing the patterns in the data?</li><li>d) What is the significance of residuals in evaluating the performance of the ets_model?</li><li>e) If you wanted to implement exponential smoothing for data with daily frequency, what changes would you make to the code?</li></ul>	
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