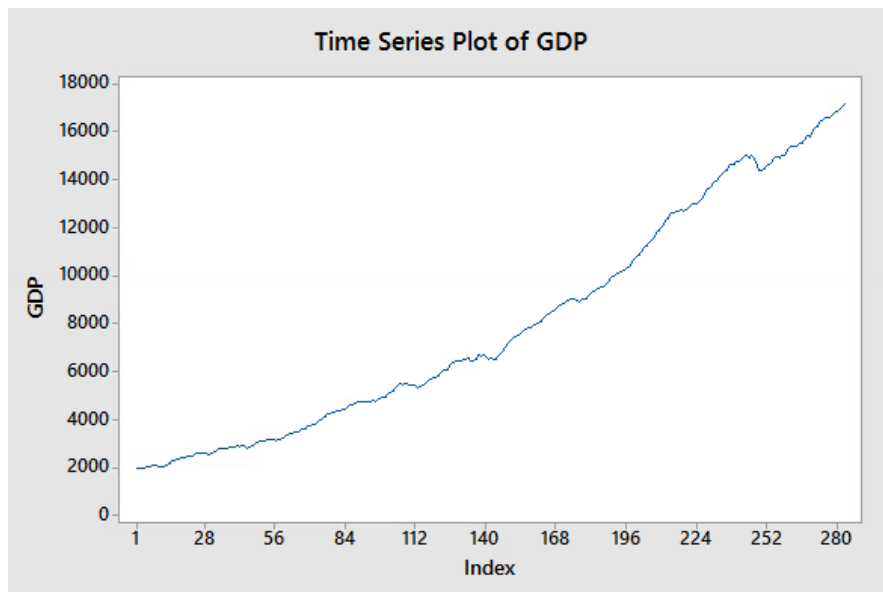


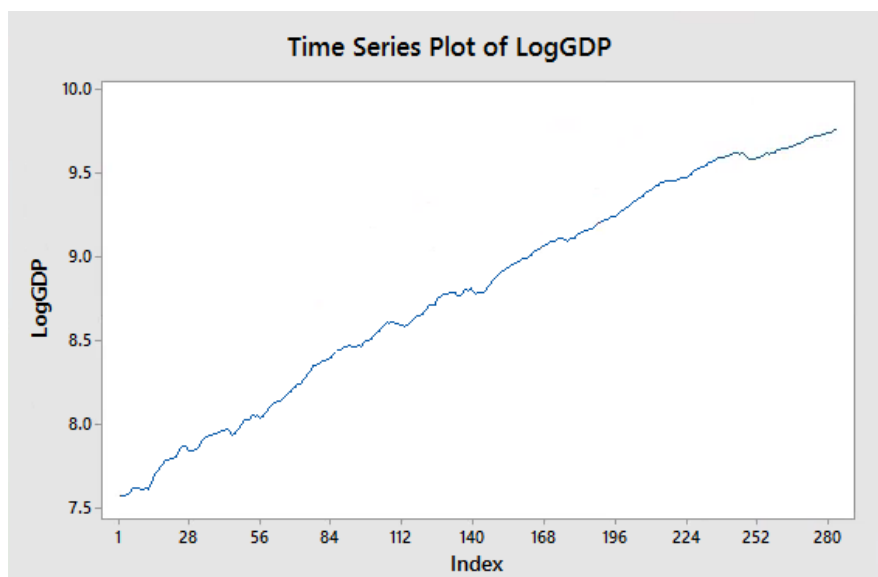
### Question 1

A.



I think GDP seems to grow linearly over time because the GDP is increasing as the time proceeds.

B.



The log GDP appears to grow linearly over time. Changes in the natural logarithm are approximately equal to percentage changes of the original data series. Log GDP increases by 2 (7.6-9.6) over past 70 years, so the average growth of GDP is about  $2/70$  per year, equivalent to 2.86% per year.

C.

#### Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	7.66121	0.00795	964.12	0.000	
time	0.007938	0.000049	163.65	0.000	1.00

#### Regression Equation

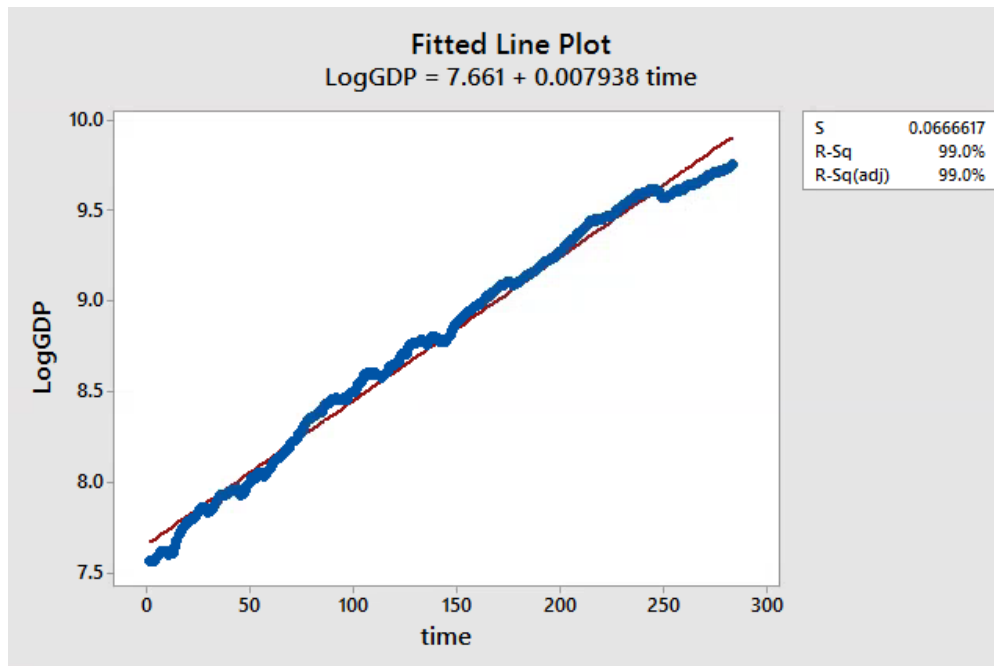
$$\text{LogGDP} = 7.66121 + 0.007938 \text{ time}$$

## Prediction

Fit	SE Fit	95% CI	95% PI
9.92355	0.0079884	(9.90783, 9.93928)	(9.79140, 10.0557)

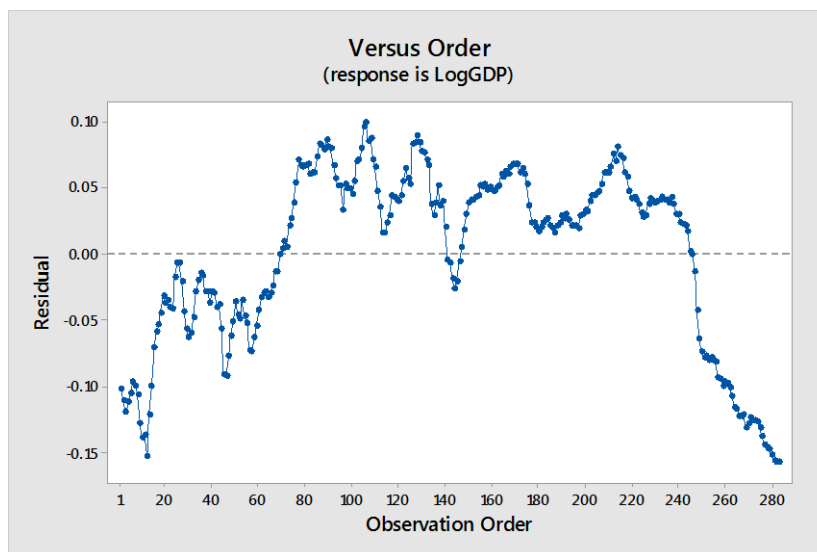
The 95% prediction interval is (9.79140, 10.0557). I don't think this prediction is effective because this prediction interval is too narrow.

**D.**



There are some data points below the fitted line in the beginning and the end of time data series, and this may be caused by trend or autocorrelation. However, the  $R^2$  is 99%, which means the model explains 99% of the variability of the response data around its mean. Therefore, the line fits well.

**E.**

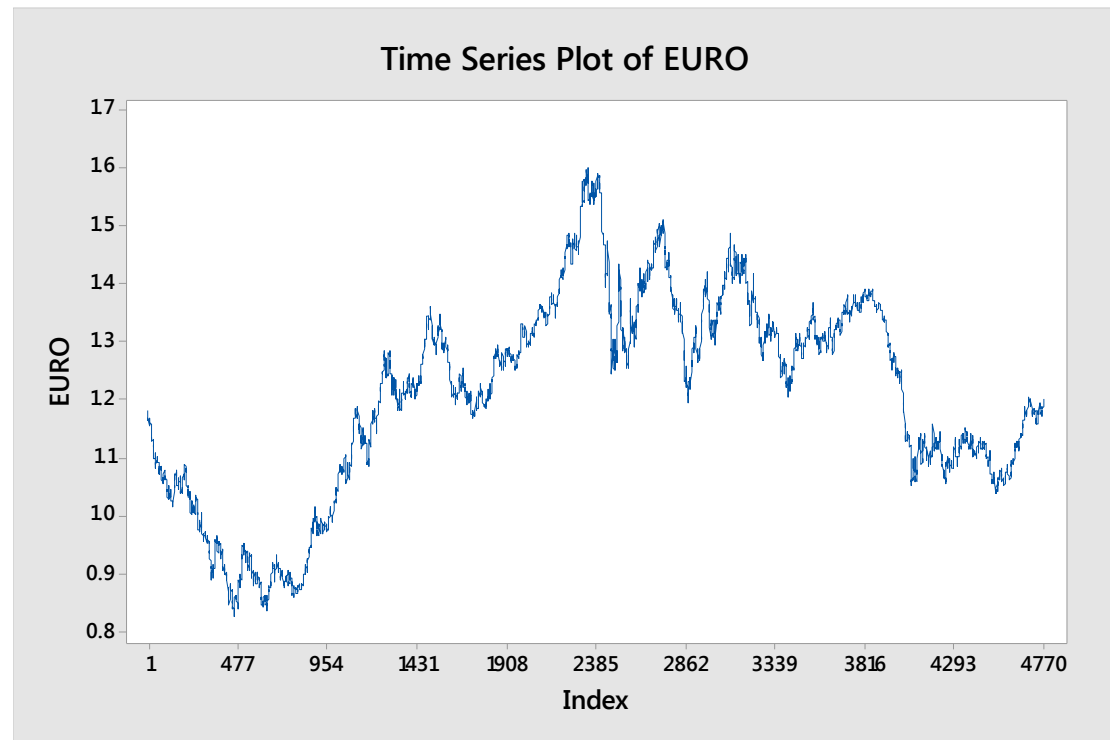


From the residual plot, we see the residuals are not randomly distributed around 0,

and they are not independent and follow some pattern This is caused by autocorrelation. These problems could spoil the validity of the forecast interval.

## Question 2

A.



Clearly, the straight-line model doesn't seem appropriate.

B.

Model of the first 700 data points:

### Regression Analysis: EURO\_2 versus Time2

#### Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	1	4.1180	4.11800	2937.14	0.000
Time2	1	4.1180	4.11800	2937.14	0.000
Error	698	0.9786	0.00140		
Total	699	5.0966			

#### Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
0.0374439	80.80%	80.77%	80.68%

#### Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	1.09976	0.00283	388.12	0.000	
Time2	-0.000380	0.000007	-54.20	0.000	1.00

## Regression Equation

$$\text{EURO\_2} = 1.09976 - 0.000380 \text{ Time2}$$

## Settings

Variable	Setting
Time2	4773

## Prediction

Fit	SE Fit	95% CI	95% PI
-0.711912	0.0310060	(-0.772788, -0.651036)	(-0.807361, -0.616463) XX

XX denotes an extremely unusual point relative to predictor levels used to fit the model.

Model of 701-4772:

## Regression Analysis: EURO\_1 versus Time1

### Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	1	1.429	1.42876	65.79	0.000
Time1	1	1.429	1.42876	65.79	0.000
Error	4070	88.386	0.02172		
Total	4071	89.815			

### Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
0.147365	1.59%	1.57%	1.48%

### Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	1.20613	0.00585	206.14	0.000	
Time1	0.000016	0.000002	8.11	0.000	1.00

## Regression Equation

$$\text{EURO\_1} = 1.20613 + 0.000016 \text{ Time1}$$

## Settings

Variable	Setting
Time1	4773

## Prediction

Fit	SE Fit	95% CI	95% PI
1.28219	0.0046196	(1.27314, 1.29125)	(0.993135, 1.57125)

The forecast interval from first model doesn't succeed, but the second one is successful. The first predication interval fails since 1) When regression is performed on time series data, errors are autocorrelated, so errors will follow some pattern, which violates the assumptions of linear regression. 2) In the regression equation, the slope term is negative, which means the USD/Euro exchange rate will always decrease as time increases. Therefore, when the time gets really large, the exchange rate will become negative, which is impossible in real-life situations.