

ISA 444: Business Forecasting

14 - Autocorrelation and Seasonality

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Fall 2020

Outline

1 Preface

2 Partial Autocorrelation

3 Stationarity

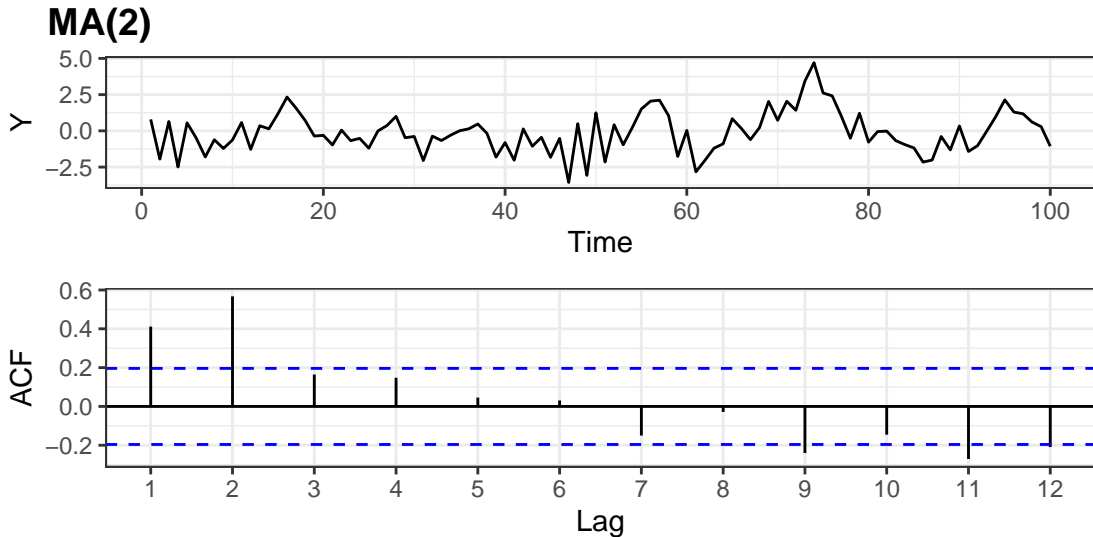
4 Recap

Quick Refresher on What we Covered in Chapter 06 so Far

Main Learning Outcomes Discussed in Chapter 06

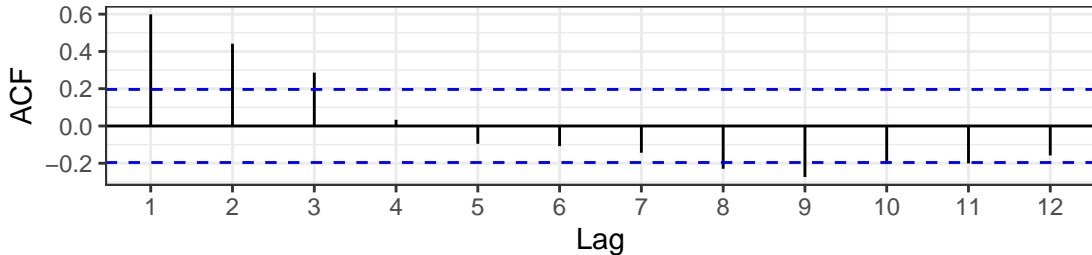
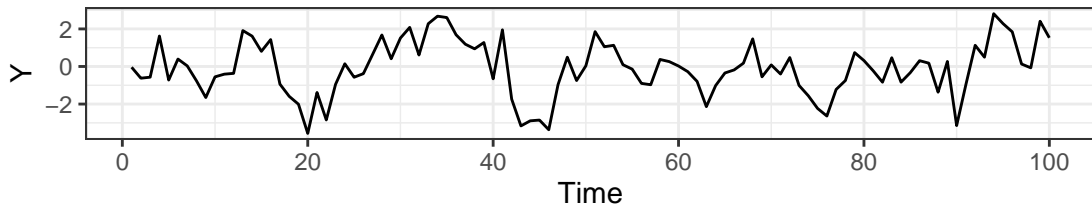
- ✓ Define the population mean, and variance of a random variable.
- ✓ Define the population covariance, and correlation between two random variables.
- ✓ Define the population autocovariance and autocorrelation of a random variable.
- ✓ Use sample estimates of the population mean, variance, covariance, and correlation.
- ✓ Explain the properties of the large sample distribution of the sample ACF.
- ✓ Use the large sample distribution of the sample ACF to identify significant autocorrelation in a time series.
- ✗ Determine if a sample ACF plot “cuts off” or “dies down”.

Some Time Series and their ACF Plots [1]



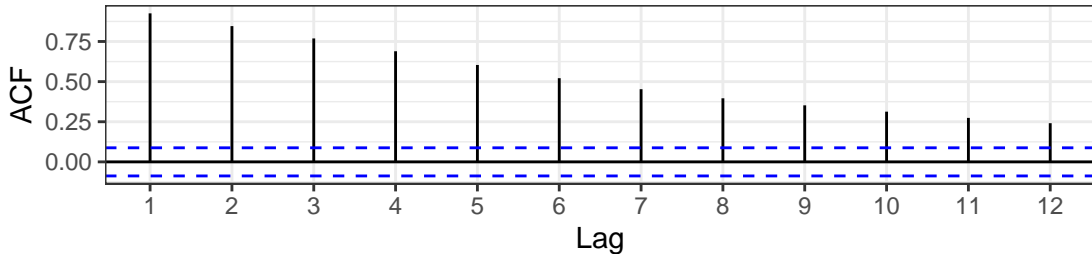
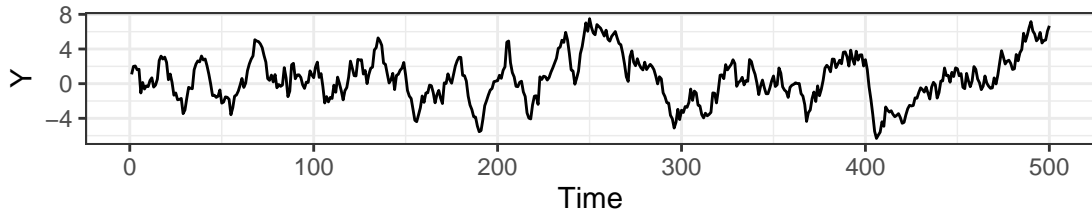
Some Time Series and their ACF Plots [2]

MA(3)



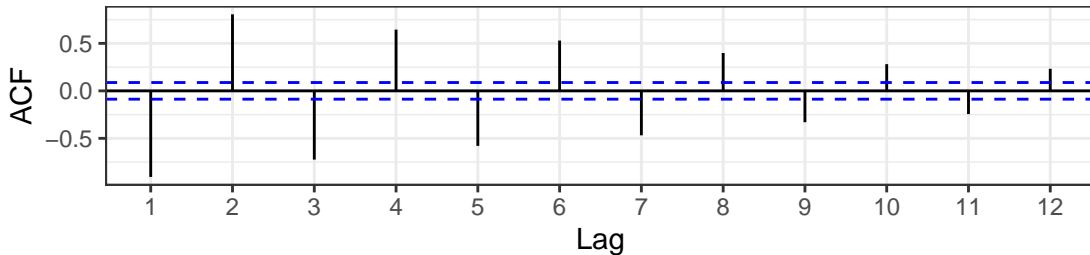
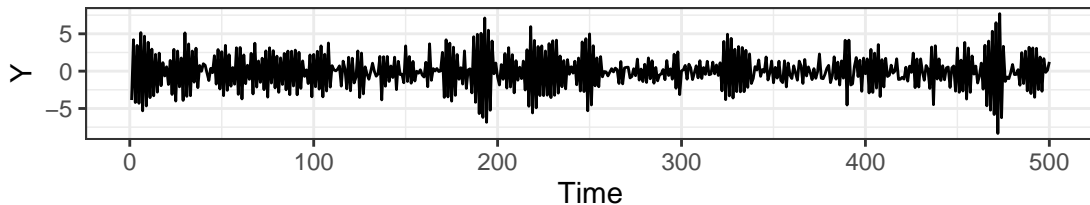
Some Time Series and their ACF Plots [3]

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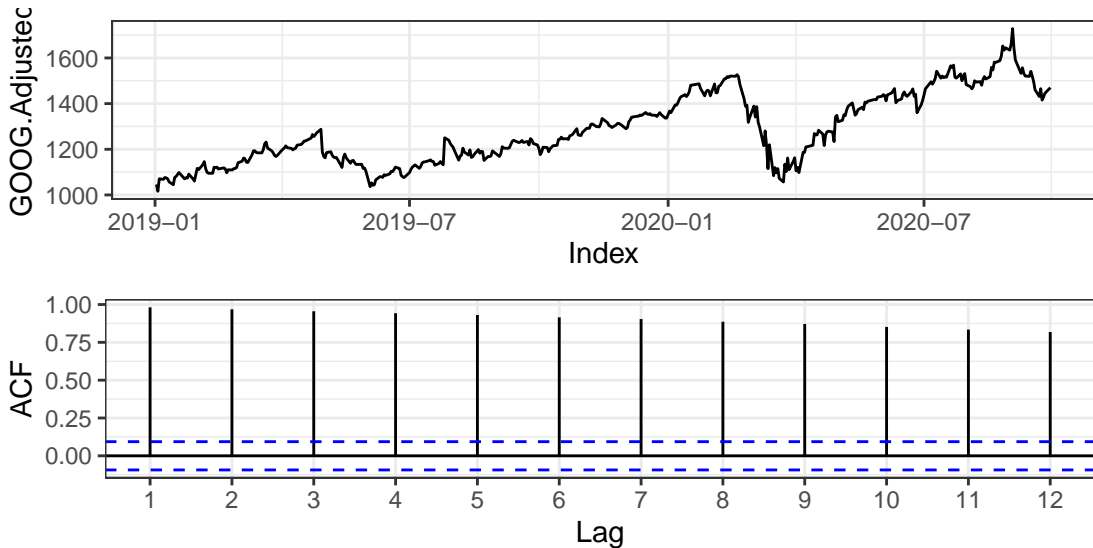


Some Time Series and their ACF Plots [4]

AR(1): -0.9



Some Time Series and their ACF Plots [5]



Learning Objectives for Today's Class

Main Learning Outcomes

- Determine if a sample ACF plot “cuts off” or “dies down”.
- Explain how sample partial autocorrelation is calculated.
- Define the term “weakly stationary” and recognize time series that do not fit this definition.
- Use the sample ACF plot to identify a nonstationary time series.

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Definition: General

Statistical Definition: Let us say that we have three variables, X , Y , and Z , all correlated, and we want to know how X and Y are correlated after we remove the effects of Z on each.

Computation Approach:

$$\hat{X} = a_1 + b_1 Z; \quad X^* = X - \hat{X}$$

$$\hat{Y} = a_2 + b_2 Z; \quad Y^* = Y - \hat{Y}$$

$Corr(X^*, Y^*)$ is the Partial Autocorrelation between X and Y . It is the correlation that remains after we remove the effect of Z .

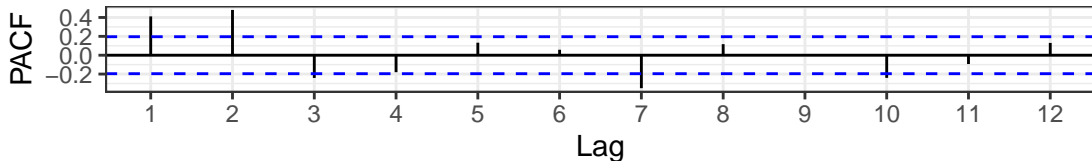
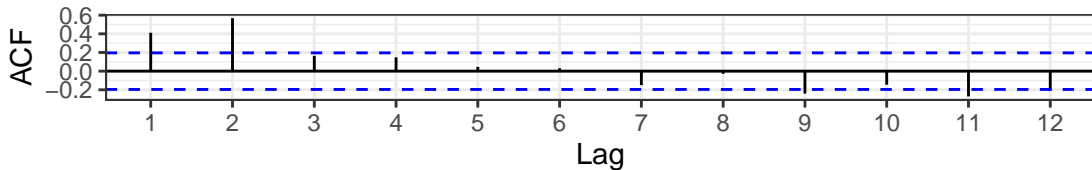
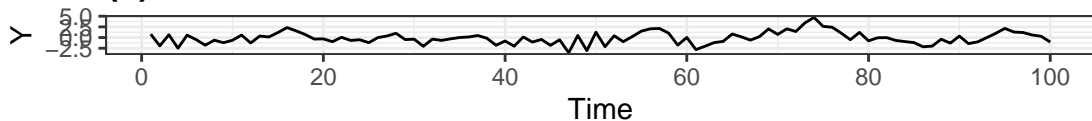
PACF in the Context of Time-Series Analysis

The Partial Autocorrelation between Y_t and Y_{t+k} is the correlation between Y_t and Y_{t+k} after removing the effects of $Y_{t+1}, Y_{t+2}, Y_{t+3}, \dots, Y_{t+k-1}$.

- We plot the partial autocorrelation over multiple lags just like the autocorrelation function (ACF).
- We refer to the plotted partial autocorrelations as the PACF.

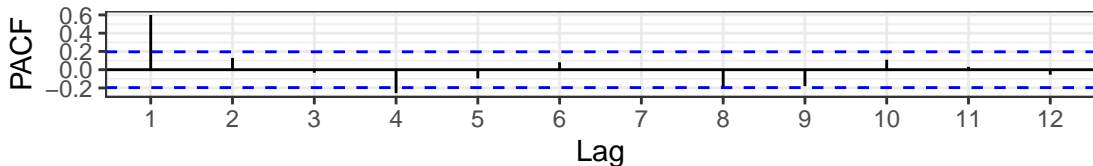
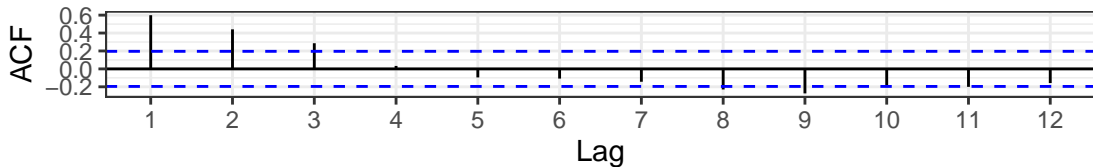
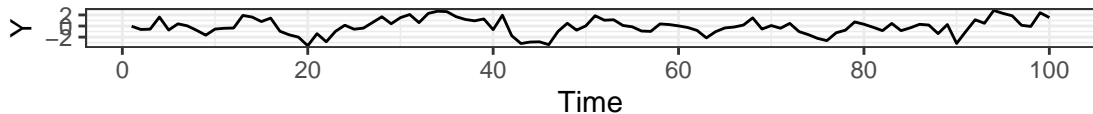
Some Time Series and their PACF Plots [1]

MA(2)



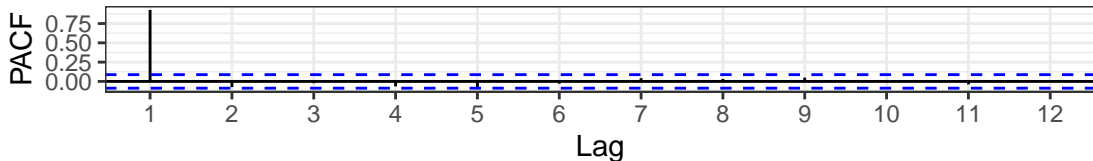
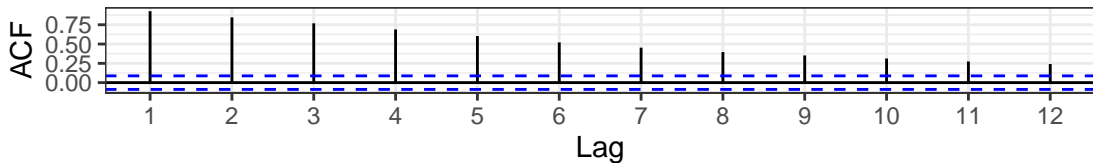
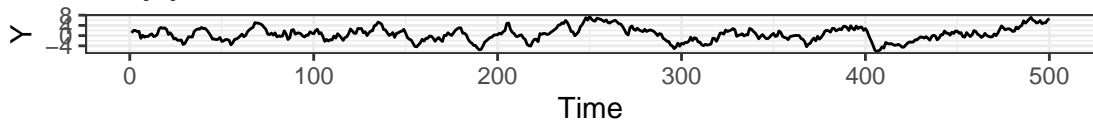
Some Time Series and their PACF Plots [2]

MA(3)



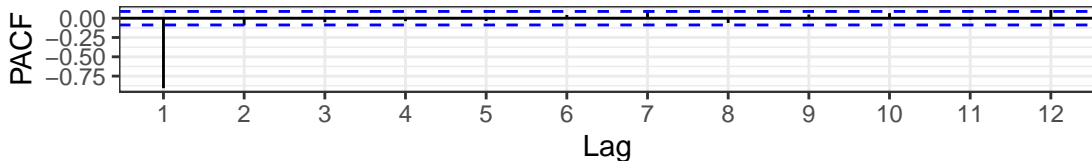
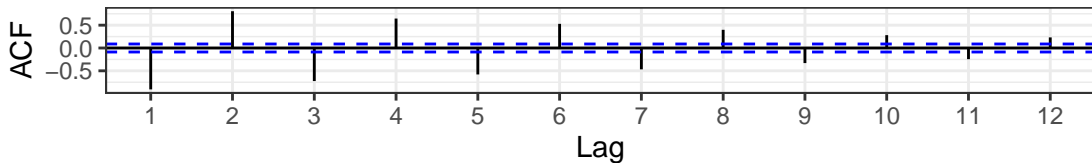
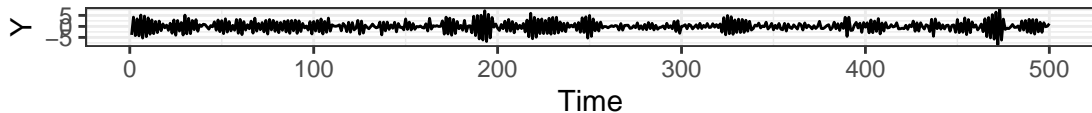
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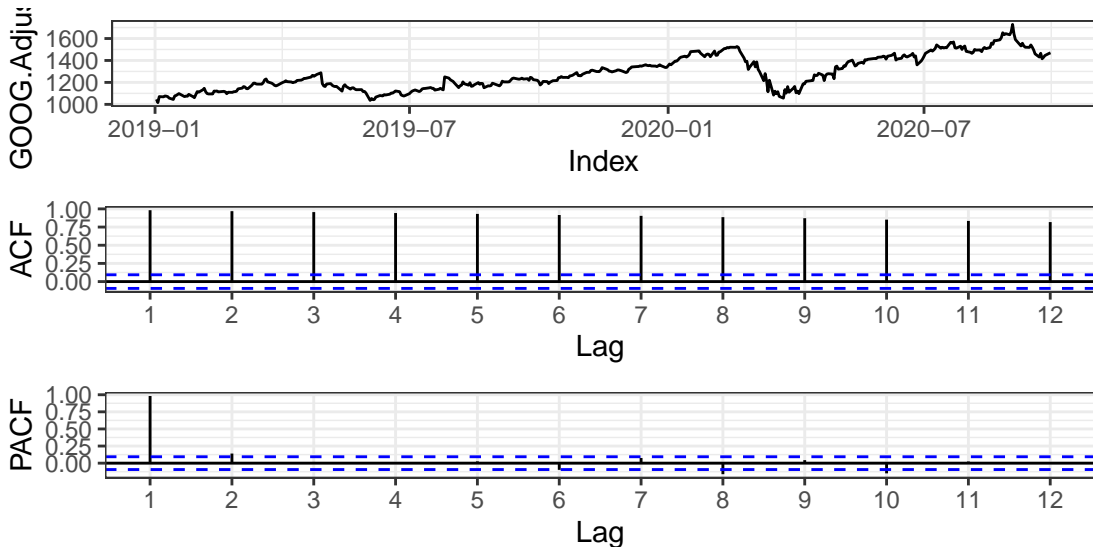


Some Time Series and their PACF Plots [4]

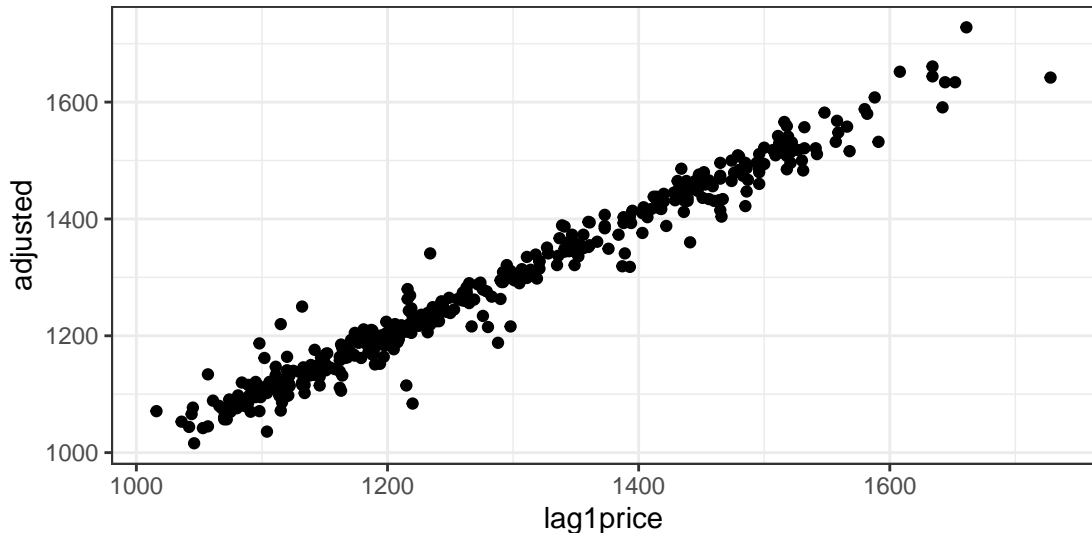
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Some Time Series and their PACF Plots [5]



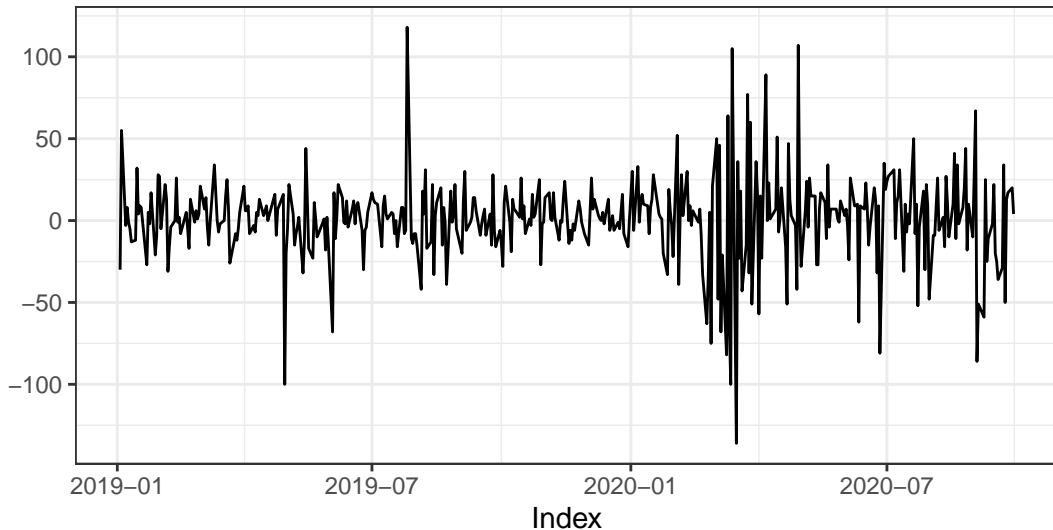
What we have learned from the GOOG Example [1]



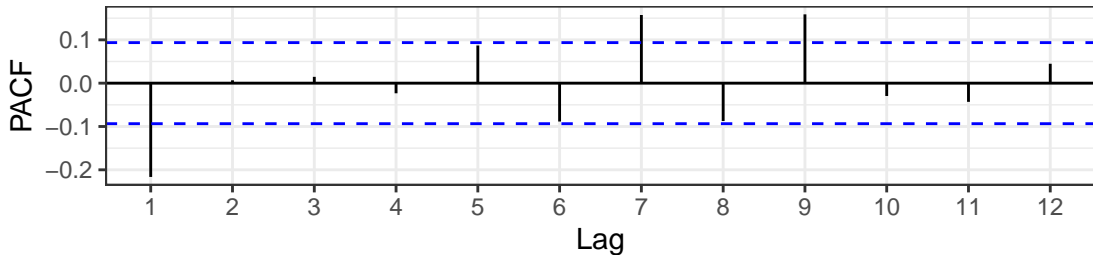
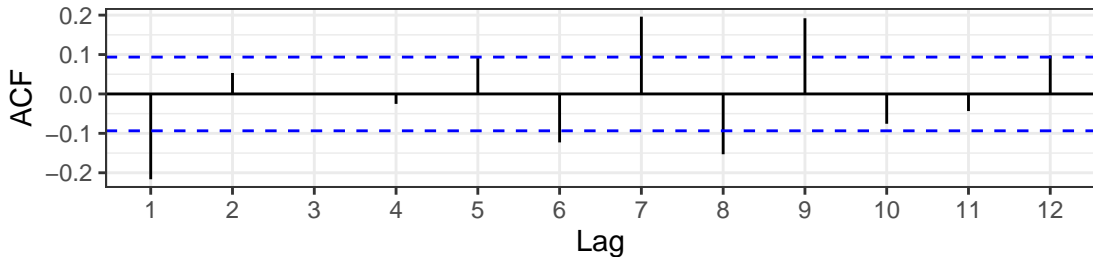
What we have learned from the GOOG Example [2]

	<i>Dependent variable:</i>
	adjusted
lag1price	0.985*** (0.008)
Constant	20.607** (10.390)
Observations	439
R ²	0.972
Adjusted R ²	0.972
Residual Std. Error	25.897 (df = 437)
F Statistic	15,001.350*** (df = 1; 437)
<i>Note:</i>	*p<0.1; **p<0.05; ***p<0.01

What we have learned from the GOOG Example [3]



What we have learned from the GOOG Example [4]



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Stationarity: A Formal Definition

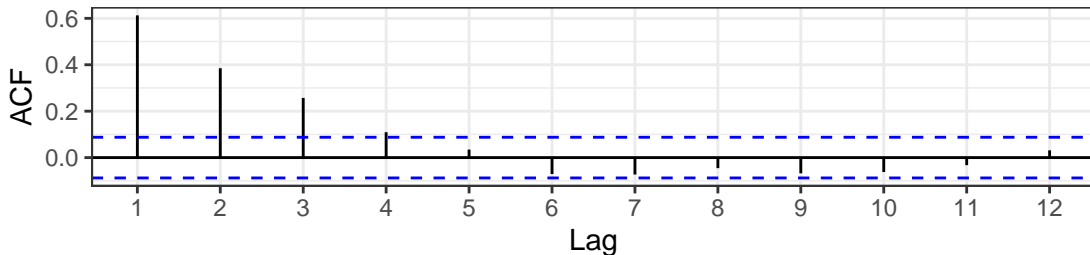
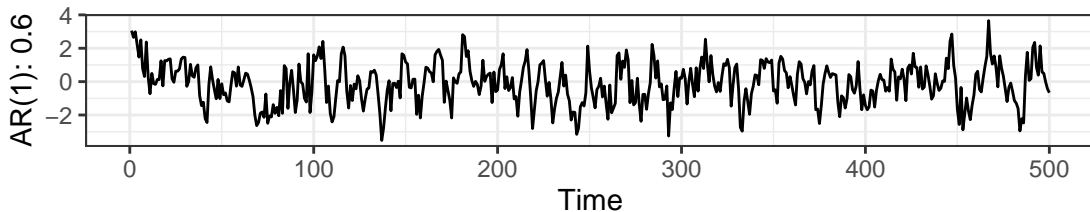
Weak Stationarity: A weakly stationary time series is a finite variance process such that:

- ➊ The mean, μ_t , is constant and does not depend on the time t ; and
- ➋ The autocovariance function, $\gamma(s, t)$ depends on s and t only through their difference $|s - t|$.

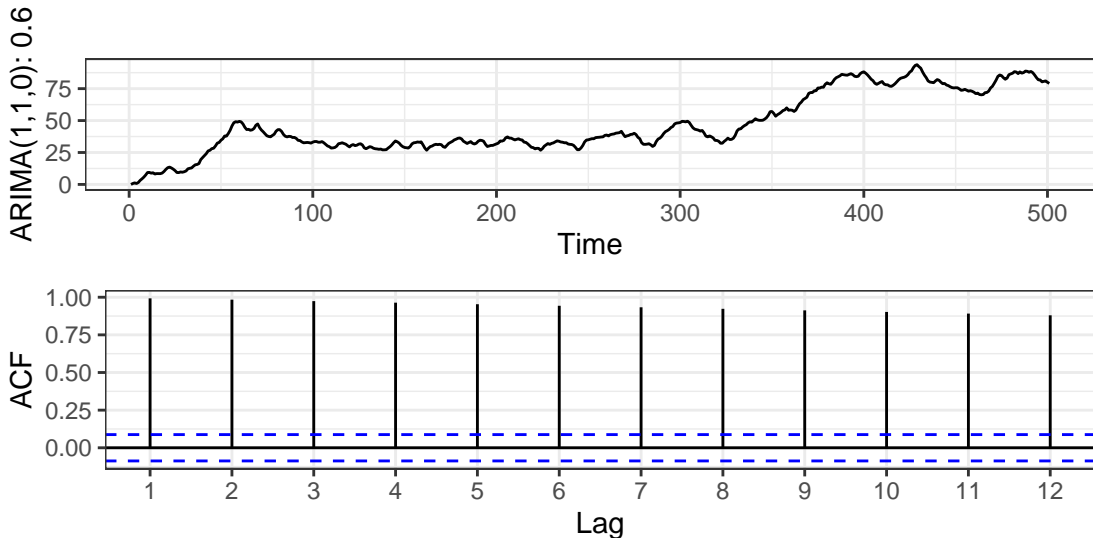
We will use the term “stationary” to refer to weak stationarity.

- The concept of weak stationarity forms the basis of much of the foundation for time series modeling.
- The fundamental properties (1 & 2) required for weak stationarity are satisfied by many of the models that are widely used.

Stationarity: A Visual Explanation



Non-Stationarity: A Visual Explanation



Stationarity: Some Thoughts??

Based on the five examples depicted in Slides 4 – 8 (or alternatively 13 – 17), please identify which of the five series are stationary and which are not!!

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Summary of Main Points

Main Learning Outcomes

- Determine if a sample ACF plot “cuts off” or “dies down”.
- Explain how sample partial autocorrelation is calculated.
- Define the term “weakly stationary” and recognize time series that do not fit this definition.
- Use the sample ACF plot to identify a nonstationary time series.

Things to Do for Next Class

- Thoroughly read Chapter 6.1 of our textbook.
- Go through the slides, examples and make sure you have a good understanding of what we have covered.

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