Midterm 2 - Fall 2024

ECON 4753 — University of Arkansas

1. (10 pts) Say you have the following time-series observations from $t = 1, \dots, 7$:

i. What is the 1-lag autocorrelation coefficient, $\hat{\rho}_1$?

Answer: The sample mean equals 1.

The variance is

$$\frac{1}{7-1} \left((0.63-1)^2 + (0.79-1)^2 + (0.93-1)^2 + (1.1-1)^2 + (1.05-1)^2 + (1.2-1)^2 + (1.3-1)^2 \right)$$
$$= 0.0547$$

The autocovariance is given by

$$\frac{1}{6} ((0.63 - 1)(0.79 - 1) + (0.79 - 1)(0.93 - 1) + (0.93 - 1)(1.1 - 1) + (1.1 - 1)(1.05 - 1) + (1.05 - 1)(1.2 - 1) + (1.2 - 1)(1.3 - 1))$$

$$= 0.0267$$

Finally, the autocorrelation is 0.0267/0.0547 = 0.488.

ii. By hand, forecast into period 8 using a one-sided 3-period rolling average.

Answer:

$$\hat{y}_8 = \frac{1}{3}(1.05 + 1.2 + 1.3) = 1.183$$

2. (10 pts) Say you are working at a company with time-series data and you want to use some smoothing method for analyzing the patterns.

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i. Say you use a two-sided moving average. Your boss asks why you do not have a value of \hat{y}_t on the ends of the time-series. Please explain why

Answer: A two-sided moving average uses observations in the future and in the past to form the forecast. At the ends of our time-series we do not have enough observations in the past or in the future, so we can not form the average.

ii. Your boss replies that they really want to predict into next month. How would you adjust your choice of smoothing method to do this?

Answer: Instead, I would use either a one-sided moving average or a simple exponential smoothing to form my forecast. Since these only use past observations, I can forecast into the future period.

3. (10 pts) Say you are trying to forecast a time-series data into the future. You notice that there is a non-linear trend in the data. Explain in 1, maybe 2, sentences why you should not try to model this using higher-order polynomial terms

Answer: Polynomials shoot off to $\pm \infty$ and so they can be unstable. This is especially true as I extrapolate outwards in the time-series.

For the following questions, we will look at monthly US data on the median sales price of housing (see Figure 1). In addition to the raw data, Figure 1 plots the estimated linear trend.

4. (15 pts) Consider a simple exponential smoothing model for inference on this time-series with $\alpha=0.1$. For this time-series, how do you think this method would perform? Please explain why.

Answer: This method would perform poorly. With $\alpha = 0.1$, this imposes a lot of smoothing on the model and hence it would likely miss the short-term peaks and valleys we see between 2019 and 2023. Moreover, since the data is trending, it will fail to capture these trends.

5. (10 pts) How do you think the linear time trend performs at describing this time series? What might you do to improve this time-series model?

Answer: The linear time trend does an okay job; it picks up on the general growth of housing prices, but does not capture the intense growth starting in 2020. I might modify this by using a piecewise trends model.

For the following questions, we will be looking at daily time-series on taxis in NYC (see Figure 2). The outcome variable is the number of taxis out at noon each day. I have highlighted on the plot the days around Christmas where a large number of people typically leave the city.

Use the results of the following time-series regression to answer the following questions

```
OLS estimation, Dep. Var.: n_taxis_at_noon
Observations: 140
Standard-errors: Heteroskedasticity-robust
                 Estimate Std. Error t value
                                                Pr(>|t|)
(Intercept)
                19857.05
                            496.397 40.00235 < 2.2e-16 ***
                            530.684 -6.25447 5.0606e-09 ***
day_of_week::Mon -3319.15
day_of_week::Tue -2357.85
                            543.931 -4.33483 2.8561e-05 ***
day_of_week::Wed -1677.25
                            532.226 -3.15139 2.0080e-03 **
day_of_week::Thu -2176.00
                            729.170 -2.98422 3.3840e-03 **
day_of_week::Fri -1939.30
                            594.769 -3.26059 1.4122e-03 **
day_of_week::Sat 1343.60
                            751.638 1.78756 7.6124e-02 .
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

6. (10 pts) What is the omitted category in this regression?

Answer: The omitted category is Sunday.

7. (10 pts) Which day of the week has the most taxis available?

Answer: The day with the most taxis available is Saturday since it has the largest coefficient estimate and is positive (relative to Sunday).

8. (15 pts) Form a 95% confidence interval for the average number of taxis available at noon on Sunday (round your answer to the nearest whole unit)

Answer: With 95% confidence, the number of taxis available at noon on Sundays is $19857 \pm 1.96 * 496.39 = (18884, 20829)$.

9. (10 pts) Do you think an adjustment should be made in the regression for the fact that the number of taxis has a sizeable drop around Christmas/New Years? What adjustment could you make?

Answer: I would include an indicator variable in my time-series regression for being in the interval of dates highlighted by the gray bar.

Figure 1 – Median Housing Price of Monthly US Sales

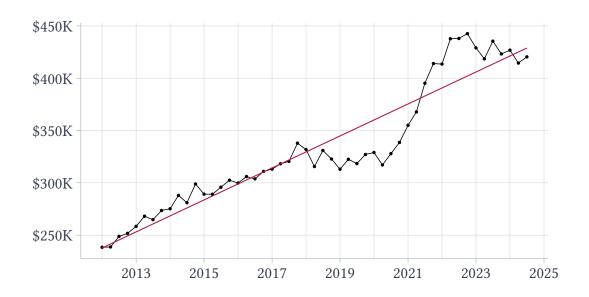


Figure 2 – Daily data on the Number of Taxis available at Noon in NYC

