## Midterm Economic Modelling and Simulation

## Instructions

- There is one and only one correct answer per question
- Fill out the answers sheet: only that sheet will be graded
- You will need to return the question sheet, but whatever you write on it will be disregarded
- All questions are worth the same
- Correct answers are worth +1
- Wrong answers are worth -1
- If you are unsure about an answer, leave it blank and it will be worth zero
- For the sake of readability, the semicolon; has been used to combine several lines of code into a single one, when appropriate

## Questions

- 1. How would you measure the accuracy of a classification problem (e.g. diagnosing a disease) when using logistic regression?
- a) Dividing the true positives by the true negatives
- b) Dividing the true positives by the false positives
- c) Dividing the sum of the true positives and the true negatives by the total
- d) Dividing the sum of the false positives and the true negatives by the sum of the true negatives
- 2. Choose the right import statement of the 'car\_crashes' dataset from the seaborn package:
- a) import seaborn as sns; df = sns.load\_dataset('car\_crashes')
- b) from seaborn import car\_crashes
- c) import seaborn as pd; car\_crashes = sns.get\_dataset\_names('pandas')
- d) import diamonds from seaborn

We will now work with a dataset called diamonds. It contains properties of diamonds such as the depth, the price, the color, etc. of a collection of diamonds.

- 3. In order to get the column names of the 'diamonds' dataset, the proper command is:
- a) diamonds.columns
- b) diamonds.names
- c) diamonds.values

d) diamonds[columns]

Below you can find a sample of the diamonds dataset:

```
cut color clarity
       carat
                                         depth
                                                 table
                                                        price
        0.23
                                          61.5
                                                  55.0
                                                                            2.43
                   Ideal
                             Ε
                                    SI2
                                                          326
                                                                3.95
                                                                      3.98
        0.21
                Premium
                             Ε
                                    SI1
                                          59.8
                                                  61.0
                                                          326
                                                                3.89
                                                                      3.84
                                                                            2.31
        0.23
                    Good
                             Ε
                                    VS1
                                          56.9
                                                  65.0
                                                          327
                                                                4.05
                                                                      4.07
        0.29
                                    VS2
                                          62.4
                                                  58.0
                                                          334
                                                                4.20
                 Premium
        0.31
                                                  58.0
                    Good
                                    SI2
                                          63.3
                                                          335
                                                                4.34
                                                                            2.75
53935
        0.72
                   Ideal
                             D
                                    SI1
                                          60.8
                                                  57.0
                                                         2757
                                                                5.75
                                                                      5.76
                                                                            3.50
53936
        0.72
                    Good
                             D
                                          63.1
                                                  55.0
                                                                5.69
                                                         2757
                                                                      5.75
                                                                            3.61
53937
        0.70
                   Good
                                          62.8
                                                  60.0
                                                         2757
                                                                5.66
                                                                      5.68
                                                                            3.56
53938
        0.86
                Premium
                                    SI2
                                          61.0
                                                  58.0
                                                               6.15
                                                                      6.12
                                                                            3.74
                                                         2757
53939
        0.75
                   Ideal
                             D
                                    SI2
                                          62.2
                                                  55.0
                                                         2757
                                                                5.83
                                                                      5.87
                                                                            3.64
[53940 rows x 10 columns]
Index(['carat', 'cut', 'color', 'clarity', 'depth', 'table', 'price', 'x', 'y',
       'z'],
      dtype='object')
```

Figure 1: Excerpt from diamonds dataset

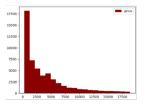
- 4. Which command would subset the dataset 'diamonds' when the price is lower than 300:
- a) diamonds.query("price > 300")
- b) diamonds.query("cut == Ideal")
- c) diamonds.query("price == 300")
- d) diamonds.query("price < 300")
- 5. Now we decided to look at the price statistics when the cut variable equals Ideal using the following filter: diamonds.query("cut == 'Ideal'")['price'].describe(). Based on the output, which is the median price when cut is 'Ideal'?

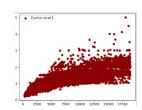
```
count
         21551.000000
          3457.541970
mean
std
          3808.401172
            326.000000
min
25%
           878.000000
50%
          1810.000000
          4678.500000
75%
         18806.000000
Name: price, dtype: float64
```

Figure 2: Output of the describe() command

- a) 3456.5
- b) 326.0
- c) 1810.0
- d) 4678.5
- 6. We want to understand how many diamonds have color E, how many have color J, and so on. Which command achieves this?
- a) diamonds.color.value\_counts()
- b) diamonds.value\_counts().color
- c) diamonds.color\_values()
- d) diamonds.value counts(color)
- 7. Based on the summary statistic of Question 5, what would be the value corresponding to the 30th percentile?
- a) Less than 326
- b) Between 326.0 and 878.0
- c) Between 4678.5 and 18806.0
- d) Between 878.0 and 1810.0
- 8. Now we want to subset the dataset by the 2 first indexes. The command would be:
- a) diamonds.loc[0:2]
- b) diamonds.query("0:2")
- c) diamonds.iloc[0:2]
- d) diamonds 0:2
- Create a column named volume based on the dimensions x, y and z of the diamonds:
- a) diamonds["volume"] = diamonds["x"] \*\* diamonds["y"] \*\* diamonds['z']
- b) diamonds["volume"] = product(diamonds["x"] \* diamonds["y"] \* diamonds['z'])
- c) diamonds.volume = diamonds["x"] + diamonds["y"] + diamonds['z']
- d) diamonds = diamonds.assign(volume = lambda df: df.x \* df.y \* df.z)
- 10. Obtain a descriptive analysis of the price column that includes the number of elements, quantiles, etc:

- a. diamonds.price.stats()
- b. diamonds.price.describe()
- c. diamonds['price'].analyse()
- d. diamonds.price
- 11. The following is one row extracted from the diamonds dataset. What would be the result of the sum of columns x and y for this row?
  - a) NaN
  - b) 3.95
  - c) 0
  - d) 6.38
- 12. Which of the following charts display a histogram of the price variable?





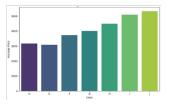


Figure 3: Output of the describe() command

- a) the left one
- b) the one in the center
- c) the right one
- d) none
- 13. Based on the above histogram, which is the most frequent price range?
  - a) Between 7500 and 10000
  - b) Between 17500 and 20000
  - c) Between 5000 and 7500
  - d) Between 0 and 2500
- 14. Which of the following best defines the term "Least Squares" in the context of statistical modeling?
  - a) A method for minimizing the sum of squared differences between observed and predicted values
  - b) A technique used to maximize the variance between data points

- c) A statistical approach focused on maximizing the absolute differences between observed and predicted values
- d) A method that prioritizes minimizing the maximum deviation between observed and predicted values
- 15. Choose the right definition of the LeastSquares function that returns the intercept and slope after getting as arguments an x and y vectors:
- a)

  def LeastSquares(xs, ys):
   mean\_x = np.mean(xs)
   mean\_y = np.mean(ys)
   cov = np.dot(xs mean\_x, ys mean\_y) / len(xs)
   slope = cov
   inter = mean\_y slope \* mean\_x
   return inter, slope

   b)

  def LeastSquares(xs, ys):
   mean\_x = np.mean(xs)
   var\_x = np.var(xs)
  - var\_x = np.var(xs)
    mean\_y = np.mean(ys)
    inter = mean\_x / mean\_y
    slope = mean\_x var\_x
    return inter, slope

    c)

```
def LeastSquares(xs, ys):
    mean_x = np.mean(xs)
    var_x = np.var(xs)
    mean_y = np.mean(ys)
    cov = np.dot(xs - mean_x, ys - mean_y) / len(xs)
    slope = cov / var_x
    inter = mean_y - slope * mean_x
    return inter, slope
```

- d) None are correct
- 16. The above function is applied to the 'carat' and 'price' columns in order to find a correlation between them: inter, slope = LeastSquares(diamonds.carat, diamonds.price). Based on the above estimated intercept & slope for the columns carat and price, let's build a column called fit\_carat that contains the model fit for price for the datapoints in the dataset:
  - a) diamonds["fit\_carat"] = inter \* diamonds['carat'] \* slope

- b) diamonds["fit\_carat"] = inter \* diamonds['carat'] + slope
- c) diamonds["fit\_carat"] = inter + slope \* diamonds['carat']
- d) diamonds["fit\_carat"] = inter + slope + diamonds['carat']
- 17. Which of the following is correct:
- a. The residuals are the differences between the observed values and the mean of the dependent variable.
- b. The residuals are the sum of the observed values and the predicted values
- c. The residuals of a linear model are the differences between the observed values and the values predicted
- d. The residuals are the differences between the predicted values and the mean of the independent variable.
- 18. Now we want to define a function that is capable to estimate the residuals of the linear fit. Select the correct definition:

- d) None is correct
- 19. Select the correct code-snippet that would allow to draw a scatterplot that includes price in the y-axis, and carat in the x-axis; plotting both the actual dataset and the fits:

```
• a) plt.scatter(x="carat", y="price", data=diamonds, color="blue", label="carat") plt.scatter(x="carat", y="fit_carat", data=diamonds, color="cyan", marker="s") plt.xlabel("carat") plt.ylabel("price") plt.legend()
```

```
b) plt.scatter(x="price", y="carat", data=diamonds, color="blue", label="carat")
    plt.scatter(x="fit_carat", y="carat", data=diamonds, color="cyan", marker="s")
    plt.xlabel("carat")
    plt.ylabel("price")
    plt.legend()
```

```
plt.line(x="price", y="cut", data=diamonds, color="blue", label="carat")
plt.line(x="fit_carat", y="cut", data=diamonds, color="cyan", marker="s")
plt.xlabel("carat")
plt.ylabel("price")
plt.legend()
```

• d) None is correct