

1. KNN is

- 12/13 ☒ A data-driven method  
1/13 ☐ B model-driven method  
0/13 ☐ C I do not know

2. The dependent variable of the classification is

- 12/13 ☒ A categorical  
1/13 ☐ B numeric  
0/13 ☐ C I do not know

3. KNN can be used for regression

- 13/13 ☒ A Yes  
0/13 ☐ B No  
0/13 ☐ C I do not know

4. In the case of KNN classification we use

- 2/13 ☐ A average of outcomes  
10/13 ☒ B majority voting scheme  
1/13 ☐ C I do not know

5. Which of these errors will increase constantly by increasing k?

- 7/13 ☒ A train error  
5/13 ☐ B test error  
1/13 ☐ C both  
0/13 ☐ D I do not know

6. This function can be used to perform KNN classification in R

12/13 ☒ A knn()

0/13 ☐ B k\_nn()

0/13 ☐ C knnreg()

0/13 ☐ D knearneib()

1/13 ☐ E I do not know

7. With the increase of k, the decision boundary will be

9/13 ☒ A simplified

4/13 ☐ B more complex

0/13 ☐ C I do not know

0/13 ☐ D unchanged

8. KNN algorithm is sensitive to outliers

12/13 ☒ A True

1/13 ☐ B False

0/13 ☐ C I do not know

9. KNN

12/13 ☒ A is a supervised learning algorithm.

1/13 ☐ B is an unsupervised learning algorithm.

0/13 ☐ C I do not know

10. In the case of small k we have

9/13 ☒ A overfitting

3/13 ☐ B underfitting

1/13 ☐ C it depends on the situation

0/13 ☐ D I do not know

11. Why do we need scaling in KNN?

3/13 ☐ A to avoid overfitting

2/13 ☐ B to avoid underfitting

6/13 ☒ C to have "equal" weights for variables

2/13 ☐ D I do not know

**12.** Let  $k = n$ , ( $n$  - number of observations), K-NN is same as

- 4/13 ☒ A random guessing
- 8/13 ☐ B everything will be classified as the most probable class (in total)
- 1/13 ☐ C everything will be classified as the least probable class (in total)
- 0/13 ☐ D I do not know

**13.** This function can be used to perform K-NN regression in R

- 11/13 ☒ A knn.reg
- 0/13 ☐ B knnforreg
- 1/13 ☐ C regknn
- 0/13 ☐ D knnforregression
- 1/13 ☐ E I do not know

**14.** Do you need to worry about scaling with one explanatory variable?

- 6/13 ☒ A No
- 6/13 ☐ B Yes
- 1/13 ☐ C I do not know

**15.**  $n$  - the number of observation,  $m$  - the number of explanatory variables When  $n=k$ ,  $m=1$ , the decision boundary for regression is

- 7/13 ☒ A a line
- 4/13 ☐ B a stepwise constant function
- 1/13 ☐ C a stepwise quadratic function
- 1/13 ☐ D I do not know

**16.** Which of these algorithms can be used to fill the missing values

- 1/13 ☐ A KNN for regression
- 3/13 ☐ B KNN for classification
- 9/13 ☒ C both
- 0/13 ☐ D I do not know

**17.** Which one is better: KNN regression or Linear regression?

- 2/13** ☐ **A** KNN outperform LR if the parametric form that has been selected is close to the true linear form
- 11/13** ☒ **B** LR outperform KNN if the parametric form that has been selected is close to the true linear form
- 0/13** ☐ **C** KNN will always outperform the LR
- 0/13** ☐ **D** I do not know

**18.** Which one is the Disadvantage of KNN?

- 3/13** ☐ **A** required assumptions
- 0/13** ☐ **B** cannot be applied for regression
- 1/13** ☐ **C** difficult to perform
- 9/13** ☒ **D** the problem of high dimensional data
- 0/13** ☐ **E** I do not know

**19.** The best k for train set equals to

- 7/13** ☒ **A** 1
- 6/13** ☐ **B** 2
- 0/13** ☐ **C** 0
- 0/13** ☐ **D** I do not know