## Tut 2: kNN

## Jan 2022

## kNN Classifier

kNN is discriminative, non-parametric predictive model

• For kNN classifier, the mathematical formulation is

$$\hat{h}(\boldsymbol{x}) = \underset{j \in \{1, \dots, K}{\operatorname{argmax}} \frac{1}{k} \sum_{\boldsymbol{x}_i \in N(\boldsymbol{x})} I(y_i = j)$$

• For kNN regressor, the mathematical formulation is

$$\hat{h}(\boldsymbol{x}) = \frac{1}{k} \sum_{(\boldsymbol{x}'', y'') \in N(\boldsymbol{x})} y''.$$

One popular choice of distance in kNN is the *Minkowski distance*:

$$d(\boldsymbol{x}, \boldsymbol{z}) = \|\boldsymbol{x} - \boldsymbol{z}\|_r = \left(\sum_{i=1}^p |x_i - z_i|^r\right)^{\frac{1}{r}}, \quad \boldsymbol{x}, \ \boldsymbol{z} \in \mathbb{R}^p.$$
 (2.1)

Note that  $\|\cdot\|^r$  is called the  $\ell^r$  norm.

When r = 1, we have the Manhattan distance:

$$\|\boldsymbol{x} - \boldsymbol{z}\|_1 = |x_1 - z_1| + |x_2 - z_2| + \dots + |x_p - z_p|.$$

When r = 2, we have the Euclidean distance:

$$\|\boldsymbol{x} - \boldsymbol{z}\|_2 = \sqrt{(x_1 - z_1)^2 + (x_2 - z_2)^2 + \dots + (x_p - z_p)^2}.$$

There are other distance / dissimilarity functions which are used in specific cases:

- Gower
- Tanimoto
- Jaccard
- Mahalanobis

1. The given table provides a training data set containing six observations, three predictors and one qualitative response variable. Suppose we wish to use this data set to make a prediction for Y when  $X_1 = X_2 = X_3 = 0$  using k-nearest neighbours.

| Obs. | $X_1$ | $X_2$ | $X_3$ | Y     |
|------|-------|-------|-------|-------|
| 1    | 0     | 3     | 0     | Red   |
| 2    | 2     | 0     | 0     | Red   |
| 3    | 0     | 1     | 3     | Red   |
| 4    | 0     | 1     | 2     | Green |
| 5    | -1    | 0     | 1     | Green |
| 6    | 1     | 1     | 1     | Red   |

| (a) | Compute the Euclidean distance between each observation and the test point (TP).   |
|-----|--|
|     |  |
|     |  |
|     |  |
|     |  |
|     |  |
| (b) | What is our prediction with $k = 1$ ? Why?   |
|     |  |
| (c) | What is our prediction with $k = 3$ ? Why?   |
|     |  |
|     |  |
|     |  |
| (d) | If the Bayes decision boundary in this problem is highly non-linear, then would we expect the optimum value for $k$ to be large or small? Why? |
|     |  |
|     |  |
| (e) | By considering $X_1$ and $X_2$ only, sketch the 3-nearest neighbours decision boundary for range   |
|     | $-1 \le X_1 \le 3$ and $-1 \le X_2 \le 3$ , with the distance measure used in (a). Assume that $X_1$ and $X_2$ can only take integer values.   |
|     | The same and some states.  |

| More   | Peri   | formance Evaluation   |
|--------|--------|---|
| 2. Wh  | at are | e the advantages of $k$ -fold cross validation relative to  |
| (a)    | Valid  | ation set approach  |
| (b)    | Leave  | e-one-out cross validation (LOOCV)  |
| 3. (Ma | ay 201 | 9 Final Q3)   |
| (a)    |        | rvised learning includes classification and regression.  State the difference between classification and regression in term of response variable.  (1 mark) |
|        | (ii)   | Explain the sampling methods used in splitting data for classification and regression respectively.  (4 marks)  |
| (b)    | (i)    | State an issue that comes along with split validation, which can be overcome by using cross validation.  (1 mark)   |
|        | (ii)   | Describe the process of a 5-fold cross validation. (4 marks)  |

| *     | nple of 500 males and 800 females had been collected to test on a model of gender prediction model resulted that 380 males and 510 females were predicted correctly.                         |
|-------|--|
|       | Assume male as positive class and female as negative class, calculate the count of trapositive (TP), true negative (TN), false positive (FP) and false negative (FN) for the model's result. |
|       | (2 mar)  |
| (ii)  | Construct the confusion matrix for the model. State the classification error, specific and sensitivity of the model. (4 mar  |
| (iii) | Compare the recall and precision for both male and female. Interpret your results.   |
| (111) | (4 ma  |
|       |  |