

## 4: Classification

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$ echo "Data Science Institute"
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# What is Classification?

Classification involves predicating a qualitative response by assigning it to a category. The methods that are used to classify observations are called **classifiers** and most of them work by following two steps:

- Compute the probability that an observation belongs to a category.
- Classify the observation based on some probability threshold (i.e. if the probability that an observation belongs to some category is greater than 0.5 then assign the observation to that category)

# Breakout Room

What are some classification methods?

# Why use Classification?

We need to predict a qualitative response.

# Example

On the basis of DNA sequence data for a number of patients with and without a given disease, a biologist would like to figure out which DNA mutations are deleterious (disease-causing) and which are not.

Let's categorize this as a class!

## Breakout Room

A person is donating a blood test. How would you qualify this?

# Why not use linear regression?

Suppose we are trying to diagnose a patient with either a *stroke*, *drug overdose*, or *epileptic seizure* based on their symptoms. We can code this response as follows

$$Y = \begin{cases} 1 & \text{if stroke;} \\ 2 & \text{if drug overdose;} \\ 3 & \text{if epileptic seizure.} \end{cases}$$

At this point we could use linear regression to predict  $Y$  based on a set of predictors. However there are several problems with this coding. One of them is we cannot use linear regression.

## Breakout Room: Why can't we use linear regression?

$$Y = \begin{cases} 1 & \text{if stroke;} \\ 2 & \text{if drug overdose;} \\ 3 & \text{if epileptic seizure.} \end{cases}$$



# Why not use linear regression?

Other problems:

- Implies an ordering of the outcomes.
- The difference between epileptic seizure and stroke versus stroke and drug overdose is assumed to be the same.

# Why not use linear regression?

Suppose we are trying to diagnose a patient with either a *stroke*, *drug overdose*, or *epileptic seizure* based on their symptoms. We can code this response as follows

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At this point we could use linear regression to predict  $Y$  based on a set of predictors. However there are several problems with this coding

A different ordering would give completely different results for the linear regression. ♦

***There is no convenient way to code a qualitative response with more than two levels so that linear regression can be used.*** ♦

# Why not use linear regression?

The 0/1 coding for a binary qualitative response variable does not suffer the same problems. However the probabilities we obtain will be difficult to interpret

- negative probabilities
- probabilities above 1

So, linear regression only able to give ♦ ***crude estimates of the probabilities for a binary response.*** ♦

In summary, we don't use linear regression for classification since:

- It does not work for a qualitative response variable with more than 2 classes.
- With 2 classes, the probability estimates are not meaningful.

# $K$ -Nearest Neighbours

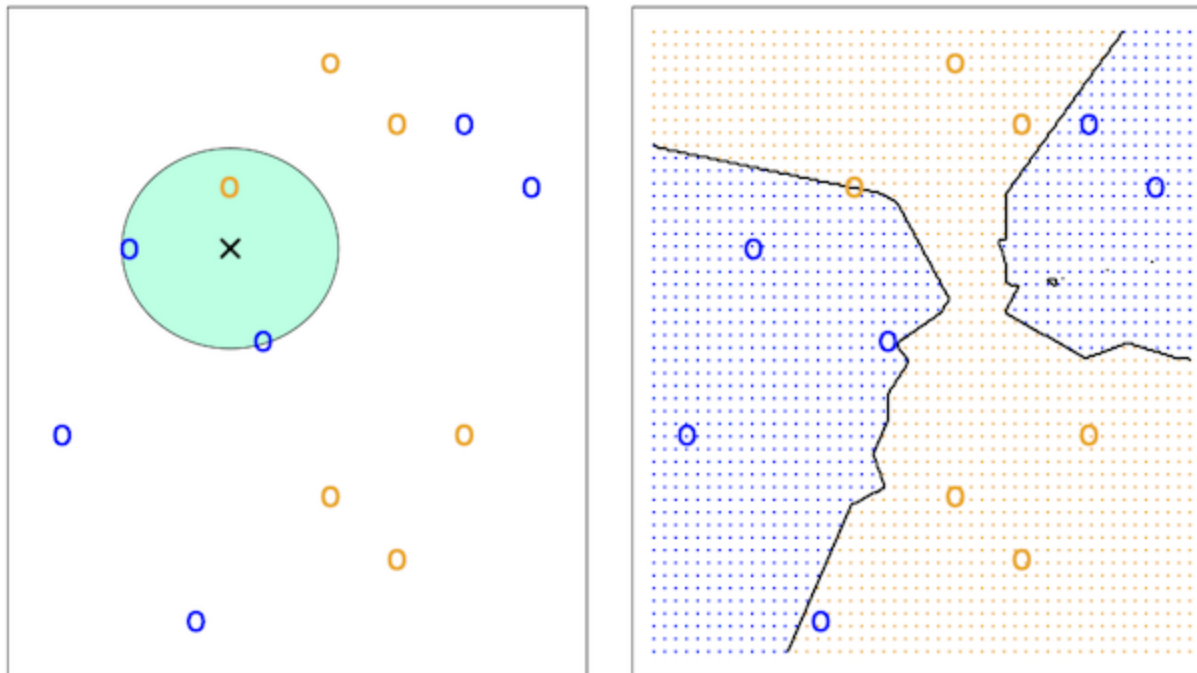
The  $K$ -nearest neighbors (KNN) classifier works very differently than any of the previous classification methods. For a test observation  $x_0$ , it identifies  $K$  training data points that are closest to  $x_0$  (represented by  $\mathcal{N}_0$ ) and estimates the conditional probability for class  $j$  as

$$\Pr(Y = j \mid X = x_0) = \frac{1}{K} \sum_{i \in \mathcal{N}_0} I(y_i = j)$$

where  $I(y_i = j)$  is an **indicator variable** that equals 1 if  $y_i = j$  and 0 otherwise. The KNN classifier classifies the test observation  $x_0$  to the class for which the above probability is the largest.

# $K$ -Nearest Neighbours

These figures illustrate the KNN approach with  $K = 3$ . To the left we see the 3 closest points to  $x$  are 1 orange and 2 blue so this observation will be classified as blue. The right figure shows the decision boundaries where an observation will be classified as blue or orange.



# Exercise: K-Nearest Neighbours

Open the Classification Exercises Jupyter Notebook file.

- Go over the "K-Nearest Neighbours" section together as a class.
- 5 minutes for students to complete the questions from "K-Nearest Neighbours".
- Questions should be completed at home if time does not allow.

# References

Chapter 4 and section 2.2.3 of the ISLP book:

James, Gareth, et al. "Classification." An Introduction to Statistical Learning: with Applications in Python, Springer, 2023.