



# Introduction to Data Analysis in R (Day 2)

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T-test

Fisher's exact test

Wilcoxon rank sum test

# Hypothesis Testing

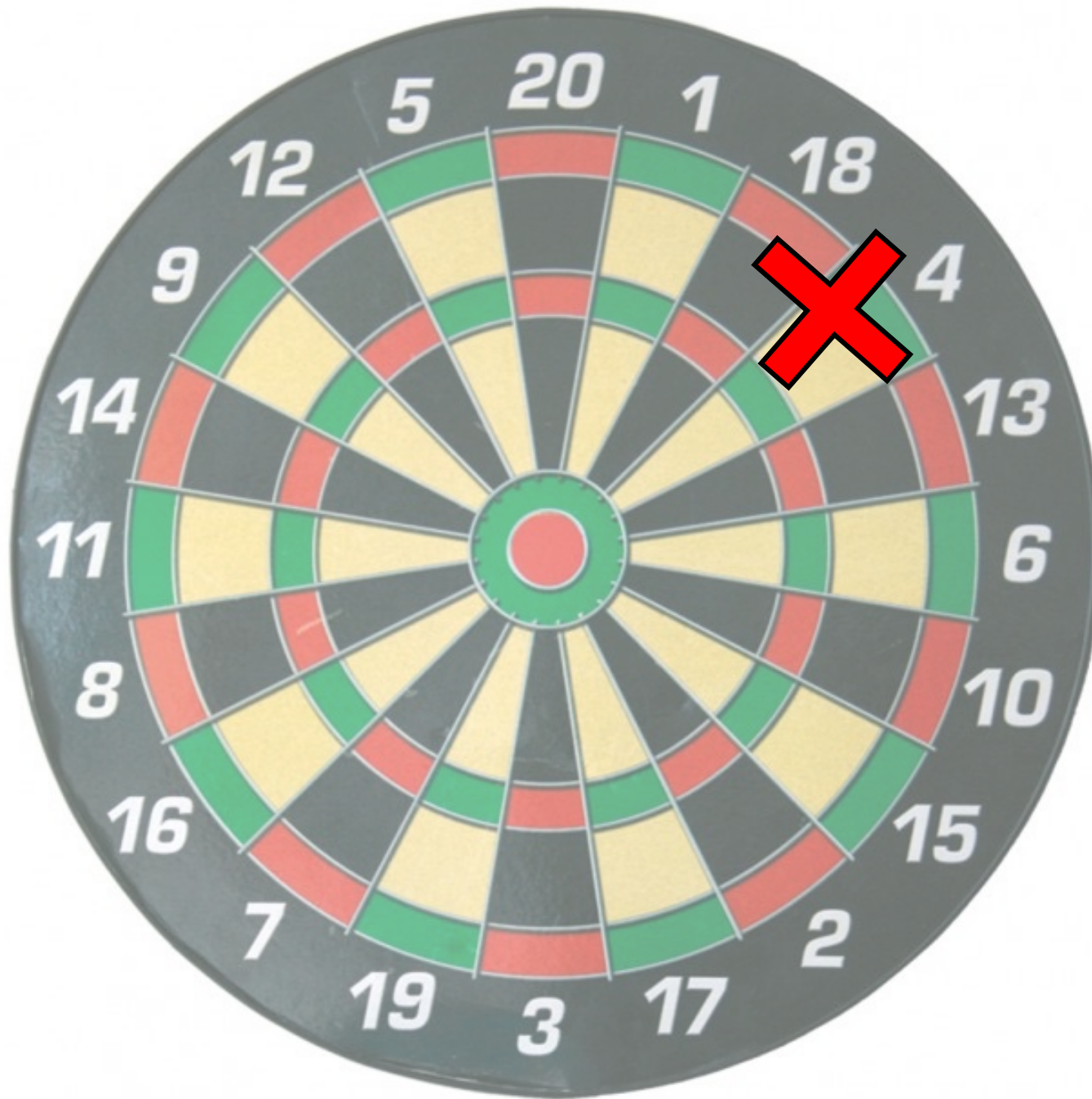
Correlation test

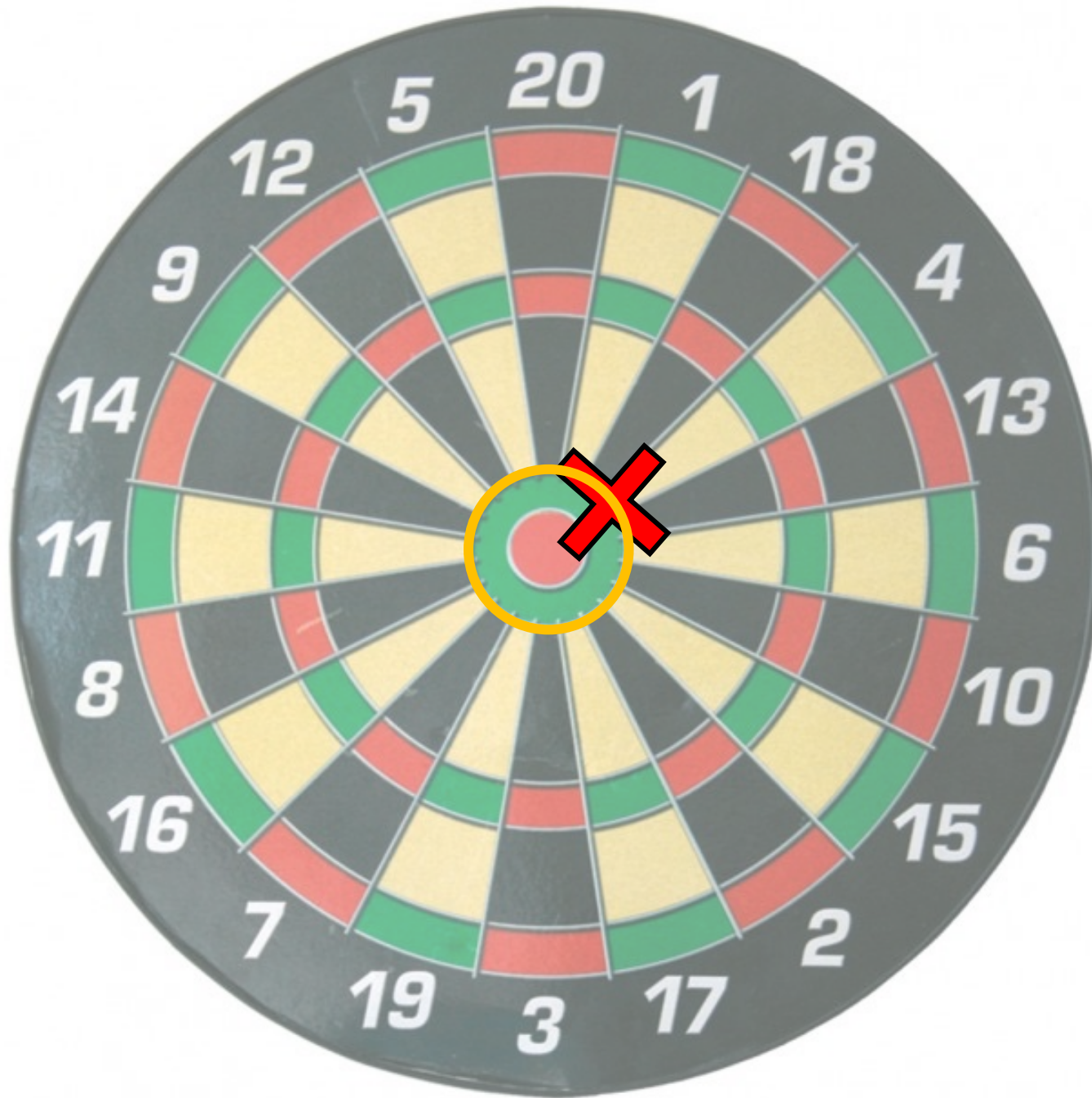
Hypergeometric test

Chi squared test

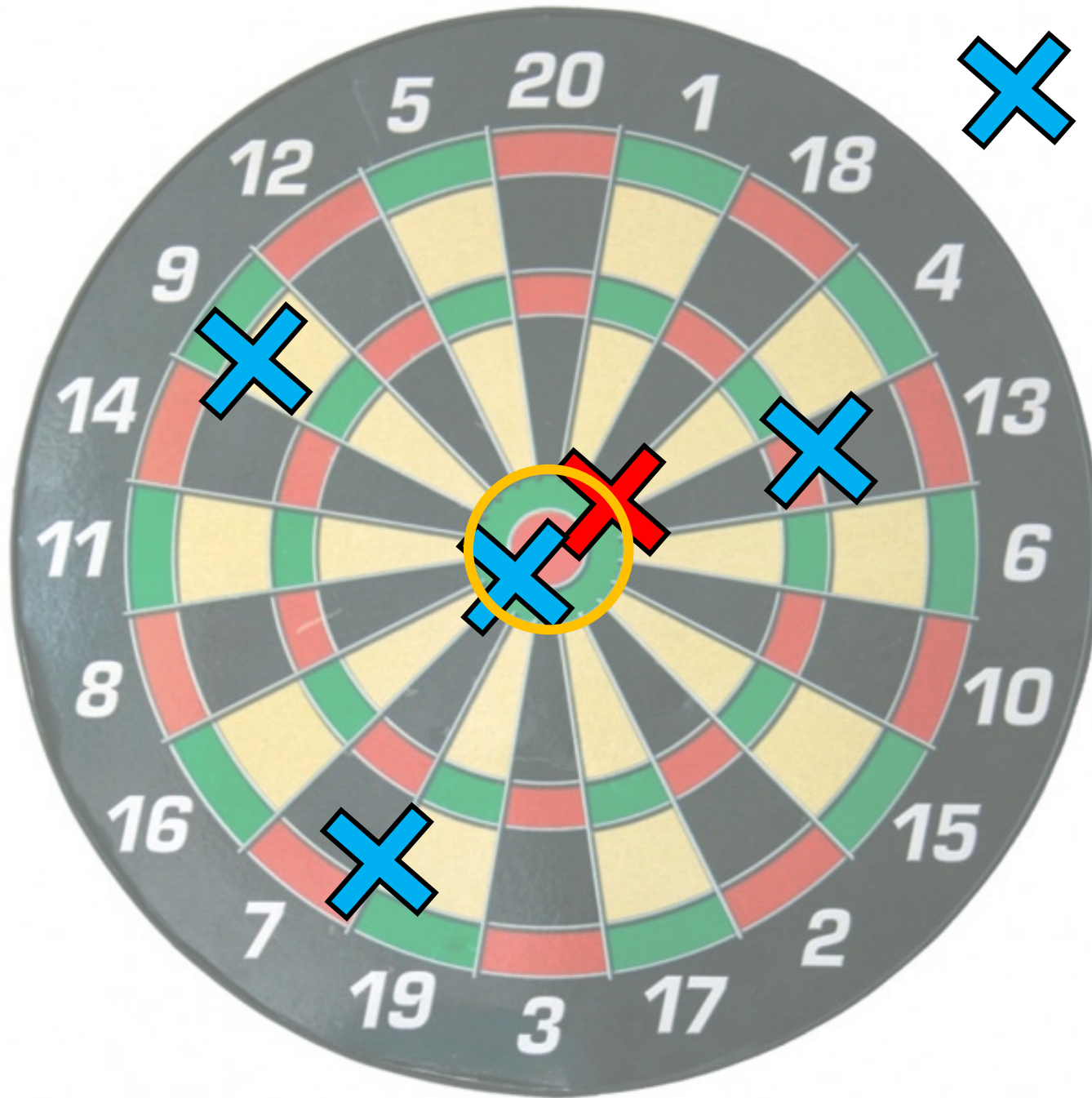










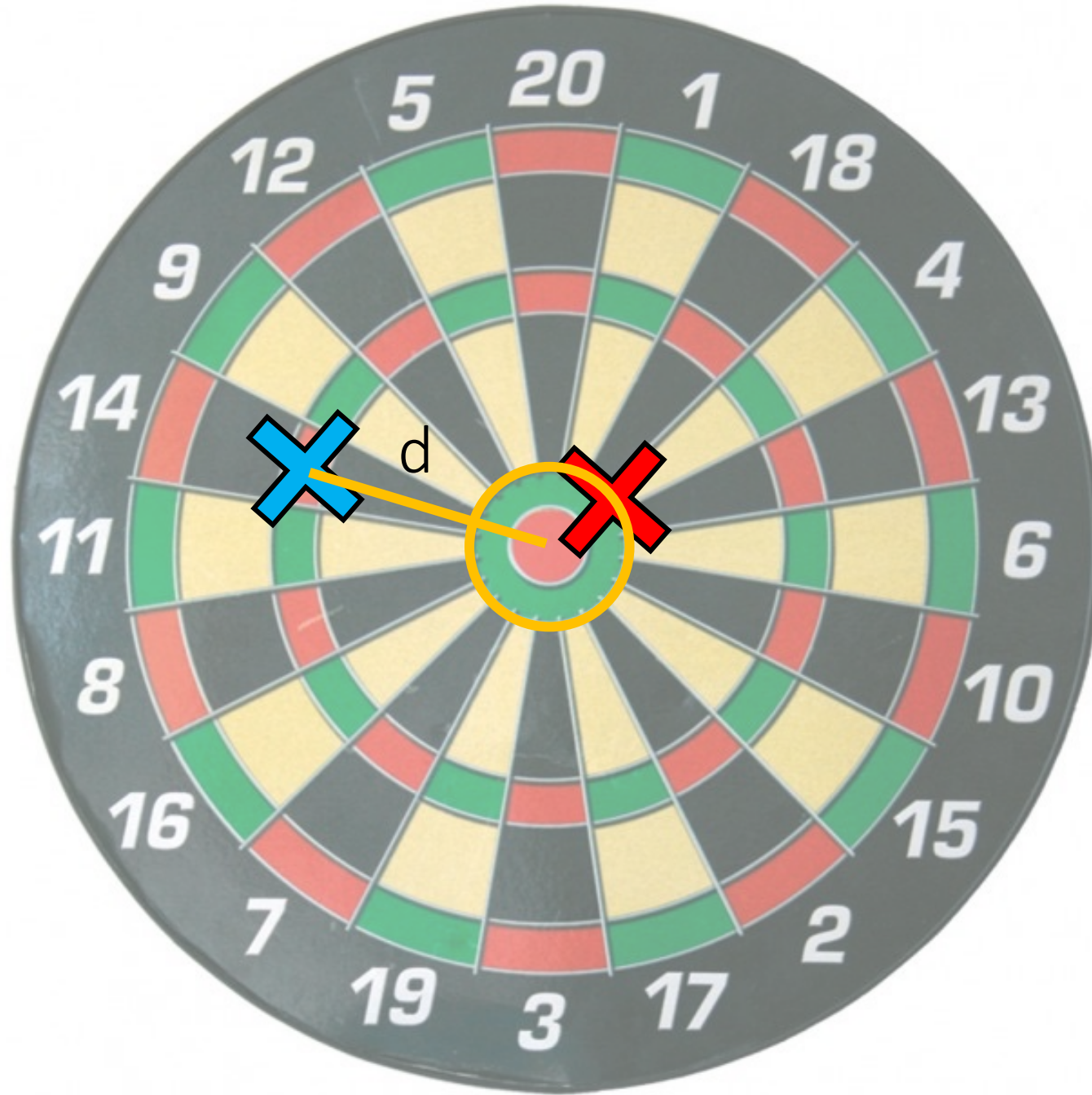
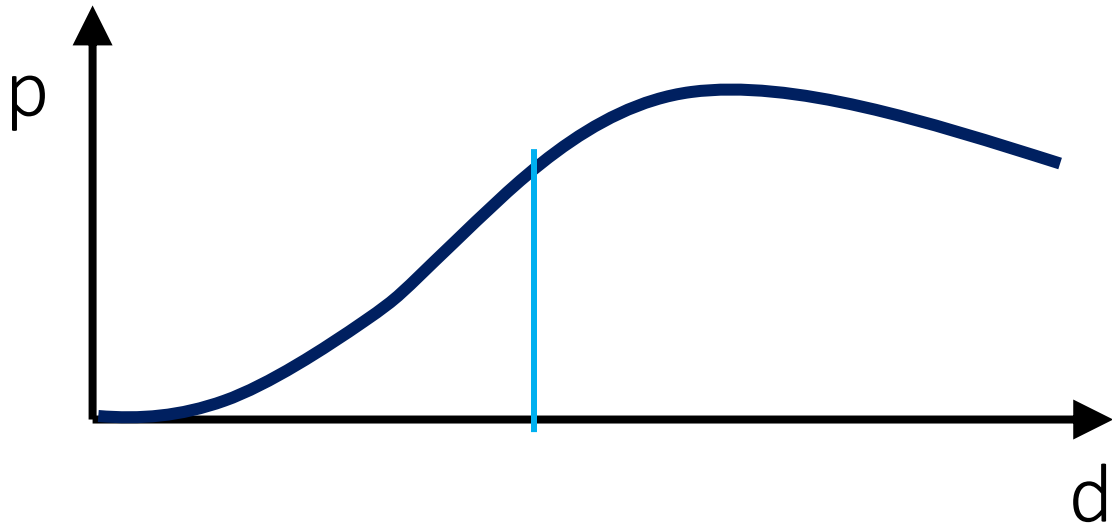


1. Experiment

2. Being pessimistic:

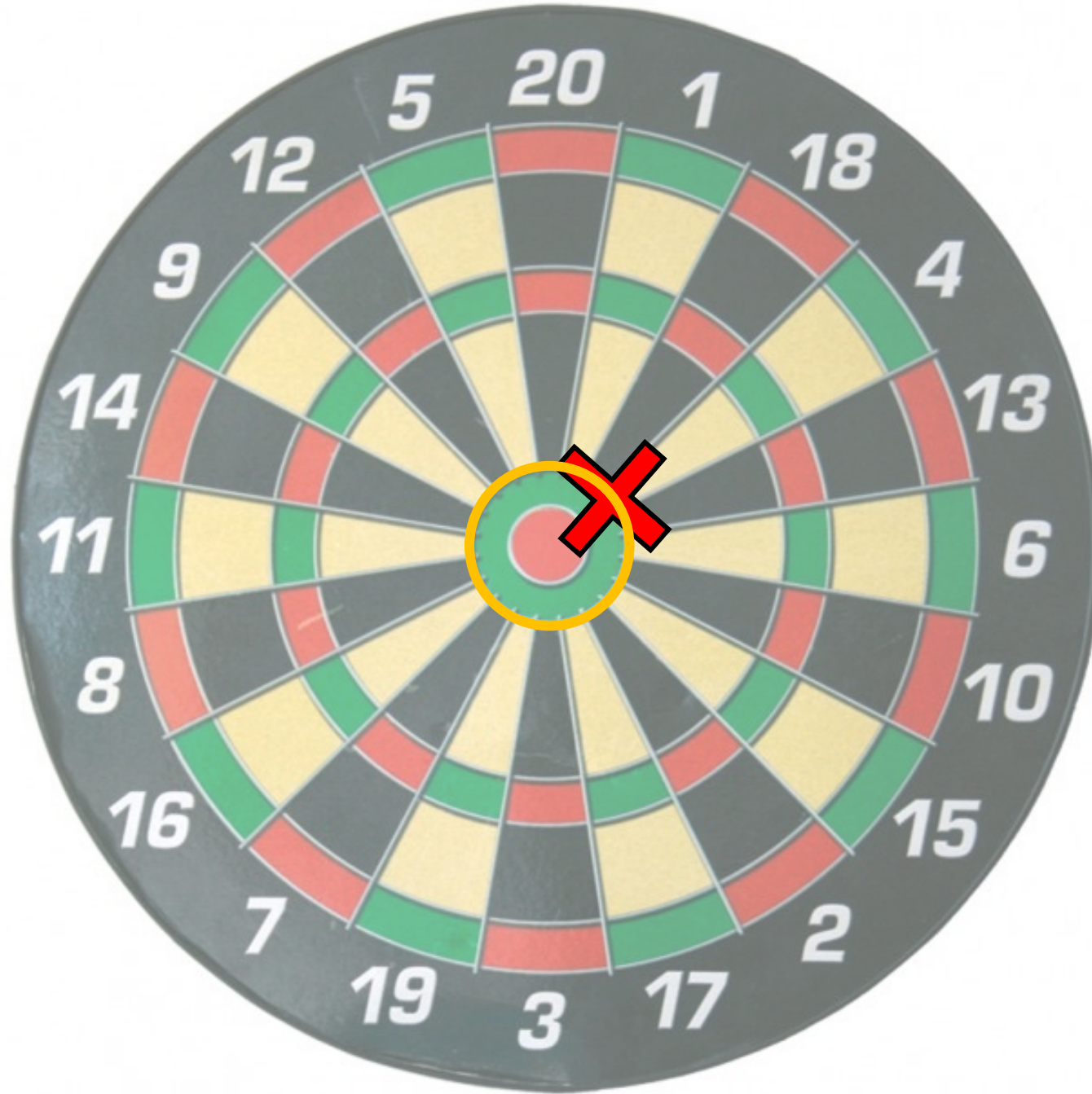
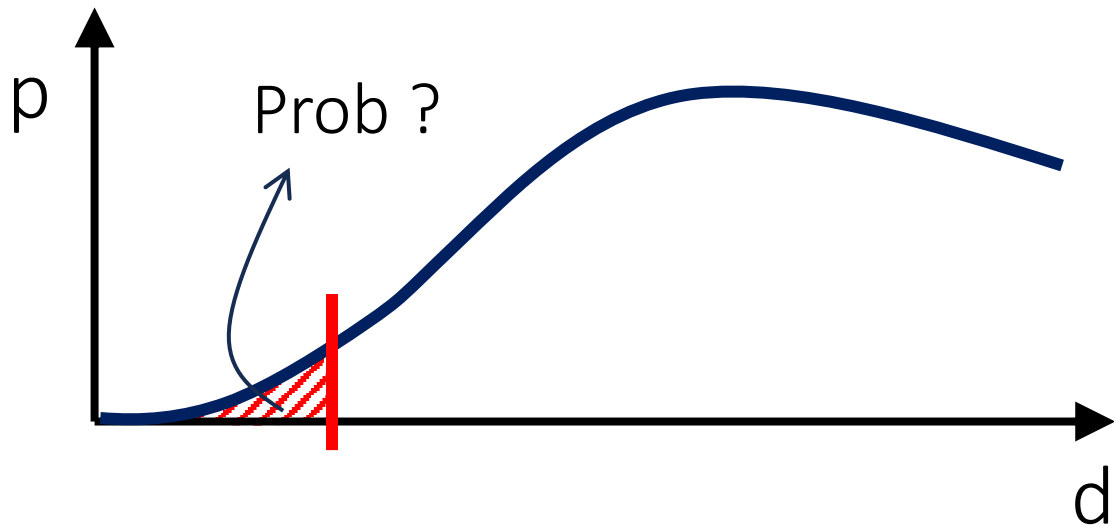
You got there only by chance  
(null hypothesis)

3. Obtain the null distribution





1. Experiment
2. Being pessimistic:  
You got there only by chance  
(null hypothesis)
3. Obtain the null distribution
4. Hypothesis testing





# Hypothesis testing

- **What does P value Mean?**

The P value is defined as the probability under the assumption of no effect or no difference (null hypothesis), of obtaining a result equal to or more extreme than what was actually observed.

# Hypothesis testing

$$p - value < 0.05 \quad (\alpha)$$

We can **reject** the null hypothesis and accept the **alternative hypothesis**

$$p - value > 0.05 \quad (\alpha)$$

We do not have **enough evidenced to reject** the null hypothesis.

# Hypothesis Testing

- Null Hypothesis:  $H_0$
- Alternative Hypothesis:  $H_1$
- Significance level ( $\alpha$ )



# T-test: one sample test

- Null Hypothesis ( $H_0$ ):  $\mu_1 = 0$
- Alternative Hypothesis ( $H_1$ )  $\mu_1 \neq 0$

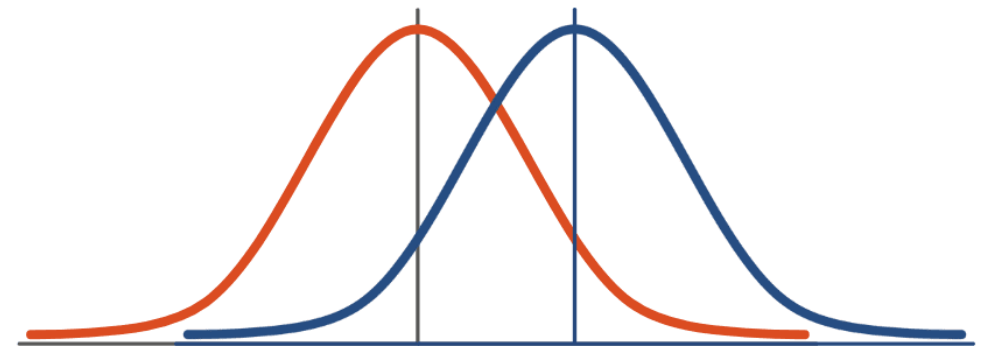
# T-test: two sample test

- Null Hypothesis ( $H_0$ ):

$$\mu_1 = \mu_2$$

- Alternative Hypothesis ( $H_1$ )

**T-TEST**



# For Loops

```
> for (i in 1:10){  
>   # your code  
> }
```

```
> set = c(1, 10, 20)  
> for (i in set){  
>   # your code  
> }
```



# Adjusting p-value

- Bonferroni correction
- Benjamini-Hochberg correction

# Fisher's Exact Test

# Fisher's Exact Test

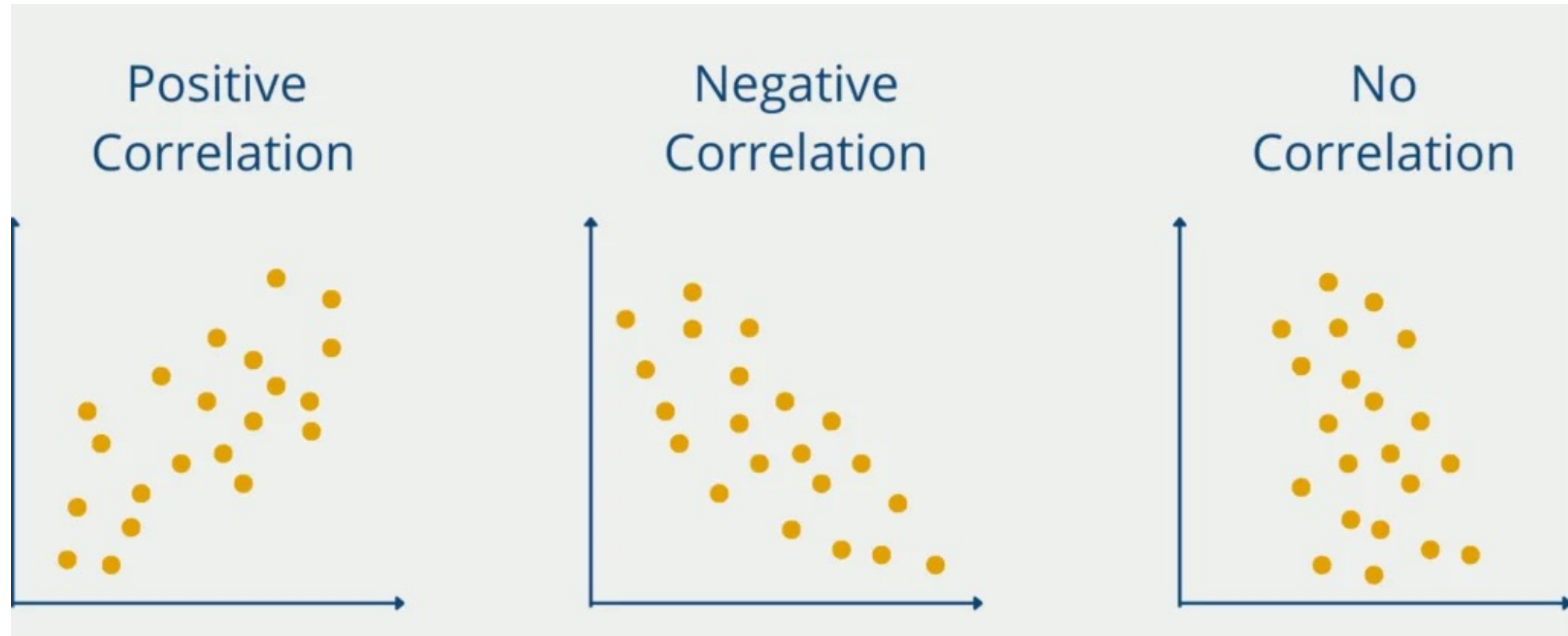
	Group 1	Group 2
Condition 1	$a$	$b$
Condition 2	$c$	$d$

$$H_0: \frac{a}{b} = \frac{c}{d}$$



# Covariance and Correlation

# Covariance and Correlation



# Covariance and Correlation

Variance

$$\text{var}(x) = \frac{1}{N-1} \sum_{i=1}^N (x_i - \bar{x})^2$$

Covariance

$$\text{cov}(x, y) = \frac{1}{N-1} \sum_{i=1}^N (x_i - \bar{x})(y_i - \bar{y})$$

Pearson  
Correlation

$$\text{cor}(x, y) = \frac{\text{cov}(x, y)}{\sigma_x \sigma_y}$$

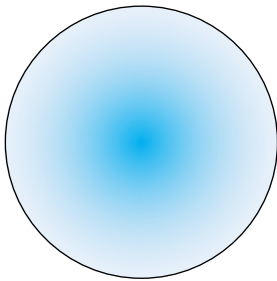
# Covariance Matrix

$$C_{xy} = \begin{bmatrix} var(x) & cov(x, y) \\ cov(y, x) & var(y) \end{bmatrix}$$

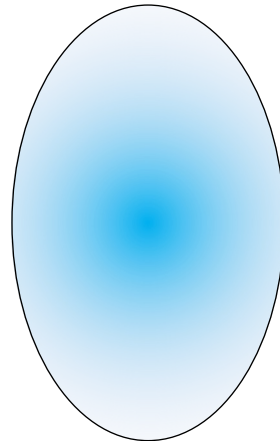
# Covariance Matrix

$$C_{xy} = \begin{bmatrix} \text{var}(x) & \text{cov}(x, y) \\ \text{cov}(y, x) & \text{var}(y) \end{bmatrix}$$

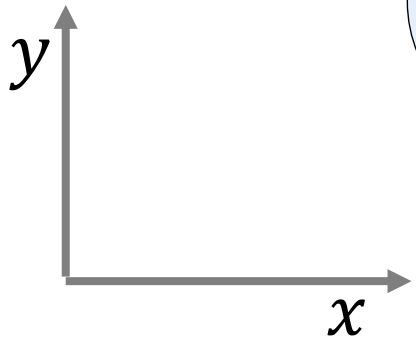
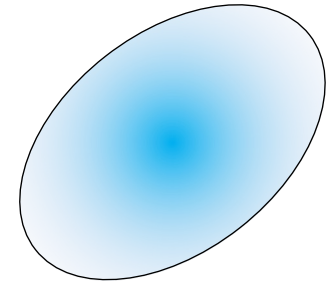
$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$



$$\begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}$$



$$\begin{bmatrix} 1 & 0.5 \\ 0.5 & 1 \end{bmatrix}$$



# Principle Component Analysis

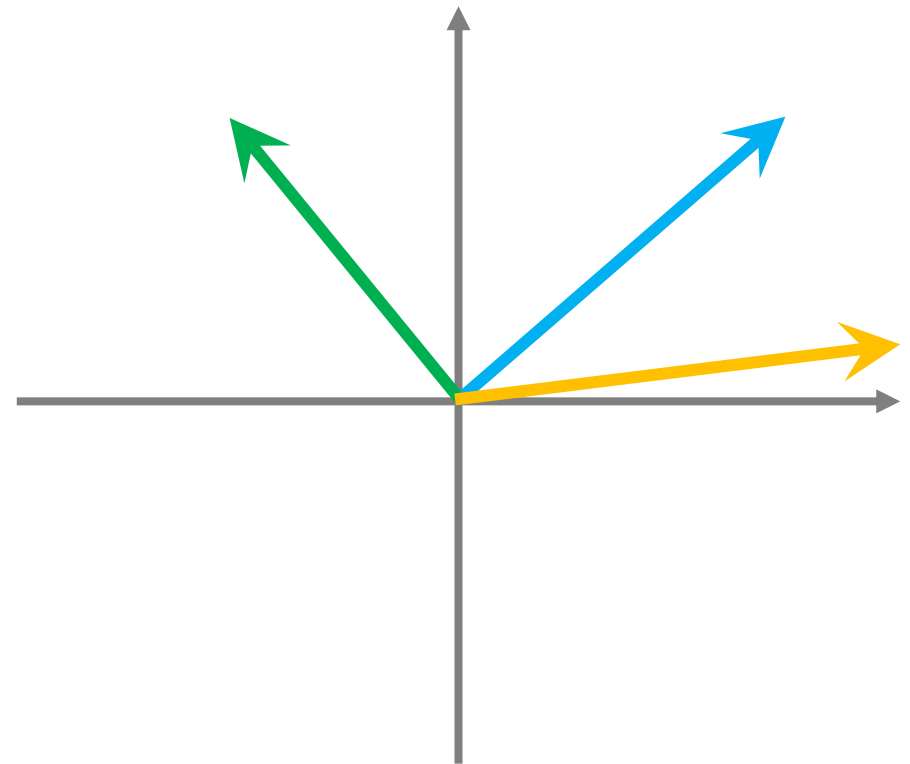
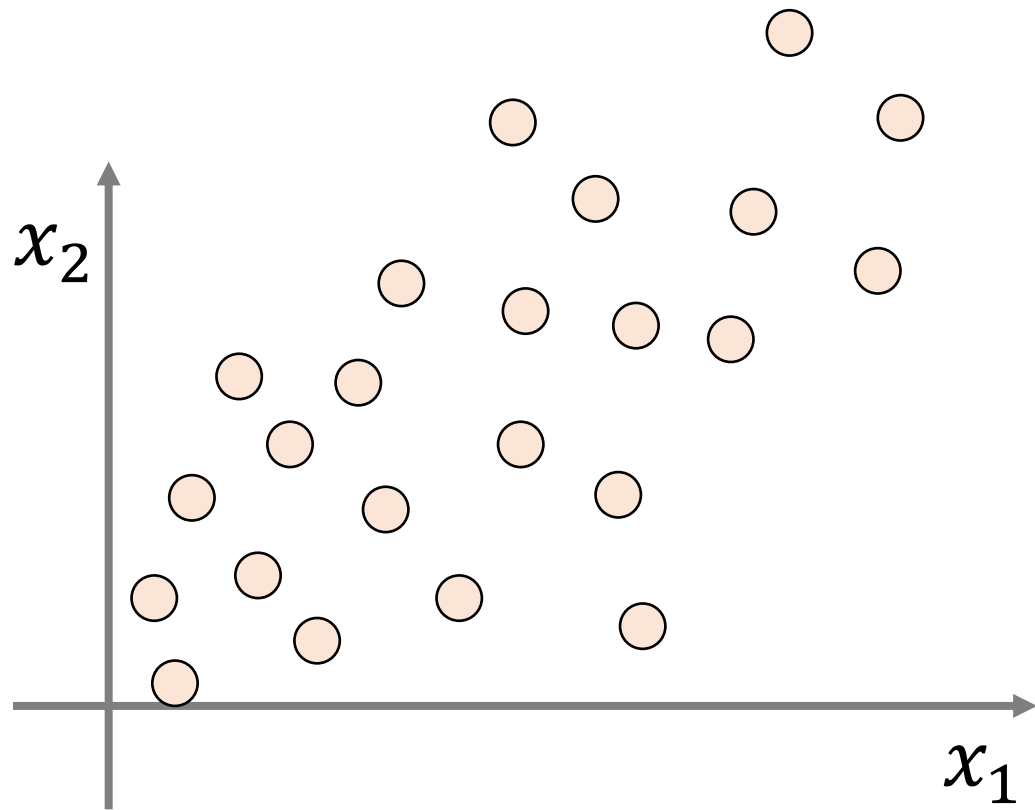


# Principle Component Analysis

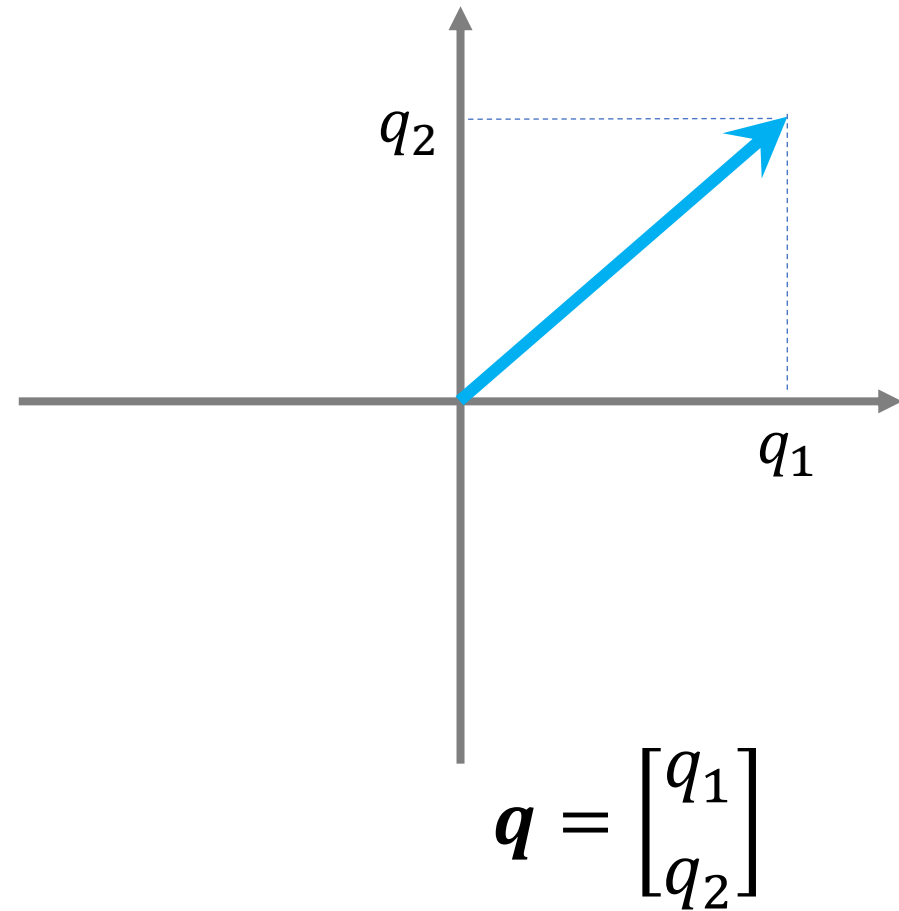
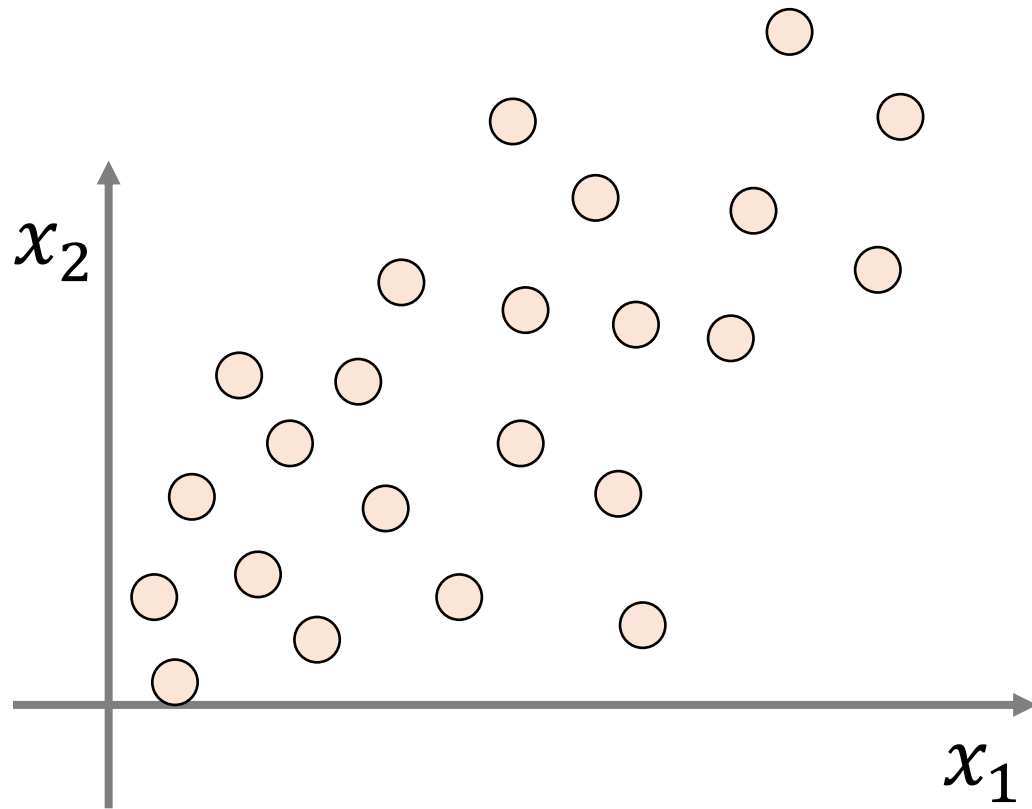
What animal does this shadow belong to?



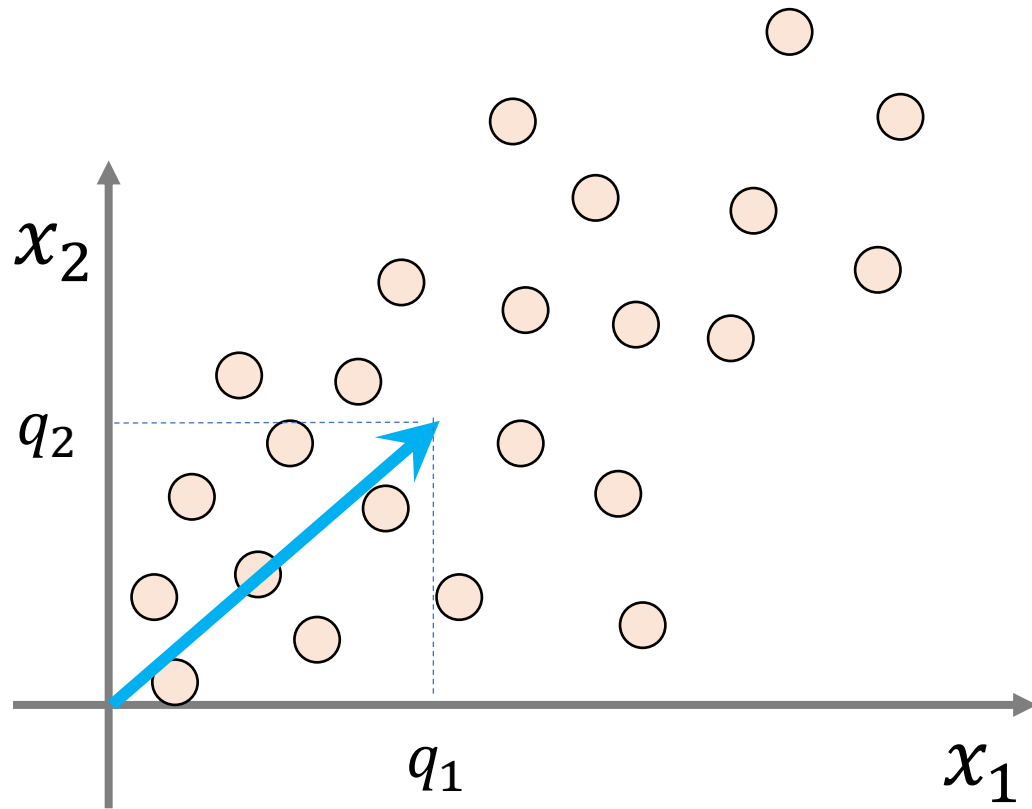
# Principle Component Analysis



# Principle Component Analysis



# Principle Component Analysis



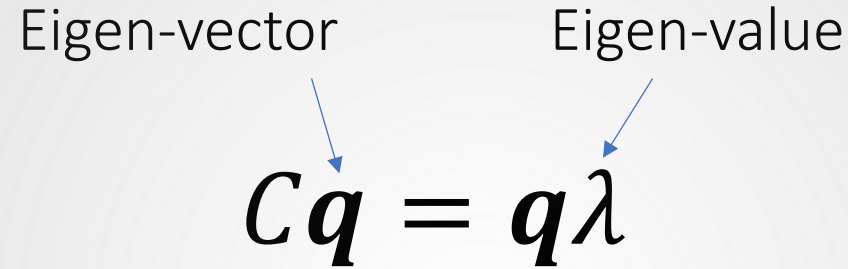
$q = \begin{bmatrix} q_1 \\ q_2 \end{bmatrix}$  is an **eigen-vector** of

$$C = \begin{bmatrix} \text{var}(x_1) & \text{cov}(x_1, x_2) \\ \text{cov}(x_1, x_2) & \text{var}(x_2) \end{bmatrix}$$

# Eigen Analysis

Eigen-vector

Eigen-value


$$C\mathbf{q} = \mathbf{q}\lambda$$

$$C\mathbf{q}_1 = \mathbf{q}_1\lambda_1$$

$$C\mathbf{q}_2 = \mathbf{q}_2\lambda_2$$

...

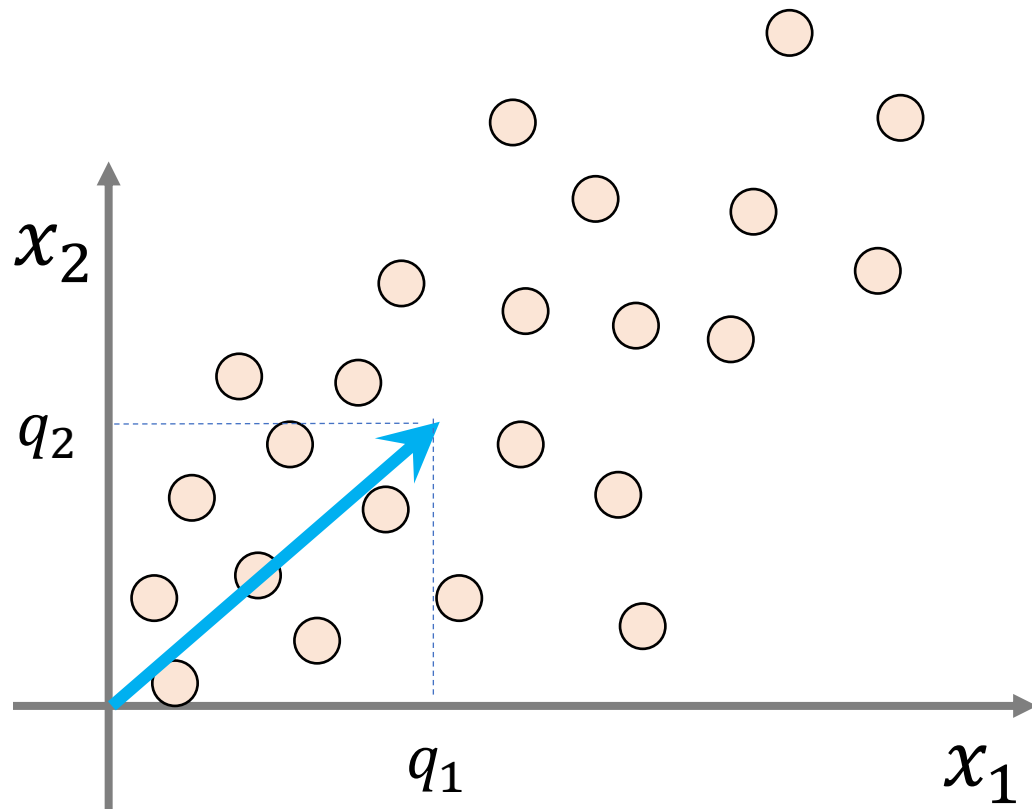
$\lambda_1$   
v

$\lambda_2$

v

...

# Principle Component Analysis



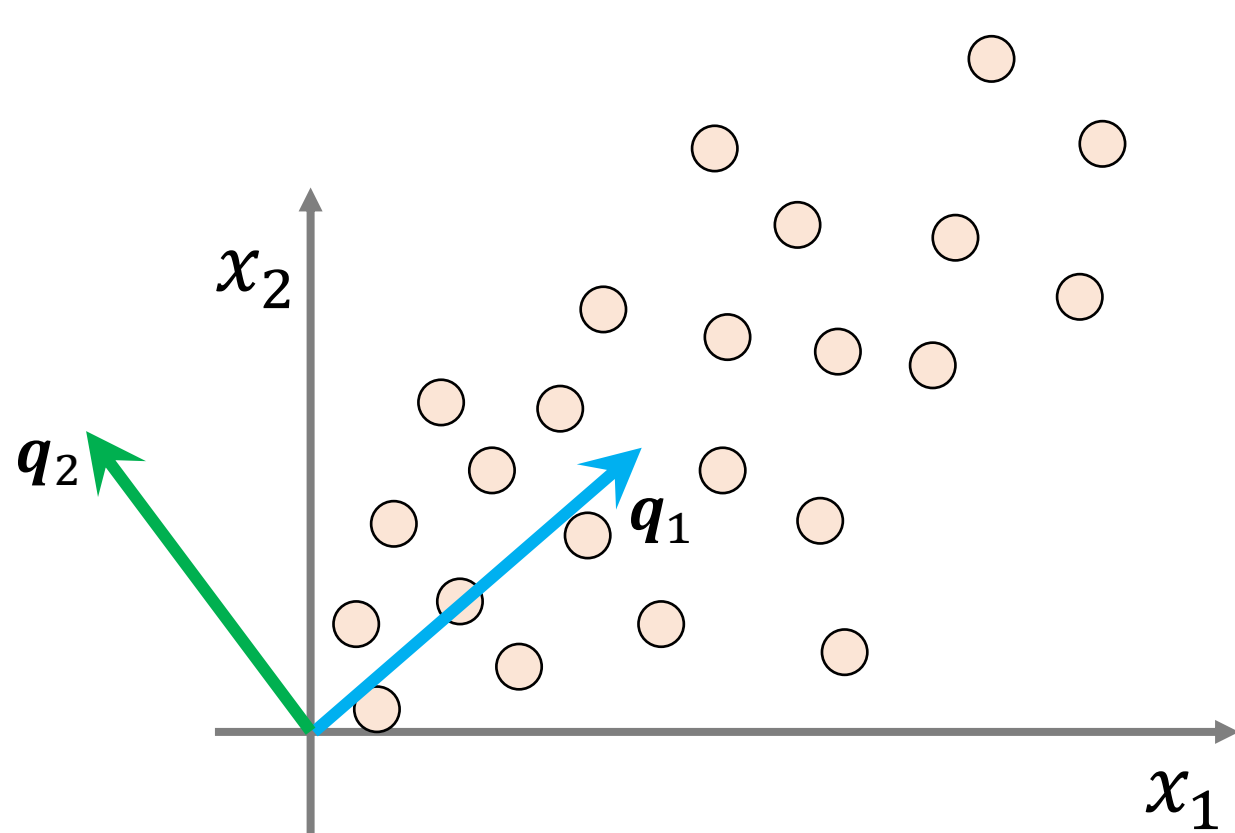
$q = \begin{bmatrix} q_1 \\ q_2 \end{bmatrix}$  is the **eigen-vector** of

$$C = \begin{bmatrix} \text{var}(x_1) & \text{cov}(x_1, x_2) \\ \text{cov}(x_1, x_2) & \text{var}(x_2) \end{bmatrix}$$

that corresponds to the largest **eigen-value**

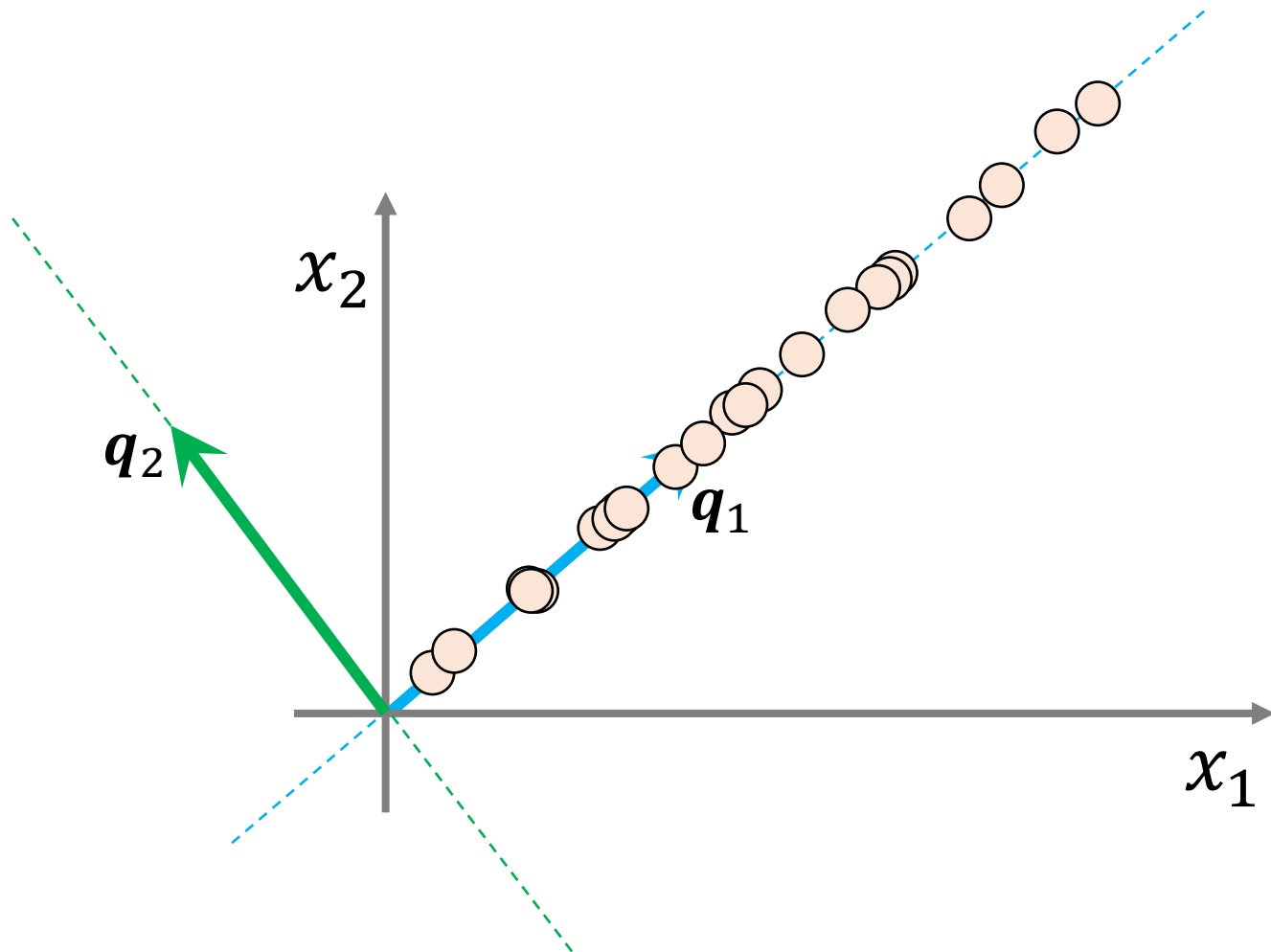


# Principle Component Analysis



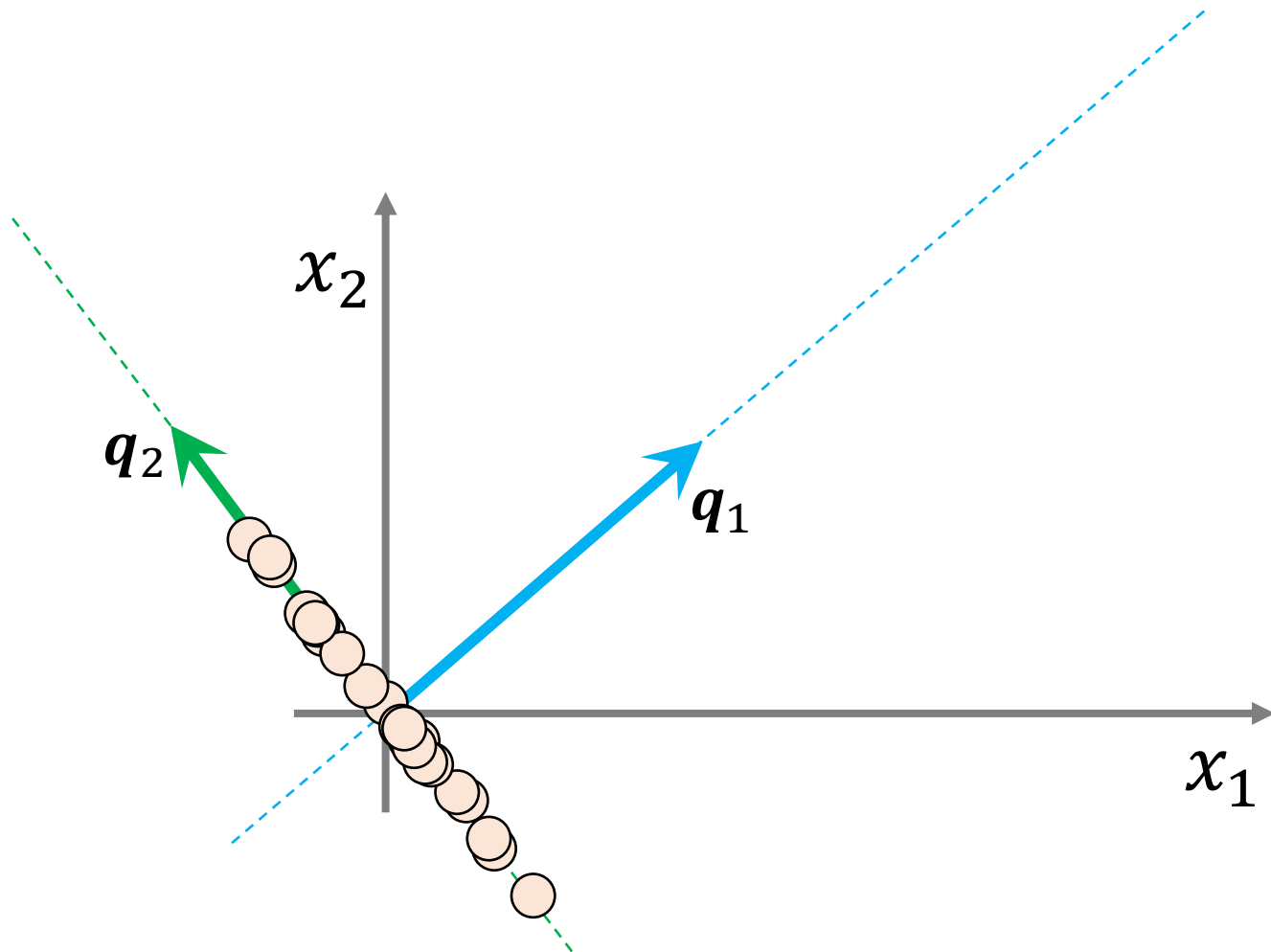
$$q_1 \perp q_2 \perp \dots$$

# Principle Component Analysis



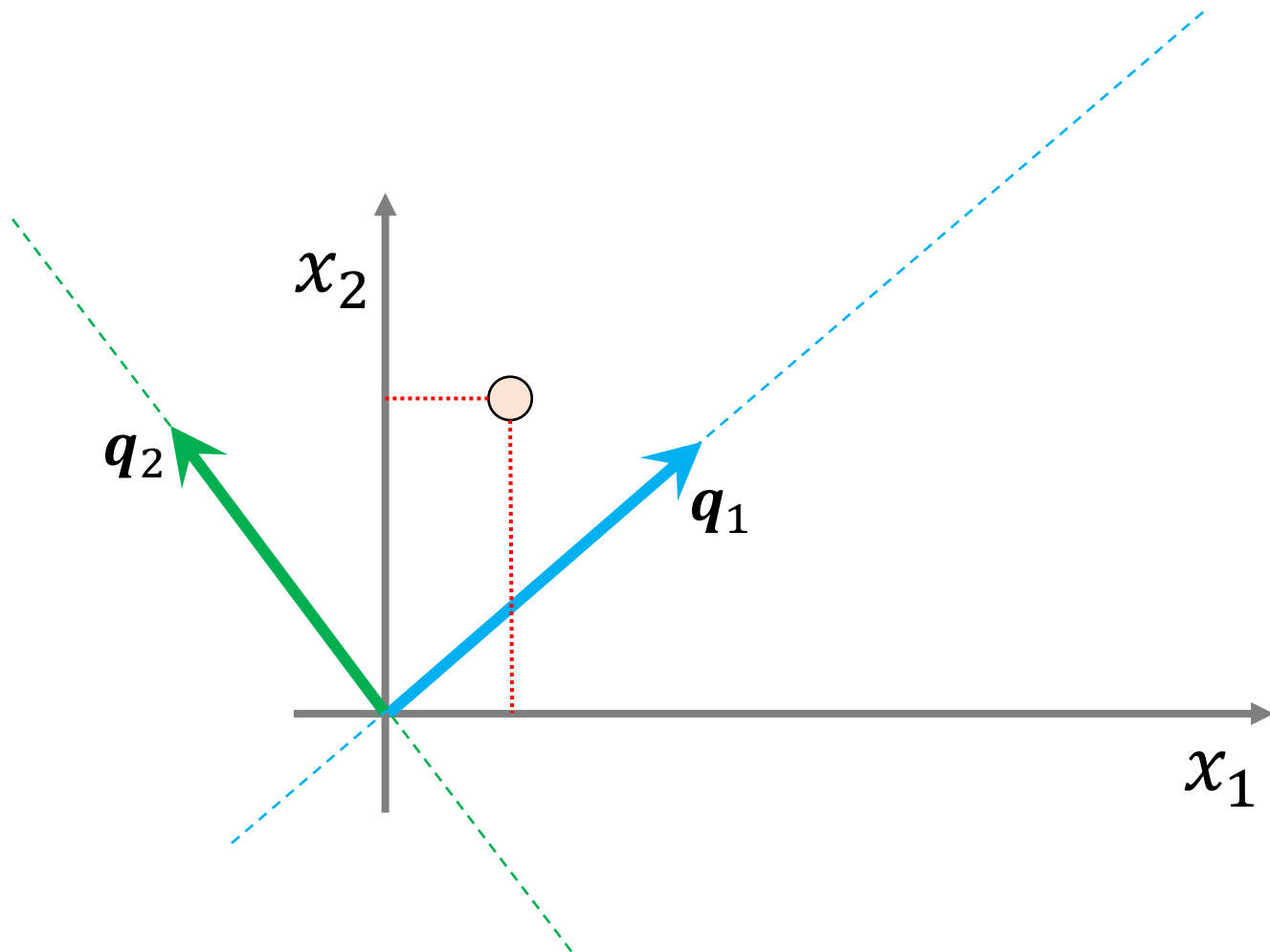
$$pc_1 = Xq_1$$

# Principle Component Analysis

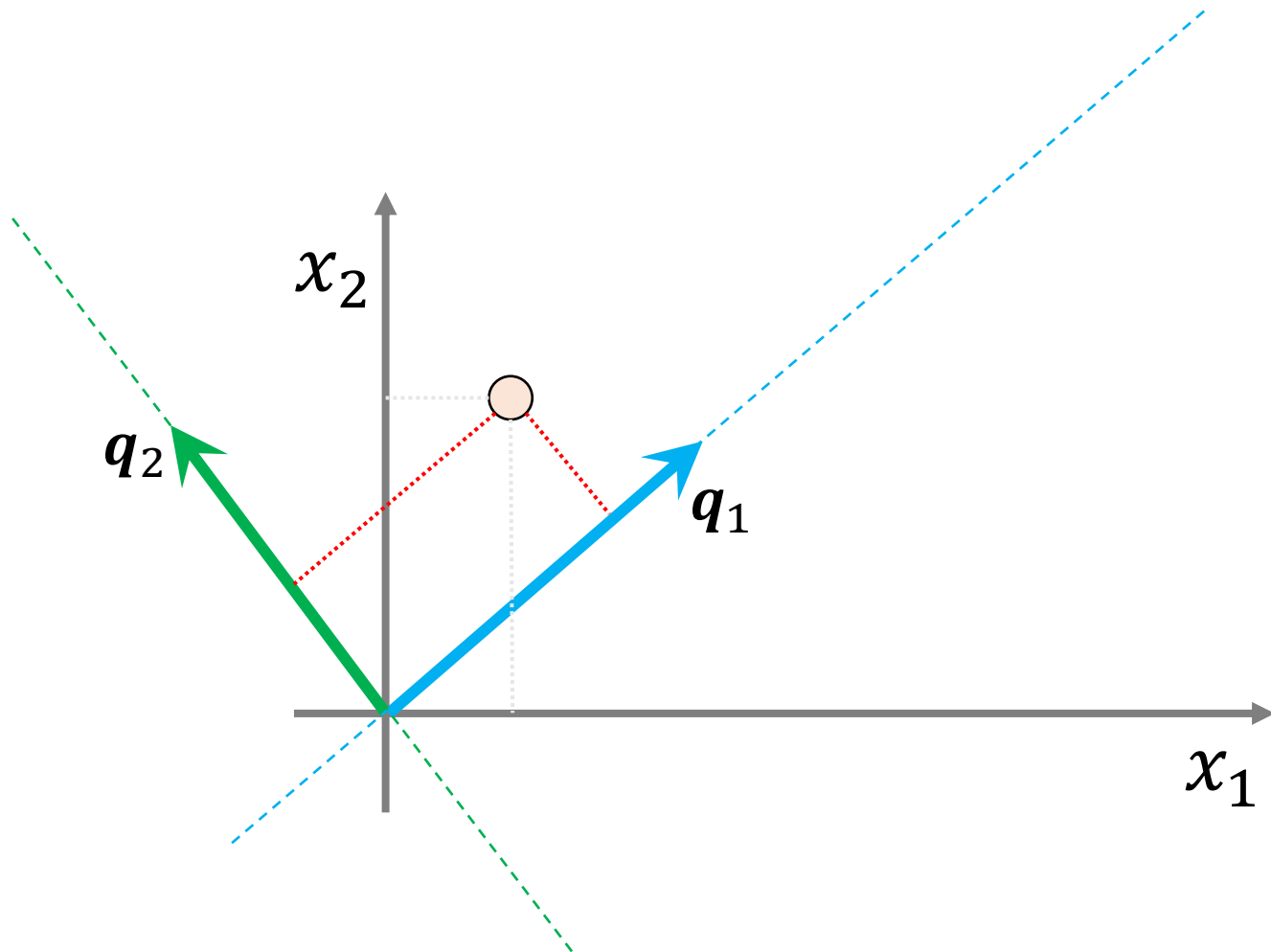


$$pc_2 = Xq_2$$

# Principle Component Analysis



# Principle Component Analysis

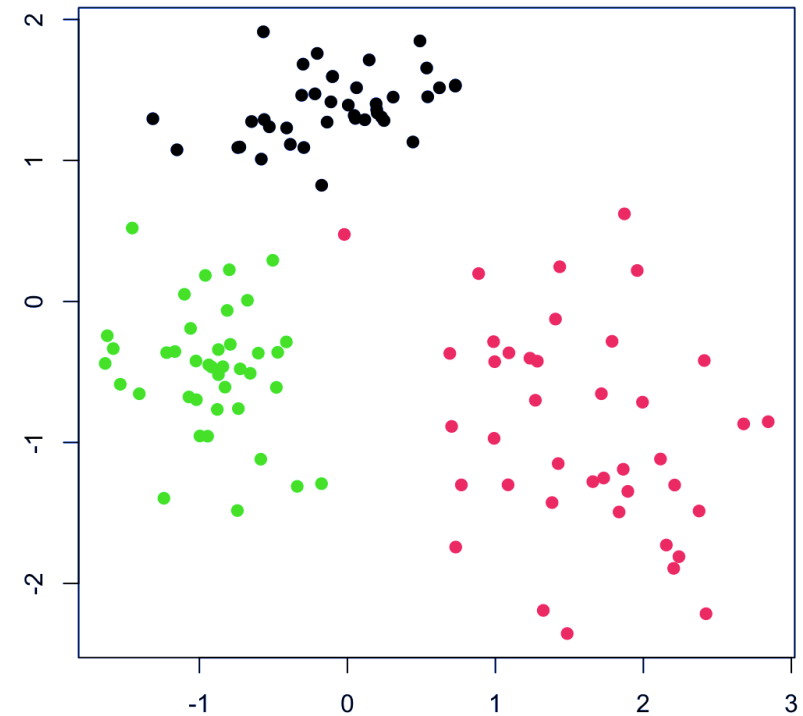
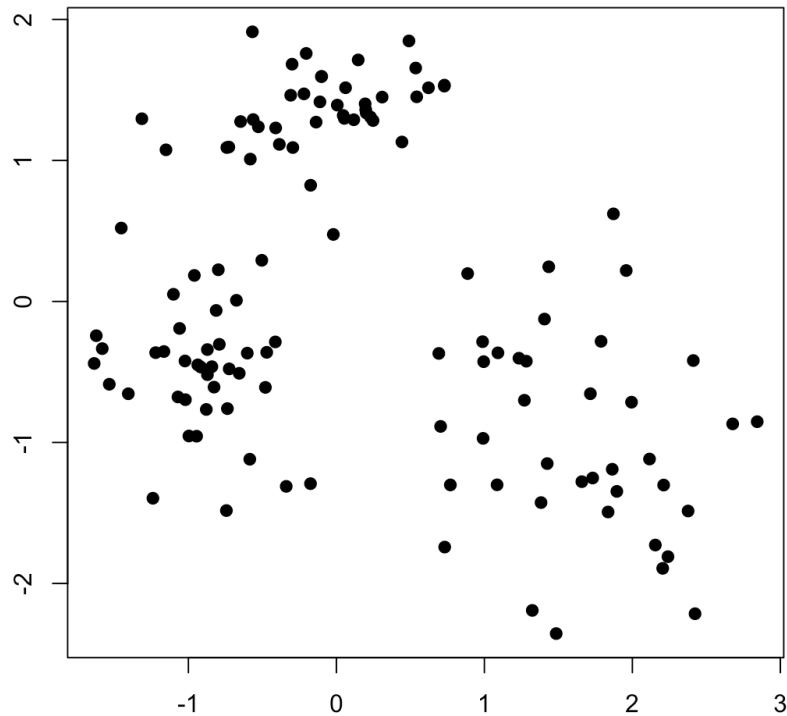


# Clustering



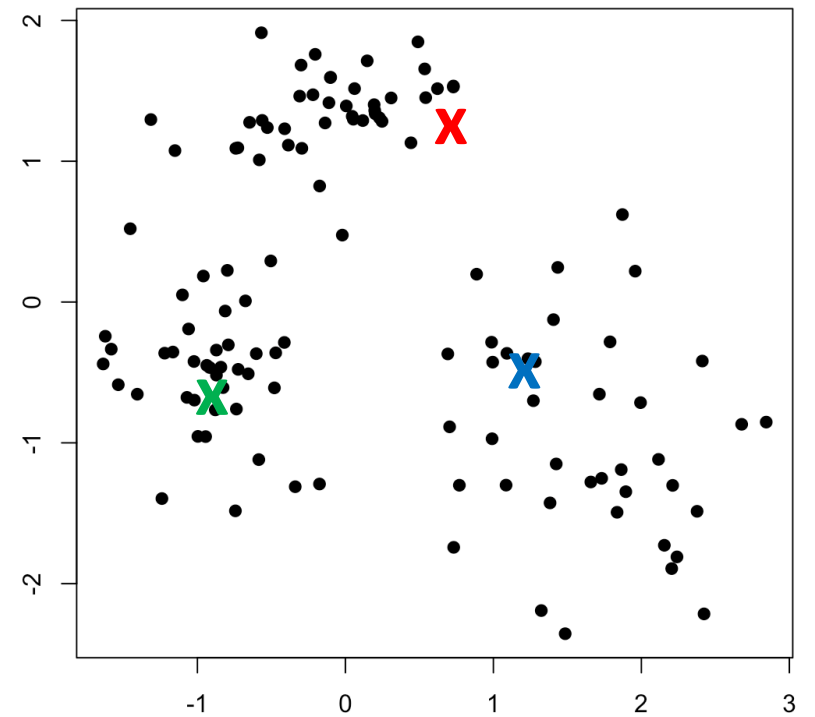
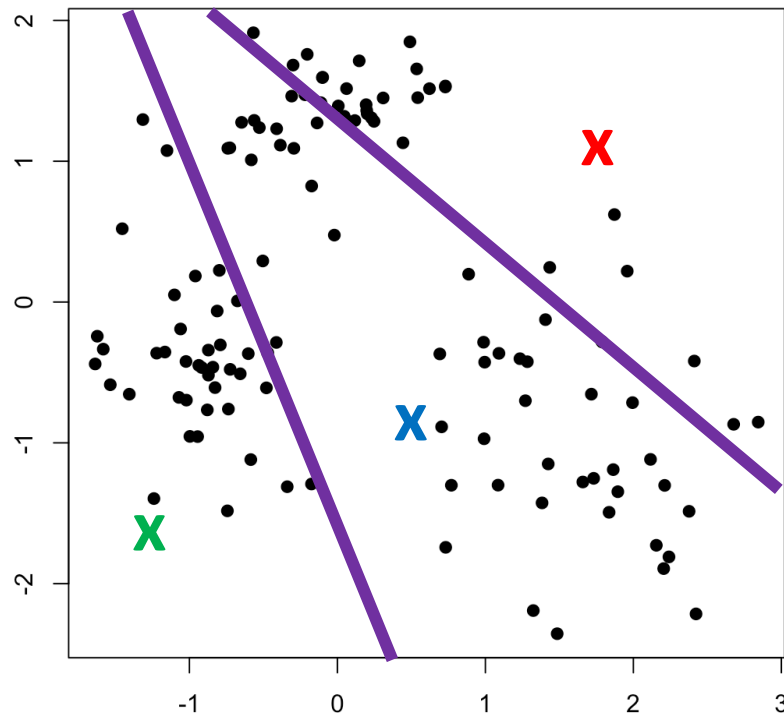
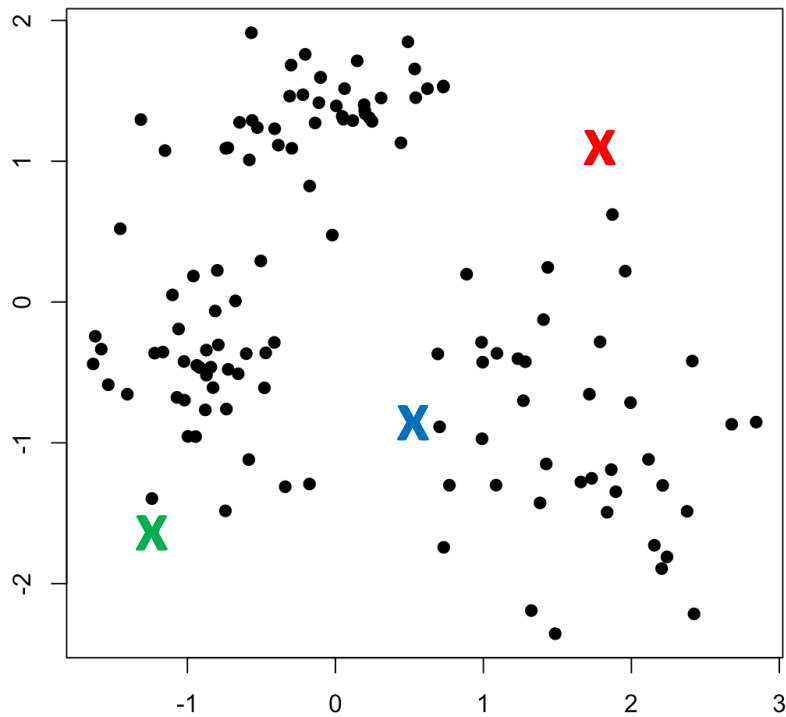
# Clustering

Identify groups of similar data points in an unsupervised manner.



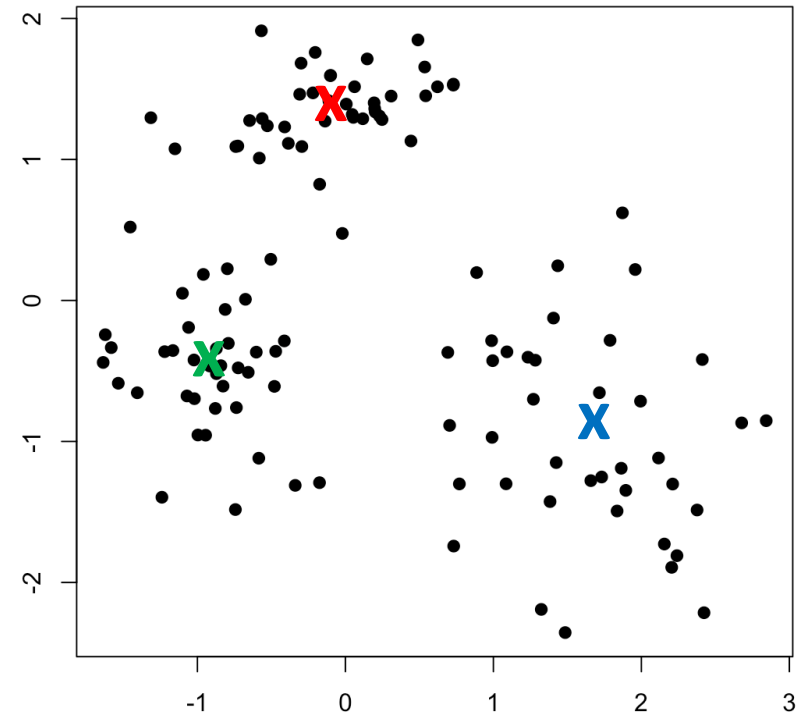
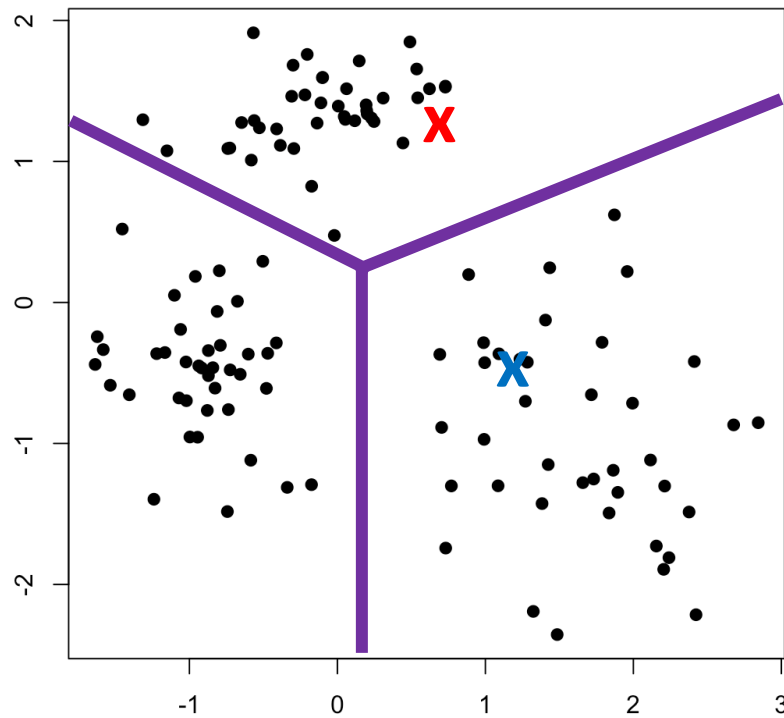
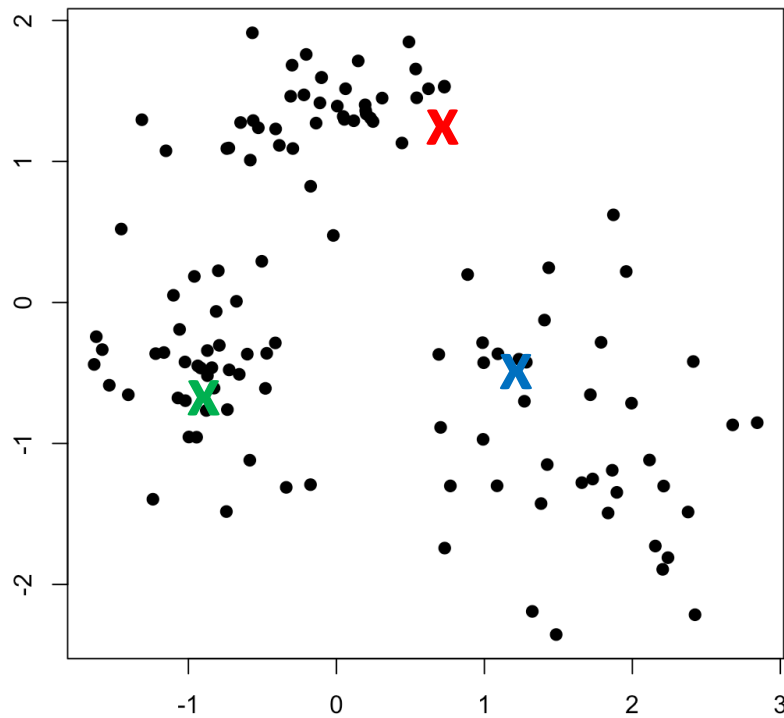
# Clustering

## K-means clustering



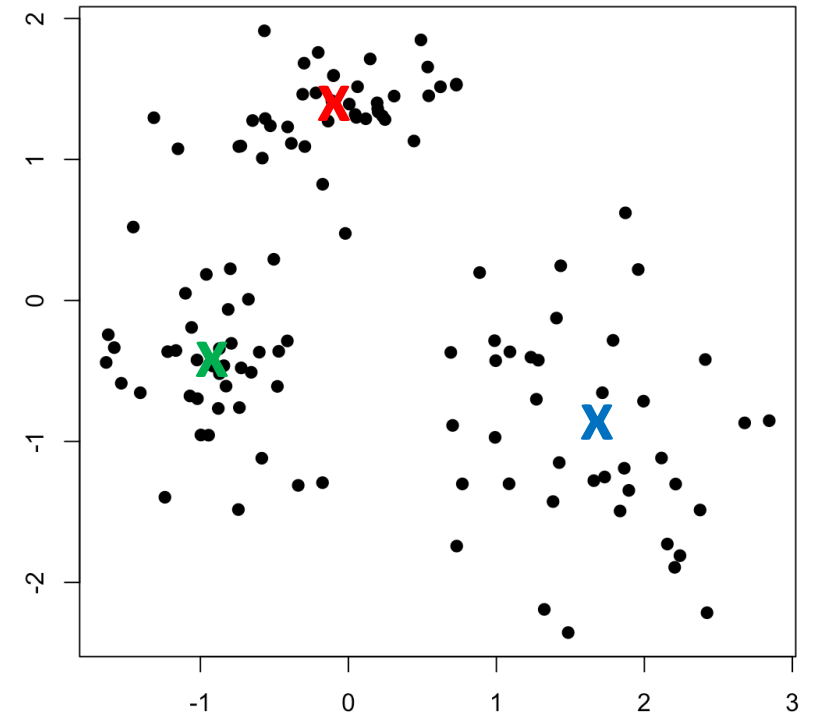
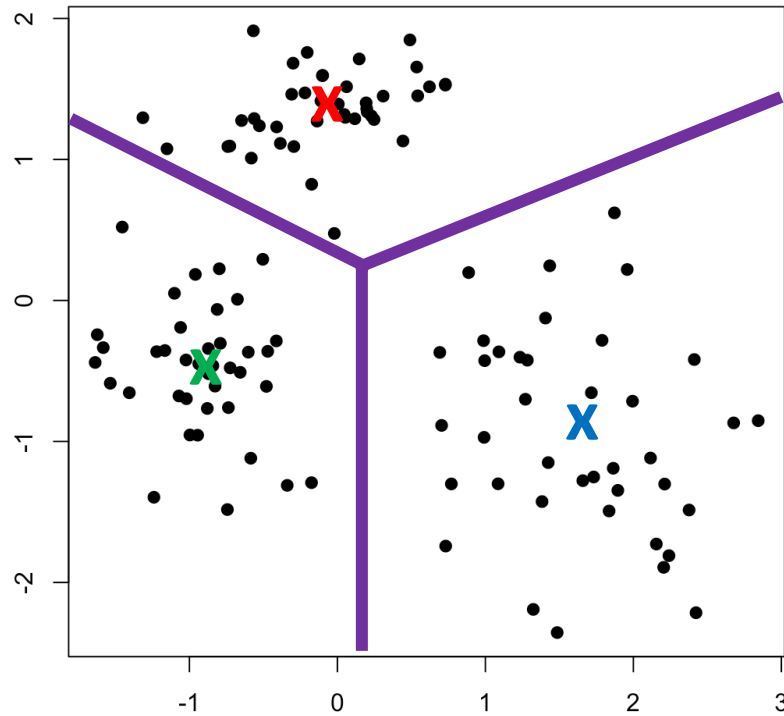
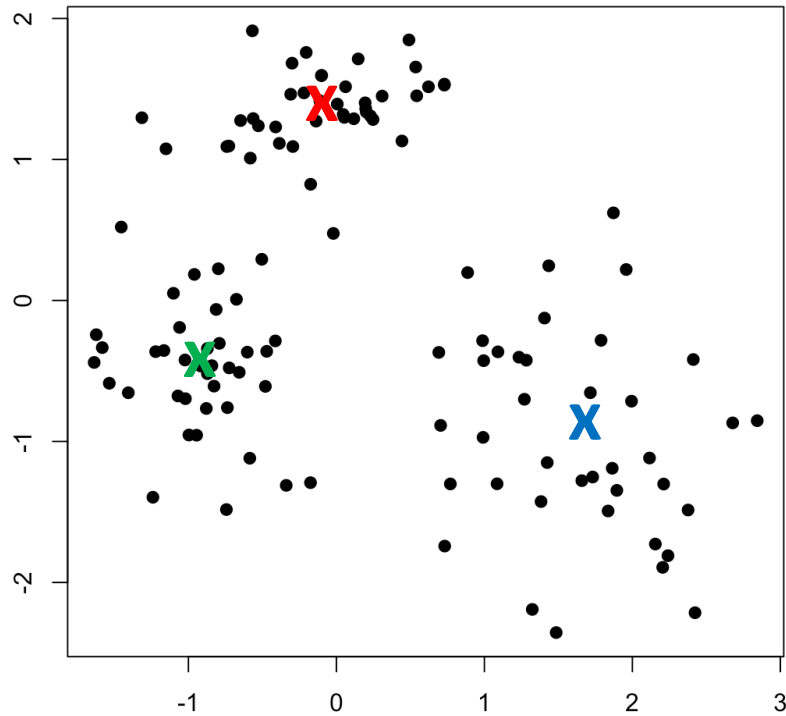
# Clustering

## K-means clustering



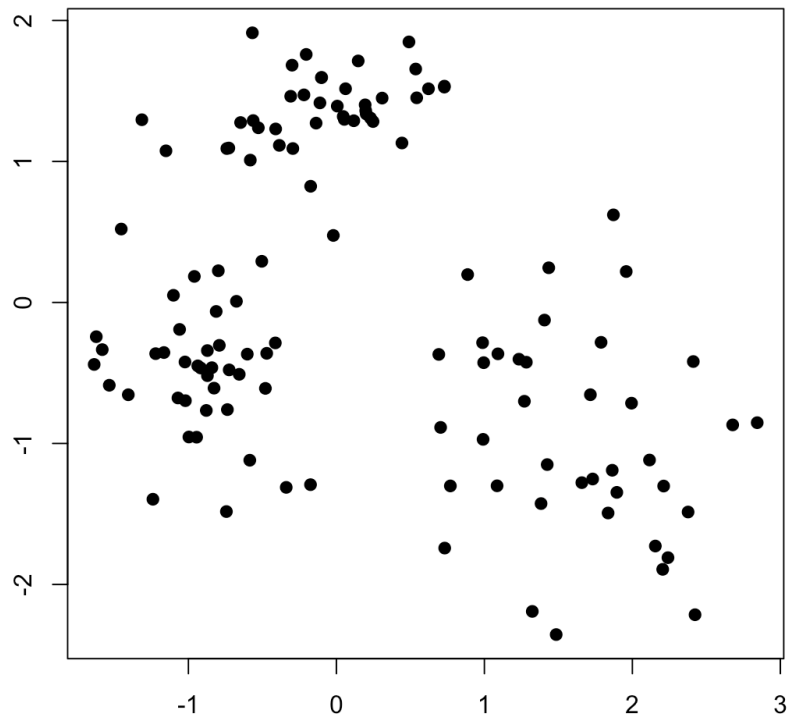
# Clustering

## K-means clustering

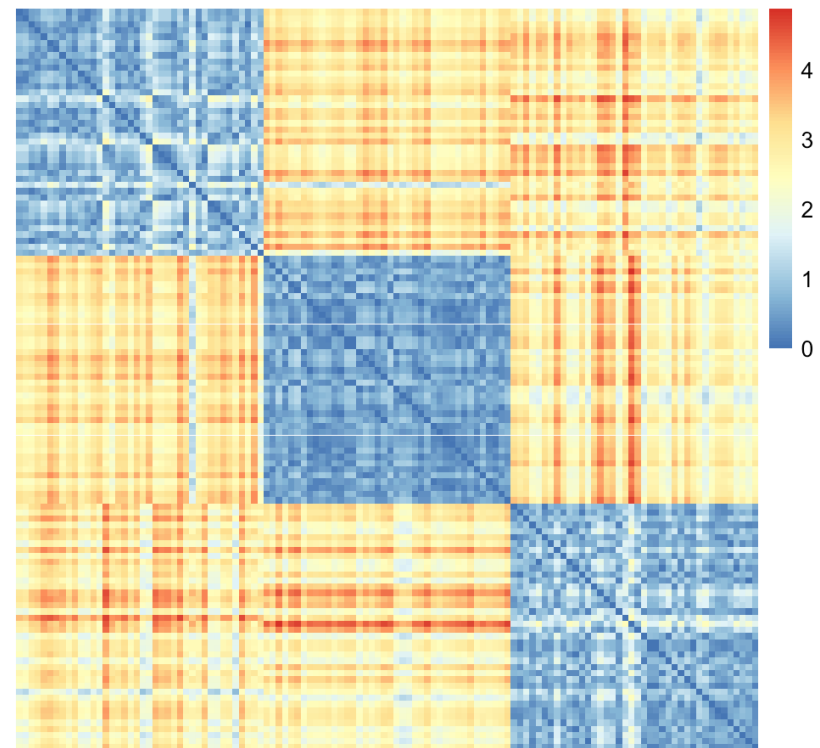


# Clustering

## Hierarchical clustering

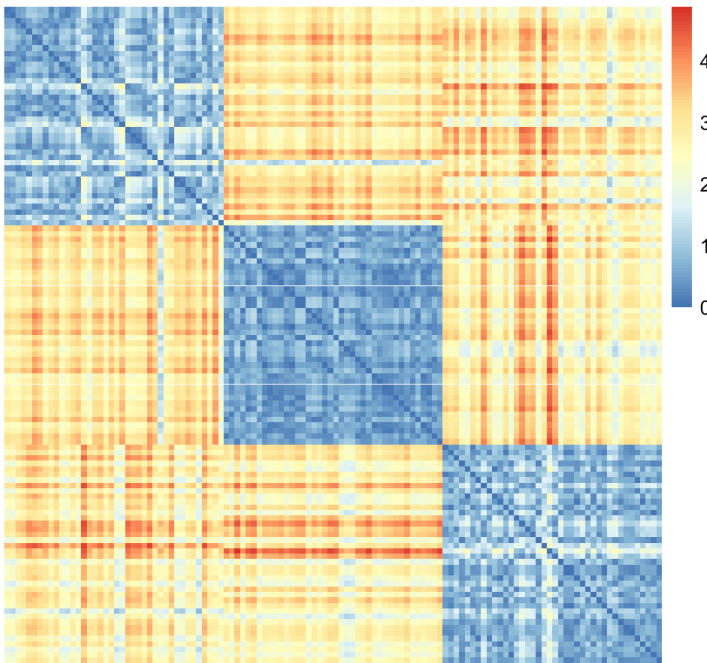


distances

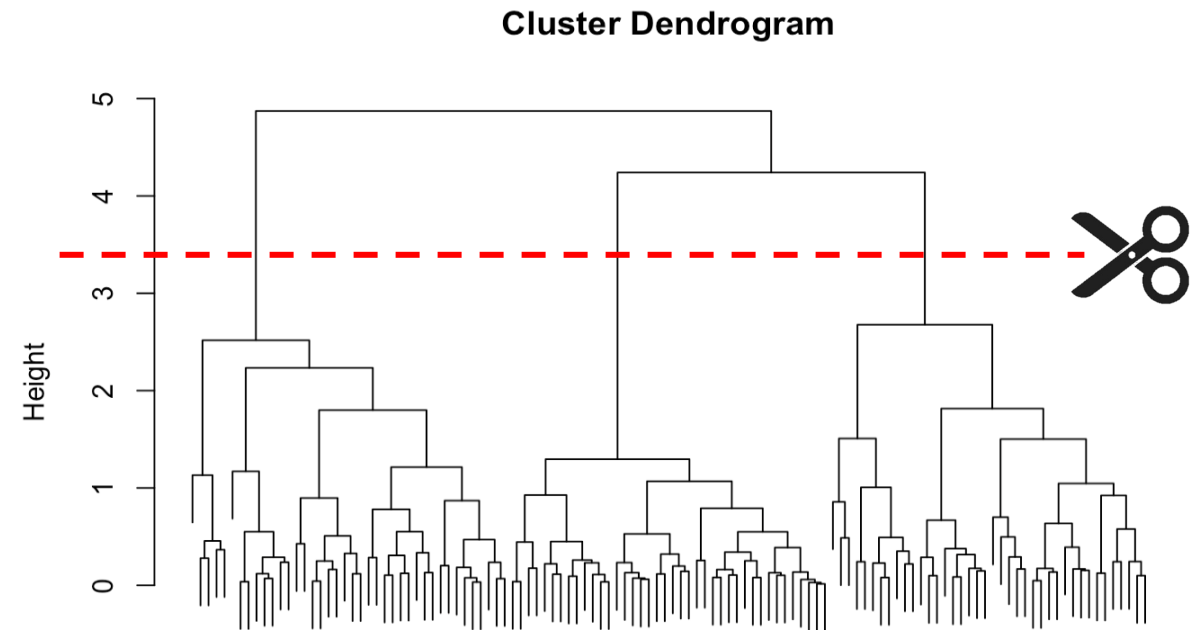


# Clustering

## Hierarchical clustering



Make tree





Thank You :)



Please give us feedback!

