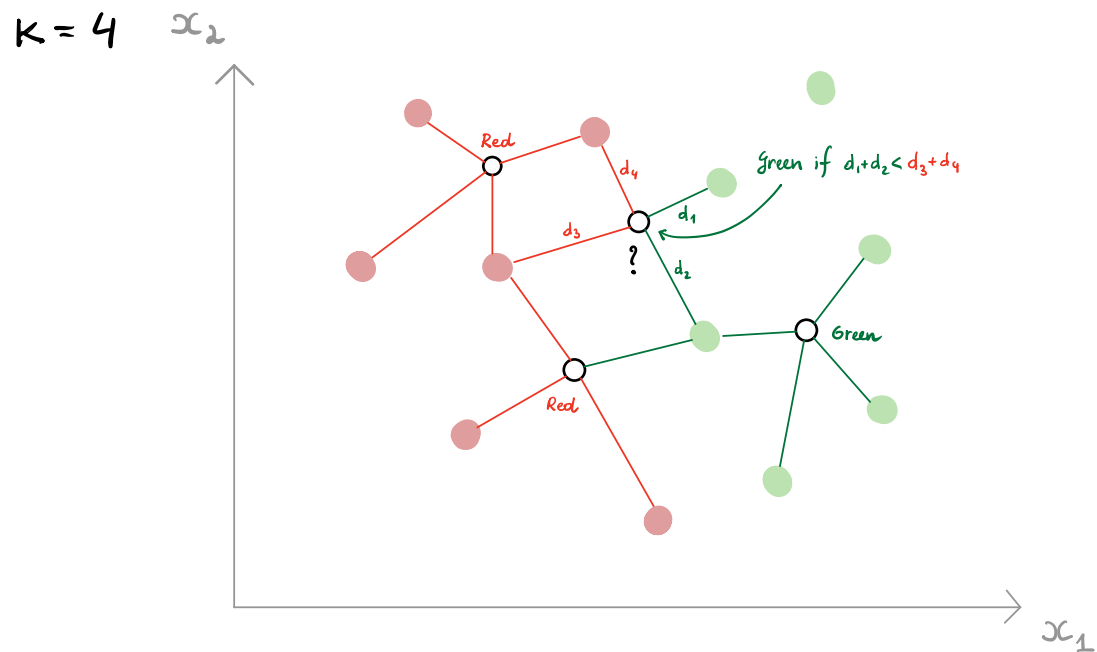
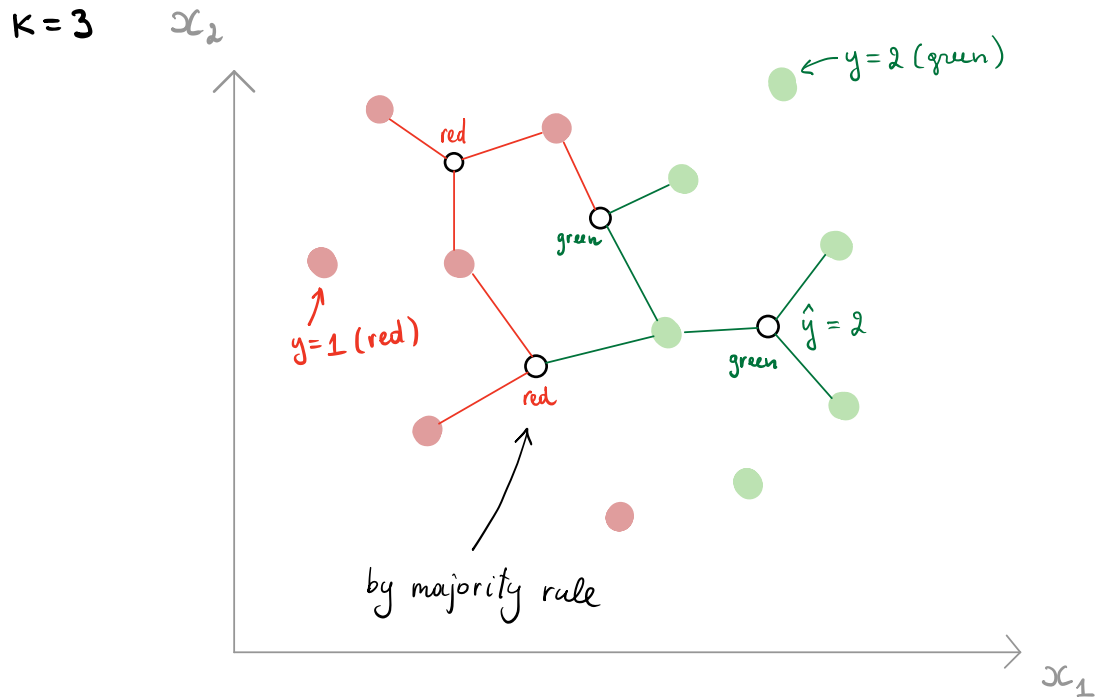
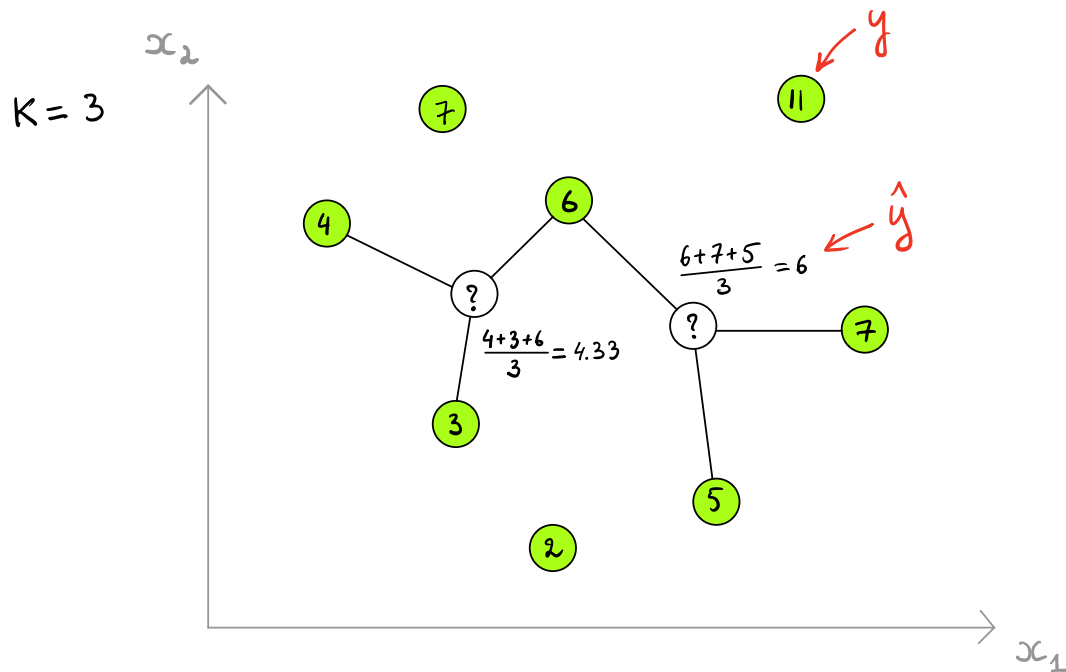


KNN classification

$$\begin{array}{ccc} i & \text{---} & j \\ (x_{i1}, x_{i2}) & & (x_{j1}, x_{j2}) \\ \text{distance } d_{ij} = \sqrt{(x_{i1} - x_{j1})^2 + (x_{i2} - x_{j2})^2} \end{array}$$



KNN Regression



Basic k-mean:

$$\hat{y} = \frac{1}{K} \sum_{i=1}^K y_i$$

Weighted k-mean:

$$\hat{y} = \frac{1}{K} \sum_{i=1}^K w_i y_i \quad \text{where} \quad w_i = \frac{\frac{1}{d_i}}{\sum_{j=1}^K \frac{1}{d_j}} \quad \sum_i w_i = 1$$

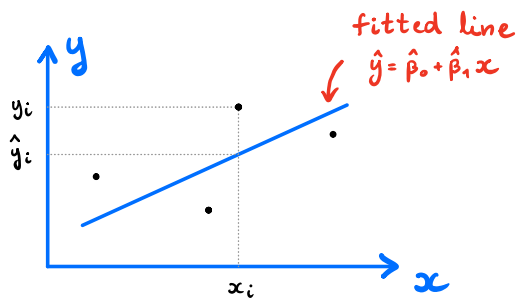
classification:

$[1, 2, 2]$ $[0.8, 0.2, 0.2]$

use `numpy.bincount()`

Linear regression:

- Single predictor $y_i = \beta_0 + \beta_1 x_i + \epsilon_i$
 - y_i : response
 - x_i : predictor
 - β_0, β_1 : regression coefficients
 - ϵ_i : random error



- Multiple predictors $y_i = \beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \dots + \epsilon_i$

Matrix form $\vec{y} = X \vec{\beta} + \vec{\epsilon}$

$\vec{y} = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix}$ (responses)
 $X = \begin{pmatrix} 1 & x_{11} & x_{12} & \dots \\ 1 & x_{21} & x_{22} & \dots \\ \vdots & \vdots & \vdots & \vdots \\ 1 & x_{n1} & x_{n2} & \dots \end{pmatrix}$ (design matrix)
 $\vec{\beta} = \begin{pmatrix} \beta_0 \\ \beta_1 \\ \vdots \end{pmatrix}$ (regression coefficients)
 $\vec{\epsilon} = \begin{pmatrix} \epsilon_1 \\ \epsilon_2 \\ \vdots \\ \epsilon_n \end{pmatrix}$ (errors)

Prediction $\hat{\vec{y}} = X^{(test)} \hat{\vec{\beta}}$

$\hat{\vec{y}} = \begin{pmatrix} \hat{y}_1 \\ \hat{y}_2 \\ \vdots \\ \hat{y}_m \end{pmatrix}$
 $X^{(test)} = \begin{pmatrix} 1 & x_{11} & x_{12} \\ 1 & x_{21} & x_{22} \\ \vdots & \vdots & \vdots \\ 1 & x_{m1} & x_{m2} \end{pmatrix}$
 $\hat{\vec{\beta}} = \left((X^{(train)})^T X^{(train)} \right)^{-1} (X^{(train)})^T \vec{y}$