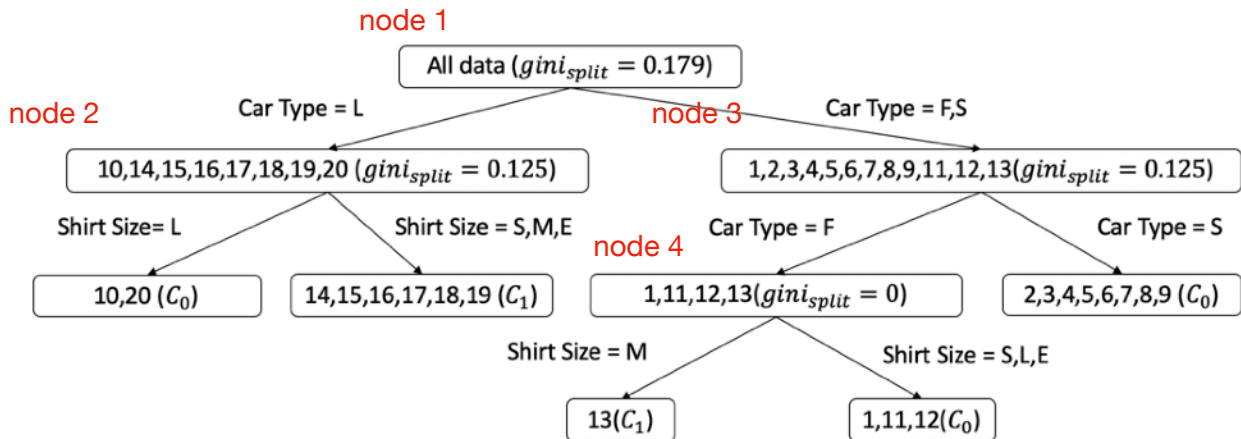


## Problem 1:

tree:



All split and its corresponding gini:

node 1:

Split with Gender (['F'], ['M']), gini: 0.39999999999999999

Split with Car Type (['Family'], ['Sport', 'Luxury']), gini: 0.46875

Split with Car Type (['Sport'], ['Family', 'Luxury']), gini: 0.26666666666666666

Split with Car Type (['Luxury'], ['Family', 'Sport']), gini: 0.17916666666666667

Split with Shirt Size (['Small'], ['Medium', 'Large', 'Extra Large']), gini: 0.48

Split with Shirt Size (['Medium'], ['Small', 'Large', 'Extra Large']), gini: 0.44835164835164837

Split with Shirt Size (['Large'], ['Small', 'Medium', 'Extra Large']), gini: 0.46875

Split with Shirt Size (['Extra Large'], ['Small', 'Medium', 'Large']), gini: 0.46875

Split with Shirt Size (['Small', 'Medium'], ['Large', 'Extra Large']), gini: 0.45

Split with Shirt Size (['Small', 'Large'], ['Medium', 'Extra Large']), gini: 0.47272727272727283

Split with Shirt Size (['Small', 'Extra Large'], ['Medium', 'Large']), gini: 0.47272727272727283

node 2:

Split with Gender (['F'], ['M']), gini: 0.21428571428571433

Split with Shirt Size (['Small'], ['Medium', 'Large', 'Extra Large']), gini: 0.20833333333333332

Split with Shirt Size (['Medium'], ['Small', 'Large', 'Extra Large']), gini: 0.19999999999999999

Split with Shirt Size (['Large'], ['Small', 'Medium', 'Extra Large']), gini: 0.125

Split with Shirt Size (['Extra Large'], ['Small', 'Medium', 'Large']), gini: 0.21428571428571433

Split with Shirt Size (['Small', 'Medium'], ['Large', 'Extra Large']), gini: 0.16666666666666666

Split with Shirt Size (['Small', 'Large'], ['Medium', 'Extra Large']), gini: 0.1875

Split with Shirt Size (['Small', 'Extra Large'], ['Medium', 'Large']), gini: 0.19999999999999999

node 3:

Split with Gender (['F'], ['M']), gini: 0.14814814814814817

Split with Car Type (['Family'], ['Sport', 'Luxury']), gini: 0.125

Split with Car Type (['Sport'], ['Family', 'Luxury']), gini: 0.125

Split with Shirt Size (['Small'], ['Medium', 'Large', 'Extra Large']), gini: 0.14814814814814817

Split with Shirt Size (['Medium'], ['Small', 'Large', 'Extra Large']), gini: 0.125

Split with Shirt Size (['Large'], ['Small', 'Medium', 'Extra Large']), gini: 0.14999999999999999

Split with Shirt Size (['Extra Large'], ['Small', 'Medium', 'Large']), gini: 0.14814814814814817

Split with Shirt Size (['Small', 'Medium'], ['Large', 'Extra Large']), gini: 0.1428571428571429

Split with Shirt Size (['Small', 'Large'], ['Medium', 'Extra Large']), gini: 0.1428571428571429

Split with Shirt Size (['Small', 'Extra Large'], ['Medium', 'Large']), gini: 0.13888888888888884

node 4:

Split with Shirt Size (['Small'], ['Medium', 'Large', 'Extra Large']), gini: 0.33333333333333337

Split with Shirt Size (['Medium'], ['Small', 'Large', 'Extra Large']), gini: 0.0

Split with Shirt Size (['Large'], ['Small', 'Medium', 'Extra Large']), gini: 0.33333333333333337

Split with Shirt Size (['Extra Large'], ['Small', 'Medium', 'Large']), gini: 0.33333333333333337

Split with Shirt Size (['Small', 'Medium'], ['Large', 'Extra Large']), gini: 0.25

Split with Shirt Size (['Small', 'Large'], ['Medium', 'Extra Large']), gini: 0.25

Split with Shirt Size (['Small', 'Extra Large'], ['Medium', 'Large']), gini: 0.25

Problem 2:

$$\begin{aligned}
 P(c = 0 | G = F, C = L, S = L) &= \frac{P(G = F, C = L, S = L | c = 0)P(c = 0)}{P(G = F, C = L, S = L)} \\
 &= \frac{P(G = F | c = 0)P(C = L | c = 0)P(S = L | c = 0)P(c = 0)}{P(G = F, C = L, S = L)} \\
 &= \frac{\frac{4}{12} \frac{1}{12} \frac{3}{12} \frac{6}{10}}{\frac{1}{10}} = \frac{72}{12^3}
 \end{aligned}$$

$$\begin{aligned}
 P(c = 1 | G = F, C = L, S = L) &= \frac{P(G = F, C = L, S = L | c = 1)P(c = 1)}{P(G = F, C = L, S = L)} \\
 &= \frac{P(G = F | c = 1)P(C = L | c = 1)P(S = L | c = 1)P(c = 1)}{P(G = F, C = L, S = L)} \\
 &= \frac{\frac{6}{8} \frac{7}{8} \frac{1}{8} \frac{4}{10}}{\frac{1}{10}} = \frac{168}{8^3}
 \end{aligned}$$

→ the prediction of Naive Bayes classifier is 1

Problem 3:

$$\text{minimize objective } \frac{|w|^2}{2}$$

subject to:

$$\begin{aligned} (w^T \cdot [4,3]^T + b) - 1 &> 0 & (w^T \cdot [2,1]^T + b) + 1 &< 0 & (w^T \cdot [-1, -2]^T + b) + 1 &< 0 \\ (w^T \cdot [7,2]^T + b) - 1 &> 0 & (w^T \cdot [2, -1]^T + b) + 1 &< 0 \\ (w^T \cdot [4,8]^T + b) - 1 &> 0 & (w^T \cdot [-1,3]^T + b) + 1 &< 0 \end{aligned}$$

→

$$\begin{aligned} ([b, w]^T \cdot [-1, -4, -3]^T) &< -1 & ([b, w]^T \cdot [1, 2, 1]^T) &< -1 \\ ([b, w]^T \cdot [-1, -7, -2]^T) &< -1 & ([b, w]^T \cdot [1, 2, -1]^T) &< -1 \\ ([b, w]^T \cdot [-1, -4, -8]^T) &< -1 & ([b, w]^T \cdot [1, -1, 3]^T) &< -1 \\ ([b, w]^T \cdot [1, -1, -2]^T) &< -1 \end{aligned}$$

$$\text{Let } P = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, q = 0 \text{ and } h = -1$$

Use python library “qpsovers” to solve the quadratic programming problem below.

$$\begin{aligned} \text{minimize}_x \quad & \frac{1}{2} x^T P x + q^T x \\ \text{subject to} \quad & Gx \leq h \\ & Ax = b \\ & lb \leq x \leq ub \end{aligned}$$

we can get the hyperplane:

$$w = \begin{bmatrix} 0.5 \\ 0.5 \end{bmatrix}$$

$$b = -2.5$$

$$f(x) = w^T x + b = \begin{bmatrix} 0.5 \\ 0.5 \end{bmatrix}^T \cdot x - 2.5$$

by calculating  $f(x)$

$$f\left(\begin{bmatrix} 4 \\ 3 \end{bmatrix}\right) = 1, f\left(\begin{bmatrix} 7 \\ 2 \end{bmatrix}\right) = 2, f\left(\begin{bmatrix} 4 \\ 8 \end{bmatrix}\right) = 3.5, f\left(\begin{bmatrix} 2 \\ 1 \end{bmatrix}\right) = -1, f\left(\begin{bmatrix} 2 \\ -1 \end{bmatrix}\right) = -2, f\left(\begin{bmatrix} -1 \\ 3 \end{bmatrix}\right) = -1, f\left(\begin{bmatrix} -1 \\ -2 \end{bmatrix}\right) = -4$$

we know two support vectors are:

$$\begin{bmatrix} 4 \\ 3 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$