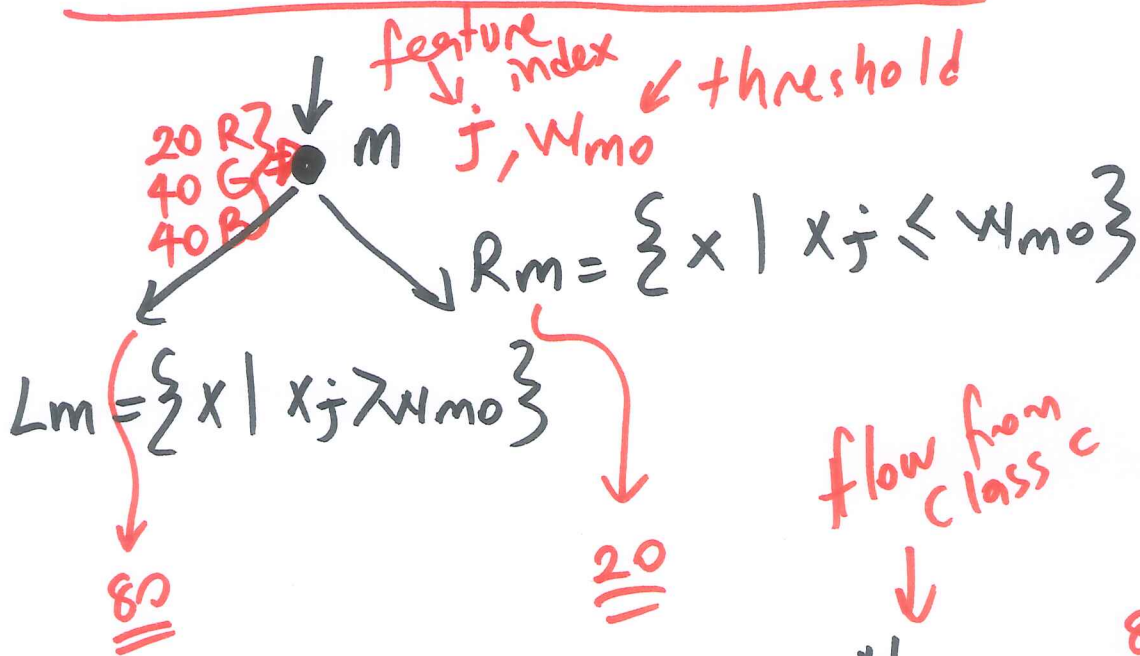


Univariate Decision Trees:

$N_m = \#$ of data points that reach node m



$$P_{mc} = \hat{P}(y=c \mid \mathcal{X}, m) = \frac{N_{mc}}{N_m}$$

$$P_{mR} = \frac{20}{100}$$

$$P_{mG} = \frac{40}{100}$$

$$P_{mB} = \frac{40}{100}$$

flow from class c

$$\left. \begin{matrix} 20 N_{m1} \\ 40 N_{m2} \\ \vdots \\ 40 N_{mk} \end{matrix} \right\}$$

$N_{m\epsilon} = \#$ of data points that reach node m from class c

$$\left. \begin{matrix} 80 N_{m1} \\ 20 N_{m2} \\ \vdots \\ N_{ms} \end{matrix} \right\}$$

$N_{ms} = \#$ of data points that reach node m and take split s

total flow

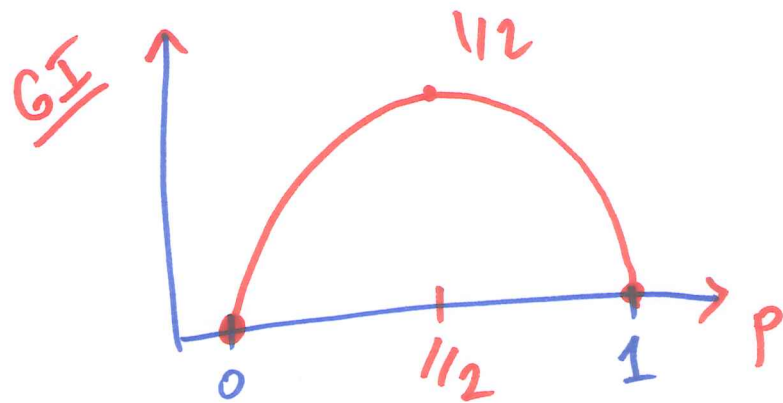
$$I_m = - \sum_{c=1}^K p_{mc} \cdot \log_2(p_{mc}) \quad \left. \vphantom{\sum_{c=1}^K} \right\} \text{impurity of node } m.$$

$$I_m = - \sum_{s=1}^S \left[\underbrace{\frac{N_{ms}}{N_m}}_{\text{weight of child node } L_m, R_m} \left[\sum_{c=1}^K p_{m\epsilon c} \cdot \log_2(p_{m\epsilon c}) \right] \right] \quad \left. \vphantom{\sum_{s=1}^S} \right\} \text{impurity of the split at node } m.$$

$$\hat{I}_m' = \frac{80}{100} \cdot \hat{I}_m(L_m) + \frac{20}{100} \cdot \hat{I}_m(R_m)$$

Entropy: $-p \cdot \log_2(p) - (1-p) \cdot \log_2(1-p)$ N_+ N_-

Gini Index: $2 \cdot p(1-p)$



p : ratio of positive data points
 $1-p$: ratio of negative data points

$$p = \frac{N_+}{N_+ + N_-}$$

p_1 0.3 p_2 0.6 p_3 0.1

Misclassification Error:

$1 - \max(p, 1-p)$ or $\min(p, 1-p)$
 $1 - \max(p_1, p_2, \dots, p_K)$

Regression Trees:

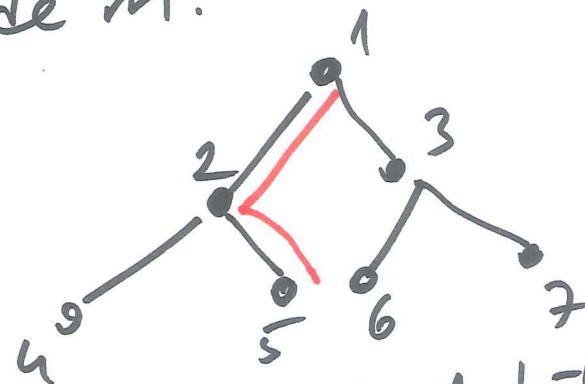
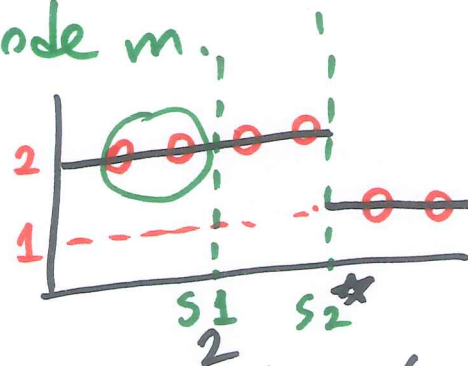
$$b_m(x) = \begin{cases} 1 & \text{if } x \in X_m: x \text{ reaches node } m. \\ 0 & \text{otherwise.} \end{cases}$$

$$E_m = \frac{1}{N_m} \sum_{i=1}^N (y_i - g_m)^2 b_m(x_i)$$

→ # of data points that reach to node m.

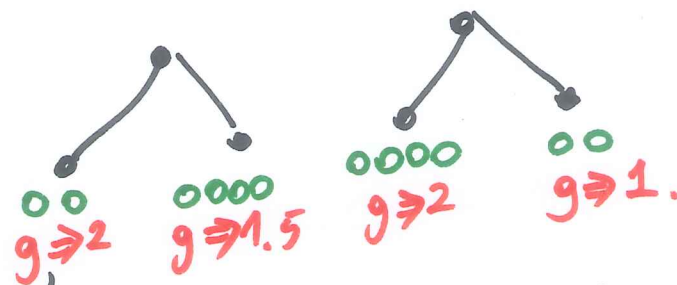
g_m : predicted value at node m.

$$g_m = \frac{\sum_{i=1}^N b_m(x_i) \cdot y_i}{\sum_{i=1}^N b_m(x_i)}$$



$$\begin{aligned} b_1(x) &= 1 & b_2(x) &= 1 & b_5(x) &= 1 \\ b_3(x) &= 0 & b_4(x) &= 0 & b_6(x) &= 0 \\ b_7(x) &= 0 \end{aligned}$$

$$N_m = \sum_{i=1}^N b_m(x_i)$$



$$E_m = \frac{1}{N_m} \sum_{s=1}^S \sum_{i=1}^N (y_i - g_{ms})^2 b_{ms}(x_i)$$

$$\begin{aligned} \bigcirc \nabla E_{S1} &= \frac{(2-2)^2 + (2-2)^2 + (2-1.5)^2 + (2-1.5)^2 + (1-1.5)^2 + (1-1.5)^2}{6} \\ \bigcirc = E_{S2} &= \frac{(2-2)^2 + (2-2)^2 + (2-2)^2 + (2-2)^2 + (1-1)^2 + (1-1)^2}{6} \end{aligned}$$

Multivariate Trees:

$$f_m(x): x_j > w_{m0} \Leftarrow \text{univariate split}$$

$$f_m(x): w_m^T \cdot x + w_{m0} > 0 \Leftarrow \text{multivariate split.}$$

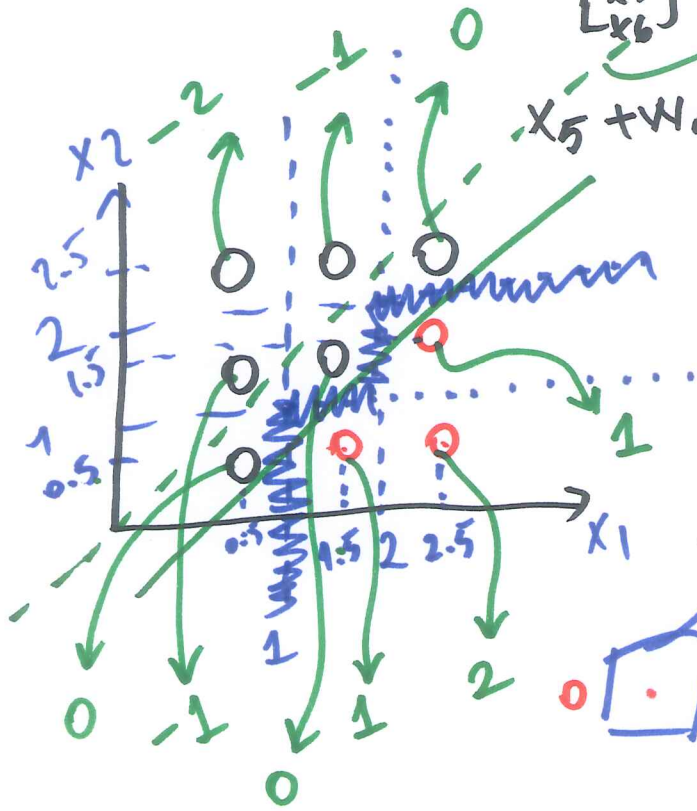
$$\begin{bmatrix} 0 & 0 & 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \end{bmatrix}$$

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

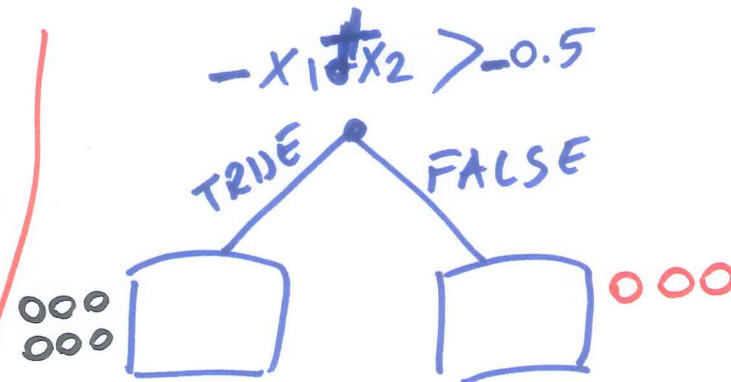
$$x_1 + x_2 + x_3 + x_4 + x_5 - 2x_6 + w_{m0} > 0$$

$$x_1 - x_6 = 0$$

$$x_5 + w_{m0} > 0 \Rightarrow x_5 > -w_{m0}$$



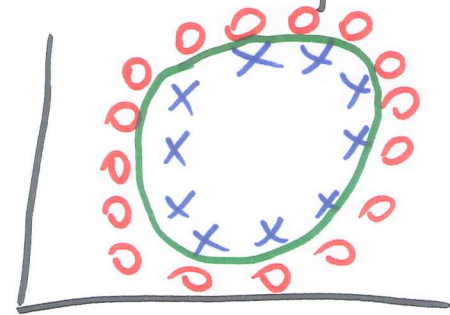
9 nodes
4 decision / 5 terminal nodes



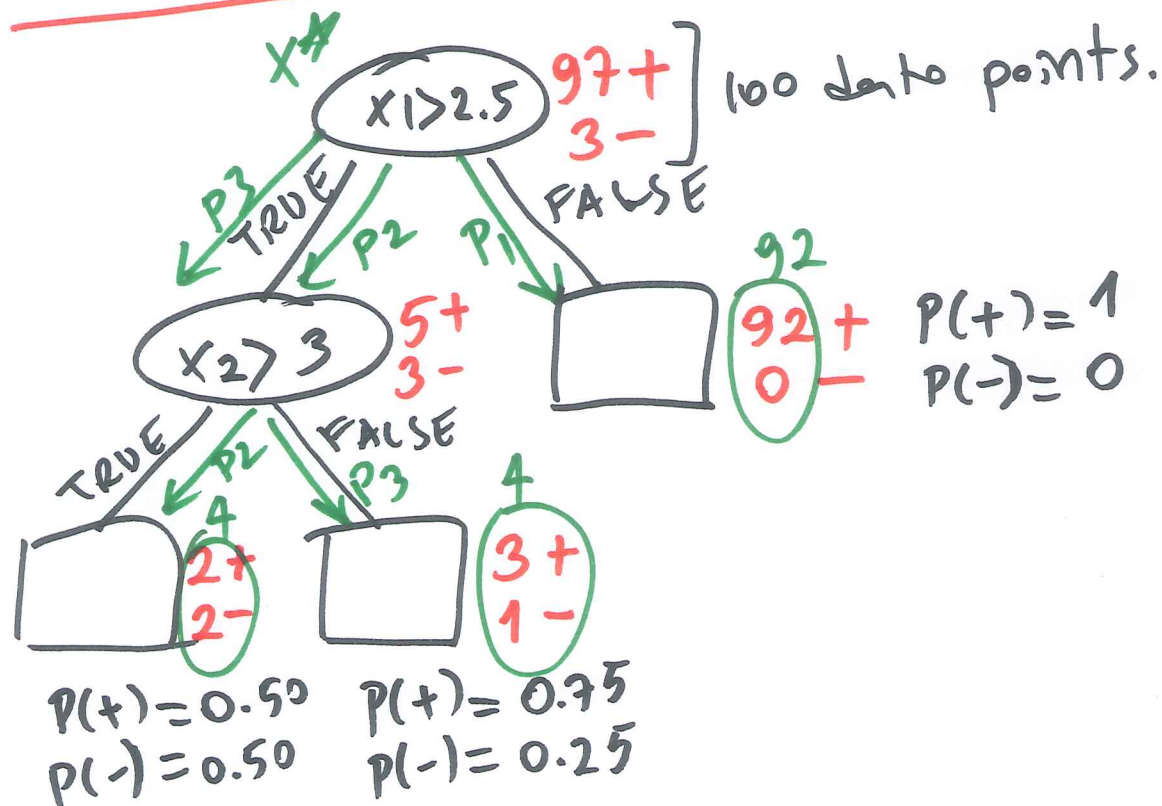
3 nodes
1 decision / 2 terminal nodes.

$$f_m(x) \Rightarrow w_m^T \cdot x + w_{m0} > 0$$

$$f_m(x) \Rightarrow \underbrace{x^T \cdot W_m \cdot x}_{[D \times D]} + \underbrace{w_m^T \cdot x}_{D \times 1} + \underbrace{w_{m0}}_{1 \times 1} > 0 \Rightarrow \text{non linear / multivariable split.}$$



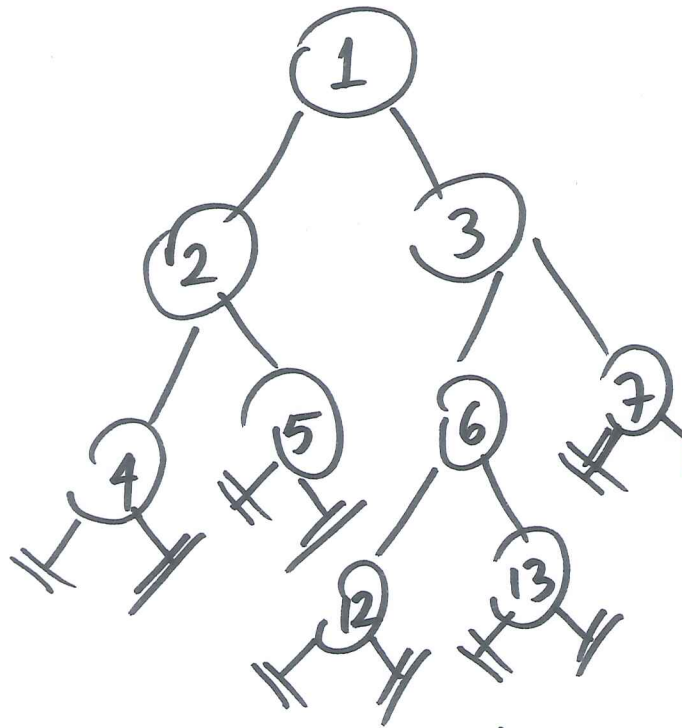
Rule Extraction:



P1: $x_1 \leq 2.5$

P2: $x_1 > 2.5 \wedge x_2 > 3$

P3: $x_1 > 2.5 \wedge x_2 \leq 3$



terminal.

left child $\rightarrow 2 * \text{parent}$
 right child $\rightarrow 2 * \text{parent} + 1$

