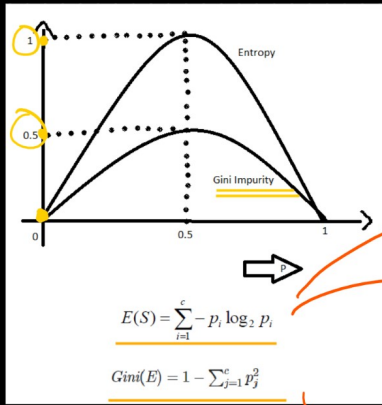


Gini vs Entropy

↳ Impurity



Highest Impure → 1

logarithmic → computationally expensive

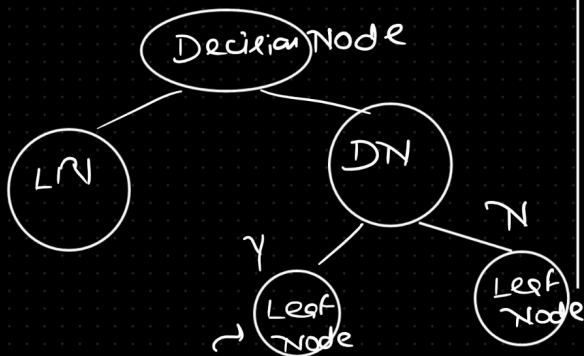
Highest Impure → 0.5

Default (Documentation)

Gini

Best case

Sample (Highest Pure)



$$\left\{ \begin{array}{l} P(Y) = 1 \\ P(N) = 0 \end{array} \right\} \rightarrow \underline{\underline{\text{certain}}}$$

$$1 - \sum_{i=1}^c p_i^2 = \underline{\underline{1 - 1 = 0}} \quad (\underline{\underline{\text{Minimum}}})$$

worst case

Sample (Highest Impure)

$$\left. \begin{array}{l} P(Y) = \frac{1}{2} \\ P(N) = \frac{1}{2} \end{array} \right\} \begin{array}{l} \text{Highly} \\ \text{Confused} \end{array}$$

$$\begin{aligned} 1 - \sum_{i=1}^c p_i^2 &= 1 - \left(\frac{1}{2}\right)^2 - \left(\frac{1}{2}\right)^2 \\ &= 1 - \frac{1}{4} - \frac{1}{4} \\ &= \underline{\underline{1 - \frac{1}{2} = 0.5}} \end{aligned}$$

Entropy

best
case

Highest Purity

$$P(Y) = 1$$

$$P(N) = 0$$

$$E = - \sum_{i=1}^n p_i \log p_i$$

$$= -1 \log 1$$

$$\Rightarrow 0$$

Highest Impure

$$P(Y) = \frac{1}{2}$$

$$P(N) = \frac{1}{2}$$

$$E = -\frac{1}{2} \log_2 \frac{1}{2} - \frac{1}{2} \log_2 \frac{1}{2}$$

$$E = 1$$

worst
case

Real time use-case

→ Gini Index → Slower training time

→ Entropy → balanced decision tree

MSE → ↓↓↓ } Decision Tree

Regressor

Leaf Node → Prediction of

the

Continuous

Value