

## Week 8 – Decision Tree Lab (Manual Calculations)

### Dataset Description

The dataset contains 10 records. The target variable is Accident (Yes/No). The input attributes are Weather Condition, Road Condition, Traffic Condition, and Engine Problem. There are 5 instances with Accident = Yes and 5 instances with Accident = No.

### 1. Entropy of the Dataset

Entropy is used to measure the uncertainty in the dataset.

Total instances = 10

Yes = 5, No = 5

$$p(\text{Yes}) = 5/10 = 0.5$$

$$p(\text{No}) = 5/10 = 0.5$$

$$\text{Entropy}(D) = -[0.5 \log_2(0.5) + 0.5 \log_2(0.5)] = 1.0$$

This means the dataset is perfectly balanced.

### 2. Information Gain Calculation

Information Gain measures how much an attribute reduces uncertainty.

$\text{Gain}(A) = \text{Entropy}(D) - \text{Weighted Entropy after split}$

#### Weather Condition

Clear: 4 instances (2 Yes, 2 No)  $\rightarrow$  Entropy = 1.000

Rain: 3 instances (1 Yes, 2 No)  $\rightarrow$  Entropy  $\approx$  0.918

Snow: 3 instances (2 Yes, 1 No)  $\rightarrow$  Entropy  $\approx$  0.918

Weighted Entropy = 0.951

Information Gain =  $1.000 - 0.951 = 0.049$

#### Road Condition

Bad: 4 instances (3 Yes, 1 No)  $\rightarrow$  Entropy  $\approx$  0.811

Good: 4 instances (1 Yes, 3 No)  $\rightarrow$  Entropy  $\approx$  0.811

Average: 2 instances (1 Yes, 1 No)  $\rightarrow$  Entropy = 1.000

Weighted Entropy = 0.849

Information Gain =  $1.000 - 0.849 = 0.151$

### **Traffic Condition**

High: 4 instances (3 Yes, 1 No)  $\rightarrow$  Entropy  $\approx 0.811$

Light: 3 instances (1 Yes, 2 No)  $\rightarrow$  Entropy  $\approx 0.918$

Normal: 3 instances (1 Yes, 2 No)  $\rightarrow$  Entropy  $\approx 0.918$

Weighted Entropy = 0.875

Information Gain =  $1.000 - 0.875 = 0.125$

### **Engine Problem**

Yes: 4 instances (3 Yes, 1 No)  $\rightarrow$  Entropy  $\approx 0.811$

No: 6 instances (2 Yes, 4 No)  $\rightarrow$  Entropy  $\approx 0.918$

Weighted Entropy = 0.875

Information Gain =  $1.000 - 0.875 = 0.125$

Based on Information Gain, Road Condition has the highest value and is selected as the root node.

### **3. Gini Impurity**

Gini Impurity measures the probability of incorrect classification.

$$\text{Gini}(D) = 1 - (0.5^2 + 0.5^2) = 0.5$$

Weighted Gini values after splitting:

Weather = 0.467

Road = 0.400

Traffic = 0.417

Engine = 0.417

The lowest Gini value is for Road Condition, confirming it as the best split.

#### **4. Final Decision Tree**

The final decision tree uses Road Condition as the root node. Traffic Condition is used for the next level of splitting, resulting in clear decision outcomes for predicting whether an accident will occur.