# XGBoost: A Beginner's Tutorial EV Battery Life Estimation Example

#### What is XGBoost?

- Gradient Boosting = Sequentially improve weak learners
- XGBoost = Efficient, regularized implementation
- Combines prediction trees using gradient descent principles
- Excellent for tabular data (e.g. EV sensor data)

### Example: EV Battery Life Estimation

#### Features:

- $\triangleright$   $X_1$ : Charge cycles
- $\triangleright$   $X_2$ : Average temperature (°C)
- $\triangleright$   $X_3$ : Battery age (years)

#### Data:

Sample	$X_1$	$X_2$	$X_3$
1	500	35	2
2	600	30	3
3	700	40	4

**Target:** y = [85, 80, 75]

# Step 1: Initial Prediction

$$\hat{y}^{(0)} = \frac{1}{3}(85 + 80 + 75) = 80$$

$$\hat{y}^{(0)} = \begin{bmatrix} 80\\80\\80 \end{bmatrix}$$

## Step 2: Compute Gradients and Hessians

#### **Squared Error Loss:**

$$L = (y - \hat{y})^2$$

**Gradient:** 

$$g_i = \hat{y}_i - y_i \quad \Rightarrow \quad \begin{bmatrix} -5\\0\\5 \end{bmatrix}$$

Hessian:

$$h_i = 1 \quad \Rightarrow \quad \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

## Step 3: Try Tree Split

#### Split on $X_1 < 600$ :

- ▶ Left Node: Sample  $1 \rightarrow g = -5$ , h = 1
- ▶ Right Node: Samples 2,  $3 \rightarrow g = 5$ , h = 2

$$\mathsf{Gain} = \frac{1}{2} \left[ \frac{(-5)^2}{1+1} + \frac{5^2}{2+1} - \frac{(0)^2}{3+1} \right] = 10.42$$

## Step 4: Compute Leaf Weights

$$w_j = -\frac{\sum g_j}{\sum h_j + \lambda}$$

▶ Left Leaf:  $w = \frac{5}{2} = 2.5$ 

▶ Right Leaf:  $w = \frac{-5}{3} \approx -1.67$ 

Learning rate:  $\eta = 0.1$ 

$$\hat{y}^{(1)} = \hat{y}^{(0)} + \eta \cdot f_1(x) \Rightarrow \begin{bmatrix} 80.25 \\ 79.83 \\ 79.83 \end{bmatrix}$$

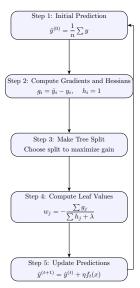
## Summary of Step 1

Sample	True y	Init $\hat{y}$	Update	New ŷ
1	85	80	+0.25	80.25
2	80	80	-0.167	79.83
3	75	80	-0.167	79.83

Each tree improves on previous predictions using gradient-based splits.

#### XGBoost Flowchart

#### XGBoost Flowchart: EV Battery Life Estimation



## Key Takeaways

- XGBoost builds trees sequentially using gradient descent.
- ▶ Uses Taylor expansion: gradient + Hessian.
- Learns leaf values analytically from gradients.
- Extremely effective on tabular prediction tasks.