Gradient Boosting for Classification

Sunday, January 26, 2025 4:16 PM

Steps:

- 1. Initialize $f_0(x) = \arg\min_{\gamma} \sum_{i=1}^{N} L(y, \gamma)$
- 2. For m = 1 to M

a. For
$$i=1,2,3,\ldots,N$$
 Compute $r_{im}=\left[\frac{\delta L(y,\gamma)}{\delta f(x_i)}\right]$

- b. Fit the regression tree to the targets r_{im} giving terminal regions R_{jm} , $j=1,2,3,...,J_m$
- c. For $j = 1, 2, 3, ..., J_m$ Compute $\gamma_{jm} = \arg\min_{\gamma} \sum_{x_i \in R_{jm}} L(y_i, f_{m-1}(x_i) + \gamma)$
- d. Update $f_m(x) = f_{m-1}(x) + \sum_{j=1}^{J_m} \gamma_{jm} I(x \in R_{jm})$
- 3. Update $\hat{f}(x) = f_M(x)$

| CGPA | IQ | Placed |
|------|-----|--------|
| 6.82 | 118 | 0 |
| 6.36 | 125 | 1 |
| 5.39 | 99 | 1 |
| 5.50 | 106 | 1 |
| 6.39 | 148 | 0 |
| 9.13 | 148 | 1 |
| 7.17 | 147 | 1 |
| 7.72 | 72 | 0 |

Step 1:

1. Calculate $\log_e odds$

$$\log_e odds = \log_e \frac{5}{3} = \frac{\log \frac{5}{3}}{\log e} = \frac{\log 1.6667}{1} = 0.51$$

2. Calculate the probability
$$\frac{1}{1 + e^{-\log odds}} = \frac{1}{1 + e^{-0.51}} = 0.625$$

| CGPA | IQ | Placed | Log(odds) 1 | Probability 1 |
|------|-----|--------|-------------|---------------|
| 6.82 | 118 | 0 | 0.51 | 0.625 |
| 6.36 | 125 | 1 | 0.51 | 0.625 |
| 5.39 | 99 | 1 | 0.51 | 0.625 |
| 5.50 | 106 | 1 | 0.51 | 0.625 |
| 6.39 | 148 | 0 | 0.51 | 0.625 |
| 9.13 | 148 | 1 | 0.51 | 0.625 |

| 7.17 | 147 | 1 | 0.51 | 0.625 |
|------|-----|---|------|-------|
| 7.72 | 72 | 0 | 0.51 | 0.625 |

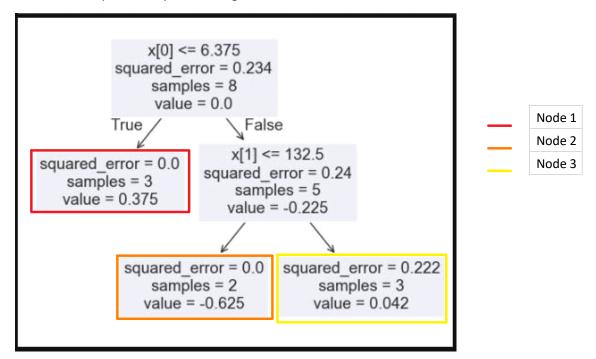
Step 2:

Calculate pseudo-residuals (Target - Probability)

| CGPA | IQ | Placed | Log(odds) 1 | Probability 1 | Residuals 1 |
|------|-----|--------|-------------|---------------|-------------|
| 6.82 | 118 | 0 | 0.51 | 0.625 | -0.625 |
| 6.36 | 125 | 1 | 0.51 | 0.625 | 0.375 |
| 5.39 | 99 | 1 | 0.51 | 0.625 | 0.375 |
| 5.50 | 106 | 1 | 0.51 | 0.625 | 0.375 |
| 6.39 | 148 | 0 | 0.51 | 0.625 | -0.625 |
| 9.13 | 148 | 1 | 0.51 | 0.625 | 0.375 |
| 7.17 | 147 | 1 | 0.51 | 0.625 | 0.375 |
| 7.72 | 72 | 0 | 0.51 | 0.625 | -0.625 |

Step 3:

Train a Decision Tree (it has to be a weak regression learner since the outputs are continuous) with features as the previous inputs and targets as residuals



Step 4:

| CGPA | IQ | Placed | Log(odds) 1 | Probability 1 | Residuals 1 | Leaf entry 1 |
|------|-----|--------|-------------|---------------|-------------|--------------|
| 6.82 | 118 | 0 | 0.51 | 0.625 | -0.625 | 2 |
| 6.36 | 125 | 1 | 0.51 | 0.625 | 0.375 | 1 |
| 5.39 | 99 | 1 | 0.51 | 0.625 | 0.375 | 1 |
| 5.50 | 106 | 1 | 0.51 | 0.625 | 0.375 | 1 |
| 6.39 | 148 | 0 | 0.51 | 0.625 | -0.625 | 3 |
| 9.13 | 148 | 1 | 0.51 | 0.625 | 0.375 | 3 |
| 7.17 | 147 | 1 | 0.51 | 0.625 | 0.375 | 3 |
| 7.72 | 72 | 0 | 0.51 | 0.625 | -0.625 | 2 |

$$\log odds \ (Node \ 1) = \frac{0.375 + 0.375 + 0.375}{0.625 * 0.375 + 0.625 * 0.375 + 0.625 * 0.375} = \frac{3 * 0.375}{3 * 0.625 * 0.375} = 1.6$$

$$\log odds \ (Node \ 2) = \frac{-0.625 - 0.625}{0.625 * 0.375 + 0.625 * 0.375} = \frac{-2 * 0.625}{2 * 0.625 * 0.375} = -2.667$$

$$\log odds \ (Node \ 3) = \frac{-0.625 + 0.375 + 0.375}{0.625 * 0.375 + 0.625 * 0.375} = \frac{0.125}{3 * 0.625 * 0.375} = 0.178$$

Step 5:

Calculate $\log odds = \log odds \ 1 + \log odds \ (respective \ node)$ Calculate Probability

| CGPA | IQ | Placed | Log(odds) 1 | Probability 1 | Residuals 1 | Leaf entry 1 | Log(odds) 2 | Probability 2 |
|------|-----|--------|-------------|---------------|-------------|--------------|-------------|---------------|
| 6.82 | 118 | 0 | 0.51 | 0.625 | -0.625 | 2 | -2.175 | 0.102 |
| 6.36 | 125 | 1 | 0.51 | 0.625 | 0.375 | 1 | 2.11 | 0.8919 |
| 5.39 | 99 | 1 | 0.51 | 0.625 | 0.375 | 1 | 2.11 | 0.8919 |
| 5.50 | 106 | 1 | 0.51 | 0.625 | 0.375 | 1 | 2.11 | 0.8919 |
| 6.39 | 148 | 0 | 0.51 | 0.625 | -0.625 | 3 | 0.688 | 0.6655 |
| 9.13 | 148 | 1 | 0.51 | 0.625 | 0.375 | 3 | 0.688 | 0.6655 |
| 7.17 | 147 | 1 | 0.51 | 0.625 | 0.375 | 3 | 0.688 | 0.6655 |
| 7.72 | 72 | 0 | 0.51 | 0.625 | -0.625 | 2 | -2.175 | 0.102 |

Step 6:

Calculate residuals

| CGPA | IQ | Placed | Log(odds) 1 | Probability 1 | Residuals 1 | Leaf entry 1 | Log(odds) 2 | Probability 2 | Residuals 2 |
|------|-----|--------|-------------|---------------|-------------|--------------|-------------|---------------|-------------|
| 6.82 | 118 | 0 | 0.51 | 0.625 | -0.625 | 2 | -2.175 | 0.102 | -0.102 |
| 6.36 | 125 | 1 | 0.51 | 0.625 | 0.375 | 1 | 2.11 | 0.8919 | 0.1081 |
| 5.39 | 99 | 1 | 0.51 | 0.625 | 0.375 | 1 | 2.11 | 0.8919 | 0.1081 |
| 5.50 | 106 | 1 | 0.51 | 0.625 | 0.375 | 1 | 2.11 | 0.8919 | 0.1081 |
| 6.39 | 148 | 0 | 0.51 | 0.625 | -0.625 | 3 | 0.688 | 0.6655 | -0.6655 |
| 9.13 | 148 | 1 | 0.51 | 0.625 | 0.375 | 3 | 0.688 | 0.6655 | 0.3345 |
| 7.17 | 147 | 1 | 0.51 | 0.625 | 0.375 | 3 | 0.688 | 0.6655 | 0.3345 |
| 7.72 | 72 | 0 | 0.51 | 0.625 | -0.625 | 2 | -2.175 | 0.102 | -0.102 |

Step 7:

Repeat the steps until the residuals are zero