Regression Loss Functions:

Mean Squared Error Loss:

Iteration: 1

MSE =
$$\frac{1}{4} \times \left[(3-2)^2 + (6-5)^2 + (9-8)^2 + (12-11)^2 \right]$$
.

Iteration: 2

Predicted values closer to adual.

Actual values -> y = [3,6,9,12].

predicted values: [2.5,5.5,8.5,11.5]

MSE = 1 x \(\frac{1}{y} - \frac{1}{y} \).

MSE = 1 x [(3-2.5)2+(6-5.5)2+(9-8.5)2+(12-11.5)2]

MSE = + x [(0.5) + (0.5) + (0.5) + (0.5)

MSE = 1 x [0.25 + 0.25 + 0.25 + 0.25]

MSE = 1

MSE = 0.25

lean Squared Logarithmic Error Loss: Actual value -> y=[3,6,9,12] Predicted = [2,5,8,11] MSLE = I x \(\int \left[ln \left(actual + I) - ln \left(y \cdot predict + i) \right]^2

MSLE =
$$\frac{1}{4} \times \left[\left(2n(3+1) - 2n(2+1) \right)^{2} + \left(2n(6+1) - 2n(8+1) \right)^{2} + \left(2n(6+1) - 2n(8+1) \right)^{2} + \left(2n(6+1) - 2n(6) \right)^{2} + \left(2n(6+1) - 2n(6) \right)^{2} + \left(2n(6-2n) \right)^{2} + \left(2n(6-2$$

Iteration: 2

Predicted values closer to actual.

Actual values > y = [3,6,9,12].

Predicted values = [2.5,5.5,8.5,11.5].

MSLE =
$$\frac{1}{4} \times \left[\left(\ln \left(\frac{1}{3} + 1 \right) - \ln \left(\frac{1}{3} \cdot \operatorname{predict} + 1 \right) \right]^{\frac{1}{4}} \right]$$

MSLE = $\frac{1}{4} \times \left[\left(\ln \left(\frac{3}{3} + 1 \right) - \ln \left(\frac{2}{3} + 1 \right) \right]^{\frac{1}{4}} \right]$

($\ln \left(\frac{6}{4} + 1 \right) - \ln \left(\frac{3}{3} + 1 \right) \right]^{\frac{1}{4}} + \left(\ln \left(\frac{1}{4} + 1 \right) - \ln \left(\frac{8}{3} + 1 \right) \right)^{\frac{1}{4}} + \left(\ln \left(\frac{1}{3} + 1 \right) - \ln \left(\frac{1}{3} + 1 \right) \right)^{\frac{1}{4}} \right]$

MSLE = $\frac{1}{4} \times \left[\left(\frac{1}{3} \cdot \frac{3}{3} \cdot \frac{1}{4} + \left(\frac{1}{3} \cdot \frac{3}{3} \cdot \frac{1}{3} + \frac{1}{3} \cdot \frac{3}{3} \cdot \frac{1}{3} + \left(\frac{1}{3} \cdot \frac{3}{3} \cdot \frac{1}{3} + \frac{1}{3} \cdot \frac{3}{3} \cdot \frac{1}{3} + \frac{1}{3} \cdot \frac{3}{3} + \left(\frac{1}{3} \cdot \frac{3}{3} \cdot \frac{1}{3} + \frac{1}{3} \cdot \frac{3}{3} + \frac{1}{3} \cdot \frac{3$

(c) Mean Absolute Error Loss:

Iteration: 1

$$MAE = \frac{1}{4} \times \left[(3-2) + (6-5) + (9-8) + (12-1) \right]$$

$$MAE = \frac{1}{4} \times (2+1+1+1)$$

Iteration: 2

Predicted values closes to adual

MAE =
$$\frac{1}{4} \times \left[(3-2.5) + (6-5.5) + (9-8.5) + (12-11.5) \right]$$

MAE = $\frac{1}{4} \times \left[0.5 + 0.5 + 0.5 + 0.5 \right]$.

MAE = $\frac{2}{4}$

[MAE = 0.5]

Binary Classification loss functions:

Binary Cross-Entropy:

Heration: 1

BCE = $-2 \left[y \times \log (y \cdot p \text{ redict}) + (1-y) \times \log (1-y \cdot p \text{ redict}) \right]$

Actual values = $y = [1,0,0,1]$

Predicted values = $[1,0,1,0]$.

- $[1 \times \log(1) + (1-1) \times \log(1-1)] + [0 \times \log(1) + (1-0) \times \log(1-1)] + [0 \times \log(1) + (1-0) \times \log(1-1)] + [1 \times \log(1) + (1-0) \times \log(1-1)] + [1 \times \log(1) + (1-0) \times \log(1-1)] + [1 \times \log(1) + (1-1) \times \log(1-1)] + [1$

BCE=0

Iteration: 2 Predicted values closes to actual. Actual values -> y=[1,0,0,1]. Predicted values = [0.9,0.1,0.9,0.1]. BCE - 2[JxJog(J. predict) + (1-y) x Dog(1-y. predict)]. = - [(1 x Jug(0.9) + (1-1) x Jug (1-0.9)) + (0 x log(0.1) + (1-0) x log (1-0.1)) + (0×Jog(0.9) + (1-0) × Jog (1-0.9))+ (1×Jog(0.1) + (1-1) × Jog (1-0.9)) + 0 + 0 + (1 x -1) BCE = - [(1 x -0.045) + 0 = (-0.045-1) BCE = 1.0) (b) Hinge Loss: Iteration: 1 HL = max(on1-(y x y.predict)). Actual values 7 7 2 [19-191,-1] Predicted valles = [0.8, -0.2, 0.7, -0.3].

Iteration: 2 Predicted values closes to actual values. predicted values = [0.9, -0.9, 0.9, -0.9] HL = max (0,1- (y x y . predict))2 HL=[max(0,1-(1x0.9))2+max(0,1-(-1x-0-9))+ max(0,1-(-1x0-9))2+ max (0,1-(1+-0.9))2) HL = [max (0,1-0.9)2 + max (0,1-0.9)2 + max (0,1+0-9)2 + max (0,1+0-9)2) HL = [max (0,0.1)3 + max (0,0.1)3 + max (0,1.9)2 + max (0,1.9)2). HL = [max (0,0.01) + max (0,0.01) + max (0,3.61) + max (0,3.61)]. HL= max (0.01 + 0.01 + 3.61 + 3.61)

[HL = 7:24]

3). Multi-class Classification Loss functions: (9). Multi-class Cross Entropy Loss: Iteration: 1 MCE = - = (actual x log (predict)) Actual values -> 1 = [0,10,0]. Predicted values = [0.1,0.6,0.2,0.1] MCE = - (0 x Jog(0.1) + 1 x Jog(0.6) + 0 x Jog(0.2) + 0 x Dog (0.1)). MCE = - (0 x -1 + 1 x - 0.22 + 0x - 0.698 + MCE = - (0 - 0.22 +0+0) MCE = 0.22

Iteration: 2

Predicted values closes to actual

Actual value=> y = [0,1,0,0].

Predicted values=> y= [-1,0.9,-1,-1]

Iteration: 1

KL Divergence = E (actual * Jog (actual predicted)

Actual values > y = [3,6,9,12).

Predicted values = [2,5,8,11].

KLD= (3 x Jog (3/2)) + (6 x Jog (6/5)) + (9 x Jog (9/8)) + (12 x Jog (12/11))

KLD: (3 × 0.17) + (6 × 0.07) + (9 × 0.05)+
(12 × 0.03)

KLD: 0.51 + 0.42 + 0.45 + 0.36.

[KLD=1.79

Iteration: 2 Predicted closer to adual values Actual values > y = [3,6,9,12]. Predicted values = [2.5, 5-5, 8-5, 11.5]. KLD: 2 (actual x log (actual predicted)) KLD: (3 x Jog (3/2.5)) + (6x Jog (6/5.5))+ (9 x log (9/8.5)) + (12 x log (12/11.5)) KLD= (3 × 0-079) + (6 × 0-037) + (9 × 0-248)+ (12 x 0.184) KLD = (0-237 + 0.222 + 2.232 + 2.208)