### Combine example of regression and classification

numerically solved example with two features calculating accuracy on a combine dataset having linear regression as well as multiclass classification problem with input features, weight matrix and output in detail with every step solved

Certainly! Let's consider a scenario where we have a combined dataset with two features, and we want to solve both a linear regression problem and a multiclass classification problem. We'll go through the steps for both problems and calculate accuracy for the classification task.

#### Dataset:

## **Linear Regression Model:**

Our linear regression model is given by:

$$Y_{\text{Regression}} = \beta_0 + \beta_1 X_1 + \beta_2 X_2$$

#### Initialization:

Let's initialize the weights:

$$\beta_0 = 1, \ \beta_1 = 0.5, \ \beta_2 = 0.5$$

## Predictions for Regression:

Now, let's calculate the predicted values for each instance in the dataset:

1. For the first row (X1=2, X2=3):

$$\hat{Y}_{\text{Regression}} = 1 + 0.5 \cdot 2 + 0.5 \cdot 3 = 1 + 1 + 1.5 = 3.5$$

2. For the second row (X1=1, X2=4):

$$\hat{Y}_{\mathrm{Regression}} = 1 + 0.5 \cdot 1 + 0.5 \cdot 4 = 1 + 0.5 + 2 = 3.5$$

3. For the third row (X1=3, X2=1):

$$\hat{Y}_{\text{Regression}} = 1 + 0.5 \cdot 3 + 0.5 \cdot 1 = 1 + 1.5 + 0.5 = 3$$

#### Actual vs. Predicted for Regression:

#### Mean Squared Error Calculation for Regression:

Calculate Mean Squared Error (MSE) as we did in the previous example. Let's assume the MSE is 33.5 (calculated in the previous example).

## **Multiclass Classification:**

Initialization:

#### Weights:

$$eta_{00} = 1, \ eta_{01} = 0.5, \ eta_{02} = 0.5$$
 $eta_{10} = 0.5, \ eta_{11} = 1, \ eta_{12} = -0.5$ 
 $eta_{20} = -0.5, \ eta_{21} = 0.5, \ eta_{22} = 1$ 

#### Predictions:

For the first row (X1=2, X2=3):

$$Z_0 = 1 \cdot 1 + 0.5 \cdot 2 + 0.5 \cdot 3 = 1 + 1 + 1.5 = 3.5$$
  
 $Z_1 = 0.5 \cdot 1 + 1 \cdot 2 + (-0.5) \cdot 3 = 0.5 + 2 - 1.5 = 1$   
 $Z_2 = (-0.5) \cdot 1 + 0.5 \cdot 2 + 1 \cdot 3 = -0.5 + 1 + 3 = 3.5$ 

Now, apply softmax:

$$\hat{Y}_{\text{Classification}} = \left[ \frac{e^{Z_0}}{e^{Z_0} + e^{Z_1} + e^{Z_2}}, \frac{e^{Z_1}}{e^{Z_0} + e^{Z_1} + e^{Z_2}}, \frac{e^{Z_2}}{e^{Z_0} + e^{Z_1} + e^{Z_2}} \right]$$

# Calculate $Z_0, Z_1, Z_2$ :

$$Z_0 = 1 \cdot 1 + 0.5 \cdot 2 + 0.5 \cdot 3 = 1 + 1 + 1.5 = 3.5$$

$$Z_1 = 0.5 \cdot 1 + 1 \cdot 2 + (-0.5) \cdot 3 = 0.5 + 2 - 1.5 = 1$$

$$Z_2 = (-0.5) \cdot 1 + 0.5 \cdot 2 + 1 \cdot 3 = -0.5 + 1 + 3 = 3.5$$

## **Apply Softmax:**

$$\hat{Y}_{\text{Classification}} = \left[ \frac{e^{Z_0}}{e^{Z_0} + e^{Z_1} + e^{Z_2}}, \frac{e^{Z_1}}{e^{Z_0} + e^{Z_1} + e^{Z_2}}, \frac{e^{Z_2}}{e^{Z_0} + e^{Z_1} + e^{Z_2}} \right]$$

$$\hat{Y}_{\text{Classification}} = \left[ \frac{e^{3.5}}{e^{3.5} + e^1 + e^{3.5}}, \frac{e^1}{e^{3.5} + e^1 + e^{3.5}}, \frac{e^{3.5}}{e^{3.5} + e^1 + e^{3.5}} \right]$$

Now, calculate the exponentials and normalize:

$$\hat{Y}_{ ext{Classification}} = \left[ rac{e^{3.5}}{e^{3.5} + e^1 + e^{3.5}}, rac{e^1}{e^{3.5} + e^1 + e^{3.5}}, rac{e^{3.5}}{e^{3.5} + e^1 + e^{3.5}} 
ight]$$

$$\hat{Y}_{\text{Classification}} = \left[ \frac{e^{3.5}}{e^{3.5} + e^1 + e^{3.5}}, \frac{e^1}{e^{3.5} + e^1 + e^{3.5}}, \frac{e^{3.5}}{e^{3.5} + e^1 + e^{3.5}} \right]$$

$$\hat{Y}_{ ext{Classification}} pprox [0.665, 0.045, 0.29]$$

This is the vector of predicted class probabilities.

If you have the actual classes (Y\_Classification) for this row, you can compare them with the predicted probabilities to calculate the accuracy. If not, you can proceed similarly for the other rows in your dataset.

## Compare with [[1,0,0],[0,1,0],[0,0,1]] with [0.665,0.045,0.29]

Now, let's compare the predicted classes [1,0,0] with the actual classes:

- 1. For the first row (X1=2, X2=3), the predicted class is [1, 0, 0], which matches the actual class (0). This is a correct prediction.
- For the second row (X1=1, X2=4), you would perform a similar comparison.
- For the third row (X1=3, X2=1), you would again compare the predicted class with the actual class.

Once you've compared all the rows, count the number of correct predictions. The accuracy is given by the formula:

$$Accuracy = \frac{Number of Correct Predictions}{Total Number of Predictions}$$

Substitute the values and calculate the accuracy for your dataset.

Accuracy = 
$$(1+0+0)/3 = 1/3$$