

COMP/INDR 421/521 HW 02: Multiclass Linear Discrimination – Report

Data generation in this project were done by the exact same function `mvnrm` using the same mean and covariance matrixes that were used in the first homework.

We learned the linear discrimination rule using the softmax function for the multiclass classification problem.

We define **softmax** function which takes weight w and w_0 along with the data point x to predict corresponding y value.

$$y_i = \hat{P}(C_i|x) = \frac{\exp[w_i^t x + w_{i0}]}{\sum_{j=1}^K \exp[w_j^t x + w_{j0}]}, i = 1, 2 \dots K \text{ where } K = \text{class number}$$

In the **softmax(w,w0,x)** function we calculated softmax value for each class with the given data point and weights. Then the function returns column vector with these 3 values.

In order to find the correct weights for each attribute of each class type we initialize w and w_0 matrices using `runif` function around 0 and define `gradient_w` and `gradient_w0` functions for updating the weights.

`Gradient_w` function takes y , y_{pred} and X to return updated w matrix of which each column corresponds to weight of each class.

$$\begin{aligned} \text{gradient_wj} &= \text{stepSize} * \sum_t (r_j^t - y_j^t) x^t \\ \text{gradien}_{w_{j0}} &= \text{stepSize} * \sum_t (r_j^t - y_j^t) \end{aligned}$$

Similarly `gradient_w0` function takes y , y_{pred} to return updated w_0 matrix of which each column corresponds to `weight0` of each class.

You can see the resulting w and w_0 matrixes after the training data.

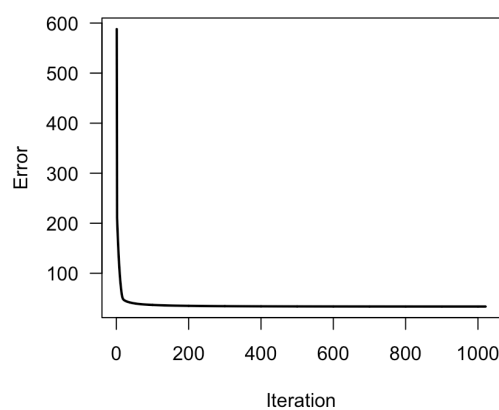
```
> print(w)
           w1          w2          w3
[1,] 0.2819587 -2.747212  2.219320
[2,] 1.2462069 -2.763820 -2.420236
> print(w0)
           [,1]      [,2]      [,3]
[1,] 7.653239 -3.036776 -1.168778
> |
```

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In order to train data I defined a while loop in which firstly each y_{pred} points are predicted using the softmax function and weight found in previous step.

Then using the gradient_w and gradient_w0 new weights are calculated according to y_{pred} generated. The while loop continues until weight converges.

In order to see the objective value change over time in the while loop we store each objective value in a vector and then make a graph of objective value vs iteration number.



In order to make a confusion matrix I defined $y_{\text{pred_subs}}$ and y_{subs} which are 300×1 matrices that store the corresponding class value as 1, 2 or 3. Using the table function with $y_{\text{pred_subs}}$ and y_{subs} I created the confusion matrix that you can see below.

```
> print(confusion_matrix)
      y_pred_subs
y_subs  1   2   3
1 100   0   0
2   0  98   2
3   0   3  97
> |
```

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After creating the `y_pred_subs` and `y_subs` drawing the decision boundaries were similar to the steps used in homework 1. You can find the commented code in the R file.

Here is thre graph with the decision boundaries.

