Multiclass Classification

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Binary Class

• There are only two classes in binary class data

| x ₁ | x ₂ | x _m | у |
|-----------------------|----------------|--------------------|---|
| | | | 0 |
| | | | 1 |
| | | | 0 |

Multiclass

More than two classes.

| X ₁ | x ₂ | x _m | у |
|-----------------------|----------------|--------------------|---|
| | | | 0 |
| | | | 1 |
| | | | 2 |
| | | | 0 |
| | | | 2 |

For examples,

- Iris data
 - https://archive.ics.uci.edu/ml/machine-learning-databases/iris/
 - 3 classes
 - 4 features
 - 150 samples (50 samples per class)
- MNIST data
 - http://yann.lecun.com/exdb/mnist/
 - 10 classes
 - 28x28 features
 - o 60K training samples and 10K test samples

One-Hot Encoding

Convert the class column to binary code (One-Hot label) like below.

| у |
|---|
| 0 |
| 1 |
| 2 |
| 0 |
| 2 |
| 1 |

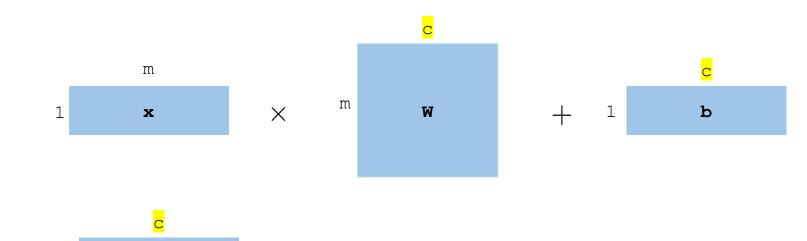


| Class 0 | Class 1 | Class 2 |
|---------|---------|---------|
| 1 | 0 | 0 |
| 0 | 1 | 0 |
| 0 | 0 | 1 |
| 1 | 0 | 0 |
| 0 | 0 | 1 |
| 0 | 1 | 0 |

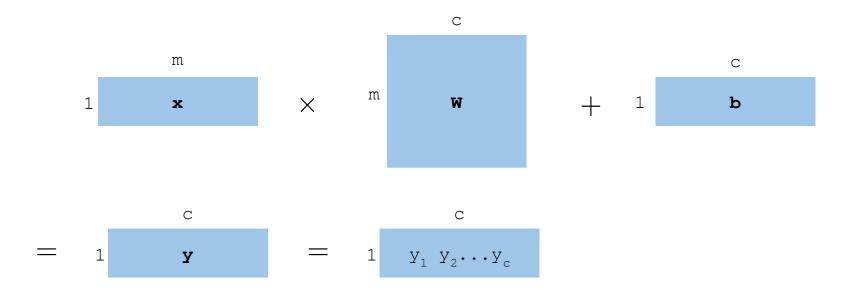
Linear Classifier

Linear Multiclass Classification

- c is # of classes.
- m is # of features.
- maxIndex() is a function that returns a index of element which is the max element given a vector.



Linear Multiclass Classification



Linear_Classifier(x) = maxIndex(y)

Linear Multiclass Classification

```
trainX = np.array([[0.1, 0.2, 0.3, 0.2],
                   [0.5, 0.4, 0.3, 0.7],
                   [0.3, 0.7, 0.4, 0.1],
                   [0.2, 0.8, 0.9, 0.3],
                   [1.1, 0.5, 0.2, 0.9],
                   [4.3, 5.3, 4.7, 4.2],
                   [4.5, 5.1, 5.3, 4.4],
                   [5.1, 4.8, 5.1, 4.6],
                   [4.9, 4.6, 4.9, 4.3],
                   [5.4, 5.5, 4.3, 4.7],
                   [10.1, 10.2, 10.3, 11.3],
                   [11.3, 11.2, 11.1, 10.3],
                   [12.5, 12.3, 12.1, 11.4],
                   [11.7, 11.8, 11.2, 12.8],
                   [13.1, 10.2, 12.4, 11.7]])
trainY = np.array([[1, 0, 0],
                   [1, 0, 0],
                   [1, 0, 0],
                   [1, 0, 0],
                   [1, 0, 0],
                   [0, 1, 0],
                   [0, 1, 0],
                   [0, 1, 0],
                   [0, 1, 0],
                   [0, 1, 0],
                   [0, 0, 1],
                   [0, 0, 1],
                   [0, 0, 1],
                   [0, 0, 1],
                   [0, 0, 1]])
```

```
testX = np.array([[0.5, 0.4, 0.6, 0.5],
                  [5.4, 5.6, 5.5, 5.2],
                  [11.7, 11.6, 11.5, 11.4]])
testY = np.array([[1, 0, 0],
                  [0, 1, 0],
                  [0, 0, 1]])
W = np.zeros((4, 3))
B = np.zeros(3)
alpha = 0.005
N = trainX.shape[0]
c = W.shape[1]
```

print(sum(np.argmax(np.dot(testX, W) + B, axis=1) == np.argmax(testY, axis=1))/testX.shape[0])



 $[0.05690523 - 0.23662129 \ 0.12713622]][0.69839576 \ 0.2862392 - 0.14555836]$

[[0.00136612 -0.03102624 0.01597588] [-0.07523478 0.2177921 -0.06253688] [-0.05620483 0.05216062 0.00881165]

1.0

Linear classifier for multiclass

Please be aware that this multiclass classifier is founded on a linear function and MSE (cost function). In the realm of multiclass classification, we typically use the softmax function and cross-entropy. We'll delve into this topic and learn more about them in our upcoming lecture on neural networks.