Softmax Classification

Multinomial classification

hypothesis = tf.nn.softmax(tf.matmul(X,W)+b)

```
In [1]:
```

```
from tqdm import tqdm_notebook
```

다항분류

In [2]:

```
import tensorflow as tf
import numpy as np
tf.set_random_seed(777) # for reproducibility
x_{data} = [[10,5], [9,5], [3,2], [2,4], [11,1]] # 5 by 2
y_{data} = [[1, 0, 0],
         [1, 0, 0],
          [0, 1, 0],
          [0, 1, 0],
          [0, 0, 1]] #5 by 3
X = tf.placeholder("float", [None, 2]) #5 by 2
Y = tf.placeholder("float", [None, 3]) #5 by 3
nb_classes = 3 #분류 종류의 수 A,B,C
W = tf.Variable(tf.random_normal([2, nb_classes]), name='weight')
b = tf.Variable(tf.random_normal([nb_classes]), name='bias')
# tf.nn.softmax computes softmax activations
# softmax = exp(logits) / reduce_sum(exp(logits), dim)
hypothesis = tf.nn.softmax(tf.matmul(X, W) + b)
# Cross entropy cost/loss
cost = tf.reduce_mean(-tf.reduce_sum(Y * tf.log(hypothesis), axis=1))
optimizer = tf.train.GradientDescentOptimizer(learning_rate=0.1).minimize(cost)
```

In [3]:

predict index: [0 2 1]

```
# Launch the graph in a session.
sess = tf.Session()
 # Initializes global variables in the graph.
sess.run(tf.global_variables_initializer())
 for step in tqdm_notebook(range(2001)):
          sess.run(optimizer, feed_dict={X: x_data, Y: y_data})
          if step % 200 == 0:
                   # print(step, sess.run(cost, feed_dict={X: x_data, Y: y_data}))
                   print("Step: {}, \text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\ti}\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\texitex{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\te}\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\te}\tint{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\tiex{\text{\text{\text{\text{\text{\text{\tiex{\tiexi{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\te
HBox(children=(IntProgress(value=0, max=2001), HTML(value='')))
                                        Cost : 5.044024467468262
Step: 0,
Step: 200,
                                        Cost: 0.1112934947013855
Step: 400,
                                        Cost: 0.05863405019044876
Step : 600,
                                        Cost: 0.03917425870895386
Step: 800,
                                        Cost: 0.02923898957669735
Step: 1000,
                                        Cost: 0.023261133581399918
Step: 1200,
                                        Cost : 0.019285501912236214
Step: 1400,
                                        Cost: 0.0164570901542902
Step: 1600,
                                        Cost: 0.0143448980525136
                                        Cost: 0.01270892471075058
Step: 1800,
Step: 2000,
                                        Cost: 0.011405272409319878
In [4]:
 # Testing & One-hot encoding
 test_data = [[9.5, 5.5],
                               [9.9, 1.5],
                                [3.1, 2.1]]
pred_val = sess.run(hypothesis, feed_dict={X: test_data})
pred_idx = sess.run(tf.argmax(pred_val, 1))
# print("predict value : Wn {} WnWnpredict index : {}".format(pred_val, pred_idx))
print("test data : {} \text{Wn\text\wnpredict value : \text\wnfty} the test data : {}\text{".format(test_data, pred_val, pred_val)}
test data : [[9.5, 5.5], [9.9, 1.5], [3.1, 2.1]]
predict value:
   [[9.89088714e-01 1.07102245e-02 2.01089031e-04]
   [2.56168302e-02 3.51880693e-08 9.74383175e-01]
  [3.70497666e-02 9.61320758e-01 1.62941683e-03]]
```

```
In [5]:
```

```
# grade로 예측값 표기
grade = ['A', 'B', 'C'] #nb_classes = 3

arg_val = sess.run(tf.arg_max(pred_val, 1))

p_grade = [ grade[val] for val in arg_val ]
print(p_grade)
```

WARNING:tensorflow:From <ipython-input-5-e0f0102ca5bf>:4: arg_max (from tensorflow.p ython.ops.gen_math_ops) is deprecated and will be removed in a future version. Instructions for updating:
Use `tf.math.argmax` instead
['A', 'C', 'B']

In []:

EX. Animal classification with softmax_cross_entropy_with_logits

In [17]:

import pandas as pd

df = pd.read_csv("./data/data-04-zoo.csv", header=None)
df[19:40]

Out[17]:

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
19	1	0	0.0	1.0	0.0	0	1.0	1.0	1.0	1.0	0.0	0.0	4.0	0.0	0.0	1.0	0.0
20	1	0	0.0	1.0	0.0	0	0.0	1.0	1.0	1.0	0.0	0.0	4.0	1.0	0.0	1.0	0.0
21	0	0	1.0	0.0	0.0	1	1.0	1.0	1.0	0.0	0.0	1.0	0.0	1.0	0.0	0.0	3.0
22	1	0	0.0	1.0	0.0	0	1.0	1.0	1.0	1.0	0.0	0.0	4.0	0.0	0.0	1.0	0.0
23	1	0	0.0	1.0	0.0	0	1.0	1.0	1.0	1.0	0.0	0.0	4.0	1.0	0.0	1.0	0.0
24	1	0	0.0	1.0	0.0	0	0.0	1.0	1.0	1.0	0.0	0.0	4.0	1.0	0.0	1.0	0.0
25	1	0	0.0	1.0	0.0	0	0.0	1.0	1.0	1.0	0.0	0.0	4.0	1.0	1.0	1.0	0.0
26	0	0	1.0	0.0	0.0	1	0.0	1.0	1.0	0.0	0.0	1.0	0.0	1.0	1.0	0.0	3.0
27	0	0	1.0	0.0	0.0	1	1.0	1.0	1.0	0.0	0.0	1.0	0.0	1.0	0.0	0.0	3.0
28	1	0	0.0	1.0	0.0	0	0.0	1.0	1.0	1.0	0.0	0.0	4.0	0.0	1.0	0.0	0.0
29	1	0	0.0	1.0	0.0	0	1.0	1.0	1.0	1.0	0.0	0.0	4.0	1.0	0.0	1.0	0.0
30	0	1	1.0	0.0	1.0	0	0.0	0.0	1.0	1.0	0.0	0.0	2.0	1.0	1.0	0.0	1.0
31	0	0	1.0	0.0	0.0	1	1.0	1.0	1.0	0.0	0.0	1.0	0.0	1.0	0.0	0.0	3.0
32	0	0	1.0	0.0	0.0	0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	6.0
33	0	0	1.0	0.0	0.0	1	1.0	0.0	0.0	0.0	0.0	0.0	4.0	0.0	0.0	0.0	6.0
34	0	0	1.0	0.0	0.0	1	1.0	0.0	0.0	0.0	0.0	0.0	6.0	0.0	0.0	0.0	6.0
35	0	1	1.0	0.0	1.0	0	1.0	0.0	1.0	1.0	0.0	0.0	2.0	1.0	0.0	0.0	1.0
36	1	0	0.0	1.0	0.0	0	0.0	1.0	1.0	1.0	0.0	0.0	4.0	1.0	0.0	1.0	0.0
37	0	0	1.0	0.0	0.0	1	1.0	1.0	1.0	0.0	0.0	1.0	0.0	1.0	0.0	1.0	3.0
38	0	0	0.0	1.0	0.0	1	1.0	1.0	1.0	1.0	0.0	1.0	0.0	1.0	0.0	1.0	0.0
39	0	1	1.0	0.0	1.0	0	0.0	0.0	1.0	1.0	0.0	0.0	2.0	1.0	1.0	0.0	1.0

In [18]:

```
import tensorflow as tf
import numpy as np
tf.set_random_seed(777) # for reproducibility
# Predicting animal type based on various features
xy = np.loadtxt('./data/data-04-zoo.csv', delimiter=',', dtype=np.float32)
x_{data} = xy[:, 0:-1]
y_{data} = xy[:, [-1]]
print(x_data.shape, y_data.shape)
print('\mx_data :\m', x_data)
print('\my_data :\m', y_data)
nb_classes = 7 \# 0 \sim 6
X = tf.placeholder(tf.float32, [None, 16])
Y = tf.placeholder(tf.int32, [None, 1]) # 0 \sim 6
Y_one_hot = tf.one_hot(Y, nb_classes) # one hot
print("\underset none_hot", Y_one_hot)
#Y_one_hot ≥ 0 ≥ CH [1,0,0,0,0.0]
             1 일때 [0,1,0,0,0,0,0]
#
#
             2 일때 [0,0,1,0,0,0,0]
Y_one_hot = tf.reshape(Y_one_hot, [-1, nb_classes])
print("₩nreshape", Y_one_hot)
# Y_one_hot에서 제일 큰 값
W = tf.Variable(tf.random_normal([16, nb_classes]), name='weight')
b = tf.Variable(tf.random_normal([nb_classes]), name='bias')
# tf.nn.softmax computes softmax activations
# softmax = exp(logits) / reduce_sum(exp(logits), dim)
logits = tf.matmul(X, W) + b
hypothesis = tf.nn.softmax(logits)
# Cross entropy cost/loss
cost_i = tf.nn.softmax_cross_entropy_with_logits(logits=logits,
                                                 labels=Y_one_hot)
cost = tf.reduce_mean(cost_i)
optimizer = tf.train.GradientDescentOptimizer(learning_rate=0.1).minimize(cost)
prediction = tf.argmax(hypothesis, 1)
correct_prediction = tf.equal(prediction, tf.argmax(Y_one_hot, 1))
accuracy = tf.reduce_mean(tf.cast(correct_prediction, tf.float32))
(101, 16) (101, 1)
x_data:
```

```
(101, 16) (101, 1)

x_data:
[[1. 0. 0. ... 0. 0. 1.]
[1. 0. 0. ... 1. 0. 1.]
[0. 0. 1. ... 1. 0. 0.]
...
```

```
[1. 0. 0. ... 1. 0. 1.]
[0. 0. 1. ... 0. 0. 0.]
 [0. 1. 1. ... 1. 0. 0.]]
y_data :
 [[0.]
[0.]
 [3.]
 [0.]
 [0.]
[0.]
 [0.]
 [3.]
 [3.]
 [0.]
 [0.]
 [1.]
 [3.]
 [6.]
 [6.]
 [6.]
 [1.]
 [0.]
[3.]
 [0.]
[1.]
 [1.]
 [0.]
 [1.]
 [5.]
 [4.]
 [4.]
 [0.]
 [0.]
[0.]
 [5.]
 [0.]
 [0.]
 [1.]
 [3.]
 [0.]
 [0.]
 [1.]
 [3.]
 [5.]
[5.]
 [1.]
[5.]
 [1.]
 [0.]
 [0.]
 [6.]
 [0.]
 [0.]
```

[0.] [5.] [4.] [6.]

[0.] [0.]

```
[1.]
[1.]
[1.]
[1.]
[3.]
 [3.]
 [2.]
 [0.]
 [0.]
 [0.]
 [0.]
 [0.]
 [0.]
 [0.]
 [0.]
 [1.]
 [6.]
 [3.]
 [0.]
 [0.]
 [2.]
 [6.]
 [1.]
 [1.]
 [2.]
 [6.]
 [3.]
 [1.]
 [0.]
 [6.]
 [3.]
 [1.]
 [5.]
 [4.]
[2.]
 [2.]
 [3.]
 [0.]
 [0.]
[1.]
 [0.]
[5.]
[0.]
[6.]
[1.]]
one_hot Tensor("one_hot_1:0", shape=(?, 1, 7), dtype=float32)
reshape Tensor("Reshape_1:0", shape=(?, 7), dtype=float32)
```

In [19]:

```
Acc: 16.83%
Step:
          0,
                 Loss: 2.473,
Step:
        100,
                 Loss: 0.517,
                                  Acc: 80.20%
                 Loss: 0.342,
                                  Acc: 91.09%
Step:
        200,
Step:
        300,
                 Loss: 0.260,
                                  Acc: 95.05%
Step:
                 Loss: 0.211,
                                  Acc: 95.05%
        400,
Step:
        500,
                 Loss: 0.178,
                                  Acc: 95.05%
                 Loss: 0.154,
                                  Acc: 96.04%
Step:
        600,
Step:
        700,
                 Loss: 0.135,
                                  Acc: 98.02%
Step:
        800,
                 Loss: 0.120,
                                  Acc: 98.02%
                                  Acc: 99.01%
Step:
        900,
                 Loss: 0.108,
Step:
       1000,
                 Loss: 0.098,
                                  Acc: 100.00%
                 Loss: 0.090,
Step:
                                  Acc: 100.00%
       1100,
Step:
       1200,
                 Loss: 0.083,
                                  Acc: 100.00%
Step:
       1300,
                 Loss: 0.077,
                                  Acc: 100.00%
Step:
                 Loss: 0.072,
                                  Acc: 100.00%
       1400,
Step:
       1500,
                 Loss: 0.068,
                                  Acc: 100.00%
Step:
       1600,
                 Loss: 0.064,
                                  Acc: 100.00%
Step:
                                  Acc: 100.00%
       1700,
                 Loss: 0.060,
Step:
                 Loss: 0.057,
                                  Acc: 100.00%
       1800,
Step:
       1900,
                 Loss: 0.054,
                                  Acc: 100.00%
```

```
In [20]:
# Let's see if we can predict
pred = sess.run(prediction, feed_dict={X: x_data})
# y_data: (N,1) = flatten => (N, ) matches pred.shape
for p, y in zip(pred, y_data.flatten()):
   print("[{}] Prediction : {}, True Y : {}".format(p == int(y), p, int(y)))
   \# print("[\{\}\}) Prediction: \{\}, True Y: \{\}, y_data: \{\}".format(p == int(y), p, int(y), y_data)
[True]
       Prediction: 0,
                      True Y : 0
[True]
       Prediction: 0.
                      True Y:0
[True]
       Prediction: 3. True Y: 3
[True]
       Prediction: 0,
                       True Y:0
[True]
       Prediction: 0,
                      True Y:0
[True]
       Prediction: 0,
                      True Y:0
       Prediction: 0, True Y: 0
[True]
[True]
       Prediction: 3,
                      True Y: 3
[True]
       Prediction: 3,
                      True Y: 3
[True]
       Prediction: 0. True Y: 0
       Prediction: 0, True Y: 0
[True]
       Prediction: 1,
[True]
                       True Y:1
[True]
       Prediction: 3,
                      True Y: 3
[True]
       Prediction: 6, True Y: 6
[True]
       Prediction: 6, True Y: 6
[True]
       Prediction: 6,
                      True Y: 6
[True]
       Prediction: 1, True Y: 1
[True]
       Prediction: 0, True Y: 0
[True]
       Prediction: 3,
                       True Y:3
[True]
       Prediction: 0,
                      True Y : 0
[True]
       Prediction: 1. True Y: 1
[True]
       Prediction: 1, True Y: 1
[True]
       Prediction: 0,
                      True Y:0
[True]
       Prediction: 1, True Y: 1
[True]
       Prediction: 5, True Y: 5
[True]
       Prediction: 4, True Y: 4
[True]
       Prediction: 4,
                      True Y:4
[True]
       Prediction : 0, True Y : 0
       Prediction: 0, True Y: 0
[True]
[True]
       Prediction: 0.
                      True Y:0
[True]
       Prediction: 5,
                      True Y: 5
       Prediction: 0.
                      True Y: 0
[True]
[True]
       Prediction: 0.
                       True Y:0
[True]
       Prediction: 1.
                       True Y:1
[True]
       Prediction: 3,
                      True Y: 3
[True]
       Prediction: 0, True Y: 0
       Prediction: 0, True Y: 0
[True]
[True]
       Prediction: 1,
                      True Y: 1
[True]
       Prediction: 3, True Y: 3
       Prediction: 5, True Y: 5
```

[True] [True]

[True]

[True]

[True] [True]

[True]

[True]

[True]

[True]

[True]

Prediction: 5.

Prediction: 1,

Prediction: 5,

Prediction: 1.

Prediction: 0.

Prediction: 0.

Prediction: 6,

Prediction: 0,

[True] Prediction: 0, True Y: 0

Prediction: 0, True Y: 0

Prediction: 0,

True Y: 5

True Y: 1

True Y: 5

True Y: 1

True Y:0

True Y:0

True Y: 6

True Y:0

True Y:0

```
[True]
       Prediction: 5,
                      True Y: 5
[True]
       Prediction: 4,
                       True Y:4
[True]
       Prediction: 6,
                       True Y:6
[True]
       Prediction: 0.
                       True Y:0
[True]
       Prediction: 0,
                       True Y:0
[True]
       Prediction: 1,
                       True Y:1
                       True Y:1
[True]
       Prediction: 1,
[True]
       Prediction: 1,
                       True Y:1
[True]
       Prediction: 1,
                       True Y:1
[True]
       Prediction: 3,
                       True Y:3
[True]
       Prediction: 3,
                       True Y:3
[True]
       Prediction: 2,
                       True Y:2
       Prediction: 0,
                       True Y:0
[True]
[True]
       Prediction: 0,
                       True Y:0
[True]
       Prediction: 0,
                       True Y : 0
[True]
       Prediction: 0,
                       True Y:0
[True]
       Prediction: 1,
                       True Y:1
[True]
       Prediction: 6,
                       True Y:6
[True]
       Prediction: 3,
                       True Y:3
[True]
       Prediction: 0,
                       True Y:0
[True]
       Prediction: 0,
                       True Y:0
                       True Y:2
[True]
       Prediction: 2,
[True]
       Prediction: 6,
                       True Y:6
[True]
       Prediction: 1,
                       True Y:1
[True]
       Prediction: 1,
                       True Y:1
[True]
       Prediction: 2,
                       True Y:2
[True]
       Prediction: 6,
                       True Y:6
[True]
       Prediction: 3,
                       True Y:3
[True]
       Prediction: 1,
                       True Y:1
[True]
       Prediction: 0,
                       True Y:0
[True]
       Prediction: 6.
                       True Y:6
[True]
       Prediction: 3,
                       True Y:3
[True]
       Prediction: 1,
                       True Y:1
[True]
       Prediction: 5,
                       True Y: 5
[True]
       Prediction: 4,
                       True Y:4
[True]
       Prediction: 2.
                       True Y: 2
[True]
       Prediction: 2,
                       True Y: 2
[True]
       Prediction: 3.
                       True Y:3
       Prediction: 0,
[True]
                       True Y:0
[True]
       Prediction: 0,
                       True Y:0
[True]
                       True Y: 1
       Prediction: 1,
[True]
       Prediction: 0.
                       True Y:0
[True]
       Prediction: 5,
                       True Y: 5
[True]
       Prediction: 0,
                       True Y:0
[True]
       Prediction: 6,
                       True Y: 6
[True]
       Prediction: 1,
                      True Y: 1
```