Feedforward-OneCase-IncompleteCode1

March 25, 2021

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```
[1]: import numpy as np
     #Create loop.
     # 5 [H1 H2] 2
     # +
     # + [+
     # + [+ +]
     # + [+ +] +
     # + [+ +] +
       black box
     def feed_forward(features,w1,b1,w2,b2,w3,b3,act):
         Olayer1 = np.matmul(w1, features) + b1 #Computing next row of nodes/
     → features (4) [Hidden Layer 1]
         Olayer2 = np.matmul(w2, Olayer1) + b2 #Computing next row of nodes/
     → features (3) [Hidden Layer 2]
         Olayer3 = np.matmul(w3, Olayer2) + b3 #Computing next row of nodes/
     \rightarrow features (2) Node/Features
         #Giving the option to choose either sigmoid and/or RELU
         if int(act) > 0:
             Output1 = sigmoid(Olayer3)
         else:
             Output1 = RELU(Olayer3)
         return(Output1)
     #Definitions for the Sigmoid or RELU
     def sigmoid(x): #Creating a definition for Sigmoid function
         return 1/(1+np.exp(-x)) #Sigmoid function
     def RELU(x):
```

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if all(x) > 0: #all() function returns True if all items in an iterable are
 \rightarrow true
        return np.around(x,decimals=7) #around() function rounds the given_
→number to the amount of decimals given
    else:
        return 0
return targets_predicted
print('Starting ...')
## Set up training data with just one case
#Given the features (the 5 starting X1, X2, X3, X4, X5). We are starting out with a_{\square}
\hookrightarrow (5x1) matrix (Input layer).
features = np.array([[0],
                     [0],
                     [0],
                     [0],
                     [0]])
targets_observed = np.array([0,1]) #The numbers we want to achieve after black_
→box calculations. The final answers for the last 2 features.
number_of_features, number_of_cases = 5,1 #Starting with 5 nodes (features) and__
\hookrightarrow 1 case (basically stating we have a (5x1)).
weights_1 = np.random.rand(4,5) #Setting the dimensions of the matrix for our_
→ random weights (rows by columns)
biases_1 = np.random.rand(4,1) #Setting the dimensions of the matrix for our__
→baises (rows by columns)
#print("Weights_1 = " , weights_1)
#Input Layer E R5 ----> Output Layer E R4
                              (5 \ x \ 1) \qquad (4 \ x \ 1)
\# \qquad (4 x 5)
                             x1 b1 w1,1x1 + w1,2x2 + w1,3x3 + 1
#w1,1 w1,2 w1,3 w1,4 w1,5
\rightarrow w1,4x4 + w1,5x5 + b1
\#w2,1 w2,2 w2,3 w2,4 w2,5 * x2 + b2 = w2,1x1 + w2,2x2 + w2,3x3 +
\rightarrow w^2, 4x^4 + w^2, 5x^5 + b^2 = (4 \times 1) matrix with new features after calculations
→ for Hidden Layer 1
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```
#w3,1 w3,2 w3,3 w3,4 w3,5
                                             b3
                                                       w3,1x1 + w3,2x2 + w3,3x3 + 
                                  x3
\rightarrow w3,4x4 + w3,5x5 + b3
#w4,1 w4,2 w4,3 w4,4 w4,5
                                             b4
                                                       w4,1x1 + w4,2x2 + w4,3x3 + 
                                  x4
\rightarrow w4,4x4 + w4,5x5 + b4
                                  x5
#This process continues until we hit our destination of 2 nodes/features.
weights_2 = np.random.rand(3,4)
biases_2 = np.random.rand(3,1)
\#(3 \times 4) weights *(4 \times 1) features +(3 \times 1) biases =(3 \times 1) matrix for new 
→ features after calculations for Hidden Layer 2
       (3 \ x \ 4)
                         (4 \ x \ 1)
                                    (3 \ x \ 1)
                                              w1,1x1 + w1,2x2 + w1,3x3 + w1,4x4
#w1,1 w1,2 w1,3 w1,4
                                     b1
                          x1

→ + b1

\#w2,1 \ w2,2 \ w2,3 \ w2,4 \ * \ x2 \ + \ b2 = \ w2,1x1 + w2,2x2 + w2,3x3 + w2,4x41
\rightarrow+ b2 = (3 x 1) matrix with new features after calculations for Hidden Layer 2
#w3,1 w3,2 w3,3 w3,4
                                      Ъ3
                                               w3,1x1 + w3,2x2 + w3,3x3 + w3,4x4
                           x3
→+ b3
                            x4
weights_3 = np.random.rand(2,3)
biases_3 = np.random.rand(2,1)
#print("Weight_3 =" , weights_3)
\#(2 \times 3) weights *(3 \times 1) features +(2 \times 1) = (2 \times 1) matrix for new features.
→after calculations for 2 features (our ending)
                  (3 \ x \ 1) \quad (2 \ x \ 1)
\# (2 x 3)
#w1,1 w1,2 w1,3
                               b1
                                     w1,1x1 + w1,2x2 + w1,3x3 + b1
                      x1
\#w2,1 w2,2 w2,3 * x2 + b2 = w2,1x1 + w2,2x2 + w2,3x3 + b2 = (2 \times 1)_{11}
→matrix with new features after calculations for Hidden Layer 2
                      x3
act = input("sigmoid(1) or RELU(-1)?") #Using an input to allow you to call ∪
⇔either Sigmoid or RELU
#Calling the "feed_foward" definition and saving the result as_
→ "Targets Predicted"
Targets_Predicted = feed_forward(features, weights_1, biases_1,
                                       weights_2,biases_2,
                                       weights_3,biases_3,act)
```

```
print('Features: ',features) #our beginning features that we are given to us
print(' Targets: ', targets_observed) #printing the "target_observed"

\( \times variables [0,1] <- This is what we want our results to be. \)

print(' Targets predicted: ', Targets_Predicted) #Prints out our predict_
\( \times values from either the Sigmoid and/or RELU function. \)
```

```
Starting ...
sigmoid(1) or RELU(-1)? 1

Features : [[0]
  [0]
  [0]
  [0]
  [0]]
  Targets : [0 1]
  Targets predicted : [[0.97627951]
  [0.98872001]]
```