Unit - 2

Feed Forward Newral Network

Food Forward Network:

It is an artificial neural network in which the connection between nodes does not form a cycle. Information is processed in one direction.

## Components:

- i) Layer of input
- ii) Hidden layor
- iii) Output layer
- iv) Weights

## Cost function:

C(w,b), =  $\frac{1}{an} \leq ||y(x)-a||^2$  MSE - Mean Square Error Loss function:

Cross entropy loss:

$$L(\bullet) = \begin{cases} -\log(9) & \text{if} \\ -\log(1-9) \end{cases}$$

## Back propagation:

It is a method of fine tuning the weights of the newral network based on the evror rate obtained in the previous iteration (or) epoch.

Peroper turing of the weights allows you to reduce everor rates and make the model reliable by increasing its generalizing. is territored using correctitive learning. Regularisation: overfitting. Overfitting - It gives good result during training process but fails in testing. Underfitting - Fails in training process Regularisation techniques: icolinis ere strain out fi i) La Regularation ii) LI Regularisation tout stugal out prinsus no chose to each other on the may. iv) Data Augmentation N) Early stopping nothouse philomogenomic (in at dimension was at iv) repluse viap. It setusure the footunes sometaile priquere judgmiz bone tugini fo phicaliniz per betasioni en emberment lastineer each other.

It refers to a newal network, which is trained using competitive learning.

- i) Competetive learning Select a wining processing element based on some criticia (Eucliean distance), after the winning processing element is selected, its weight is adjusted according to learning law.
  - ii) Topology Ondered Map It implies that if two inputs are similar characteristics, the most active processing element answering two inputs that are brated close to each other on the map.
- ·iii) Dimensionality reduction High dimension to low dimension.
- iv) Feature Map. It returns the features of input and simply grouping distance themselves as indicated by similarity lecturen each other.

occure without supervision because the nodes are self organising.

Structure

1 D 2D 3D vector rectangular heagonal

SOM training algorithm: Model tourstand

Step1: Initialisation

i) wij → weight matrix . I a of book

i → input neuron of catalla for all

j → cluster neuron 20 for along princesel

ii) Check the topology radius R.

R=0: means only winning unit weight updated R=1: nearby neuron weights also updated

: sale &a

iii) Alpha -> Learning Rate

Step 1: While the stopping critoria is not satisfied.

Step a: For each, every input pattern do 3-5

Step 8: Find distance lecturen input and cluster

D[i] = \(\frac{2}{5}\), \(\frac{2}{5}\),

Step 4: Select winning unit

Step: Weight updation

wij (new) = wij (old) + & [xi - wij (old)]

Step 6: Update & situe 20 20 - MOZ (v Step 71: Reduce Rod. vaisinreque tuantien ernon Step8: Check for stopping critoria Structure In Co Co de l'adord section de Co Construct ksom to cluster 4 given vectors [0 0 11], [1 0 0 0], [0 1 1 and [o o o i]. soietom theire - jiw (i No of cluster formed is 2. Assume learning rate of 0.5. novemen restends ii) Check the the topology nadius R. Soln: batallo offeist vettors in wino anom: 0 = 81 hallow of cluster m= 2 navera person ! = 9 iii) Alpha - Leasuring Rater it would will pro gro iiin step 1: While the stopping w Triboripro Francet solviefied. Wij = w31 0.6 0.5 W22 at 6 gets to eget 8-60 8: Fez each, eurous explus. postion 3-5 steps: Find distant [ Deption ] inprotes buit sque 可到一系是一个一一一 Step 4: . Salast winning unit Step; Weight updation [(bid); (w-ix) > + (bid); (cust); (cust); (cust)

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D[i] = \frac{1}{2} \sum_{i=1}^{2} \left[ x_i - w_{ij} \right]^{2^0} = 0 
      = (x1-w1)2+ (x2-w21)2+ (x3-w31)2+ (x4-w41)2
       = (0-0.2)2+ (0-0.4)2+ (1-0.6)2+ (1-0.8)2
= 0.04+0.16+0.04=0.4

(P.O-0) + 40.16+(8.0.0) + (1.0-1) =
        D[i] = 0:4 18.0+ 40.0+ 40.0 + 18.0=
 D[a] = \sum_{i=1}^{4} \sum_{j=a}^{4} \left[x_i^2 - w_{ij}\right]^2 : \infty 
   = (x_{1} - w_{12})^{2} + (x_{2} - w_{22})^{2} + (x_{3} - w_{32})^{2} + (x_{4} - w_{42})^{2}
         = (0-0.9)^{2}+(0-0.7)^{2}+(1-0.5)^{2}+(1-0.3)^{2}
           +3.0 = PO.07-722.0+PD.0+10.0=
          = 0.81+0.49+0.25+0.49
         = 2.04
                                  [s) a < [1] $
     D[2] = 2.04
        D[i] < D[2] = (P.O i) (2.0) + P.O = 8100
         0.4 (2.048.0 - (0.0-0) (3.0) + 7.0 = 2200
      D[1] is the winning element of sew
  W11 = 0.2+ (0.5) (0=0.2) = 8), (2.0) + 8.0 = 6,00
  W21 = 0.4+ (0.5) (0.0.4) = 0.2
  W31 = 0.6+(0.5) (1-0.6) = 0.8 5.01
  W41 = 0.8 + (0.5) (1-0.8)=0.9 3.0
       Wij = \begin{bmatrix} 0.1 & 0.9 \\ 0.2 & 0.7 \\ 0.8 & 0.5 \end{bmatrix}
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2nd i/p vector = [1 0 0 0]
    P. Co. - B. E. Exis - WHJ. and + "Cow on) =
        D[i] = 2 ( . [x1 - Wi]? 0 - 0) + (6.0 - 0) =
              = (1-0.1)+ (0.0.2)+ (0-0.8)+ (0-0.9)+
              = 0.81 + 0.04 + 0.64 + 0.81 = 2.3
        力[2] = 差 是 [x; -wij] ? [] = [] []
= [1-0.9) + (0-0.7) + (0-0.5) + (0-0.8) + (0-0.8)
        = 0.01 + 0.09 + 0.05 + 0.09 = 0.84
               2.3 > 0.84
              か[2] 人(2]
                                           D (3) = 2.04
         W12 = 0.9 + (0.5) (1-0.9) = 0.95 > [1] C.
         Was = 0.7 + (0.5) (0-0.7) = 0.350.5) +.0
        W32 = 0.5 + (0.5) (0-0.5) = 0.25 2 [j]
        W42 = 0.3+ (0.5) (0-0.5) = 0.15.0) + 8.0 = 16
                     [6.1 60.95 (4.0.0) (3.0) + 4.0 = 1000
         Wij = \begin{cases} 0.1 & 0.15 \\ 0.8 & 0.35 & 0.01 \\ 0.8 & 0.85 & 0.01 \\ 0.9 & 0.15 \end{cases} \begin{pmatrix} 2.0 & 0.15 \\ 0.0 & 0.05 \\ 0.0 & 0.05 \end{pmatrix}
```

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39d i/p vector: [0-1010]: restroy 11/ 100
 D[j] = E E Exiswwiji?
DCIJ-18 5 [0-0.1] 4 [1-0.2] 7 [1-0.2] 7
         = 0.01+0.64 +0.04 +0.81 0
  D[2] = 2 [0-0.95] + [1-0.35] + [1-0.25] +
                             [0-0.15]a
      [21.0-1] + [30.0-0]
       = 0.90 + 0.42 + 0.56 + 0.02
        D[i] (D[2]
          1.5 (1.9
       D[i] is the winning element
 w_{11} = 0.1 + (0.5)(0.-0.01) = 0.05 21 [1] (
 Wal = 0.2+ (0.5) (1-0.2) = 0.6.0) + 30.0 = 10
 W31 = 0.8 + (0.5). (1-0.8) 0= 0.9(2.0) + d.0 = 1000
 Wat = 0.9+ (0.5) (0=0.9)0=0.45.0) + 1.0=180
              5127 = (940-1) (5.0) + 54.0 = 14M
            W11 0.05 0.95 W12
           Wal 0.6 0.35 Was
     Wij =
           W31 0.9 0.25 W32
           W41 10.45
                   0.15 J.W42
                               1.0
                        21.0 81.0
```

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4 nd 1/p vector: [0,00,01]; patoov 4/i has
                        D[1]= 2, 2, [0-0.05] 7 [0-0.6] 7 [0-0.9] 8
                                                    = 0.002+0.36+0.81+0.30
 D[a] = [2 0 0 0 95] $ [0-0.35] $ [0-0.35] $ [0-0.5] $ [0-0.5] $
                                                                                   = 0.90+0.42+0.56+0.0=
                                                       = 0.90 + 0.12 + 0.06 + 0.72
                                                        = 1.8
                                                                                                                                      D[1] < D[2]
                                        D[1] is the winning elestine 71.47
                                       D[i] is the winning element
               w_{11} = 0.05 + (0.5)(0-0.05) = 0.02 + 6.0 = 1800
              Wal = 0.6 + (0.5) (0-0.6) =10.8 2.0) + 3.0 = 1800
              W31 = 0.9 + (0.5) (0=0(9))=00.45.0) + P.0 = 10W
              W41 = 0.45 + (0.5) (1-0:45) = 0:72
                         w_{ij} = \begin{bmatrix} 0.2 & 0.95 & 20.0 & 20.0 \\ 0.3 & 0.35 & 20.0 & 20.0 \\ 0.4 & 0.35 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 0.15 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 & 20.0 \\ 0.72 & 20.0 & 20.0 &
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