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```
import numpy as np
import pandas as pd
```

Loading and Procession of dataset

```
In [2]:
         data = pd.read csv("C:\\Users\\ASUS\\titanic processed.csv")
         x = data.drop("Survived",axis=1).values
         y = data["Survived"].values
In [3]:
         hidden layers = [5,3]
         1r = 30
         n = 20
         batch_size = 25
         n = len(hidden layers)+1
In [4]:
         s = int(0.8*x.shape[0])
         x_{train} = x[:s].T
         y_train = y[:s]
         outputs = [None]*4
         delta = [None]*3
         x_{test} = x[s:]
         y_{test} = y[s:]
         n_input = x_train.shape[0]
         z = y_train.reshape(-1,1)
         classes = np.unique(y_train)
         target = (z==classes).T
         n_output = target.shape[0]
         m = x train.shape[1]
In [5]:
         np.random.seed(0)
         weights = [np.random.standard_normal((o,i+1))*0.01]
                    for i,o in zip([n_input]+hidden_layers,hidden_layers+[n_output])]
In [6]:
         def act(x):
             return 1/(1+np.exp(-x))
In [7]:
         def dact(x):
             return x*(1-x)
In [8]:
         def pad(x):
             pad_w = [(1,0),(0,0)]
             return np.pad(x,pad_w,constant_values=1)
In [9]:
         def forward propagate(n,input):
             outputs[-1] = input
             for i in range(n):
```

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```
outputs[i] = act(weights[i]@pad(outputs[i-1]))
              return outputs[n-1]
In [10]:
          def error(actual,predicted):
              error = actual - predicted
              return np.sum(error*error)
In [11]:
          def back_propagate(n,target):
              for i in range(n-1,-1,-1):
                   if i == n-1:
                       errors = outputs[i]-target
                   else:
                       errors = weights[i+1][:,1:].T@delta[i+1]
                   delta[i] = errors*dact(outputs[i])
In [12]:
          def update weight(n,lr,batch size):
              for i in range(n):
                  weights[i] -= lr*delta[i]@pad(outputs[i-1]).T/batch_size
In [13]:
          def predict(n,inputs):
              pred = forward propagate(n,inputs.T).T.argmax(axis=-1)
              return pred
In [14]:
          def accuracy(x_test,y_test,n):
              return (predict(n,x test)==y test).mean()
```

Implementing BPNN

```
In [15]:
          verbose = True
          if verbose:
              print("Initital Weights:")
              print(*weights,sep="\n")
              print("Training:")
          for epoch in range(n_epochs):
              sum error = 0
              fs = range(m+batch size)
              for f,t in zip(fs,fs[1:]):
                   forwd out = forward propagate(n,x train[:,f:t])
                   sum_error += error(target[:,f:t],forwd_out)
                   back_propagate(n,target[:,f:t])
                  update weight(n,lr,batch size)
              if verbose:
                  print(f'> epoch={epoch+1}, lrate={lr:.1f}, error={sum error/m:.5f}')
          if verbose:
              print("Trained Weights:")
              print(*weights, sep="\n")
              print("Evaluation:")
              print(f"Accuracy of the classifer:{accuracy(x_test,y_test,n)*100:.2f}%")
```

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```
Initital Weights:
[[ 0.01764052  0.00400157  0.00978738  0.02240893  0.01867558 -0.00977278
  0.00950088 -0.00151357 -0.00103219 0.00410599]
 0.01494079 -0.00205158 0.00313068 -0.00854096]
 0.00045759 -0.00187184 0.01532779 0.01469359]
 0.00154947 0.00378163 -0.00887786 -0.01980796 -0.00347912 0.00156349
  0.01230291 0.0120238 -0.00387327 -0.00302303]
 [-0.01048553 -0.01420018 -0.0170627
                                   0.01950775 -0.00509652 -0.00438074
  [[-0.00895467 \quad 0.00386902 \quad -0.00510805 \quad -0.01180632 \quad -0.00028182 \quad 0.00428332]
 [ 0.00066517  0.00302472  -0.00634322  -0.00362741  -0.0067246   -0.00359553]
 [[-0.00907298  0.00051945  0.00729091  0.00128983]
 [ 0.01139401 -0.01234826  0.00402342 -0.0068481 ]]
Training:
> epoch=1, lrate=30.0, error=0.49509
> epoch=2, lrate=30.0, error=0.49015
> epoch=3, lrate=30.0, error=0.42328
> epoch=4, lrate=30.0, error=0.32519
> epoch=5, lrate=30.0, error=0.31955
> epoch=6, lrate=30.0, error=0.31736
> epoch=7, lrate=30.0, error=0.31585
> epoch=8, lrate=30.0, error=0.31467
> epoch=9, lrate=30.0, error=0.31368
> epoch=10, lrate=30.0, error=0.31279
> epoch=11, lrate=30.0, error=0.31195
> epoch=12, lrate=30.0, error=0.31113
> epoch=13, lrate=30.0, error=0.31029
> epoch=14, lrate=30.0, error=0.30946
> epoch=15, lrate=30.0, error=0.30865
> epoch=16, lrate=30.0, error=0.30789
> epoch=17, lrate=30.0, error=0.30718
> epoch=18, lrate=30.0, error=0.30652
> epoch=19, lrate=30.0, error=0.30591
> epoch=20, lrate=30.0, error=0.30533
Trained Weights:
[[ 2.59377565 -3.30489323 -1.36582057 0.55108638 0.54484425 -3.2269304
  0.40831845 -0.37603816 -0.49050768 -3.62382812]
 [ \ 2.55310358 \ -3.23245764 \ -1.35879432 \ \ 0.50578107 \ \ 0.51561022 \ -3.1856521 
  0.4094337 -0.39589713 -0.53693558 -3.54442153]
 [ 2.4621099 -3.18672024 -1.30013933 0.49196919 0.55699735 -3.27845112
  0.4317014 -0.42330159 -0.44407793 -3.51849201]
 [ 2.56214525 -3.26541624 -1.36150036  0.51779539  0.50778396 -3.20546279
  0.42429237 -0.38049015 -0.51475034 -3.5775894 ]
 [ 2.62232846 -3.30355771 -1.44113874 0.54280011 0.47086607 -2.99803274
  0.35224981 -0.37702196 -0.64849008 -3.48690878]]
[[ 0.41666756 -2.29532261 -2.27306919 -2.26138852 -2.27356199 -2.2601175 ]
  2.14367882 -2.05891705 -2.00511851 -2.01911792 -2.02489772 -1.93553795]
 [ 1.5681226 -2.13860288 -2.06685704 -2.07847008 -2.09969811 -2.01787684]]
[[-2.63598113 1.55565258 1.8048985
                                   1.630234 ]
 [ 2.63438194 -1.56389923 -1.79391282 -1.63487603]]
Evaluation:
Accuracy of the classifer:85.96%
```