

Assignment 7:Artificial Neural Network

A)import libraries you think you'll need

```
In [1]: import pandas as pd
import seaborn as sb
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report
from sklearn.model_selection import train_test_split
from sklearn import tree
from sklearn.tree import DecisionTreeClassifier
import matplotlib.pyplot as plt
%matplotlib.inline
```

UsageError: Line magic function `%matplotlib.inline` not found.

B)read dataset

```
In [2]: from sklearn.datasets import load_iris
```

```
In [3]: iris=load_iris()
```

```
In [4]: iris.data
```

```
Out[4]: array([[5.1, 3.5, 1.4, 0.2],
 [4.9, 3. , 1.4, 0.2],
 [4.7, 3.2, 1.3, 0.2],
 [4.6, 3.1, 1.5, 0.2],
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 [5.5, 3.5, 1.3, 0.2],
```

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[6.3, 2.9, 5.6, 1.8],  
[6.5, 3. , 5.8, 2.2],  
[7.6, 3. , 6.6, 2.1],
```

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[7.3, 2.9, 6.3, 1.8],
[6.7, 2.5, 5.8, 1.8],
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[6.8, 3. , 5.5, 2.1],
[5.7, 2.5, 5. , 2.],
[5.8, 2.8, 5.1, 2.4],
[6.4, 3.2, 5.3, 2.3],
[6.5, 3. , 5.5, 1.8],
[7.7, 3.8, 6.7, 2.2],
[7.7, 2.6, 6.9, 2.3],
[6. , 2.2, 5. , 1.5],
[6.9, 3.2, 5.7, 2.3],
[5.6, 2.8, 4.9, 2.],
[7.7, 2.8, 6.7, 2.],
[6.3, 2.7, 4.9, 1.8],
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[6.2, 2.8, 4.8, 1.8],
[6.1, 3. , 4.9, 1.8],
[6.4, 2.8, 5.6, 2.1],
[7.2, 3. , 5.8, 1.6],
[7.4, 2.8, 6.1, 1.9],
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[6.3, 2.8, 5.1, 1.5],
[6.1, 2.6, 5.6, 1.4],
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[6.9, 3.1, 5.4, 2.1],
[6.7, 3.1, 5.6, 2.4],
[6.9, 3.1, 5.1, 2.3],
[5.8, 2.7, 5.1, 1.9],
[6.8, 3.2, 5.9, 2.3],
[6.7, 3.3, 5.7, 2.5],
[6.7, 3. , 5.2, 2.3],
[6.3, 2.5, 5. , 1.9],
[6.5, 3. , 5.2, 2.],
[6.2, 3.4, 5.4, 2.3],
[5.9, 3. , 5.1, 1.8]]

```
In [5]: iris.target
```

```
Out[5]: array([[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
               [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
               [0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1],
               [1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1],
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               [2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2])
```

```
In [6]: X = iris.data
        Y = iris.target
```

C)split train test datasets

```
In [7]: X_train ,X_test,Y_train,Y_test = train_test_split(X,Y,random_state=42,test_size=0.30)
```

```
In [8]: X_train
```

```
Out[8]: array([[5.5, 2.4, 3.7, 1. ],
               [6.3, 2.8, 5.1, 1.5],
```

[6.4, 3.1, 5.5, 1.8],
[6.6, 3. , 4.4, 1.4],
[7.2, 3.6, 6.1, 2.5],
[5.7, 2.9, 4.2, 1.3],
[7.6, 3. , 6.6, 2.1],
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[7.7, 2.8, 6.7, 2.],
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[6. , 3. , 4.8, 1.8],
[5.8, 2.7, 5.1, 1.9],
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[5. , 3.4, 1.5, 0.2],
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[5.1, 3.8, 1.6, 0.2],
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```
[5.9, 3. , 4.2, 1.5],
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[5.8, 4. , 1.2, 0.2],
[5.8, 2.6, 4. , 1.2],
[7.1, 3. , 5.9, 2.1]])
```

In [9]: Y_train

Out[9]: array([[1, 2, 2, 1, 2, 1, 2, 1, 0, 2, 1, 0, 0, 0, 1, 2, 0, 0, 0, 1, 0, 1,
2, 0, 1, 2, 0, 2, 2, 1, 1, 2, 1, 0, 1, 2, 0, 0, 1, 1, 0, 2, 0, 0,
1, 1, 2, 1, 2, 2, 1, 0, 0, 2, 2, 0, 0, 0, 1, 2, 0, 2, 2, 0, 1, 1,
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2, 0, 1, 2, 2, 1, 2, 1, 1, 2, 2, 0, 1, 2, 0, 1, 2])

D)feature scaling

```
In [10]: from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.fit_transform(X_test)
```

E)import keras

```
In [11]: import keras
from keras.models import Sequential
from keras.layers import Dense
```

F)initialize ann,adding input layer,hidden layer,output layer,compile the ann.

```
In [12]: ann = Sequential()
```

```
In [13]: ann.add(Dense(units =5, kernel_initializer = 'uniform' , activation = 'relu', input_
```

```
In [14]: ann.add(Dense(units =5, kernel_initializer = 'uniform' , activation = 'relu', input_
In [15]: ann.add(Dense(units =1, kernel_initializer = 'uniform' , activation = 'softmax', inp
In [16]: ann.compile(optimizer = 'adam', loss = 'categorical_crossentropy', metrics = ['accur
In [17]: ann.fit(X_train, Y_train, batch_size = 32, epochs = 100)
```

```
Epoch 1/100
4/4 [=====] - 1s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 2/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 3/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 4/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 5/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 6/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 7/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 8/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 9/100
4/4 [=====] - 0s 6ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 10/100
4/4 [=====] - 0s 6ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 11/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 12/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 13/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 14/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 15/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 16/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 17/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
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Epoch 18/100
4/4 [=====] - 0s 6ms/step - loss: 0.0000e+00 - accuracy: 0.
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Epoch 19/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
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Epoch 20/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
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3524
Epoch 21/100
4/4 [=====] - 0s 9ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 22/100
4/4 [=====] - 0s 1ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 23/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 24/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
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Epoch 25/100
4/4 [=====] - 0s 2ms/step - loss: 0.0000e+00 - accuracy: 0.
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Epoch 26/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
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Epoch 27/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
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Epoch 28/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
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Epoch 29/100
4/4 [=====] - 0s 6ms/step - loss: 0.0000e+00 - accuracy: 0.
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Epoch 30/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
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Epoch 31/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 32/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
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Epoch 33/100
4/4 [=====] - 0s 2ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 34/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 35/100
4/4 [=====] - 0s 2ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 36/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 37/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 38/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 39/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 40/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 41/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 42/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 43/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
```

```
3524
Epoch 44/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 45/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 46/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 47/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 48/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 49/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 50/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 51/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 52/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 53/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 54/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 55/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 56/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 57/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 58/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 59/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 60/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 61/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 62/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 63/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 64/100
4/4 [=====] - 0s 2ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 65/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 66/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
```



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3524
Epoch 67/100
4/4 [=====] - 0s 2ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 68/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 69/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 70/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 71/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 72/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 73/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 74/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 75/100
4/4 [=====] - 0s 2ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 76/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 77/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 78/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 79/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 80/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 81/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 82/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 83/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 84/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 85/100
4/4 [=====] - 0s 2ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 86/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 87/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 88/100
4/4 [=====] - 0s 6ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 89/100
4/4 [=====] - 0s 2ms/step - loss: 0.0000e+00 - accuracy: 0.
```

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3524
Epoch 90/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 91/100
4/4 [=====] - 0s 3ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 92/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 93/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 94/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 95/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 96/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 97/100
4/4 [=====] - 0s 5ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 98/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 99/100
4/4 [=====] - 0s 4ms/step - loss: 0.0000e+00 - accuracy: 0.
3524
Epoch 100/100
4/4 [=====] - 0s 2ms/step - loss: 0.0000e+00 - accuracy: 0.
3524

```

Out[17]: <keras.callbacks.History at 0x29eba1875b0>

G)predict the test results

```

In [18]: Y_pred = ann.predict(X_test)

2/2 [=====] - 0s 0s/step

```

```

In [19]: Y_pred = (Y_pred > 0.5)

```

H)accuracy_score

```

In [20]: from sklearn.metrics import confusion_matrix, accuracy_score
cm = confusion_matrix(Y_test, Y_pred)
print(cm)
accuracy_score(Y_test, Y_pred)
acc=accuracy_score(Y_test, Y_pred)
print(acc)

[[ 0 19  0]
 [ 0 13  0]
 [ 0 13  0]]
0.28888888888888886

```