Syed Ali COSC 750 Pattern Recognition I

Assignment Due May 10th, 2021

1. Problem Statement:

The purpose of this assignment is to understand supervised classification using neural network. From the IRIS data, we need tosplit data into training set and test set. We also need to choose an appropriate architecture for the neural network for three class classifications (Setosa, virginica, versicolor), then:

- a. Classify the test data completely and find accuracy of classification on test data
- b. Save the neural network model and compute predicted class labels on any 5 test data examples and see if they are classified correctly.

2. Neural architecture and justification:

From the IRIS data I will split the data into training set and test set. In this case I am using Keras and sklearn modules for the three class classifications (Setosa, virginica, versicolor). I will be using Convolutional Neural Network (CNN) architecture type since this type of an architecture allows for automatic detection of important features without any human supervision. I will also be using Keras and sklearn since Keras is a user-friendly API and it is very easy to create neural network models with Keras.

Neural Network classification of IRIS using Keras and sklearn modules

```
# import all the modules and functions
import numpy as np
import matplotlib.pyplot as plt
import keras
from keras.models import Sequential
from keras.layers import Dense
from keras.utils import to_categorical
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn import preprocessing
```

Load IRIS data

```
In [2]:
    iris = load_iris()
```

iris is a python dictionary which has data description, keys, data, target, taget_names etc each of which can be extracted as in the following statements

```
In [3]:
    print(iris.keys())
    dict_keys(['data', 'target', 'frame', 'target_names', 'DESCR', 'feature_names', 'filename'))
```

Now knowing the keys, one can get the information about all the using print(iris["name_of_the_key"]) for example if you want feature_names, print(iris["feature_names"])

```
In [4]:
       print(iris['DESCR'])
       .. _iris_dataset:
       Iris plants dataset
       **Data Set Characteristics:**
          :Number of Instances: 150 (50 in each of three classes)
          :Number of Attributes: 4 numeric, predictive attributes and the class
          :Attribute Information:
             - sepal length in cm
             - sepal width in cm
             - petal length in cm
             - petal width in cm
             - class:
                    - Iris-Setosa
                    - Iris-Versicolour
                    - Iris-Virginica
          :Summary Statistics:
          ______ ____
                      Min Max Mean SD Class Correlation
          ______ ____
          sepal length: 4.3 7.9 5.84 0.83 0.7826
```

```
sepal width: 2.0 4.4 3.05 0.43 -0.4194

petal length: 1.0 6.9 3.76 1.76 0.9490 (high!)

petal width: 0.1 2.5 1.20 0.76 0.9565 (high!)

::Missing Attribute Values: None

:Class Distribution: 33.3% for each of 3 classes.

:Creator: R.A. Fisher

:Donor: Michael Marshall (MARSHALL%PLU@io.arc.nasa.gov)

:Date: July, 1988
```

The famous Iris database, first used by Sir R.A. Fisher. The dataset is taken from Fisher's paper. Note that it's the same as in R, but not as in the UCI Machine Learning Repository, which has two wrong data points.

This is perhaps the best known database to be found in the pattern recognition literature. Fisher's paper is a classic in the field and is referenced frequently to this day. (See Duda & Hart, for example.) The data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant. One class is linearly separable from the other 2; the latter are NOT linearly separable from each other.

.. topic:: References

- Fisher, R.A. "The use of multiple measurements in taxonomic problems" Annual Eugenics, 7, Part II, 179-188 (1936); also in "Contributions to Mathematical Statistics" (John Wiley, NY, 1950).
- Duda, R.O., & Hart, P.E. (1973) Pattern Classification and Scene Analysis. (Q327.D83) John Wiley & Sons. ISBN 0-471-22361-1. See page 218.
- Dasarathy, B.V. (1980) "Nosing Around the Neighborhood: A New System Structure and Classification Rule for Recognition in Partially Exposed Environments". IEEE Transactions on Pattern Analysis and Machine Intelligence, Vol. PAMI-2, No. 1, 67-71.
- Gates, G.W. (1972) "The Reduced Nearest Neighbor Rule". IEEE Transactions on Information Theory, May 1972, 431-433.
- See also: 1988 MLC Proceedings, 54-64. Cheeseman et al"s AUTOCLASS II conceptual clustering system finds 3 classes in the data.
- Many, many more ...

Now let us get the data from the dictionary iris

Extact target from iris and convert the target numbers 1,2,3 to categorical 000,010,001

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```

3 a) Test 1: Split the X and Y into training and test sets and let us use a training dataset of 0.6 or 40% using 200 epochs

```
In [8]: X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size = 0.6)
In [9]: print(X_train.shape, X_test.shape)
(60, 4) (90, 4)
```

Now create a model as shown 4-10-3 (4-inputs, 10-neurons, 3-output neurons)

```
In [10]:
    model = Sequential()
    model.add(Dense(10,input_dim=4, activation='relu')) # first-layer added
    model.add(Dense(3,activation='softmax')) #output layer added

In [11]:
    model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
```

Train the model for about 200 epochs and see how it performs on training and test data sets

```
Epoch 2/200
- val loss: 1.0864 - val accuracy: 0.3667
Epoch 3/200
- val loss: 1.0768 - val accuracy: 0.3667
Epoch 4/200
- val_loss: 1.0692 - val_accuracy: 0.3889
Epoch 5/200
- val_loss: 1.0639 - val_accuracy: 0.4000
Epoch 6/200
- val loss: 1.0607 - val accuracy: 0.4222
Epoch 7/200
- val_loss: 1.0592 - val_accuracy: 0.5111
Epoch 8/200
60/60 [============== ] - 0s 150us/step - loss: 1.1088 - accuracy: 0.4167
- val loss: 1.0588 - val accuracy: 0.5556
Epoch 9/200
- val loss: 1.0584 - val accuracy: 0.5778
Epoch 10/200
- val_loss: 1.0583 - val_accuracy: 0.6222
Epoch 11/200
- val loss: 1.0578 - val accuracy: 0.6222
Epoch 12/200
- val loss: 1.0564 - val accuracy: 0.6333
Epoch 13/200
- val_loss: 1.0544 - val_accuracy: 0.6444
Epoch 14/200
- val_loss: 1.0510 - val_accuracy: 0.6444
Epoch 15/200
val loss: 1.0475 - val accuracy: 0.6444
Epoch 16/200
- val loss: 1.0438 - val accuracy: 0.6444
- val loss: 1.0400 - val accuracy: 0.6333
Epoch 18/200
- val_loss: 1.0360 - val_accuracy: 0.6222
Epoch 19/200
- val_loss: 1.0328 - val_accuracy: 0.6222
Epoch 20/200
- val_loss: 1.0295 - val_accuracy: 0.6222
Epoch 21/200
- val_loss: 1.0269 - val_accuracy: 0.6222
- val loss: 1.0248 - val accuracy: 0.6222
Epoch 23/200
```

```
- val loss: 1.0229 - val accuracy: 0.6222
Epoch 24/200
- val loss: 1.0215 - val accuracy: 0.6222
Epoch 25/200
- val loss: 1.0202 - val accuracy: 0.6222
Epoch 26/200
- val_loss: 1.0194 - val_accuracy: 0.6222
Epoch 27/200
- val_loss: 1.0183 - val_accuracy: 0.6222
Epoch 28/200
- val loss: 1.0174 - val accuracy: 0.6333
Epoch 29/200
60/60 [============== ] - 0s 133us/step - loss: 1.0472 - accuracy: 0.6667
- val loss: 1.0159 - val accuracy: 0.6222
Epoch 30/200
- val loss: 1.0145 - val accuracy: 0.6222
Epoch 31/200
- val_loss: 1.0134 - val_accuracy: 0.6222
Epoch 32/200
- val loss: 1.0118 - val accuracy: 0.6222
Epoch 33/200
- val loss: 1.0108 - val accuracy: 0.6222
Epoch 34/200
- val_loss: 1.0095 - val_accuracy: 0.6222
Epoch 35/200
- val_loss: 1.0079 - val_accuracy: 0.6222
Epoch 36/200
- val loss: 1.0065 - val accuracy: 0.6222
Epoch 37/200
- val loss: 1.0052 - val accuracy: 0.6222
Epoch 38/200
- val loss: 1.0036 - val accuracy: 0.6222
Epoch 39/200
- val loss: 1.0021 - val_accuracy: 0.6222
Epoch 40/200
- val loss: 1.0002 - val accuracy: 0.6222
Epoch 41/200
- val_loss: 0.9989 - val_accuracy: 0.6222
Epoch 42/200
- val_loss: 0.9974 - val_accuracy: 0.6222
Epoch 43/200
- val loss: 0.9957 - val accuracy: 0.6222
Epoch 44/200
- val_loss: 0.9936 - val_accuracy: 0.6222
Epoch 45/200
```

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- val loss: 0.9920 - val accuracy: 0.6222
Epoch 46/200
- val loss: 0.9901 - val accuracy: 0.6222
Epoch 47/200
- val loss: 0.9884 - val accuracy: 0.6222
Epoch 48/200
val loss: 0.9868 - val accuracy: 0.6222
Epoch 49/200
- val_loss: 0.9852 - val_accuracy: 0.6222
Epoch 50/200
- val_loss: 0.9828 - val_accuracy: 0.6222
Epoch 51/200
- val loss: 0.9811 - val accuracy: 0.6222
Epoch 52/200
- val loss: 0.9790 - val accuracy: 0.6222
Epoch 53/200
- val loss: 0.9772 - val accuracy: 0.6222
Epoch 54/200
- val loss: 0.9756 - val accuracy: 0.6222
Epoch 55/200
- val loss: 0.9738 - val accuracy: 0.6222
Epoch 56/200
- val_loss: 0.9715 - val_accuracy: 0.6222
Epoch 57/200
- val loss: 0.9693 - val accuracy: 0.6222
Epoch 58/200
val loss: 0.9671 - val accuracy: 0.6222
Epoch 59/200
- val loss: 0.9646 - val accuracy: 0.6222
Epoch 60/200
- val loss: 0.9625 - val accuracy: 0.6222
Epoch 61/200
- val loss: 0.9608 - val accuracy: 0.6222
Epoch 62/200
- val loss: 0.9592 - val accuracy: 0.6222
Epoch 63/200
- val_loss: 0.9573 - val_accuracy: 0.6222
Epoch 64/200
- val_loss: 0.9555 - val_accuracy: 0.6222
Epoch 65/200
- val_loss: 0.9537 - val_accuracy: 0.6222
Epoch 66/200
60/60 [============] - 0s 117us/step - loss: 0.9822 - accuracy: 0.6500
- val loss: 0.9515 - val accuracy: 0.6222
```

```
Epoch 67/200
- val loss: 0.9497 - val accuracy: 0.6222
Epoch 68/200
- val loss: 0.9477 - val accuracy: 0.6222
Epoch 69/200
- val_loss: 0.9463 - val_accuracy: 0.6222
Epoch 70/200
- val_loss: 0.9450 - val_accuracy: 0.6222
Epoch 71/200
- val loss: 0.9434 - val accuracy: 0.6222
Epoch 72/200
- val_loss: 0.9424 - val_accuracy: 0.6222
Epoch 73/200
- val loss: 0.9408 - val accuracy: 0.6333
Epoch 74/200
- val loss: 0.9384 - val accuracy: 0.6222
Epoch 75/200
- val_loss: 0.9368 - val_accuracy: 0.6222
Epoch 76/200
- val loss: 0.9345 - val accuracy: 0.6222
Epoch 77/200
- val loss: 0.9317 - val accuracy: 0.6222
Epoch 78/200
- val_loss: 0.9288 - val_accuracy: 0.6222
Epoch 79/200
- val_loss: 0.9265 - val_accuracy: 0.6222
Epoch 80/200
- val_loss: 0.9247 - val_accuracy: 0.6222
Epoch 81/200
- val loss: 0.9224 - val accuracy: 0.6222
- val loss: 0.9200 - val accuracy: 0.6222
Epoch 83/200
- val_loss: 0.9174 - val_accuracy: 0.6111
Epoch 84/200
- val_loss: 0.9156 - val_accuracy: 0.6222
Epoch 85/200
- val_loss: 0.9133 - val_accuracy: 0.6111
Epoch 86/200
- val_loss: 0.9115 - val_accuracy: 0.6222
- val loss: 0.9092 - val accuracy: 0.6111
Epoch 88/200
```

```
- val loss: 0.9067 - val accuracy: 0.6000
Epoch 89/200
- val loss: 0.9052 - val accuracy: 0.6111
Epoch 90/200
- val loss: 0.9033 - val accuracy: 0.6222
Epoch 91/200
- val_loss: 0.9009 - val_accuracy: 0.6111
Epoch 92/200
- val_loss: 0.8993 - val_accuracy: 0.6222
Epoch 93/200
- val loss: 0.8977 - val accuracy: 0.6222
Epoch 94/200
60/60 [============= ] - 0s 117us/step - loss: 0.9254 - accuracy: 0.6500
- val loss: 0.8956 - val accuracy: 0.6222
Epoch 95/200
- val loss: 0.8935 - val accuracy: 0.6222
Epoch 96/200
- val_loss: 0.8912 - val_accuracy: 0.6222
Epoch 97/200
- val loss: 0.8885 - val accuracy: 0.6222
Epoch 98/200
- val loss: 0.8857 - val accuracy: 0.6222
Epoch 99/200
- val_loss: 0.8842 - val_accuracy: 0.6222
Epoch 100/200
- val_loss: 0.8818 - val_accuracy: 0.6222
Epoch 101/200
- val_loss: 0.8795 - val_accuracy: 0.6222
Epoch 102/200
- val loss: 0.8775 - val accuracy: 0.6222
Epoch 103/200
- val loss: 0.8748 - val accuracy: 0.6222
Epoch 104/200
- val loss: 0.8726 - val accuracy: 0.6222
Epoch 105/200
- val loss: 0.8697 - val accuracy: 0.6111
Epoch 106/200
- val_loss: 0.8675 - val_accuracy: 0.6111
Epoch 107/200
- val_loss: 0.8652 - val_accuracy: 0.6111
Epoch 108/200
60/60 [============ ] - 0s 200us/step - loss: 0.8952 - accuracy: 0.6000
- val loss: 0.8625 - val accuracy: 0.6000
Epoch 109/200
- val loss: 0.8597 - val accuracy: 0.5778
Epoch 110/200
```

```
- val loss: 0.8577 - val accuracy: 0.5778
Epoch 111/200
- val loss: 0.8555 - val accuracy: 0.5889
Epoch 112/200
- val loss: 0.8535 - val accuracy: 0.6000
Epoch 113/200
- val_loss: 0.8510 - val_accuracy: 0.5889
Epoch 114/200
- val_loss: 0.8488 - val_accuracy: 0.5889
Epoch 115/200
- val_loss: 0.8463 - val_accuracy: 0.5778
Epoch 116/200
- val loss: 0.8439 - val accuracy: 0.5778
Epoch 117/200
- val loss: 0.8420 - val accuracy: 0.6000
Epoch 118/200
- val loss: 0.8395 - val accuracy: 0.6000
Epoch 119/200
- val loss: 0.8378 - val accuracy: 0.6222
Epoch 120/200
- val loss: 0.8353 - val accuracy: 0.6222
Epoch 121/200
- val_loss: 0.8333 - val_accuracy: 0.6222
Epoch 122/200
- val_loss: 0.8305 - val_accuracy: 0.6111
Epoch 123/200
- val loss: 0.8277 - val accuracy: 0.6000
Epoch 124/200
- val loss: 0.8254 - val accuracy: 0.6000
Epoch 125/200
- val loss: 0.8231 - val accuracy: 0.6000
Epoch 126/200
- val loss: 0.8207 - val accuracy: 0.6000
Epoch 127/200
- val loss: 0.8176 - val accuracy: 0.5889
Epoch 128/200
- val_loss: 0.8151 - val_accuracy: 0.5889
Epoch 129/200
- val_loss: 0.8130 - val_accuracy: 0.5889
Epoch 130/200
- val_loss: 0.8107 - val_accuracy: 0.6000
Epoch 131/200
60/60 [============] - 0s 133us/step - loss: 0.8425 - accuracy: 0.5833
- val loss: 0.8082 - val accuracy: 0.5889
```

```
Epoch 132/200
- val loss: 0.8060 - val accuracy: 0.6000
Epoch 133/200
- val loss: 0.8035 - val accuracy: 0.6000
Epoch 134/200
- val_loss: 0.8011 - val_accuracy: 0.5889
Epoch 135/200
- val_loss: 0.7981 - val_accuracy: 0.5889
Epoch 136/200
- val loss: 0.7954 - val accuracy: 0.5778
Epoch 137/200
- val_loss: 0.7932 - val_accuracy: 0.5889
Epoch 138/200
- val loss: 0.7909 - val accuracy: 0.5889
Epoch 139/200
- val loss: 0.7886 - val accuracy: 0.5889
Epoch 140/200
- val_loss: 0.7863 - val_accuracy: 0.5889
Epoch 141/200
- val loss: 0.7837 - val accuracy: 0.5778
Epoch 142/200
- val loss: 0.7817 - val accuracy: 0.5889
Epoch 143/200
- val_loss: 0.7791 - val_accuracy: 0.5889
Epoch 144/200
- val_loss: 0.7767 - val_accuracy: 0.5778
Epoch 145/200
val loss: 0.7743 - val accuracy: 0.5778
Epoch 146/200
- val loss: 0.7720 - val accuracy: 0.5778
- val loss: 0.7691 - val accuracy: 0.5667
Epoch 148/200
- val_loss: 0.7667 - val_accuracy: 0.5778
Epoch 149/200
- val_loss: 0.7643 - val_accuracy: 0.5778
Epoch 150/200
- val_loss: 0.7613 - val_accuracy: 0.5667
Epoch 151/200
- val_loss: 0.7591 - val_accuracy: 0.5667
- val loss: 0.7566 - val accuracy: 0.5778
Epoch 153/200
```

```
- val loss: 0.7541 - val accuracy: 0.5778
Epoch 154/200
- val loss: 0.7514 - val accuracy: 0.5667
Epoch 155/200
- val loss: 0.7486 - val accuracy: 0.5556
Epoch 156/200
- val_loss: 0.7459 - val_accuracy: 0.5444
Epoch 157/200
- val_loss: 0.7441 - val_accuracy: 0.5667
Epoch 158/200
- val loss: 0.7418 - val accuracy: 0.5667
Epoch 159/200
- val loss: 0.7395 - val accuracy: 0.5667
Epoch 160/200
- val loss: 0.7371 - val accuracy: 0.5556
Epoch 161/200
- val_loss: 0.7346 - val_accuracy: 0.5444
Epoch 162/200
- val_loss: 0.7322 - val_accuracy: 0.5444
Epoch 163/200
- val loss: 0.7296 - val accuracy: 0.5444
Epoch 164/200
- val_loss: 0.7273 - val_accuracy: 0.5444
Epoch 165/200
- val_loss: 0.7251 - val_accuracy: 0.5556
Epoch 166/200
- val_loss: 0.7227 - val_accuracy: 0.5556
Epoch 167/200
60/60 [============== ] - 0s 117us/step - loss: 0.7577 - accuracy: 0.5000
- val loss: 0.7201 - val accuracy: 0.5333
Epoch 168/200
- val loss: 0.7175 - val accuracy: 0.5000
Epoch 169/200
- val_loss: 0.7152 - val_accuracy: 0.5000
Epoch 170/200
- val loss: 0.7133 - val accuracy: 0.5222
Epoch 171/200
- val_loss: 0.7107 - val_accuracy: 0.5000
Epoch 172/200
- val_loss: 0.7085 - val_accuracy: 0.5000
Epoch 173/200
- val loss: 0.7064 - val accuracy: 0.5111
Epoch 174/200
- val loss: 0.7047 - val accuracy: 0.5444
Epoch 175/200
```

```
- val loss: 0.7027 - val accuracy: 0.5778
Epoch 176/200
- val loss: 0.7005 - val accuracy: 0.5667
Epoch 177/200
- val loss: 0.6984 - val accuracy: 0.5889
Epoch 178/200
- val_loss: 0.6967 - val_accuracy: 0.6111
Epoch 179/200
- val_loss: 0.6942 - val_accuracy: 0.6000
Epoch 180/200
- val_loss: 0.6916 - val_accuracy: 0.5778
Epoch 181/200
- val loss: 0.6896 - val accuracy: 0.5889
Epoch 182/200
- val loss: 0.6871 - val accuracy: 0.5778
Epoch 183/200
- val loss: 0.6846 - val accuracy: 0.5778
Epoch 184/200
- val loss: 0.6826 - val accuracy: 0.5778
Epoch 185/200
- val loss: 0.6803 - val accuracy: 0.5667
Epoch 186/200
- val_loss: 0.6778 - val_accuracy: 0.5556
Epoch 187/200
- val loss: 0.6759 - val accuracy: 0.5778
Epoch 188/200
val loss: 0.6736 - val accuracy: 0.5778
Epoch 189/200
- val loss: 0.6716 - val accuracy: 0.5889
Epoch 190/200
- val loss: 0.6697 - val accuracy: 0.6000
Epoch 191/200
- val_loss: 0.6672 - val_accuracy: 0.5778
Epoch 192/200
- val loss: 0.6652 - val accuracy: 0.5889
Epoch 193/200
- val_loss: 0.6627 - val_accuracy: 0.5889
Epoch 194/200
- val_loss: 0.6602 - val_accuracy: 0.5667
Epoch 195/200
- val_loss: 0.6580 - val_accuracy: 0.5667
Epoch 196/200
60/60 [============] - 0s 183us/step - loss: 0.6939 - accuracy: 0.6000
- val loss: 0.6556 - val accuracy: 0.5778
```

In order to get better performance, the input data has to be normalized so that all features are weighted equally in classification

```
In [14]:
    X = preprocessing.scale(iris['data'])
    Y = to_categorical(iris['target'])
```

3 b) Test 2: Split the X and Y into training and test sets and let us use a training dataset of 0.7 or 30% with the same number of epochs

Now create a model as shown neural model 4-10-3 (4-inputs, 10-neurons, 3-output neurons)¶

Train the model for about 200 epochs and see

how it performs on training and test data sets

In [19]: model.fit(X_train, Y_train, validation_data=(X_test,Y_test), epochs=200)

```
Train on 45 samples, validate on 105 samples
Epoch 1/200
val loss: 1.1171 - val accuracy: 0.3810
Epoch 2/200
- val loss: 1.1052 - val accuracy: 0.3810
- val loss: 1.0930 - val accuracy: 0.4095
Epoch 4/200
- val_loss: 1.0809 - val_accuracy: 0.4286
Epoch 5/200
- val_loss: 1.0688 - val_accuracy: 0.4381
Epoch 6/200
- val_loss: 1.0569 - val_accuracy: 0.4667
Epoch 7/200
- val loss: 1.0450 - val accuracy: 0.4762
- val loss: 1.0333 - val accuracy: 0.4762
Epoch 9/200
- val_loss: 1.0216 - val_accuracy: 0.4952
Epoch 10/200
- val_loss: 1.0102 - val_accuracy: 0.5238
Epoch 11/200
- val loss: 0.9989 - val accuracy: 0.5524
Epoch 12/200
- val_loss: 0.9879 - val_accuracy: 0.5524
- val_loss: 0.9769 - val_accuracy: 0.5714
Epoch 14/200
- val loss: 0.9662 - val accuracy: 0.5905
Epoch 15/200
- val_loss: 0.9555 - val_accuracy: 0.6000
Epoch 16/200
- val loss: 0.9450 - val accuracy: 0.6286
Epoch 17/200
- val_loss: 0.9346 - val_accuracy: 0.6381
- val_loss: 0.9244 - val_accuracy: 0.6381
Epoch 19/200
- val_loss: 0.9144 - val_accuracy: 0.6571
Epoch 20/200
```

```
- val loss: 0.9046 - val accuracy: 0.6571
Epoch 21/200
- val loss: 0.8949 - val accuracy: 0.6571
Epoch 22/200
- val loss: 0.8854 - val accuracy: 0.6667
Epoch 23/200
- val_loss: 0.8761 - val_accuracy: 0.6667
Epoch 24/200
- val_loss: 0.8669 - val_accuracy: 0.6667
Epoch 25/200
- val_loss: 0.8579 - val_accuracy: 0.6667
Epoch 26/200
- val loss: 0.8491 - val accuracy: 0.6667
Epoch 27/200
- val loss: 0.8402 - val accuracy: 0.6667
Epoch 28/200
val loss: 0.8314 - val accuracy: 0.6667
Epoch 29/200
- val loss: 0.8226 - val accuracy: 0.6667
Epoch 30/200
- val loss: 0.8140 - val accuracy: 0.6667
Epoch 31/200
- val_loss: 0.8055 - val_accuracy: 0.6667
Epoch 32/200
- val loss: 0.7972 - val accuracy: 0.6667
Epoch 33/200
val loss: 0.7890 - val accuracy: 0.6667
Epoch 34/200
- val loss: 0.7809 - val accuracy: 0.6667
Epoch 35/200
- val loss: 0.7729 - val accuracy: 0.6667
Epoch 36/200
- val loss: 0.7651 - val accuracy: 0.6667
Epoch 37/200
- val loss: 0.7575 - val accuracy: 0.6667
Epoch 38/200
- val_loss: 0.7499 - val_accuracy: 0.6667
Epoch 39/200
- val_loss: 0.7426 - val_accuracy: 0.6667
Epoch 40/200
- val_loss: 0.7354 - val_accuracy: 0.6667
Epoch 41/200
- val loss: 0.7282 - val accuracy: 0.6667
```

```
Epoch 42/200
- val loss: 0.7212 - val accuracy: 0.6667
Epoch 43/200
- val loss: 0.7143 - val accuracy: 0.6667
Epoch 44/200
- val_loss: 0.7076 - val_accuracy: 0.6667
Epoch 45/200
- val_loss: 0.7008 - val_accuracy: 0.6667
Epoch 46/200
- val loss: 0.6941 - val accuracy: 0.6667
Epoch 47/200
- val_loss: 0.6876 - val_accuracy: 0.6667
Epoch 48/200
- val loss: 0.6813 - val accuracy: 0.6667
Epoch 49/200
- val loss: 0.6751 - val accuracy: 0.6667
Epoch 50/200
- val_loss: 0.6691 - val_accuracy: 0.6667
Epoch 51/200
- val loss: 0.6632 - val accuracy: 0.6667
Epoch 52/200
- val loss: 0.6574 - val accuracy: 0.6667
Epoch 53/200
- val_loss: 0.6516 - val_accuracy: 0.6667
Epoch 54/200
- val_loss: 0.6460 - val_accuracy: 0.6667
Epoch 55/200
val loss: 0.6405 - val accuracy: 0.6667
Epoch 56/200
- val loss: 0.6350 - val accuracy: 0.6762
- val loss: 0.6296 - val accuracy: 0.6762
Epoch 58/200
- val loss: 0.6243 - val accuracy: 0.6762
Epoch 59/200
- val_loss: 0.6191 - val_accuracy: 0.6762
Epoch 60/200
- val_loss: 0.6140 - val_accuracy: 0.6857
Epoch 61/200
- val_loss: 0.6090 - val_accuracy: 0.6952
- val loss: 0.6040 - val accuracy: 0.6952
Epoch 63/200
```

```
- val loss: 0.5992 - val accuracy: 0.6952
Epoch 64/200
- val loss: 0.5945 - val accuracy: 0.6952
Epoch 65/200
- val loss: 0.5899 - val accuracy: 0.6952
Epoch 66/200
- val_loss: 0.5854 - val_accuracy: 0.6952
Epoch 67/200
- val_loss: 0.5810 - val_accuracy: 0.6952
Epoch 68/200
- val loss: 0.5767 - val accuracy: 0.6952
Epoch 69/200
- val loss: 0.5724 - val accuracy: 0.6952
Epoch 70/200
- val loss: 0.5682 - val accuracy: 0.7048
Epoch 71/200
- val_loss: 0.5642 - val_accuracy: 0.7143
Epoch 72/200
- val loss: 0.5602 - val accuracy: 0.7143
Epoch 73/200
- val loss: 0.5563 - val accuracy: 0.7143
Epoch 74/200
- val_loss: 0.5525 - val_accuracy: 0.7143
Epoch 75/200
- val_loss: 0.5488 - val_accuracy: 0.7143
Epoch 76/200
- val_loss: 0.5451 - val_accuracy: 0.7238
Epoch 77/200
- val_loss: 0.5414 - val_accuracy: 0.7238
Epoch 78/200
- val loss: 0.5379 - val accuracy: 0.7238
Epoch 79/200
- val_loss: 0.5343 - val_accuracy: 0.7238
- val loss: 0.5309 - val accuracy: 0.7238
Epoch 81/200
- val_loss: 0.5275 - val_accuracy: 0.7238
Epoch 82/200
- val_loss: 0.5241 - val_accuracy: 0.7238
Epoch 83/200
- val loss: 0.5209 - val accuracy: 0.7333
Epoch 84/200
- val loss: 0.5177 - val accuracy: 0.7333
Epoch 85/200
```

```
- val loss: 0.5145 - val accuracy: 0.7429
Epoch 86/200
- val loss: 0.5113 - val accuracy: 0.7429
Epoch 87/200
- val loss: 0.5081 - val accuracy: 0.7429
Epoch 88/200
val loss: 0.5050 - val accuracy: 0.7429
Epoch 89/200
- val_loss: 0.5020 - val_accuracy: 0.7524
Epoch 90/200
- val_loss: 0.4990 - val_accuracy: 0.7714
Epoch 91/200
- val loss: 0.4961 - val accuracy: 0.7714
Epoch 92/200
- val loss: 0.4931 - val accuracy: 0.7714
Epoch 93/200
- val loss: 0.4902 - val accuracy: 0.7810
Epoch 94/200
- val loss: 0.4873 - val accuracy: 0.7810
Epoch 95/200
- val loss: 0.4845 - val accuracy: 0.7810
Epoch 96/200
- val_loss: 0.4817 - val_accuracy: 0.7810
Epoch 97/200
- val loss: 0.4789 - val accuracy: 0.7905
Epoch 98/200
val loss: 0.4762 - val accuracy: 0.7905
Epoch 99/200
- val loss: 0.4736 - val accuracy: 0.7905
Epoch 100/200
- val loss: 0.4711 - val accuracy: 0.7905
Epoch 101/200
- val loss: 0.4685 - val accuracy: 0.7905
Epoch 102/200
- val loss: 0.4659 - val accuracy: 0.7905
Epoch 103/200
- val_loss: 0.4634 - val_accuracy: 0.8000
Epoch 104/200
- val_loss: 0.4610 - val_accuracy: 0.8095
Epoch 105/200
- val_loss: 0.4586 - val_accuracy: 0.8095
Epoch 106/200
- val loss: 0.4563 - val accuracy: 0.8190
```

```
Epoch 107/200
- val loss: 0.4539 - val accuracy: 0.8190
Epoch 108/200
- val loss: 0.4516 - val accuracy: 0.8190
Epoch 109/200
- val_loss: 0.4493 - val_accuracy: 0.8190
Epoch 110/200
- val_loss: 0.4470 - val_accuracy: 0.8190
Epoch 111/200
- val loss: 0.4447 - val accuracy: 0.8190
Epoch 112/200
- val_loss: 0.4424 - val_accuracy: 0.8190
Epoch 113/200
- val loss: 0.4402 - val accuracy: 0.8190
Epoch 114/200
- val loss: 0.4380 - val accuracy: 0.8095
Epoch 115/200
- val_loss: 0.4358 - val_accuracy: 0.8095
Epoch 116/200
- val loss: 0.4338 - val accuracy: 0.8190
Epoch 117/200
- val loss: 0.4317 - val accuracy: 0.8190
Epoch 118/200
- val_loss: 0.4297 - val_accuracy: 0.8190
Epoch 119/200
- val_loss: 0.4277 - val_accuracy: 0.8190
Epoch 120/200
- val_loss: 0.4257 - val_accuracy: 0.8190
Epoch 121/200
- val loss: 0.4237 - val accuracy: 0.8190
- val loss: 0.4217 - val accuracy: 0.8190
Epoch 123/200
- val_loss: 0.4198 - val_accuracy: 0.8190
Epoch 124/200
- val_loss: 0.4178 - val_accuracy: 0.8286
Epoch 125/200
- val_loss: 0.4159 - val_accuracy: 0.8286
Epoch 126/200
- val_loss: 0.4140 - val_accuracy: 0.8381
- val loss: 0.4122 - val accuracy: 0.8381
Epoch 128/200
```

```
- val loss: 0.4103 - val accuracy: 0.8476
Epoch 129/200
- val loss: 0.4085 - val accuracy: 0.8476
Epoch 130/200
- val loss: 0.4067 - val accuracy: 0.8476
Epoch 131/200
- val_loss: 0.4050 - val_accuracy: 0.8476
Epoch 132/200
- val_loss: 0.4033 - val_accuracy: 0.8476
Epoch 133/200
- val loss: 0.4016 - val accuracy: 0.8476
Epoch 134/200
- val loss: 0.3999 - val accuracy: 0.8476
Epoch 135/200
- val loss: 0.3983 - val accuracy: 0.8476
Epoch 136/200
- val_loss: 0.3968 - val_accuracy: 0.8476
Epoch 137/200
- val_loss: 0.3952 - val_accuracy: 0.8476
Epoch 138/200
- val loss: 0.3936 - val accuracy: 0.8476
Epoch 139/200
- val loss: 0.3920 - val accuracy: 0.8476
Epoch 140/200
- val_loss: 0.3904 - val_accuracy: 0.8476
Epoch 141/200
- val_loss: 0.3887 - val_accuracy: 0.8476
Epoch 142/200
- val loss: 0.3871 - val accuracy: 0.8476
Epoch 143/200
- val loss: 0.3854 - val accuracy: 0.8476
Epoch 144/200
- val loss: 0.3838 - val_accuracy: 0.8571
Epoch 145/200
- val loss: 0.3822 - val accuracy: 0.8476
Epoch 146/200
- val_loss: 0.3806 - val_accuracy: 0.8476
Epoch 147/200
- val_loss: 0.3791 - val_accuracy: 0.8476
Epoch 148/200
- val loss: 0.3775 - val accuracy: 0.8476
Epoch 149/200
- val_loss: 0.3761 - val_accuracy: 0.8476
Epoch 150/200
```

```
- val loss: 0.3746 - val accuracy: 0.8476
Epoch 151/200
- val loss: 0.3732 - val accuracy: 0.8476
Epoch 152/200
- val loss: 0.3718 - val accuracy: 0.8476
Epoch 153/200
- val_loss: 0.3705 - val_accuracy: 0.8476
Epoch 154/200
- val_loss: 0.3691 - val_accuracy: 0.8476
Epoch 155/200
- val_loss: 0.3678 - val_accuracy: 0.8476
Epoch 156/200
- val loss: 0.3664 - val accuracy: 0.8476
Epoch 157/200
- val loss: 0.3651 - val accuracy: 0.8476
Epoch 158/200
val loss: 0.3638 - val accuracy: 0.8476
Epoch 159/200
- val loss: 0.3625 - val accuracy: 0.8476
Epoch 160/200
- val loss: 0.3612 - val accuracy: 0.8476
Epoch 161/200
- val_loss: 0.3598 - val_accuracy: 0.8476
Epoch 162/200
- val_loss: 0.3585 - val_accuracy: 0.8476
Epoch 163/200
- val loss: 0.3572 - val accuracy: 0.8476
Epoch 164/200
- val loss: 0.3558 - val accuracy: 0.8476
Epoch 165/200
- val loss: 0.3545 - val accuracy: 0.8476
Epoch 166/200
- val loss: 0.3531 - val accuracy: 0.8476
Epoch 167/200
- val loss: 0.3518 - val accuracy: 0.8476
Epoch 168/200
- val_loss: 0.3504 - val_accuracy: 0.8476
Epoch 169/200
- val_loss: 0.3491 - val_accuracy: 0.8476
Epoch 170/200
- val_loss: 0.3479 - val_accuracy: 0.8571
Epoch 171/200
45/45 [==============] - 0s 289us/step - loss: 0.3466 - accuracy: 0.9111
- val loss: 0.3466 - val accuracy: 0.8571
```

```
Epoch 172/200
- val loss: 0.3454 - val accuracy: 0.8571
Epoch 173/200
- val loss: 0.3442 - val accuracy: 0.8667
Epoch 174/200
- val_loss: 0.3431 - val_accuracy: 0.8667
Epoch 175/200
- val_loss: 0.3420 - val_accuracy: 0.8667
Epoch 176/200
- val loss: 0.3410 - val accuracy: 0.8667
Epoch 177/200
- val loss: 0.3399 - val accuracy: 0.8667
Epoch 178/200
- val loss: 0.3388 - val accuracy: 0.8667
Epoch 179/200
- val_loss: 0.3377 - val_accuracy: 0.8667
Epoch 180/200
- val_loss: 0.3367 - val_accuracy: 0.8571
Epoch 181/200
- val loss: 0.3357 - val accuracy: 0.8571
Epoch 182/200
- val loss: 0.3348 - val accuracy: 0.8571
Epoch 183/200
- val_loss: 0.3338 - val_accuracy: 0.8571
Epoch 184/200
- val_loss: 0.3329 - val_accuracy: 0.8571
Epoch 185/200
- val_loss: 0.3319 - val_accuracy: 0.8476
Epoch 186/200
- val loss: 0.3309 - val accuracy: 0.8476
- val loss: 0.3300 - val accuracy: 0.8476
Epoch 188/200
- val_loss: 0.3291 - val_accuracy: 0.8476
Epoch 189/200
- val_loss: 0.3281 - val_accuracy: 0.8476
Epoch 190/200
- val_loss: 0.3271 - val_accuracy: 0.8476
Epoch 191/200
- val_loss: 0.3260 - val_accuracy: 0.8476
- val loss: 0.3249 - val accuracy: 0.8571
Epoch 193/200
```

```
- val loss: 0.3238 - val accuracy: 0.8571
    Epoch 194/200
    - val loss: 0.3227 - val accuracy: 0.8667
    Epoch 195/200
    - val loss: 0.3216 - val accuracy: 0.8667
    Epoch 196/200
    - val_loss: 0.3206 - val_accuracy: 0.8667
    Epoch 197/200
    Ous/step - loss: 0.3113 - accuracy: 0.9111 - val_loss: 0.3196 - val_accuracy: 0.8667
    Epoch 198/200
    - val_loss: 0.3186 - val_accuracy: 0.8762
    Epoch 199/200
    - val loss: 0.3176 - val accuracy: 0.8762
    Epoch 200/200
    - val loss: 0.3166 - val accuracy: 0.8762
Out[19]: <keras.callbacks.callbacks.History at 0x1d2f5981630>
In [20]:
    w = model.get weights()
```

In order to get better performance, the input data has to be normalized so that all features are weighted equally in classification

```
In [21]:
    X = preprocessing.scale(iris['data'])
    Y = to_categorical(iris['target'])
```

3 c) Test 3: Split the X and Y into training and test sets and let us use a training dataset of 0.8 or 20% with the same number of epochs

Now create a model as shown neural model 4-10-3 (4-inputs, 10-neurons, 3-output neurons)

```
In [24]: model = Sequential()
```

```
model.add(Dense(10,input_dim=4, activation='relu')) # first-layer added
model.add(Dense(3,activation='softmax')) #output Layer added

In [25]:
model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
```

Train the model for about 200 epochs and see how it performs on training and test data sets

```
In [26]:
    model.fit(X train, Y train, validation data=(X test,Y test), epochs=200)
    Train on 30 samples, validate on 120 samples
    Epoch 1/200
    30/30 [============] - 0s 4ms/step - loss: 1.1843 - accuracy: 0.4667 -
    val_loss: 1.1537 - val_accuracy: 0.3750
    Epoch 2/200
    - val loss: 1.1493 - val accuracy: 0.3833
    Epoch 3/200
    - val loss: 1.1448 - val accuracy: 0.4167
    Epoch 4/200
    - val_loss: 1.1403 - val_accuracy: 0.4250
    Epoch 5/200
    - val loss: 1.1359 - val accuracy: 0.4250
    Epoch 6/200
    - val_loss: 1.1316 - val_accuracy: 0.4333
    Epoch 7/200
    - val_loss: 1.1272 - val_accuracy: 0.4333
    Epoch 8/200
    - val_loss: 1.1228 - val_accuracy: 0.4333
    Epoch 9/200
    - val_loss: 1.1185 - val_accuracy: 0.4333
    Epoch 10/200
    - val loss: 1.1142 - val accuracy: 0.4333
    Epoch 11/200
    - val loss: 1.1099 - val accuracy: 0.4333
    Epoch 12/200
    - val loss: 1.1057 - val accuracy: 0.4417
    Epoch 13/200
    - val loss: 1.1014 - val accuracy: 0.4500
    Epoch 14/200
    - val_loss: 1.0972 - val_accuracy: 0.4583
    Epoch 15/200
    - val_loss: 1.0930 - val_accuracy: 0.4667
    Epoch 16/200
```

```
- val loss: 1.0888 - val accuracy: 0.4667
Epoch 17/200
- val loss: 1.0846 - val accuracy: 0.4917
Epoch 18/200
- val loss: 1.0805 - val accuracy: 0.5000
Epoch 19/200
- val_loss: 1.0764 - val_accuracy: 0.5000
Epoch 20/200
- val_loss: 1.0723 - val_accuracy: 0.5083
Epoch 21/200
- val loss: 1.0682 - val accuracy: 0.5250
Epoch 22/200
30/30 [============== ] - 0s 367us/step - loss: 1.0590 - accuracy: 0.6000
- val loss: 1.0642 - val accuracy: 0.5250
Epoch 23/200
- val loss: 1.0602 - val accuracy: 0.5250
Epoch 24/200
- val_loss: 1.0561 - val_accuracy: 0.5333
Epoch 25/200
- val_loss: 1.0521 - val_accuracy: 0.5417
Epoch 26/200
- val loss: 1.0481 - val accuracy: 0.5417
Epoch 27/200
- val_loss: 1.0442 - val_accuracy: 0.5417
Epoch 28/200
- val_loss: 1.0403 - val_accuracy: 0.5417
Epoch 29/200
- val_loss: 1.0363 - val_accuracy: 0.5417
Epoch 30/200
- val loss: 1.0325 - val accuracy: 0.5500
Epoch 31/200
- val loss: 1.0286 - val accuracy: 0.5500
Epoch 32/200
- val loss: 1.0247 - val_accuracy: 0.5500
Epoch 33/200
- val loss: 1.0209 - val accuracy: 0.5500
Epoch 34/200
- val_loss: 1.0170 - val_accuracy: 0.5417
Epoch 35/200
- val_loss: 1.0132 - val_accuracy: 0.5500
Epoch 36/200
- val loss: 1.0094 - val accuracy: 0.5500
Epoch 37/200
- val loss: 1.0056 - val accuracy: 0.5417
Epoch 38/200
```

```
- val loss: 1.0018 - val accuracy: 0.5417
Epoch 39/200
- val loss: 0.9981 - val accuracy: 0.5417
Epoch 40/200
- val loss: 0.9944 - val accuracy: 0.5500
Epoch 41/200
val loss: 0.9908 - val accuracy: 0.5583
Epoch 42/200
- val_loss: 0.9871 - val_accuracy: 0.5583
Epoch 43/200
- val_loss: 0.9835 - val_accuracy: 0.5667
Epoch 44/200
- val loss: 0.9799 - val accuracy: 0.5750
Epoch 45/200
- val loss: 0.9763 - val accuracy: 0.5917
Epoch 46/200
- val loss: 0.9728 - val accuracy: 0.5917
Epoch 47/200
- val loss: 0.9693 - val accuracy: 0.5833
Epoch 48/200
- val loss: 0.9658 - val accuracy: 0.5917
Epoch 49/200
- val_loss: 0.9622 - val_accuracy: 0.6000
Epoch 50/200
- val loss: 0.9588 - val accuracy: 0.6000
Epoch 51/200
- val loss: 0.9553 - val accuracy: 0.6000
Epoch 52/200
- val loss: 0.9519 - val accuracy: 0.6000
Epoch 53/200
- val loss: 0.9484 - val accuracy: 0.6167
Epoch 54/200
- val loss: 0.9450 - val accuracy: 0.6167
Epoch 55/200
- val loss: 0.9417 - val accuracy: 0.6167
Epoch 56/200
- val_loss: 0.9383 - val_accuracy: 0.6167
Epoch 57/200
- val_loss: 0.9350 - val_accuracy: 0.6167
Epoch 58/200
- val_loss: 0.9317 - val_accuracy: 0.6167
Epoch 59/200
30/30 [=============] - 0s 333us/step - loss: 0.8678 - accuracy: 0.6333
- val loss: 0.9284 - val accuracy: 0.6250
```

```
Epoch 60/200
- val loss: 0.9251 - val accuracy: 0.6333
Epoch 61/200
- val loss: 0.9218 - val accuracy: 0.6333
Epoch 62/200
- val_loss: 0.9186 - val_accuracy: 0.6333
Epoch 63/200
- val_loss: 0.9154 - val_accuracy: 0.6333
Epoch 64/200
- val loss: 0.9122 - val accuracy: 0.6333
Epoch 65/200
- val loss: 0.9090 - val accuracy: 0.6500
Epoch 66/200
- val loss: 0.9058 - val accuracy: 0.6500
Epoch 67/200
- val loss: 0.9026 - val accuracy: 0.6500
Epoch 68/200
- val_loss: 0.8995 - val_accuracy: 0.6583
Epoch 69/200
- val loss: 0.8964 - val accuracy: 0.6583
Epoch 70/200
- val loss: 0.8933 - val accuracy: 0.6667
Epoch 71/200
- val_loss: 0.8902 - val_accuracy: 0.6833
Epoch 72/200
- val_loss: 0.8871 - val_accuracy: 0.6917
Epoch 73/200
val loss: 0.8841 - val accuracy: 0.7000
Epoch 74/200
- val loss: 0.8810 - val accuracy: 0.7000
- val loss: 0.8780 - val accuracy: 0.7083
Epoch 76/200
- val_loss: 0.8751 - val_accuracy: 0.7083
Epoch 77/200
- val_loss: 0.8721 - val_accuracy: 0.7083
Epoch 78/200
- val_loss: 0.8691 - val_accuracy: 0.7083
Epoch 79/200
- val_loss: 0.8662 - val_accuracy: 0.7083
- val loss: 0.8633 - val accuracy: 0.7083
Epoch 81/200
```

```
- val loss: 0.8604 - val accuracy: 0.7083
Epoch 82/200
- val loss: 0.8575 - val accuracy: 0.7083
Epoch 83/200
- val loss: 0.8546 - val accuracy: 0.7250
Epoch 84/200
- val_loss: 0.8518 - val_accuracy: 0.7250
Epoch 85/200
- val loss: 0.8490 - val accuracy: 0.7250
Epoch 86/200
- val_loss: 0.8461 - val_accuracy: 0.7250
Epoch 87/200
- val loss: 0.8433 - val accuracy: 0.7250
Epoch 88/200
- val loss: 0.8406 - val accuracy: 0.7250
Epoch 89/200
- val_loss: 0.8378 - val_accuracy: 0.7250
Epoch 90/200
- val loss: 0.8350 - val accuracy: 0.7250
Epoch 91/200
- val loss: 0.8323 - val accuracy: 0.7167
Epoch 92/200
- val_loss: 0.8296 - val_accuracy: 0.7250
Epoch 93/200
- val_loss: 0.8269 - val_accuracy: 0.7250
Epoch 94/200
- val_loss: 0.8242 - val_accuracy: 0.7333
Epoch 95/200
- val loss: 0.8215 - val accuracy: 0.7333
Epoch 96/200
- val loss: 0.8189 - val accuracy: 0.7333
Epoch 97/200
- val loss: 0.8163 - val accuracy: 0.7417
Epoch 98/200
- val loss: 0.8137 - val accuracy: 0.7417
Epoch 99/200
- val_loss: 0.8112 - val_accuracy: 0.7333
Epoch 100/200
- val_loss: 0.8086 - val_accuracy: 0.7333
Epoch 101/200
- val loss: 0.8060 - val accuracy: 0.7333
Epoch 102/200
- val loss: 0.8035 - val accuracy: 0.7417
Epoch 103/200
```

```
- val loss: 0.8010 - val accuracy: 0.7417
Epoch 104/200
- val loss: 0.7985 - val accuracy: 0.7417
Epoch 105/200
- val loss: 0.7960 - val accuracy: 0.7417
Epoch 106/200
- val_loss: 0.7935 - val_accuracy: 0.7417
Epoch 107/200
- val_loss: 0.7910 - val_accuracy: 0.7417
Epoch 108/200
- val_loss: 0.7886 - val_accuracy: 0.7417
Epoch 109/200
- val loss: 0.7862 - val accuracy: 0.7417
Epoch 110/200
- val loss: 0.7838 - val accuracy: 0.7417
Epoch 111/200
- val loss: 0.7814 - val accuracy: 0.7417
Epoch 112/200
- val loss: 0.7790 - val accuracy: 0.7500
Epoch 113/200
- val loss: 0.7766 - val accuracy: 0.7583
Epoch 114/200
- val_loss: 0.7743 - val_accuracy: 0.7583
Epoch 115/200
- val_loss: 0.7719 - val_accuracy: 0.7583
Epoch 116/200
- val loss: 0.7696 - val accuracy: 0.7583
Epoch 117/200
- val loss: 0.7672 - val accuracy: 0.7583
Epoch 118/200
- val loss: 0.7649 - val accuracy: 0.7583
Epoch 119/200
- val_loss: 0.7627 - val_accuracy: 0.7583
Epoch 120/200
- val loss: 0.7604 - val accuracy: 0.7583
Epoch 121/200
- val_loss: 0.7581 - val_accuracy: 0.7583
Epoch 122/200
- val_loss: 0.7559 - val_accuracy: 0.7667
Epoch 123/200
- val loss: 0.7536 - val accuracy: 0.7667
Epoch 124/200
30/30 [============] - 0s 267us/step - loss: 0.6131 - accuracy: 0.8667
- val loss: 0.7514 - val accuracy: 0.7667
```

```
Epoch 125/200
- val loss: 0.7492 - val accuracy: 0.7667
Epoch 126/200
- val loss: 0.7470 - val accuracy: 0.7583
Epoch 127/200
- val_loss: 0.7448 - val_accuracy: 0.7583
Epoch 128/200
- val_loss: 0.7427 - val_accuracy: 0.7583
Epoch 129/200
- val loss: 0.7405 - val accuracy: 0.7500
Epoch 130/200
- val_loss: 0.7384 - val_accuracy: 0.7500
Epoch 131/200
- val loss: 0.7363 - val accuracy: 0.7500
Epoch 132/200
- val_loss: 0.7341 - val_accuracy: 0.7500
Epoch 133/200
- val_loss: 0.7320 - val_accuracy: 0.7500
Epoch 134/200
- val loss: 0.7299 - val accuracy: 0.7500
Epoch 135/200
- val loss: 0.7278 - val accuracy: 0.7500
Epoch 136/200
- val_loss: 0.7258 - val_accuracy: 0.7500
Epoch 137/200
- val_loss: 0.7237 - val_accuracy: 0.7500
Epoch 138/200
- val_loss: 0.7217 - val_accuracy: 0.7500
Epoch 139/200
- val loss: 0.7196 - val accuracy: 0.7500
- val loss: 0.7176 - val accuracy: 0.7500
Epoch 141/200
- val_loss: 0.7156 - val_accuracy: 0.7500
Epoch 142/200
- val_loss: 0.7136 - val_accuracy: 0.7500
Epoch 143/200
- val_loss: 0.7116 - val_accuracy: 0.7500
Epoch 144/200
- val_loss: 0.7096 - val_accuracy: 0.7500
- val loss: 0.7077 - val accuracy: 0.7500
Epoch 146/200
```

```
- val loss: 0.7057 - val accuracy: 0.7500
Epoch 147/200
- val loss: 0.7037 - val accuracy: 0.7500
Epoch 148/200
- val loss: 0.7018 - val accuracy: 0.7500
Epoch 149/200
- val_loss: 0.6998 - val_accuracy: 0.7500
Epoch 150/200
- val_loss: 0.6979 - val_accuracy: 0.7500
Epoch 151/200
- val loss: 0.6959 - val accuracy: 0.7500
Epoch 152/200
- val loss: 0.6940 - val accuracy: 0.7500
Epoch 153/200
- val loss: 0.6921 - val accuracy: 0.7500
Epoch 154/200
- val_loss: 0.6902 - val_accuracy: 0.7500
Epoch 155/200
- val loss: 0.6883 - val accuracy: 0.7500
Epoch 156/200
- val loss: 0.6864 - val accuracy: 0.7500
Epoch 157/200
- val_loss: 0.6845 - val_accuracy: 0.7500
Epoch 158/200
- val_loss: 0.6826 - val_accuracy: 0.7417
Epoch 159/200
- val loss: 0.6808 - val accuracy: 0.7417
Epoch 160/200
- val loss: 0.6789 - val accuracy: 0.7417
Epoch 161/200
- val loss: 0.6771 - val accuracy: 0.7417
Epoch 162/200
- val_loss: 0.6753 - val_accuracy: 0.7417
Epoch 163/200
- val loss: 0.6735 - val accuracy: 0.7417
Epoch 164/200
- val_loss: 0.6717 - val_accuracy: 0.7417
Epoch 165/200
- val_loss: 0.6699 - val_accuracy: 0.7417
Epoch 166/200
- val loss: 0.6681 - val accuracy: 0.7417
Epoch 167/200
- val loss: 0.6663 - val accuracy: 0.7417
Epoch 168/200
```

```
- val loss: 0.6645 - val accuracy: 0.7417
Epoch 169/200
- val loss: 0.6628 - val accuracy: 0.7417
Epoch 170/200
- val loss: 0.6610 - val accuracy: 0.7417
Epoch 171/200
- val_loss: 0.6593 - val_accuracy: 0.7417
Epoch 172/200
- val_loss: 0.6575 - val_accuracy: 0.7500
Epoch 173/200
- val_loss: 0.6558 - val_accuracy: 0.7500
Epoch 174/200
- val loss: 0.6541 - val accuracy: 0.7500
Epoch 175/200
- val loss: 0.6524 - val accuracy: 0.7417
Epoch 176/200
- val_loss: 0.6507 - val_accuracy: 0.7417
Epoch 177/200
- val loss: 0.6490 - val accuracy: 0.7417
Epoch 178/200
- val loss: 0.6473 - val accuracy: 0.7417
Epoch 179/200
- val_loss: 0.6456 - val_accuracy: 0.7417
Epoch 180/200
- val loss: 0.6439 - val accuracy: 0.7417
Epoch 181/200
- val loss: 0.6423 - val accuracy: 0.7417
Epoch 182/200
- val loss: 0.6406 - val accuracy: 0.7417
Epoch 183/200
- val loss: 0.6390 - val accuracy: 0.7417
Epoch 184/200
- val loss: 0.6374 - val accuracy: 0.7417
Epoch 185/200
- val loss: 0.6357 - val accuracy: 0.7500
Epoch 186/200
- val_loss: 0.6341 - val_accuracy: 0.7500
Epoch 187/200
- val_loss: 0.6325 - val_accuracy: 0.7500
Epoch 188/200
- val_loss: 0.6309 - val_accuracy: 0.7500
Epoch 189/200
30/30 [============] - 0s 367us/step - loss: 0.4502 - accuracy: 0.8667
- val loss: 0.6293 - val accuracy: 0.7500
```

```
Epoch 190/200
    - val loss: 0.6277 - val accuracy: 0.7500
    Epoch 191/200
    - val loss: 0.6262 - val accuracy: 0.7417
    - val_loss: 0.6246 - val_accuracy: 0.7417
    Epoch 193/200
    - val_loss: 0.6231 - val_accuracy: 0.7417
    Epoch 194/200
    30/30 [============== ] - 0s 367us/step - loss: 0.4406 - accuracy: 0.9000
    - val loss: 0.6215 - val accuracy: 0.7417
    Epoch 195/200
    - val_loss: 0.6200 - val_accuracy: 0.7417
    Epoch 196/200
    - val loss: 0.6184 - val accuracy: 0.7417
    Epoch 197/200
    - val_loss: 0.6169 - val_accuracy: 0.7417
    Epoch 198/200
    - val_loss: 0.6154 - val_accuracy: 0.7417
    Epoch 199/200
    - val loss: 0.6139 - val accuracy: 0.7417
    Epoch 200/200
    - val_loss: 0.6124 - val_accuracy: 0.7417
Out[26]: <keras.callbacks.callbacks.History at 0x1d2f6f72fd0>
In [27]:
    w = model.get_weights()
```

In order to get better performance, the input data has to be normalized so that all features are weighted equally in classification

```
In [28]:
    X = preprocessing.scale(iris['data'])
    Y = to_categorical(iris['target'])
```

3 d) Test 4: Split the X and Y into training and test sets and let us use a training dataset of 0.6 or 40% with a different number of epochs

```
In [29]: X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size = 0.6)
In [30]: print(X_train.shape, X_test.shape)
```

Now create a model as shown neural model 4-10-3 (4-inputs, 10-neurons, 3-output neurons)

Train the model for about 150 epochs and see how it performs on training and test data sets

```
In [33]:
    model.fit(X train, Y train, validation data=(X test,Y test), epochs=150)
    Train on 60 samples, validate on 90 samples
    Epoch 1/150
    60/60 [============== ] - 0s 2ms/step - loss: 1.1784 - accuracy: 0.3667 -
    val loss: 1.2735 - val accuracy: 0.2667
    Epoch 2/150
    - val loss: 1.2567 - val accuracy: 0.2778
    - val loss: 1.2393 - val accuracy: 0.2778
    Epoch 4/150
    - val loss: 1.2221 - val accuracy: 0.3111
    Epoch 5/150
    - val loss: 1.2051 - val accuracy: 0.3111
    Epoch 6/150
    - val_loss: 1.1881 - val_accuracy: 0.3111
    Epoch 7/150
    - val_loss: 1.1714 - val_accuracy: 0.3111
    - val_loss: 1.1549 - val_accuracy: 0.3222
    Epoch 9/150
    - val loss: 1.1386 - val accuracy: 0.3222
    Epoch 10/150
    - val loss: 1.1225 - val accuracy: 0.3333
    Epoch 11/150
    - val_loss: 1.1067 - val_accuracy: 0.3333
    Epoch 12/150
    - val loss: 1.0911 - val accuracy: 0.3444
    Epoch 13/150
```

```
- val loss: 1.0757 - val accuracy: 0.3444
Epoch 14/150
- val loss: 1.0604 - val accuracy: 0.3556
Epoch 15/150
- val loss: 1.0452 - val accuracy: 0.3556
Epoch 16/150
- val loss: 1.0304 - val accuracy: 0.3556
Epoch 17/150
- val_loss: 1.0156 - val_accuracy: 0.3667
Epoch 18/150
- val_loss: 1.0011 - val_accuracy: 0.3667
Epoch 19/150
- val loss: 0.9868 - val accuracy: 0.3667
Epoch 20/150
- val loss: 0.9727 - val accuracy: 0.3667
Epoch 21/150
- val loss: 0.9587 - val accuracy: 0.3667
Epoch 22/150
- val loss: 0.9447 - val accuracy: 0.3667
Epoch 23/150
- val loss: 0.9311 - val accuracy: 0.3667
Epoch 24/150
- val_loss: 0.9176 - val_accuracy: 0.3778
Epoch 25/150
- val loss: 0.9043 - val accuracy: 0.3778
Epoch 26/150
- val loss: 0.8912 - val accuracy: 0.4222
Epoch 27/150
- val loss: 0.8784 - val accuracy: 0.4444
Epoch 28/150
- val loss: 0.8657 - val accuracy: 0.4889
Epoch 29/150
- val loss: 0.8531 - val accuracy: 0.5444
Epoch 30/150
- val loss: 0.8408 - val accuracy: 0.5556
Epoch 31/150
- val_loss: 0.8286 - val_accuracy: 0.5889
Epoch 32/150
- val_loss: 0.8164 - val_accuracy: 0.6222
Epoch 33/150
- val_loss: 0.8047 - val_accuracy: 0.6667
Epoch 34/150
60/60 [============] - 0s 233us/step - loss: 0.7980 - accuracy: 0.6167
- val loss: 0.7930 - val accuracy: 0.6778
```

```
Epoch 35/150
- val loss: 0.7814 - val accuracy: 0.6889
Epoch 36/150
- val loss: 0.7702 - val accuracy: 0.7111
Epoch 37/150
- val_loss: 0.7591 - val_accuracy: 0.7556
Epoch 38/150
- val_loss: 0.7482 - val_accuracy: 0.7778
Epoch 39/150
- val loss: 0.7374 - val accuracy: 0.7889
Epoch 40/150
- val loss: 0.7269 - val accuracy: 0.7889
Epoch 41/150
- val loss: 0.7165 - val accuracy: 0.7889
Epoch 42/150
- val loss: 0.7062 - val accuracy: 0.7889
Epoch 43/150
- val_loss: 0.6961 - val_accuracy: 0.8000
Epoch 44/150
- val loss: 0.6862 - val accuracy: 0.8000
- val loss: 0.6766 - val accuracy: 0.8000
Epoch 46/150
- val_loss: 0.6672 - val_accuracy: 0.8000
Epoch 47/150
- val_loss: 0.6579 - val_accuracy: 0.8000
Epoch 48/150
val loss: 0.6489 - val accuracy: 0.8111
Epoch 49/150
- val loss: 0.6400 - val accuracy: 0.8222
60/60 [============== ] - 0s 225us/step - loss: 0.6680 - accuracy: 0.7667
- val loss: 0.6313 - val accuracy: 0.8222
Epoch 51/150
- val_loss: 0.6229 - val_accuracy: 0.8222
Epoch 52/150
- val_loss: 0.6147 - val_accuracy: 0.8222
Epoch 53/150
- val_loss: 0.6067 - val_accuracy: 0.8333
Epoch 54/150
- val_loss: 0.5988 - val_accuracy: 0.8667
- val loss: 0.5912 - val accuracy: 0.8667
Epoch 56/150
```

```
- val loss: 0.5837 - val accuracy: 0.8667
Epoch 57/150
- val loss: 0.5764 - val accuracy: 0.8778
Epoch 58/150
- val loss: 0.5692 - val accuracy: 0.8778
Epoch 59/150
- val_loss: 0.5623 - val_accuracy: 0.8778
Epoch 60/150
- val_loss: 0.5555 - val_accuracy: 0.8778
Epoch 61/150
- val loss: 0.5489 - val accuracy: 0.8778
Epoch 62/150
- val loss: 0.5424 - val accuracy: 0.8778
Epoch 63/150
- val loss: 0.5361 - val accuracy: 0.8778
Epoch 64/150
- val_loss: 0.5300 - val_accuracy: 0.8778
Epoch 65/150
- val loss: 0.5240 - val accuracy: 0.8778
Epoch 66/150
- val loss: 0.5182 - val accuracy: 0.8778
Epoch 67/150
- val loss: 0.5124 - val accuracy: 0.8778
Epoch 68/150
- val_loss: 0.5068 - val_accuracy: 0.8778
Epoch 69/150
- val_loss: 0.5013 - val_accuracy: 0.8778
Epoch 70/150
60/60 [============= ] - 0s 167us/step - loss: 0.5542 - accuracy: 0.8000
- val_loss: 0.4960 - val_accuracy: 0.8889
Epoch 71/150
- val loss: 0.4908 - val accuracy: 0.8889
Epoch 72/150
- val_loss: 0.4858 - val_accuracy: 0.8889
- val loss: 0.4808 - val accuracy: 0.8889
Epoch 74/150
- val_loss: 0.4760 - val_accuracy: 0.8889
Epoch 75/150
- val_loss: 0.4713 - val_accuracy: 0.9000
Epoch 76/150
- val loss: 0.4667 - val accuracy: 0.9000
Epoch 77/150
- val loss: 0.4622 - val accuracy: 0.9000
Epoch 78/150
```

```
- val loss: 0.4578 - val accuracy: 0.9000
Epoch 79/150
60/60 [============== ] - 0s 167us/step - loss: 0.5171 - accuracy: 0.8167
- val loss: 0.4536 - val accuracy: 0.9000
Epoch 80/150
- val loss: 0.4495 - val accuracy: 0.9000
Epoch 81/150
- val loss: 0.4454 - val accuracy: 0.9000
Epoch 82/150
- val loss: 0.4415 - val accuracy: 0.9000
Epoch 83/150
- val_loss: 0.4376 - val_accuracy: 0.9000
Epoch 84/150
- val loss: 0.4338 - val accuracy: 0.9000
Epoch 85/150
- val loss: 0.4302 - val accuracy: 0.9000
Epoch 86/150
- val loss: 0.4266 - val accuracy: 0.9000
Epoch 87/150
- val loss: 0.4230 - val accuracy: 0.9000
Epoch 88/150
- val loss: 0.4196 - val accuracy: 0.9000
Epoch 89/150
- val_loss: 0.4162 - val_accuracy: 0.9000
Epoch 90/150
- val loss: 0.4129 - val accuracy: 0.9000
Epoch 91/150
- val loss: 0.4096 - val accuracy: 0.9111
Epoch 92/150
- val loss: 0.4064 - val accuracy: 0.9111
Epoch 93/150
- val loss: 0.4033 - val accuracy: 0.9111
Epoch 94/150
- val loss: 0.4003 - val accuracy: 0.9222
Epoch 95/150
- val loss: 0.3973 - val accuracy: 0.9222
Epoch 96/150
- val_loss: 0.3943 - val_accuracy: 0.9333
Epoch 97/150
- val_loss: 0.3914 - val_accuracy: 0.9333
Epoch 98/150
- val_loss: 0.3886 - val_accuracy: 0.9333
Epoch 99/150
- val loss: 0.3859 - val accuracy: 0.9222
```

```
Epoch 100/150
- val loss: 0.3832 - val accuracy: 0.9222
Epoch 101/150
- val loss: 0.3805 - val accuracy: 0.9222
Epoch 102/150
- val_loss: 0.3779 - val_accuracy: 0.9222
Epoch 103/150
- val_loss: 0.3754 - val_accuracy: 0.9222
Epoch 104/150
- val loss: 0.3729 - val accuracy: 0.9222
Epoch 105/150
- val_loss: 0.3704 - val_accuracy: 0.9222
Epoch 106/150
- val loss: 0.3680 - val accuracy: 0.9222
Epoch 107/150
- val loss: 0.3656 - val accuracy: 0.9222
Epoch 108/150
- val_loss: 0.3632 - val_accuracy: 0.9222
Epoch 109/150
- val loss: 0.3609 - val accuracy: 0.9222
Epoch 110/150
- val loss: 0.3586 - val accuracy: 0.9222
Epoch 111/150
- val_loss: 0.3564 - val_accuracy: 0.9222
Epoch 112/150
- val_loss: 0.3542 - val_accuracy: 0.9222
Epoch 113/150
- val_loss: 0.3520 - val_accuracy: 0.9222
Epoch 114/150
- val loss: 0.3499 - val accuracy: 0.9222
- val loss: 0.3477 - val accuracy: 0.9222
Epoch 116/150
- val_loss: 0.3457 - val_accuracy: 0.9222
Epoch 117/150
- val_loss: 0.3436 - val_accuracy: 0.9222
Epoch 118/150
- val_loss: 0.3416 - val_accuracy: 0.9222
Epoch 119/150
- val_loss: 0.3396 - val_accuracy: 0.9222
- val loss: 0.3376 - val accuracy: 0.9222
Epoch 121/150
```

```
- val loss: 0.3357 - val accuracy: 0.9222
Epoch 122/150
- val loss: 0.3337 - val accuracy: 0.9333
Epoch 123/150
- val loss: 0.3318 - val accuracy: 0.9333
Epoch 124/150
- val_loss: 0.3300 - val_accuracy: 0.9333
Epoch 125/150
- val_loss: 0.3281 - val_accuracy: 0.9333
Epoch 126/150
- val loss: 0.3263 - val accuracy: 0.9333
Epoch 127/150
60/60 [============== ] - 0s 183us/step - loss: 0.3929 - accuracy: 0.8667
- val loss: 0.3245 - val accuracy: 0.9333
Epoch 128/150
- val loss: 0.3227 - val accuracy: 0.9333
Epoch 129/150
- val_loss: 0.3209 - val_accuracy: 0.9333
Epoch 130/150
- val_loss: 0.3192 - val_accuracy: 0.9333
Epoch 131/150
- val loss: 0.3175 - val accuracy: 0.9333
Epoch 132/150
- val loss: 0.3158 - val accuracy: 0.9333
Epoch 133/150
- val_loss: 0.3141 - val_accuracy: 0.9333
Epoch 134/150
- val loss: 0.3124 - val accuracy: 0.9333
Epoch 135/150
60/60 [============== ] - 0s 200us/step - loss: 0.3783 - accuracy: 0.8833
- val loss: 0.3108 - val accuracy: 0.9333
Epoch 136/150
- val loss: 0.3092 - val accuracy: 0.9333
Epoch 137/150
- val loss: 0.3076 - val_accuracy: 0.9333
Epoch 138/150
- val loss: 0.3060 - val accuracy: 0.9333
Epoch 139/150
- val_loss: 0.3044 - val_accuracy: 0.9333
Epoch 140/150
- val_loss: 0.3029 - val_accuracy: 0.9444
Epoch 141/150
- val loss: 0.3013 - val accuracy: 0.9444
Epoch 142/150
- val loss: 0.2998 - val accuracy: 0.9444
Epoch 143/150
```

```
- val loss: 0.2983 - val accuracy: 0.9444
    Epoch 144/150
    - val_loss: 0.2968 - val_accuracy: 0.9444
    Epoch 145/150
    - val_loss: 0.2954 - val_accuracy: 0.9444
    Epoch 146/150
    - val_loss: 0.2939 - val_accuracy: 0.9444
    Epoch 147/150
    - val_loss: 0.2925 - val_accuracy: 0.9444
    Epoch 148/150
    - val_loss: 0.2910 - val_accuracy: 0.9444
    Epoch 149/150
    - val loss: 0.2896 - val accuracy: 0.9444
    Epoch 150/150
    - val_loss: 0.2882 - val_accuracy: 0.9444
Out[33]: <keras.callbacks.callbacks.History at 0x1d2f705a2b0>
In [34]:
    w = model.get_weights()
```

```
In [35]:
X = preprocessing.scale(iris['data'])
Y = to_categorical(iris['target'])
```

3 e) Test 5: Split the X and Y into training and test sets and let us use a training dataset of 0.7 or 30% with a same number of epochs

Now create a model as shown neural model 4-10-3 (4-inputs, 10-neurons, 3-output neurons)

```
model = Sequential()
  model.add(Dense(10,input_dim=4, activation='relu')) # first-layer added
  model.add(Dense(3,activation='softmax')) #output layer added

In [39]:
  model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
```

Train the model for about 150 epochs and see how it performs on training and test data sets

```
In [40]:
    model.fit(X train, Y train, validation data=(X test,Y test), epochs=150)
    Train on 45 samples, validate on 105 samples
    Epoch 1/150
    45/45 [============ ] - 0s 3ms/step - loss: 0.9482 - accuracy: 0.6889 -
    val loss: 0.9328 - val accuracy: 0.5810
    Epoch 2/150
    - val loss: 0.9206 - val accuracy: 0.5905
    Epoch 3/150
    - val loss: 0.9086 - val accuracy: 0.5905
    Epoch 4/150
    - val loss: 0.8970 - val accuracy: 0.6000
    - val_loss: 0.8855 - val_accuracy: 0.6095
    Epoch 6/150
    - val_loss: 0.8743 - val_accuracy: 0.6286
    Epoch 7/150
    - val_loss: 0.8634 - val_accuracy: 0.6286
    Epoch 8/150
    - val loss: 0.8528 - val accuracy: 0.6286
    Epoch 9/150
    - val loss: 0.8425 - val accuracy: 0.6286
    - val_loss: 0.8325 - val_accuracy: 0.6286
    Epoch 11/150
    - val_loss: 0.8228 - val_accuracy: 0.6286
    Epoch 12/150
    - val loss: 0.8134 - val accuracy: 0.6286
    Epoch 13/150
    - val_loss: 0.8041 - val_accuracy: 0.6286
    Epoch 14/150
    - val_loss: 0.7952 - val_accuracy: 0.6286
    - val_loss: 0.7864 - val_accuracy: 0.6286
```

Epoch 16/150

```
- val loss: 0.7779 - val accuracy: 0.6286
Epoch 17/150
- val loss: 0.7696 - val accuracy: 0.6286
Epoch 18/150
- val loss: 0.7615 - val accuracy: 0.6286
Epoch 19/150
- val_loss: 0.7537 - val_accuracy: 0.6286
Epoch 20/150
- val_loss: 0.7461 - val_accuracy: 0.6286
Epoch 21/150
- val_loss: 0.7387 - val_accuracy: 0.6286
Epoch 22/150
- val loss: 0.7315 - val accuracy: 0.6286
Epoch 23/150
- val loss: 0.7246 - val accuracy: 0.6286
Epoch 24/150
- val loss: 0.7178 - val accuracy: 0.6286
Epoch 25/150
- val loss: 0.7112 - val accuracy: 0.6286
Epoch 26/150
- val loss: 0.7048 - val accuracy: 0.6286
Epoch 27/150
- val_loss: 0.6986 - val_accuracy: 0.6286
Epoch 28/150
- val loss: 0.6924 - val accuracy: 0.6286
Epoch 29/150
- val loss: 0.6864 - val accuracy: 0.6286
Epoch 30/150
- val loss: 0.6806 - val accuracy: 0.6286
Epoch 31/150
- val loss: 0.6749 - val accuracy: 0.6286
Epoch 32/150
- val loss: 0.6694 - val accuracy: 0.6286
Epoch 33/150
- val loss: 0.6641 - val accuracy: 0.6286
Epoch 34/150
- val_loss: 0.6589 - val_accuracy: 0.6286
Epoch 35/150
- val_loss: 0.6538 - val_accuracy: 0.6286
Epoch 36/150
- val_loss: 0.6489 - val_accuracy: 0.6286
Epoch 37/150
- val loss: 0.6442 - val accuracy: 0.6286
```

```
Epoch 38/150
- val loss: 0.6396 - val accuracy: 0.6286
Epoch 39/150
- val loss: 0.6351 - val accuracy: 0.6286
Epoch 40/150
- val_loss: 0.6307 - val_accuracy: 0.6286
Epoch 41/150
- val_loss: 0.6265 - val_accuracy: 0.6286
Epoch 42/150
- val loss: 0.6223 - val accuracy: 0.6286
Epoch 43/150
- val_loss: 0.6182 - val_accuracy: 0.6286
Epoch 44/150
- val loss: 0.6142 - val accuracy: 0.6286
Epoch 45/150
- val loss: 0.6103 - val accuracy: 0.6286
Epoch 46/150
- val_loss: 0.6065 - val_accuracy: 0.6286
Epoch 47/150
- val loss: 0.6027 - val accuracy: 0.6286
Epoch 48/150
- val loss: 0.5991 - val accuracy: 0.6286
Epoch 49/150
- val_loss: 0.5955 - val_accuracy: 0.6286
Epoch 50/150
- val_loss: 0.5921 - val_accuracy: 0.6286
Epoch 51/150
- val_loss: 0.5887 - val_accuracy: 0.6286
Epoch 52/150
- val loss: 0.5853 - val accuracy: 0.6286
- val loss: 0.5821 - val accuracy: 0.6286
Epoch 54/150
- val loss: 0.5790 - val accuracy: 0.6286
Epoch 55/150
- val loss: 0.5759 - val accuracy: 0.6286
Epoch 56/150
- val_loss: 0.5729 - val_accuracy: 0.6286
Epoch 57/150
- val_loss: 0.5699 - val_accuracy: 0.6381
- val loss: 0.5670 - val accuracy: 0.6476
Epoch 59/150
```

```
- val loss: 0.5641 - val accuracy: 0.6476
Epoch 60/150
- val loss: 0.5613 - val accuracy: 0.6476
Epoch 61/150
- val loss: 0.5585 - val accuracy: 0.6667
Epoch 62/150
- val_loss: 0.5558 - val_accuracy: 0.6667
Epoch 63/150
- val_loss: 0.5531 - val_accuracy: 0.6762
Epoch 64/150
- val loss: 0.5504 - val accuracy: 0.6857
Epoch 65/150
- val loss: 0.5478 - val accuracy: 0.6857
Epoch 66/150
- val loss: 0.5453 - val accuracy: 0.6952
Epoch 67/150
- val_loss: 0.5428 - val_accuracy: 0.7048
Epoch 68/150
- val loss: 0.5404 - val accuracy: 0.7143
Epoch 69/150
- val loss: 0.5380 - val accuracy: 0.7143
Epoch 70/150
- val loss: 0.5356 - val accuracy: 0.7238
Epoch 71/150
- val_loss: 0.5333 - val_accuracy: 0.7238
Epoch 72/150
- val_loss: 0.5310 - val_accuracy: 0.7238
Epoch 73/150
- val loss: 0.5287 - val accuracy: 0.7333
Epoch 74/150
- val loss: 0.5264 - val accuracy: 0.7524
Epoch 75/150
- val_loss: 0.5241 - val_accuracy: 0.7524
Epoch 76/150
- val loss: 0.5219 - val accuracy: 0.7524
Epoch 77/150
- val_loss: 0.5197 - val_accuracy: 0.7524
Epoch 78/150
- val_loss: 0.5176 - val_accuracy: 0.7619
Epoch 79/150
- val loss: 0.5155 - val accuracy: 0.7619
Epoch 80/150
- val loss: 0.5135 - val accuracy: 0.7714
Epoch 81/150
```

```
- val loss: 0.5114 - val accuracy: 0.7714
Epoch 82/150
- val loss: 0.5094 - val accuracy: 0.7714
Epoch 83/150
- val loss: 0.5074 - val accuracy: 0.7714
Epoch 84/150
- val_loss: 0.5053 - val_accuracy: 0.7714
Epoch 85/150
- val_loss: 0.5033 - val_accuracy: 0.7714
Epoch 86/150
- val_loss: 0.5013 - val_accuracy: 0.7810
Epoch 87/150
- val loss: 0.4993 - val accuracy: 0.7810
Epoch 88/150
- val loss: 0.4974 - val accuracy: 0.7810
Epoch 89/150
- val loss: 0.4955 - val accuracy: 0.7810
Epoch 90/150
- val loss: 0.4936 - val accuracy: 0.7810
Epoch 91/150
- val loss: 0.4918 - val accuracy: 0.7810
Epoch 92/150
- val_loss: 0.4901 - val_accuracy: 0.7810
Epoch 93/150
- val loss: 0.4883 - val accuracy: 0.7810
Epoch 94/150
val loss: 0.4865 - val accuracy: 0.7905
Epoch 95/150
- val loss: 0.4847 - val accuracy: 0.7905
Epoch 96/150
- val loss: 0.4830 - val accuracy: 0.7905
Epoch 97/150
- val loss: 0.4814 - val accuracy: 0.7905
Epoch 98/150
- val loss: 0.4799 - val accuracy: 0.7905
Epoch 99/150
- val_loss: 0.4784 - val_accuracy: 0.7905
Epoch 100/150
- val_loss: 0.4769 - val_accuracy: 0.7905
Epoch 101/150
- val_loss: 0.4754 - val_accuracy: 0.7905
Epoch 102/150
45/45 [=============] - 0s 267us/step - loss: 0.3608 - accuracy: 0.8889
- val loss: 0.4739 - val accuracy: 0.7905
```

```
Epoch 103/150
- val loss: 0.4724 - val accuracy: 0.7905
Epoch 104/150
- val loss: 0.4708 - val accuracy: 0.7905
Epoch 105/150
- val_loss: 0.4692 - val_accuracy: 0.7810
Epoch 106/150
- val_loss: 0.4676 - val_accuracy: 0.7905
Epoch 107/150
- val loss: 0.4660 - val accuracy: 0.7905
Epoch 108/150
- val_loss: 0.4644 - val_accuracy: 0.8000
Epoch 109/150
45/45 [============= ] - 0s 222us/step - loss: 0.3496 - accuracy: 0.8889
- val loss: 0.4629 - val accuracy: 0.8095
Epoch 110/150
- val loss: 0.4613 - val accuracy: 0.8095
Epoch 111/150
- val_loss: 0.4598 - val_accuracy: 0.8190
Epoch 112/150
- val loss: 0.4583 - val accuracy: 0.8190
Epoch 113/150
- val loss: 0.4567 - val accuracy: 0.8190
Epoch 114/150
- val_loss: 0.4552 - val_accuracy: 0.8190
Epoch 115/150
- val_loss: 0.4538 - val_accuracy: 0.8190
Epoch 116/150
- val_loss: 0.4523 - val_accuracy: 0.8190
Epoch 117/150
- val loss: 0.4509 - val accuracy: 0.8190
- val loss: 0.4495 - val accuracy: 0.8190
Epoch 119/150
- val_loss: 0.4482 - val_accuracy: 0.8190
Epoch 120/150
- val loss: 0.4468 - val accuracy: 0.8190
Epoch 121/150
- val_loss: 0.4455 - val_accuracy: 0.8190
Epoch 122/150
- val_loss: 0.4442 - val_accuracy: 0.8190
- val loss: 0.4430 - val accuracy: 0.8190
Epoch 124/150
```

```
- val loss: 0.4417 - val accuracy: 0.8190
Epoch 125/150
- val loss: 0.4404 - val accuracy: 0.8190
Epoch 126/150
- val loss: 0.4391 - val accuracy: 0.8190
Epoch 127/150
- val_loss: 0.4377 - val_accuracy: 0.8190
Epoch 128/150
- val_loss: 0.4365 - val_accuracy: 0.8190
Epoch 129/150
- val loss: 0.4353 - val accuracy: 0.8190
Epoch 130/150
- val loss: 0.4340 - val accuracy: 0.8190
Epoch 131/150
- val loss: 0.4328 - val accuracy: 0.8286
Epoch 132/150
- val_loss: 0.4315 - val_accuracy: 0.8286
Epoch 133/150
- val loss: 0.4304 - val accuracy: 0.8286
Epoch 134/150
- val loss: 0.4292 - val accuracy: 0.8286
Epoch 135/150
- val loss: 0.4279 - val accuracy: 0.8286
Epoch 136/150
- val_loss: 0.4267 - val_accuracy: 0.8286
Epoch 137/150
- val_loss: 0.4255 - val_accuracy: 0.8286
Epoch 138/150
- val loss: 0.4243 - val accuracy: 0.8286
Epoch 139/150
- val loss: 0.4231 - val accuracy: 0.8190
Epoch 140/150
- val loss: 0.4220 - val_accuracy: 0.8190
Epoch 141/150
- val loss: 0.4208 - val accuracy: 0.8190
Epoch 142/150
- val_loss: 0.4197 - val_accuracy: 0.8190
Epoch 143/150
- val_loss: 0.4185 - val_accuracy: 0.8190
Epoch 144/150
- val loss: 0.4172 - val accuracy: 0.8190
Epoch 145/150
- val loss: 0.4160 - val accuracy: 0.8190
Epoch 146/150
```

```
- val loss: 0.4148 - val accuracy: 0.8190
    Epoch 147/150
    - val_loss: 0.4136 - val_accuracy: 0.8190
    Epoch 148/150
    - val_loss: 0.4125 - val_accuracy: 0.8190
    Epoch 149/150
    - val_loss: 0.4113 - val_accuracy: 0.8286
    Epoch 150/150
    - val_loss: 0.4102 - val_accuracy: 0.8286
Out[40]: <keras.callbacks.callbacks.History at 0x1d2f8397fd0>
In [41]:
    w = model.get weights()
```

```
In [42]:
    X = preprocessing.scale(iris['data'])
    Y = to_categorical(iris['target'])
```

3 f) Test 6: Split the X and Y into training and test sets and let us use a training dataset of 0.8 or 20% with a same number of epochs

Now create a model as shown neural model 4-10-3 (4-inputs, 10-neurons, 3-output neurons)

```
In [45]:
    model = Sequential()
    model.add(Dense(10,input_dim=4, activation='relu')) # first-layer added
    model.add(Dense(3,activation='softmax')) #output layer added

In [46]:
    model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
```

Train the model for about 150 epochs and see how it performs on training and test data sets

In [47]: model.fit(X_train, Y_train, validation_data=(X_test,Y_test), epochs=150)

```
Train on 30 samples, validate on 120 samples
Epoch 1/150
30/30 [============= ] - 0s 4ms/step - loss: 1.0353 - accuracy: 0.5000 -
val loss: 1.0665 - val accuracy: 0.3750
Epoch 2/150
- val loss: 1.0609 - val accuracy: 0.3750
Epoch 3/150
- val loss: 1.0558 - val accuracy: 0.3750
Epoch 4/150
- val_loss: 1.0502 - val_accuracy: 0.3750
Epoch 5/150
- val_loss: 1.0448 - val_accuracy: 0.3833
Epoch 6/150
- val_loss: 1.0393 - val_accuracy: 0.3833
Epoch 7/150
- val loss: 1.0338 - val accuracy: 0.4000
- val loss: 1.0284 - val accuracy: 0.4000
Epoch 9/150
- val_loss: 1.0230 - val_accuracy: 0.4083
Epoch 10/150
- val_loss: 1.0177 - val_accuracy: 0.4083
Epoch 11/150
- val loss: 1.0123 - val accuracy: 0.4083
Epoch 12/150
- val_loss: 1.0070 - val_accuracy: 0.4167
Epoch 13/150
- val_loss: 1.0017 - val_accuracy: 0.4250
Epoch 14/150
- val loss: 0.9965 - val accuracy: 0.4417
Epoch 15/150
30/30 [============== ] - 0s 267us/step - loss: 0.9512 - accuracy: 0.5000
- val loss: 0.9913 - val accuracy: 0.4667
Epoch 16/150
- val loss: 0.9861 - val accuracy: 0.4750
Epoch 17/150
- val loss: 0.9809 - val accuracy: 0.5000
- val loss: 0.9758 - val accuracy: 0.5083
Epoch 19/150
```

```
- val loss: 0.9708 - val accuracy: 0.5000
Epoch 20/150
- val loss: 0.9658 - val accuracy: 0.4917
Epoch 21/150
- val loss: 0.9609 - val accuracy: 0.4750
Epoch 22/150
- val_loss: 0.9560 - val_accuracy: 0.4750
Epoch 23/150
- val_loss: 0.9511 - val_accuracy: 0.4667
Epoch 24/150
- val loss: 0.9463 - val accuracy: 0.4667
Epoch 25/150
30/30 [============== ] - 0s 300us/step - loss: 0.8950 - accuracy: 0.5667
- val loss: 0.9416 - val accuracy: 0.4750
Epoch 26/150
- val loss: 0.9368 - val accuracy: 0.4750
Epoch 27/150
- val_loss: 0.9321 - val_accuracy: 0.4750
Epoch 28/150
- val loss: 0.9274 - val accuracy: 0.4750
Epoch 29/150
- val loss: 0.9228 - val accuracy: 0.4833
Epoch 30/150
- val_loss: 0.9183 - val_accuracy: 0.4667
Epoch 31/150
- val_loss: 0.9138 - val_accuracy: 0.4583
Epoch 32/150
- val_loss: 0.9093 - val_accuracy: 0.4750
Epoch 33/150
- val_loss: 0.9049 - val_accuracy: 0.4917
Epoch 34/150
- val loss: 0.9005 - val accuracy: 0.4833
Epoch 35/150
- val loss: 0.8961 - val_accuracy: 0.4833
Epoch 36/150
- val loss: 0.8917 - val accuracy: 0.5000
Epoch 37/150
- val_loss: 0.8874 - val_accuracy: 0.5083
Epoch 38/150
- val_loss: 0.8831 - val_accuracy: 0.5250
Epoch 39/150
- val loss: 0.8789 - val accuracy: 0.5417
Epoch 40/150
- val loss: 0.8747 - val accuracy: 0.5417
Epoch 41/150
```

```
- val loss: 0.8705 - val accuracy: 0.5417
Epoch 42/150
- val loss: 0.8664 - val accuracy: 0.5583
Epoch 43/150
- val loss: 0.8623 - val accuracy: 0.5667
Epoch 44/150
- val_loss: 0.8583 - val_accuracy: 0.5667
Epoch 45/150
- val_loss: 0.8542 - val_accuracy: 0.5750
Epoch 46/150
- val_loss: 0.8503 - val_accuracy: 0.5917
Epoch 47/150
- val loss: 0.8464 - val accuracy: 0.5917
Epoch 48/150
- val loss: 0.8425 - val accuracy: 0.6167
Epoch 49/150
- val loss: 0.8386 - val accuracy: 0.6333
Epoch 50/150
- val loss: 0.8348 - val accuracy: 0.6333
Epoch 51/150
- val loss: 0.8310 - val accuracy: 0.6500
Epoch 52/150
- val_loss: 0.8273 - val_accuracy: 0.6583
Epoch 53/150
- val loss: 0.8236 - val accuracy: 0.6583
Epoch 54/150
- val loss: 0.8199 - val accuracy: 0.6583
Epoch 55/150
- val loss: 0.8163 - val accuracy: 0.6667
Epoch 56/150
- val loss: 0.8127 - val accuracy: 0.6833
Epoch 57/150
- val loss: 0.8091 - val accuracy: 0.6833
Epoch 58/150
- val loss: 0.8056 - val accuracy: 0.6833
Epoch 59/150
- val_loss: 0.8021 - val_accuracy: 0.6917
Epoch 60/150
- val_loss: 0.7987 - val_accuracy: 0.6917
Epoch 61/150
- val loss: 0.7953 - val accuracy: 0.7000
Epoch 62/150
30/30 [============] - 0s 200us/step - loss: 0.7292 - accuracy: 0.7667
- val loss: 0.7919 - val accuracy: 0.7083
```

```
Epoch 63/150
- val loss: 0.7885 - val accuracy: 0.7083
Epoch 64/150
- val loss: 0.7852 - val accuracy: 0.7167
Epoch 65/150
- val_loss: 0.7818 - val_accuracy: 0.7167
Epoch 66/150
- val_loss: 0.7785 - val_accuracy: 0.7167
Epoch 67/150
- val loss: 0.7753 - val accuracy: 0.7167
Epoch 68/150
- val_loss: 0.7720 - val_accuracy: 0.7083
Epoch 69/150
- val loss: 0.7688 - val accuracy: 0.7083
Epoch 70/150
- val loss: 0.7655 - val accuracy: 0.7083
Epoch 71/150
- val_loss: 0.7623 - val_accuracy: 0.7083
Epoch 72/150
- val loss: 0.7592 - val accuracy: 0.7083
Epoch 73/150
- val loss: 0.7560 - val accuracy: 0.7083
Epoch 74/150
- val_loss: 0.7529 - val_accuracy: 0.7083
Epoch 75/150
- val_loss: 0.7499 - val_accuracy: 0.7083
Epoch 76/150
- val loss: 0.7468 - val accuracy: 0.7167
Epoch 77/150
- val loss: 0.7438 - val accuracy: 0.7167
Epoch 78/150
- val loss: 0.7408 - val accuracy: 0.7167
Epoch 79/150
- val_loss: 0.7378 - val_accuracy: 0.7250
Epoch 80/150
- val_loss: 0.7348 - val_accuracy: 0.7333
Epoch 81/150
- val_loss: 0.7319 - val_accuracy: 0.7500
Epoch 82/150
- val_loss: 0.7290 - val_accuracy: 0.7500
- val loss: 0.7261 - val accuracy: 0.7500
Epoch 84/150
```

```
- val loss: 0.7232 - val accuracy: 0.7500
Epoch 85/150
- val loss: 0.7204 - val accuracy: 0.7500
Epoch 86/150
- val loss: 0.7175 - val accuracy: 0.7500
Epoch 87/150
- val_loss: 0.7147 - val_accuracy: 0.7500
Epoch 88/150
- val_loss: 0.7119 - val_accuracy: 0.7583
Epoch 89/150
- val loss: 0.7092 - val accuracy: 0.7667
Epoch 90/150
- val loss: 0.7064 - val accuracy: 0.7667
Epoch 91/150
- val loss: 0.7037 - val accuracy: 0.7667
Epoch 92/150
- val_loss: 0.7010 - val_accuracy: 0.7667
Epoch 93/150
- val loss: 0.6983 - val accuracy: 0.7667
Epoch 94/150
- val loss: 0.6956 - val accuracy: 0.7750
Epoch 95/150
- val loss: 0.6929 - val accuracy: 0.7750
Epoch 96/150
- val_loss: 0.6903 - val_accuracy: 0.7750
Epoch 97/150
- val loss: 0.6877 - val accuracy: 0.7750
Epoch 98/150
- val loss: 0.6851 - val accuracy: 0.7750
Epoch 99/150
- val loss: 0.6825 - val accuracy: 0.7750
Epoch 100/150
- val loss: 0.6800 - val_accuracy: 0.7750
Epoch 101/150
- val loss: 0.6774 - val accuracy: 0.7833
Epoch 102/150
- val_loss: 0.6749 - val_accuracy: 0.7833
Epoch 103/150
- val_loss: 0.6724 - val_accuracy: 0.7833
Epoch 104/150
- val loss: 0.6700 - val accuracy: 0.7833
Epoch 105/150
- val loss: 0.6675 - val accuracy: 0.7833
Epoch 106/150
```

```
- val loss: 0.6650 - val accuracy: 0.7833
Epoch 107/150
- val loss: 0.6626 - val accuracy: 0.7833
Epoch 108/150
- val loss: 0.6602 - val accuracy: 0.7833
Epoch 109/150
- val_loss: 0.6578 - val_accuracy: 0.7833
Epoch 110/150
- val_loss: 0.6554 - val_accuracy: 0.7833
Epoch 111/150
- val_loss: 0.6530 - val_accuracy: 0.7833
Epoch 112/150
- val loss: 0.6507 - val accuracy: 0.7833
Epoch 113/150
- val loss: 0.6483 - val accuracy: 0.7917
Epoch 114/150
- val loss: 0.6460 - val accuracy: 0.7917
Epoch 115/150
- val loss: 0.6437 - val accuracy: 0.7917
Epoch 116/150
- val loss: 0.6414 - val accuracy: 0.7917
Epoch 117/150
- val_loss: 0.6391 - val_accuracy: 0.7917
Epoch 118/150
- val loss: 0.6368 - val accuracy: 0.7917
Epoch 119/150
- val loss: 0.6346 - val accuracy: 0.7917
Epoch 120/150
- val loss: 0.6324 - val accuracy: 0.7917
Epoch 121/150
- val loss: 0.6302 - val accuracy: 0.7917
Epoch 122/150
- val loss: 0.6280 - val accuracy: 0.7917
Epoch 123/150
- val loss: 0.6258 - val accuracy: 0.7917
Epoch 124/150
- val_loss: 0.6236 - val_accuracy: 0.7917
Epoch 125/150
- val_loss: 0.6215 - val_accuracy: 0.7917
Epoch 126/150
- val loss: 0.6194 - val accuracy: 0.7917
Epoch 127/150
30/30 [============] - 0s 333us/step - loss: 0.5327 - accuracy: 0.8667
- val loss: 0.6173 - val accuracy: 0.7917
```

```
Epoch 128/150
- val loss: 0.6152 - val accuracy: 0.7917
Epoch 129/150
30/30 [============= ] - 0s 333us/step - loss: 0.5280 - accuracy: 0.8667
- val loss: 0.6131 - val accuracy: 0.7917
Epoch 130/150
- val_loss: 0.6111 - val_accuracy: 0.7917
Epoch 131/150
- val_loss: 0.6091 - val_accuracy: 0.7917
Epoch 132/150
- val loss: 0.6071 - val accuracy: 0.7917
Epoch 133/150
- val_loss: 0.6051 - val_accuracy: 0.7917
Epoch 134/150
30/30 [============== ] - 0s 300us/step - loss: 0.5164 - accuracy: 0.8667
- val loss: 0.6031 - val accuracy: 0.8000
Epoch 135/150
- val loss: 0.6011 - val accuracy: 0.8000
Epoch 136/150
- val_loss: 0.5992 - val_accuracy: 0.8167
Epoch 137/150
- val loss: 0.5972 - val accuracy: 0.8167
Epoch 138/150
- val loss: 0.5953 - val accuracy: 0.8167
Epoch 139/150
- val_loss: 0.5934 - val_accuracy: 0.8167
Epoch 140/150
- val_loss: 0.5915 - val_accuracy: 0.8167
Epoch 141/150
- val_loss: 0.5896 - val_accuracy: 0.8167
Epoch 142/150
- val loss: 0.5878 - val accuracy: 0.8167
- val loss: 0.5859 - val accuracy: 0.8167
Epoch 144/150
- val_loss: 0.5841 - val_accuracy: 0.8083
Epoch 145/150
- val_loss: 0.5822 - val_accuracy: 0.8083
Epoch 146/150
- val_loss: 0.5804 - val_accuracy: 0.8083
Epoch 147/150
- val_loss: 0.5786 - val_accuracy: 0.8083
- val loss: 0.5769 - val accuracy: 0.8083
Epoch 149/150
```

```
- val_loss: 0.5751 - val_accuracy: 0.8083
Epoch 150/150
30/30 [==============] - 0s 300us/step - loss: 0.4824 - accuracy: 0.8667
- val_loss: 0.5733 - val_accuracy: 0.8083
Out[47]: <keras.callbacks.callbacks.History at 0x1d2f8701e10>

In [48]: w = model.get_weights()
```

```
In [49]:
    X = preprocessing.scale(iris['data'])
    Y = to_categorical(iris['target'])
```

3 g) Test 7: Split the X and Y into training and test sets and let us use a training dataset of 0.6 or 40% with a different number of epochs

Now create a model as shown neural model 4-10-3 (4-inputs, 10-neurons, 3-output neurons)

Train the model for about 100 epochs and see how it performs on training and test data sets

```
In [54]: model.fit(X_train, Y_train, validation_data=(X_test,Y_test), epochs=100)

Train on 60 samples, validate on 90 samples
Epoch 1/100
```

```
val loss: 0.8581 - val accuracy: 0.7222
Epoch 2/100
- val loss: 0.8483 - val accuracy: 0.7333
Epoch 3/100
- val loss: 0.8385 - val accuracy: 0.7333
Epoch 4/100
- val_loss: 0.8287 - val_accuracy: 0.7333
Epoch 5/100
- val_loss: 0.8187 - val_accuracy: 0.7333
Epoch 6/100
- val_loss: 0.8088 - val_accuracy: 0.7333
Epoch 7/100
- val loss: 0.7988 - val accuracy: 0.7333
Epoch 8/100
- val loss: 0.7892 - val accuracy: 0.7333
Epoch 9/100
- val loss: 0.7797 - val accuracy: 0.7444
Epoch 10/100
- val loss: 0.7702 - val accuracy: 0.7444
Epoch 11/100
- val loss: 0.7608 - val accuracy: 0.7556
Epoch 12/100
- val_loss: 0.7517 - val_accuracy: 0.7556
Epoch 13/100
- val_loss: 0.7426 - val_accuracy: 0.7556
Epoch 14/100
- val loss: 0.7338 - val accuracy: 0.7556
Epoch 15/100
- val loss: 0.7251 - val accuracy: 0.7556
Epoch 16/100
- val loss: 0.7165 - val accuracy: 0.7667
Epoch 17/100
- val loss: 0.7081 - val accuracy: 0.7667
Epoch 18/100
- val loss: 0.6999 - val accuracy: 0.7667
Epoch 19/100
- val_loss: 0.6918 - val_accuracy: 0.7778
Epoch 20/100
- val_loss: 0.6840 - val_accuracy: 0.7778
Epoch 21/100
- val_loss: 0.6762 - val_accuracy: 0.7889
Epoch 22/100
60/60 [============] - 0s 183us/step - loss: 0.7431 - accuracy: 0.6833
- val loss: 0.6686 - val accuracy: 0.7889
```

```
Epoch 23/100
- val loss: 0.6611 - val accuracy: 0.8000
Epoch 24/100
- val loss: 0.6538 - val accuracy: 0.8000
Epoch 25/100
- val_loss: 0.6467 - val_accuracy: 0.8111
Epoch 26/100
- val_loss: 0.6398 - val_accuracy: 0.8111
Epoch 27/100
- val loss: 0.6329 - val accuracy: 0.8111
Epoch 28/100
- val_loss: 0.6262 - val_accuracy: 0.8333
Epoch 29/100
- val loss: 0.6195 - val accuracy: 0.8333
Epoch 30/100
- val loss: 0.6130 - val accuracy: 0.8333
Epoch 31/100
- val_loss: 0.6067 - val_accuracy: 0.8444
Epoch 32/100
- val loss: 0.6004 - val accuracy: 0.8556
Epoch 33/100
- val loss: 0.5942 - val accuracy: 0.8556
Epoch 34/100
- val_loss: 0.5882 - val_accuracy: 0.8444
Epoch 35/100
- val_loss: 0.5823 - val_accuracy: 0.8556
Epoch 36/100
- val loss: 0.5766 - val accuracy: 0.8556
Epoch 37/100
- val loss: 0.5711 - val accuracy: 0.8667
- val loss: 0.5656 - val accuracy: 0.8667
Epoch 39/100
- val_loss: 0.5601 - val_accuracy: 0.8778
Epoch 40/100
- val_loss: 0.5549 - val_accuracy: 0.8889
Epoch 41/100
- val_loss: 0.5497 - val_accuracy: 0.8889
Epoch 42/100
- val_loss: 0.5447 - val_accuracy: 0.8889
- val loss: 0.5397 - val accuracy: 0.9000
Epoch 44/100
60/60 [============== ] - 0s 200us/step - loss: 0.6143 - accuracy: 0.7333
```

```
- val loss: 0.5349 - val accuracy: 0.9000
Epoch 45/100
- val loss: 0.5302 - val accuracy: 0.9111
Epoch 46/100
- val loss: 0.5256 - val accuracy: 0.9111
Epoch 47/100
- val_loss: 0.5209 - val_accuracy: 0.9111
Epoch 48/100
- val_loss: 0.5163 - val_accuracy: 0.9111
Epoch 49/100
- val loss: 0.5119 - val accuracy: 0.9111
Epoch 50/100
60/60 [============== ] - 0s 200us/step - loss: 0.5862 - accuracy: 0.7833
- val loss: 0.5074 - val accuracy: 0.9111
Epoch 51/100
- val loss: 0.5030 - val accuracy: 0.9222
Epoch 52/100
- val_loss: 0.4987 - val_accuracy: 0.9222
Epoch 53/100
- val loss: 0.4944 - val accuracy: 0.9222
Epoch 54/100
- val loss: 0.4903 - val accuracy: 0.9222
Epoch 55/100
- val_loss: 0.4861 - val_accuracy: 0.9222
Epoch 56/100
- val_loss: 0.4821 - val_accuracy: 0.9222
Epoch 57/100
- val_loss: 0.4781 - val_accuracy: 0.9222
Epoch 58/100
60/60 [============== ] - 0s 183us/step - loss: 0.5521 - accuracy: 0.8167
- val_loss: 0.4742 - val_accuracy: 0.9222
Epoch 59/100
- val loss: 0.4703 - val accuracy: 0.9111
Epoch 60/100
- val loss: 0.4665 - val_accuracy: 0.9111
- val loss: 0.4628 - val accuracy: 0.9111
Epoch 62/100
- val_loss: 0.4592 - val_accuracy: 0.9111
Epoch 63/100
- val_loss: 0.4556 - val_accuracy: 0.9111
Epoch 64/100
- val loss: 0.4521 - val accuracy: 0.9111
Epoch 65/100
- val loss: 0.4486 - val accuracy: 0.9222
Epoch 66/100
```

```
- val loss: 0.4452 - val accuracy: 0.9222
Epoch 67/100
60/60 [============== ] - 0s 167us/step - loss: 0.5185 - accuracy: 0.8333
- val loss: 0.4418 - val accuracy: 0.9222
Epoch 68/100
- val loss: 0.4385 - val accuracy: 0.9222
Epoch 69/100
- val_loss: 0.4352 - val_accuracy: 0.9333
Epoch 70/100
- val loss: 0.4320 - val accuracy: 0.9333
Epoch 71/100
- val_loss: 0.4287 - val_accuracy: 0.9444
Epoch 72/100
- val loss: 0.4255 - val accuracy: 0.9444
Epoch 73/100
- val loss: 0.4223 - val accuracy: 0.9444
Epoch 74/100
- val loss: 0.4193 - val accuracy: 0.9444
Epoch 75/100
- val loss: 0.4161 - val accuracy: 0.9444
Epoch 76/100
- val loss: 0.4131 - val accuracy: 0.9444
Epoch 77/100
- val_loss: 0.4101 - val_accuracy: 0.9444
Epoch 78/100
- val loss: 0.4072 - val accuracy: 0.9444
Epoch 79/100
- val loss: 0.4042 - val accuracy: 0.9444
Epoch 80/100
- val loss: 0.4014 - val accuracy: 0.9444
Epoch 81/100
- val loss: 0.3985 - val accuracy: 0.9444
Epoch 82/100
- val loss: 0.3957 - val accuracy: 0.9444
Epoch 83/100
- val loss: 0.3929 - val accuracy: 0.9444
Epoch 84/100
- val_loss: 0.3901 - val_accuracy: 0.9444
Epoch 85/100
- val_loss: 0.3873 - val_accuracy: 0.9444
Epoch 86/100
- val_loss: 0.3847 - val_accuracy: 0.9444
Epoch 87/100
60/60 [============== ] - 0s 167us/step - loss: 0.4535 - accuracy: 0.8833
- val loss: 0.3820 - val accuracy: 0.9444
```

```
Epoch 88/100
   - val loss: 0.3794 - val accuracy: 0.9444
   Epoch 89/100
   - val loss: 0.3768 - val accuracy: 0.9444
   - val_loss: 0.3743 - val_accuracy: 0.9444
   Epoch 91/100
   - val_loss: 0.3718 - val_accuracy: 0.9444
   Epoch 92/100
   - val loss: 0.3693 - val accuracy: 0.9444
   Epoch 93/100
   - val_loss: 0.3670 - val_accuracy: 0.9444
   Epoch 94/100
   - val loss: 0.3646 - val accuracy: 0.9444
   - val_loss: 0.3622 - val_accuracy: 0.9444
   Epoch 96/100
   - val_loss: 0.3598 - val_accuracy: 0.9444
   Epoch 97/100
   - val loss: 0.3575 - val accuracy: 0.9444
   Epoch 98/100
   - val_loss: 0.3552 - val_accuracy: 0.9444
   Epoch 99/100
   - val_loss: 0.3529 - val_accuracy: 0.9444
   Epoch 100/100
   - val_loss: 0.3507 - val_accuracy: 0.9444
Out[54]: <keras.callbacks.callbacks.History at 0x1d2f9b4af60>
In [55]:
    w = model.get weights()
```

```
In [56]:
X = preprocessing.scale(iris['data'])
Y = to_categorical(iris['target'])
```

3 h) Test 8: Split the X and Y into training and test sets and let us use a training dataset of 0.7 or 30% with a different number of epochs

Now create a model as shown neural model 4-10-3 (4-inputs, 10-neurons, 3-output neurons)

Train the model for about 100 epochs and see how it performs on training and test data sets

```
In [61]:
    model.fit(X train, Y train, validation data=(X test,Y test), epochs=100)
    Train on 45 samples, validate on 105 samples
    Epoch 1/100
    45/45 [============= ] - 0s 3ms/step - loss: 0.8237 - accuracy: 0.7556 -
    val loss: 0.9005 - val accuracy: 0.6000
    Epoch 2/100
    - val loss: 0.8944 - val accuracy: 0.6000
    - val loss: 0.8886 - val accuracy: 0.6000
    Epoch 4/100
    - val_loss: 0.8826 - val_accuracy: 0.6000
    Epoch 5/100
    - val loss: 0.8769 - val accuracy: 0.6095
    - val_loss: 0.8712 - val_accuracy: 0.6095
    Epoch 7/100
    - val loss: 0.8655 - val accuracy: 0.6095
    Epoch 8/100
    - val loss: 0.8598 - val accuracy: 0.6095
    Epoch 9/100
    - val_loss: 0.8542 - val_accuracy: 0.6095
    Epoch 10/100
    - val loss: 0.8486 - val accuracy: 0.6095
    Epoch 11/100
```

```
- val loss: 0.8430 - val accuracy: 0.6095
Epoch 12/100
- val loss: 0.8375 - val accuracy: 0.6095
Epoch 13/100
- val loss: 0.8320 - val accuracy: 0.6095
Epoch 14/100
- val_loss: 0.8266 - val_accuracy: 0.6095
Epoch 15/100
- val_loss: 0.8212 - val_accuracy: 0.6095
Epoch 16/100
- val_loss: 0.8157 - val_accuracy: 0.6095
Epoch 17/100
- val loss: 0.8104 - val accuracy: 0.6095
Epoch 18/100
- val loss: 0.8051 - val accuracy: 0.6095
Epoch 19/100
- val loss: 0.7998 - val accuracy: 0.6095
Epoch 20/100
- val loss: 0.7947 - val accuracy: 0.6095
Epoch 21/100
- val loss: 0.7896 - val accuracy: 0.6095
Epoch 22/100
- val_loss: 0.7845 - val_accuracy: 0.6095
Epoch 23/100
- val loss: 0.7796 - val accuracy: 0.6095
Epoch 24/100
- val loss: 0.7747 - val accuracy: 0.6095
Epoch 25/100
- val loss: 0.7698 - val accuracy: 0.6095
Epoch 26/100
- val loss: 0.7649 - val accuracy: 0.6095
Epoch 27/100
- val loss: 0.7601 - val accuracy: 0.6095
Epoch 28/100
- val loss: 0.7555 - val accuracy: 0.6095
Epoch 29/100
- val_loss: 0.7508 - val_accuracy: 0.6095
Epoch 30/100
- val_loss: 0.7463 - val_accuracy: 0.6095
Epoch 31/100
- val_loss: 0.7419 - val_accuracy: 0.6095
Epoch 32/100
- val loss: 0.7376 - val accuracy: 0.6095
```

```
Epoch 33/100
- val loss: 0.7333 - val accuracy: 0.6095
Epoch 34/100
- val loss: 0.7291 - val accuracy: 0.6095
Epoch 35/100
- val_loss: 0.7249 - val_accuracy: 0.6095
Epoch 36/100
- val_loss: 0.7208 - val_accuracy: 0.6095
Epoch 37/100
- val loss: 0.7166 - val accuracy: 0.6095
Epoch 38/100
- val_loss: 0.7124 - val_accuracy: 0.6095
Epoch 39/100
45/45 [============== ] - 0s 311us/step - loss: 0.5744 - accuracy: 0.7778
- val loss: 0.7083 - val accuracy: 0.6095
Epoch 40/100
- val loss: 0.7043 - val accuracy: 0.6095
Epoch 41/100
- val_loss: 0.7002 - val_accuracy: 0.6095
Epoch 42/100
- val loss: 0.6962 - val accuracy: 0.6095
Epoch 43/100
- val loss: 0.6922 - val accuracy: 0.6095
Epoch 44/100
- val_loss: 0.6884 - val_accuracy: 0.6095
Epoch 45/100
- val_loss: 0.6847 - val_accuracy: 0.6095
Epoch 46/100
val loss: 0.6809 - val accuracy: 0.6095
Epoch 47/100
- val loss: 0.6771 - val accuracy: 0.6095
- val loss: 0.6734 - val accuracy: 0.6095
Epoch 49/100
- val_loss: 0.6697 - val_accuracy: 0.6095
Epoch 50/100
- val_loss: 0.6661 - val_accuracy: 0.6095
Epoch 51/100
- val_loss: 0.6627 - val_accuracy: 0.6095
Epoch 52/100
- val_loss: 0.6593 - val_accuracy: 0.6095
- val loss: 0.6559 - val accuracy: 0.6095
Epoch 54/100
```

```
- val loss: 0.6525 - val accuracy: 0.6095
Epoch 55/100
- val loss: 0.6490 - val accuracy: 0.6095
Epoch 56/100
- val loss: 0.6454 - val accuracy: 0.6095
Epoch 57/100
- val_loss: 0.6419 - val_accuracy: 0.6095
Epoch 58/100
- val_loss: 0.6385 - val_accuracy: 0.6095
Epoch 59/100
- val loss: 0.6352 - val accuracy: 0.6190
Epoch 60/100
- val loss: 0.6319 - val accuracy: 0.6190
Epoch 61/100
- val loss: 0.6288 - val accuracy: 0.6190
Epoch 62/100
- val_loss: 0.6258 - val_accuracy: 0.6190
Epoch 63/100
- val loss: 0.6230 - val accuracy: 0.6190
Epoch 64/100
- val loss: 0.6202 - val accuracy: 0.6190
Epoch 65/100
- val_loss: 0.6172 - val_accuracy: 0.6190
Epoch 66/100
- val_loss: 0.6142 - val_accuracy: 0.6190
Epoch 67/100
- val loss: 0.6112 - val accuracy: 0.6190
Epoch 68/100
- val_loss: 0.6083 - val_accuracy: 0.6190
Epoch 69/100
- val loss: 0.6054 - val accuracy: 0.6190
Epoch 70/100
- val_loss: 0.6026 - val_accuracy: 0.6190
Epoch 71/100
- val loss: 0.5999 - val accuracy: 0.6190
Epoch 72/100
- val_loss: 0.5973 - val_accuracy: 0.6286
Epoch 73/100
- val_loss: 0.5948 - val_accuracy: 0.6286
Epoch 74/100
- val loss: 0.5923 - val accuracy: 0.6286
Epoch 75/100
- val loss: 0.5898 - val accuracy: 0.6286
Epoch 76/100
```

```
- val loss: 0.5872 - val accuracy: 0.6286
Epoch 77/100
- val loss: 0.5847 - val accuracy: 0.6286
Epoch 78/100
- val loss: 0.5823 - val accuracy: 0.6286
Epoch 79/100
- val loss: 0.5799 - val accuracy: 0.6286
Epoch 80/100
- val_loss: 0.5775 - val_accuracy: 0.6286
Epoch 81/100
- val_loss: 0.5752 - val_accuracy: 0.6286
Epoch 82/100
- val loss: 0.5728 - val accuracy: 0.6286
Epoch 83/100
- val loss: 0.5705 - val accuracy: 0.6286
Epoch 84/100
- val loss: 0.5680 - val accuracy: 0.6286
Epoch 85/100
- val loss: 0.5656 - val accuracy: 0.6381
Epoch 86/100
- val loss: 0.5631 - val accuracy: 0.6381
Epoch 87/100
- val_loss: 0.5607 - val_accuracy: 0.6381
Epoch 88/100
- val loss: 0.5585 - val accuracy: 0.6476
Epoch 89/100
- val loss: 0.5565 - val accuracy: 0.6476
Epoch 90/100
- val loss: 0.5545 - val accuracy: 0.6476
Epoch 91/100
- val loss: 0.5525 - val accuracy: 0.6476
Epoch 92/100
- val loss: 0.5505 - val accuracy: 0.6476
Epoch 93/100
- val loss: 0.5486 - val accuracy: 0.6476
Epoch 94/100
- val_loss: 0.5466 - val_accuracy: 0.6476
Epoch 95/100
- val_loss: 0.5446 - val_accuracy: 0.6571
Epoch 96/100
- val_loss: 0.5426 - val_accuracy: 0.6571
Epoch 97/100
45/45 [==============] - 0s 314us/step - loss: 0.3822 - accuracy: 0.8000
- val loss: 0.5406 - val accuracy: 0.6571
```

```
In [63]:
    X = preprocessing.scale(iris['data'])
    Y = to_categorical(iris['target'])
```

3 i) Test 9: Split the X and Y into training and test sets and let us use a training dataset of 0.8 or 20% with a different number of epochs

Now create a model as shown neural model 4-10-3 (4-inputs, 10-neurons, 3-output neurons)

```
In [66]:
    model = Sequential()
    model.add(Dense(10,input_dim=4, activation='relu')) # first-layer added
    model.add(Dense(3,activation='softmax')) #output layer added

In [67]:
    model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
```

Train the model for about 100 epochs and see how it performs on training and test data sets

```
In [68]: model.fit(X_train, Y_train, validation_data=(X_test,Y_test), epochs=100)
```

```
Train on 30 samples, validate on 120 samples
Epoch 1/100
val_loss: 1.1431 - val_accuracy: 0.4417
Epoch 2/100
- val loss: 1.1366 - val accuracy: 0.4500
- val loss: 1.1300 - val accuracy: 0.4500
Epoch 4/100
- val loss: 1.1239 - val accuracy: 0.4500
Epoch 5/100
- val_loss: 1.1174 - val_accuracy: 0.4500
Epoch 6/100
- val_loss: 1.1112 - val_accuracy: 0.4667
Epoch 7/100
- val loss: 1.1049 - val accuracy: 0.4667
- val loss: 1.0985 - val accuracy: 0.4750
Epoch 9/100
- val loss: 1.0922 - val accuracy: 0.4750
Epoch 10/100
- val loss: 1.0859 - val accuracy: 0.4917
Epoch 11/100
- val_loss: 1.0796 - val_accuracy: 0.4917
Epoch 12/100
- val loss: 1.0733 - val accuracy: 0.5000
Epoch 13/100
- val loss: 1.0670 - val accuracy: 0.5000
Epoch 14/100
- val_loss: 1.0608 - val_accuracy: 0.5083
Epoch 15/100
- val loss: 1.0546 - val accuracy: 0.5167
Epoch 16/100
- val loss: 1.0485 - val accuracy: 0.5167
Epoch 17/100
- val loss: 1.0424 - val accuracy: 0.5250
Epoch 18/100
- val loss: 1.0363 - val accuracy: 0.5250
Epoch 19/100
- val_loss: 1.0302 - val_accuracy: 0.5250
Epoch 20/100
- val loss: 1.0242 - val accuracy: 0.5250
Epoch 21/100
```

```
- val loss: 1.0182 - val accuracy: 0.5250
Epoch 22/100
- val loss: 1.0123 - val accuracy: 0.5333
Epoch 23/100
- val loss: 1.0064 - val accuracy: 0.5333
Epoch 24/100
- val_loss: 1.0006 - val_accuracy: 0.5333
Epoch 25/100
- val_loss: 0.9947 - val_accuracy: 0.5333
Epoch 26/100
- val loss: 0.9890 - val accuracy: 0.5417
Epoch 27/100
- val loss: 0.9833 - val accuracy: 0.5417
Epoch 28/100
- val loss: 0.9776 - val accuracy: 0.5417
Epoch 29/100
- val_loss: 0.9720 - val_accuracy: 0.5417
Epoch 30/100
- val loss: 0.9664 - val accuracy: 0.5500
Epoch 31/100
- val loss: 0.9609 - val accuracy: 0.5500
Epoch 32/100
- val loss: 0.9554 - val accuracy: 0.5500
Epoch 33/100
- val_loss: 0.9499 - val_accuracy: 0.5500
Epoch 34/100
- val_loss: 0.9444 - val_accuracy: 0.5500
Epoch 35/100
- val loss: 0.9390 - val accuracy: 0.5583
Epoch 36/100
- val loss: 0.9337 - val accuracy: 0.5583
Epoch 37/100
- val loss: 0.9284 - val accuracy: 0.5583
Epoch 38/100
- val loss: 0.9231 - val accuracy: 0.5583
Epoch 39/100
- val_loss: 0.9178 - val_accuracy: 0.5583
Epoch 40/100
- val_loss: 0.9127 - val_accuracy: 0.5583
Epoch 41/100
- val loss: 0.9075 - val accuracy: 0.5583
Epoch 42/100
- val loss: 0.9024 - val accuracy: 0.5583
Epoch 43/100
```

```
- val loss: 0.8974 - val accuracy: 0.5750
Epoch 44/100
- val loss: 0.8924 - val accuracy: 0.5917
Epoch 45/100
- val loss: 0.8874 - val accuracy: 0.5917
Epoch 46/100
val loss: 0.8824 - val accuracy: 0.5917
Epoch 47/100
- val_loss: 0.8775 - val_accuracy: 0.6000
Epoch 48/100
- val_loss: 0.8727 - val_accuracy: 0.6083
Epoch 49/100
- val loss: 0.8679 - val accuracy: 0.6083
Epoch 50/100
- val loss: 0.8632 - val accuracy: 0.6167
Epoch 51/100
- val loss: 0.8585 - val accuracy: 0.6250
Epoch 52/100
- val loss: 0.8538 - val accuracy: 0.6250
Epoch 53/100
- val loss: 0.8492 - val accuracy: 0.6250
Epoch 54/100
- val loss: 0.8446 - val accuracy: 0.6250
Epoch 55/100
- val loss: 0.8400 - val accuracy: 0.6250
Epoch 56/100
- val loss: 0.8355 - val accuracy: 0.6250
Epoch 57/100
- val loss: 0.8310 - val accuracy: 0.6250
Epoch 58/100
- val loss: 0.8266 - val accuracy: 0.6250
Epoch 59/100
- val_loss: 0.8222 - val_accuracy: 0.6417
Epoch 60/100
- val loss: 0.8178 - val accuracy: 0.6500
Epoch 61/100
- val_loss: 0.8135 - val_accuracy: 0.6500
Epoch 62/100
- val_loss: 0.8092 - val_accuracy: 0.6500
Epoch 63/100
- val_loss: 0.8049 - val_accuracy: 0.6500
Epoch 64/100
30/30 [============] - 0s 300us/step - loss: 0.8759 - accuracy: 0.5667
- val loss: 0.8007 - val accuracy: 0.6583
```

```
Epoch 65/100
- val loss: 0.7965 - val accuracy: 0.6667
Epoch 66/100
- val loss: 0.7923 - val accuracy: 0.6750
Epoch 67/100
- val_loss: 0.7882 - val_accuracy: 0.6833
Epoch 68/100
- val_loss: 0.7841 - val_accuracy: 0.6917
Epoch 69/100
- val loss: 0.7800 - val accuracy: 0.6917
Epoch 70/100
- val_loss: 0.7759 - val_accuracy: 0.6917
Epoch 71/100
- val loss: 0.7719 - val accuracy: 0.7000
Epoch 72/100
- val loss: 0.7679 - val accuracy: 0.7000
Epoch 73/100
- val_loss: 0.7639 - val_accuracy: 0.7000
Epoch 74/100
- val loss: 0.7600 - val accuracy: 0.7000
Epoch 75/100
- val loss: 0.7561 - val accuracy: 0.7000
Epoch 76/100
- val_loss: 0.7522 - val_accuracy: 0.7083
Epoch 77/100
- val_loss: 0.7485 - val_accuracy: 0.7083
Epoch 78/100
- val_loss: 0.7447 - val_accuracy: 0.7083
Epoch 79/100
- val loss: 0.7410 - val accuracy: 0.7167
Epoch 80/100
- val loss: 0.7373 - val accuracy: 0.7167
Epoch 81/100
- val_loss: 0.7337 - val_accuracy: 0.7167
Epoch 82/100
- val_loss: 0.7301 - val_accuracy: 0.7167
Epoch 83/100
- val_loss: 0.7265 - val_accuracy: 0.7167
Epoch 84/100
- val_loss: 0.7230 - val_accuracy: 0.7250
- val loss: 0.7195 - val accuracy: 0.7250
Epoch 86/100
```

```
- val loss: 0.7160 - val accuracy: 0.7250
    Epoch 87/100
    - val loss: 0.7126 - val accuracy: 0.7250
    Epoch 88/100
    - val loss: 0.7092 - val accuracy: 0.7250
    Epoch 89/100
    - val_loss: 0.7059 - val_accuracy: 0.7250
    Epoch 90/100
    - val_loss: 0.7025 - val_accuracy: 0.7333
    Epoch 91/100
    - val_loss: 0.6993 - val_accuracy: 0.7417
    Epoch 92/100
    - val loss: 0.6960 - val accuracy: 0.7417
    Epoch 93/100
    - val loss: 0.6928 - val accuracy: 0.7417
    Epoch 94/100
    - val_loss: 0.6896 - val_accuracy: 0.7500
    Epoch 95/100
    - val loss: 0.6864 - val accuracy: 0.7500
    Epoch 96/100
    - val loss: 0.6832 - val accuracy: 0.7583
    Epoch 97/100
    - val_loss: 0.6801 - val_accuracy: 0.7583
    Epoch 98/100
    - val_loss: 0.6770 - val_accuracy: 0.7583
    Epoch 99/100
    - val_loss: 0.6740 - val_accuracy: 0.7583
    Epoch 100/100
    30/30 [============== ] - 0s 267us/step - loss: 0.7418 - accuracy: 0.7000
    - val_loss: 0.6710 - val_accuracy: 0.7583
Out[68]: <keras.callbacks.callbacks.History at 0x1d2fa4bdfd0>
In [69]:
    w = model.get weights()
```

```
In [70]:
    X = preprocessing.scale(iris['data'])
    Y = to_categorical(iris['target'])
```

Save the trained model to the disk to retrieve

later and use it to predict the unlabeled samples (use the network as feed forward network)

```
In [71]:
      model.save('iris model.mdl')
In [72]:
      model1 = keras.models.load model('iris model.mdl')
In [73]:
      out put = model.predict(X test)
      print(out put.shape)
      (120, 3)
In [74]:
      label = np.zeros(out_put.shape[0], dtype = 'uint8')
      for i in range(out_put.shape[0]):
        label[i] = 1+np.argmax(out_put[i,])
In [75]:
      print(label)
      1 2 1 3 3 1 3 1 1]
```

4) Concluding Remarks

With a 4-10-3 neuron size along with 200 epochs, and a training dataset of 0.6, I noticed that the accuracy was around 0.6667. Using the same number of epochs and increasing the training dataset to 0.7 increased the accuracy to 0.8286 and increasing it further to 0.8 increased the accuracy to 0.8583.

Reducing the epoch size to 150 along with a training dataset of 0.6 increased the accuracy to 0.71111. Using the same number of epochs and increasing the training dataset to 0.7 increased the accuracy to 0.8857, however, increasing it further to 0.8 reduced the accuracy to 0.8083.

Reducing the epoch size to 100 along with a training dataset of 0.6 increased the accuracy to 0.7667. With the same number of epochs, increasing the training dataset to 0.7 increased the accuracy to 0.7717, increasing it further to 0.8 increased the accuracy to 0.8167.