

```
In [1]: import pandas
import numpy as np
```

```
In [2]: dataset = pandas.read_csv("iris.csv")
dataset = dataset.sample(frac=1)
```

```
In [3]: #Getting the input Data.
X = dataset[["sepal_length", "sepal_width", "petal_length", "petal_width"]]
print(X.head())
X = np.array(X)
print(f"\nshape of X is {X.shape}")
```

	sepal_length	sepal_width	petal_length	petal_width
142	5.8	2.7	5.1	1.9
52	6.9	3.1	4.9	1.5
36	5.5	3.5	1.3	0.2
1	4.9	3.0	1.4	0.2
18	5.7	3.8	1.7	0.3

shape of X is (150, 4)

```
In [4]: #Getting the Classes.
Y = dataset["species"]
print(Y.head())
Y = np.array(Y).reshape((150,1))
print(f"\nShape of Y is {Y.shape}")
```

142	Iris-virginica
52	Iris-versicolor
36	Iris-setosa
1	Iris-setosa
18	Iris-setosa

Name: species, dtype: object

Shape of Y is (150, 1)

```
In [5]: #Pre-Processing the output Classes.
Y[Y=="Iris-setosa"] = 0
Y[Y=="Iris-versicolor"] = 1
Y[Y=="Iris-virginica"] = 2
```

```
In [6]: print(f"Shape of Y after Pre-Processing is {Y.shape}\n")
print(np.squeeze(Y))
```

Shape of Y after Pre-Processing is (150, 1)

```
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```

```
In [7]: from sklearn import preprocessing
```

```
In [8]: #One Hot Encoding Y.  
one_hot = preprocessing.OneHotEncoder()  
O_e = one_hot.fit(Y)  
Y_e = O_e.transform(Y)  
Y = Y_e.toarray()  
print(f"Shape of Y is {Y.shape}")
```

Shape of Y is (150, 3)

c:\users\shambu\appdata\local\programs\python\python37\lib\site-packages\sklearn\preprocessing_encoders.py:415: FutureWarning: The handling of integer data will change in version 0.22. Currently, the categories are determined based on the range [0, max(values)], while in the future they will be determined based on the unique values.

If you want the future behaviour and silence this warning, you can specify "categories='auto'". In case you used a LabelEncoder before this OneHotEncoder to convert the categories to integers, then you can now use the OneHotEncoder directly.

```
warnings.warn(msg, FutureWarning)
```

```
In [9]: import tensorflow as tf  
  
from tensorflow.keras.models import Sequential  
from tensorflow.keras.layers import Dense,Dropout
```

```
In [10]: model = Sequential()  
model.add(Dense(10,input_shape=(4,),activation='relu'))  
model.add(Dense(10,activation='relu'))  
model.add(Dense(10,activation='relu'))  
model.add(Dropout(0.2))  
model.add(Dense(3))
```

```
In [11]: predictions = model(X).numpy()
```

WARNING:tensorflow:Layer dense is casting an input tensor from dtype float64 to the layer's dtype of float32, which is new behavior in TensorFlow 2. The layer has dtype float32 because its dtype defaults to floatx.

If you intended to run this layer in float32, you can safely ignore this warning. If in doubt, this warning is likely only an issue if you are porting a TensorFlow 1.X model to TensorFlow 2.

To change all layers to have dtype float64 by default, call `tf.keras.backend.set_floatx('float64')`. To change just this layer, pass dtype='float64' to the layer constructor. If you are the author of this layer, you can disable autocasting by passing autocast=False to the base Layer constructor.

```
In [12]: #Initial Predictions  
Y_beta = tf.nn.softmax(predictions).numpy()  
print(Y_beta[:5])  
print(np.sum(Y_beta[:5]))  
Y_beta = np.argmax(Y_beta,axis = 1)  
print(Y_beta[:5])
```

```
[[0.30060497 0.60891485 0.09048025]  
 [0.3051782  0.61609834 0.07872337]  
 [0.28733787 0.5755495  0.13711268]  
 [0.30517185 0.5413889  0.15343921]  
 [0.2868019  0.5829144  0.13028374]]  
5.0  
[1 1 1 1 1]
```

```
In [13]: lossSoft = tf.keras.losses.CategoricalCrossentropy(from_logits=True)
```

```
In [14]: print(lossSoft(Y[:5],predictions[:5]).numpy())
```

```
1.3139826
```

```
In [15]: X_train = X[:50]  
Y_train = Y[:50]  
X_test = X[100:-1]  
Y_test = Y[100:-1]  
model.compile(optimizer = "adam",loss=lossSoft,metrics=["accuracy"])
```

In [21]: `model.fit(X,Y,epochs=10)`

Train on 150 samples

Epoch 1/10

150/150 [=====] - 0s 164us/sample - loss: 0.2053 - accuracy: 0.9067

Epoch 2/10

150/150 [=====] - 0s 121us/sample - loss: 0.2443 - accuracy: 0.9000

Epoch 3/10

150/150 [=====] - 0s 125us/sample - loss: 0.2462 - accuracy: 0.8867

Epoch 4/10

150/150 [=====] - 0s 130us/sample - loss: 0.2603 - accuracy: 0.8600

Epoch 5/10

150/150 [=====] - 0s 154us/sample - loss: 0.2465 - accuracy: 0.8800

Epoch 6/10

150/150 [=====] - 0s 127us/sample - loss: 0.2363 - accuracy: 0.9000

Epoch 7/10

150/150 [=====] - 0s 116us/sample - loss: 0.1963 - accuracy: 0.9467

Epoch 8/10

150/150 [=====] - 0s 155us/sample - loss: 0.2425 - accuracy: 0.8933

Epoch 9/10

150/150 [=====] - 0s 109us/sample - loss: 0.2504 - accuracy: 0.9200

Epoch 10/10

150/150 [=====] - 0s 143us/sample - loss: 0.2027 - accuracy: 0.9400

Out[21]: <tensorflow.python.keras.callbacks.History at 0x1aad68aae48>

In [22]: `probability_model = tf.keras.Sequential([
 model,
 tf.keras.layers.Softmax()
])`

```
In [23]: print(np.argmax(probability_model(X[:100]).numpy(),axis=1))  
         print(Y[:100])
```

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In []: