

5. Deep Learning – First Neural Network with Keras

Contents

1. Implementing first neural network using keras	2
2. Dataset explanation	3
3. Input and output from the Dataset	4
3.1. Input Variables (X):.....	4
3.2. Output Variables (y):.....	4
4. Create the model.....	7
4.1. Input shape and activation functions	8
5. Compile the keras model	9
5.1. Loss function, optimiser and metrics.....	10
6. Fit the model	11
7. Evaluate the model	14
8. Prediction	17
9. verbose = 0	20
10. Model summary	23
11. Image of the model	26

5. Deep Learning – First Neural Network with Keras

1. Implementing first neural network using keras

- ✓ Importing the libraries
- ✓ Loading Dataset
- ✓ Data preparation
- ✓ Splitting the dataset
- ✓ Model creation
- ✓ Model compilation
- ✓ Model training
- ✓ Prediction

2. Dataset explanation

- ✓ We are going to work with **pima-indians-diabetes.csv**
- ✓ This Dataset is related to health care domain.
- ✓ Pima Indians are a Native American group that lives in Mexico and Arizona, USA.
- ✓ It describes patient medical record data for Pima Indians and whether they had a diabetes within five years.
- ✓ The Pima Indian Diabetes dataset consisting of Pima Indian females 21 years and older is a popular benchmark dataset.
- ✓ It is a binary classification problem (onset of diabetes as 1 or not as 0).
- ✓ All of the input variables that describe each patient are numerical.

Feature	Description	Data type	Range
Preg	Number of times pregnant	Numeric	[0, 17]
Gluc	Plasma glucose concentration at 2 Hours in an oral glucose tolerance test (GTIT)	Numeric	[0, 199]
BP	Diastolic Blood Pressure (mm Hg)	Numeric	[0, 122]
Skin	Triceps skin fold thickness (mm)	Numeric	[0, 99]
Insulin	2-Hour Serum insulin (μ h/ml)	Numeric	[0, 846]
BMI	Body mass index [weight in kg/(Height in m)]	Numeric	[0, 67.1]
DPF	Diabetes pedigree function	Numeric	[0.078, 2.42]
Age	Age (years)	Numeric	[21, 81]
Outcome	Binary value indicating non-diabetic /diabetic	Factor	[0,1]

3. Input and output from the Dataset

3.1. Input Variables (X):

- ✓ 1. Number of times pregnant
- ✓ 2. Plasma glucose concentration at 2 hours in an oral glucose tolerance test
- ✓ 3. Diastolic blood pressure (mm Hg)
- ✓ 4. Triceps skin fold thickness (mm)
- ✓ 5. 2-hour serum insulin (μ lU/ml)
- ✓ 6. Body mass index (weight in kg/(height in m))
- ✓ 7. Diabetes pedigree function
- ✓ 8. Age (years)

3.2. Output Variables (y):

- ✓ 1. Class variable (0 or 1)

Program Loading csv file
Name demo1.py
Input file pima-indians-diabetes.csv

```
from numpy import loadtxt

dataset = loadtxt('pima-indians-diabetes.csv', delimiter = ',')

print(dataset)
```

Output

```
[[ 6.    148.    72.    ...  0.627  50.    1.   ]
 [ 1.     85.    66.    ...  0.351  31.    0.   ]
 [ 8.    183.    64.    ...  0.672  32.    1.   ]
 ...
 [ 5.    121.    72.    ...  0.245  30.    0.   ]
 [ 1.    126.    60.    ...  0.349  47.    1.   ]
 [ 1.     93.    70.    ...  0.315  23.    0.   ]]
```

Program Split into input (X) and output (y) variables

Name demo2.py

Input file pima-indians-diabetes.csv

```
from numpy import loadtxt
```

```
dataset = loadtxt('pima-indians-diabetes.csv', delimiter = ',')
```

```
X = dataset[:, 0:8]
```

```
y = dataset[:, 8]
```

```
print("Splitting data done")
```

Output

```
Splitting data done
```

4. Create the model

- ✓ First step is, we need to create the model by using Sequential class
- ✓ Once model created then we need to add layers to the model

Program	Model creation
Name	demo4.py
Input file	pima-indians-diabetes.csv


```
from numpy import loadtxt
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense

dataset = loadtxt('pima-indians-diabetes.csv', delimiter = ',')

X = dataset[:, 0:8]
y = dataset[:, 8]

model = Sequential()

layer1 = Dense(12, input_shape = (8, ), activation = 'relu')
layer2 = Dense(8, activation = 'relu')
layer3 = Dense(1, activation = 'sigmoid')

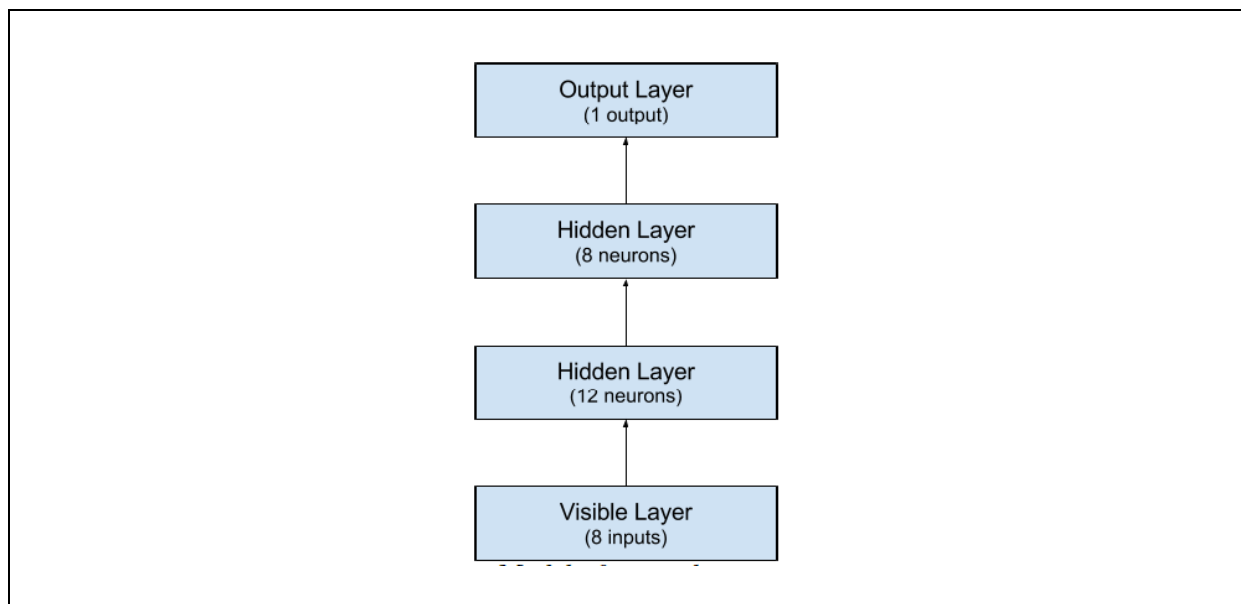
model.add(layer1)
model.add(layer2)
model.add(layer3)

print("Model created")
```


Output	Model created
---------------	---------------

4.1. Input shape and activation functions

- ✓ Fully connected layers are defined using the Dense class.
- ✓ The model expects rows of data with **8** features (the `input_shape = (8,)` argument).
- ✓ The first hidden layer has **12** nodes and uses the `relu` activation function.
- ✓ The second hidden layer has **8** nodes and uses the `relu` activation function.
- ✓ The output layer has **1** node and uses the sigmoid activation function.
- ✓ **The line of code**, Dense layer is doing two things, creating first hidden and input layers.



5. Compile the keras model

- ✓ Once the model created then we need to compile the model

Program Name	demo5.py
Input file	pima-indians-diabetes.csv

```
# importing required libraries
from numpy import loadtxt
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense

# load the dataset
dataset = loadtxt('pima-indians-diabetes.csv', delimiter = ',')

# split into input (X) and output (y) variables
X = dataset[:, 0:8]
y = dataset[:, 8]

# define the keras model
model = Sequential()

layer1 = Dense(12, input_shape = (8, ), activation = 'relu')
layer2 = Dense(8, activation = 'relu')
layer3 = Dense(1, activation = 'sigmoid')

model.add(layer1)
model.add(layer2)
model.add(layer3)

model.compile(loss = 'binary_crossentropy', optimizer = 'adam',
metrics = ['accuracy'])

print("Model compiled")
```

Output

Model created

5.1. Loss function, optimiser and metrics

- ✓ We need to provide loss function to evaluate a set of weights.
- ✓ The optimizer is used to search through different weights for the network.
- ✓ This loss is for a binary classification problems and is defined in Keras as “binary_crossentropy”.
- ✓ We have given optimizer value as adam.
 - This is more efficient gradient descent algorithm.

6. Fit the model

- ✓ Once model compiled then we need to train the model
- ✓ By using fit(...) method we can train the model.

```
Program    Fit the model
Name       demo6.py
Input file pima-indians-diabetes.csv

print("Topic: First DL Example")
print()

print("Step 1: Importing libraries")

from numpy import loadtxt
from tensorflow.keras.layers import Dense
from tensorflow.keras.models import Sequential

import warnings
warnings.filterwarnings("ignore")

print("Step 2: Loading the dataset")

dataset = loadtxt(
    'pima-indians-diabetes.csv',
    delimiter = ','
)

print("Step 3: Data preparation")

X = dataset[:, 0:8]
y = dataset[:, 8]
```

```
print("Step 4: Splitting the dataset: Optional")

print("Step 5: Model(Neural Network) creation")

model = Sequential()

print("Step 5.1: Creating layers and add to the model")

layers12 = Dense(12, input_shape = (8, ), activation = 'relu')
layer3 = Dense(8, activation = 'relu')
layer4 = Dense(1, activation = 'sigmoid')

model.add(layers12)
model.add(layer3)
model.add(layer4)

print("Step 6: Model compilation")

model.compile(
    loss = 'binary_crossentropy',
    optimizer = 'adam',
    metrics = ['accuracy']
)

print("Step 7: Model training")

model.fit(
    X,
    y,
    epochs = 150,
    batch_size = 10,
)
```

Output

```
Epoch 147/150: accuracy: 0.7211 - loss: 0.5256
Epoch 148/150: accuracy: 0.7631 - loss: 0.5104
Epoch 149/150: accuracy: 0.7820 - loss: 0.4735
Epoch 150/150: accuracy: 0.7645 - loss: 0.4969
Epoch 151/150: accuracy: 0.7715 - loss: 0.4861
Epoch 152/150: accuracy: 0.7010 - loss: 0.5844
```

7. Evaluate the model

- ✓ Once training is done then we need to evaluate the performance of the network.
- ✓ By using evaluate() method we can evaluate.
- ✓ This method return two values,
 - The first will be the loss of the model on the dataset.
 - The second will be the accuracy of the model on the dataset.

```
Program    Fit the model
Name       demo7.py
Input file pima-indians-diabetes.csv

print("Topic: First DL Example")
print()

print("Step 1: Importing libraries")

from numpy import loadtxt
from tensorflow.keras.layers import Dense
from tensorflow.keras.models import Sequential

import warnings
warnings.filterwarnings("ignore")

print("Step 2: Loading the dataset")

dataset = loadtxt(
    'pima-indians-diabetes.csv',
    delimiter = ','
)

print("Step 3: Data preparation")

X = dataset[:, 0:8]
y = dataset[:, 8]
```

```
print("Step 4: Splitting the dataset: Optional")

print("Step 5: Model(Neural Network) creation")

model = Sequential()

print("Step 5.1: Creating layers and add to the model")

layers12 = Dense(12, input_shape = (8, ), activation = 'relu')
layer3 = Dense(8, activation = 'relu')
layer4 = Dense(1, activation = 'sigmoid')

model.add(layers12)
model.add(layer3)
model.add(layer4)

print("Step 6: Model compilation")

model.compile(
    loss = 'binary_crossentropy',
    optimizer = 'adam',
    metrics = ['accuracy']
)

print("Step 7: Model training")

model.fit(
    X,
    y,
    epochs = 150,
    batch_size = 10,
)

# Evaluate the keras model

_, accuracy = model.evaluate(X, y)
print("Accuracy is:", accuracy*100)
```

Output

```
Epoch 147/150  
←[1m77/77←[0m ←[32m ←[0m←[37m←[0m ←[1m0s←[0m 1ms/step - accuracy: 0.7728 - loss: 0.4662  
Epoch 148/150  
←[1m77/77←[0m ←[32m ←[0m←[37m←[0m ←[1m0s←[0m 1ms/step - accuracy: 0.8010 - loss: 0.4503  
Epoch 149/150  
←[1m77/77←[0m ←[32m ←[0m←[37m←[0m ←[1m0s←[0m 1ms/step - accuracy: 0.7617 - loss: 0.4846  
Epoch 150/150  
←[1m77/77←[0m ←[32m ←[0m←[37m←[0m ←[1m0s←[0m 1ms/step - accuracy: 0.7996 - loss: 0.4454  
←[1m24/24←[0m ←[32m ←[0m←[37m←[0m ←[1m0s←[0m 679us/step - accuracy: 0.7554 - loss: 0.4755  
Accuracy is: 78.25520634651184
```


8. Prediction

- ✓ By using predict(...) method we can do the prediction.
- ✓ We are using sigmoid activation function on the output layer.
 - The predictions will be a probability in the range between 0 and 1

```
Program      Model prediction
Name         demo8.py
Input file   pima-indians-diabetes.csv

print("Topic: First DL Example")
print()

print("Step 1: Importing libraries")

import numpy as np
from numpy import loadtxt
from tensorflow.keras.layers import Dense
from tensorflow.keras.models import Sequential

import warnings
warnings.filterwarnings("ignore")

print("Step 2: Loading the dataset")

dataset = loadtxt(
    'pima-indians-diabetes.csv',
    delimiter = ','
)

print("Step 3: Data preparation")

X = dataset[:, 0:8]
y = dataset[:, 8]
```

```
print("Step 4: Splitting the dataset: Optional")

print("Step 5: Model(Neural Network) creation")

model = Sequential()

print("Step 5.1: Creating layers and add to the model")

layers12 = Dense(12, input_shape = (8, ), activation = 'relu')
layer3 = Dense(8, activation = 'relu')
layer4 = Dense(1, activation = 'sigmoid')

model.add(layers12)
model.add(layer3)
model.add(layer4)

print("Step 6: Model compilation")

model.compile(
    loss = 'binary_crossentropy',
    optimizer = 'adam',
    metrics = ['accuracy']
)

print("Step 7: Model training")

model.fit(
    X,
    y,
    epochs = 150,
    batch_size = 10,
)
```

```
print("Step 8: Prediction")

ip = [[6, 148, 72, 35, 0, 33.6, 0.627, 50]]

input_data = np.array(ip)
result = model.predict(input_data)

print()
print(result)
```

Output

```
Epoch 148/150
+1m77/77+0m +32m-----+0m+37m+0m +1m0s+0m 1ms/step - accuracy: 0.7216 - loss: 0.5420
Epoch 149/150
+1m77/77+0m +32m-----+0m+37m+0m +1m0s+0m 2ms/step - accuracy: 0.7353 - loss: 0.5331
Epoch 150/150
+1m77/77+0m +32m-----+0m+37m+0m +1m0s+0m 1ms/step - accuracy: 0.6918 - loss: 0.5576
Step 8: Prediction
+1m1/1+0m +32m-----+0m+37m+0m +1m0s+0m 78ms/step
[[0.7929609]]
```

Note

- ✓ Here model done prediction for first five rows compared to the expected class value.
- ✓ You can see that most rows are correctly predicted.
- ✓ We got 79.2% accuracy with good model performance.

9. verbose = 0

- ✓ If we provide verbose = 0 then progress bar will be not displayed.

```
Program      Model prediction
Name         demo9.py
Input file   pima-indians-diabetes.csv

print("Topic: First DL Example")
print()

print("Step 1: Importing libraries")

from numpy import loadtxt
from tensorflow.keras.layers import Dense
from tensorflow.keras.models import Sequential

import warnings
warnings.filterwarnings("ignore")

print("Step 2: Loading the dataset")

dataset = loadtxt(
    'pima-indians-diabetes.csv',
    delimiter = ','
)

print("Step 3: Data preparation")

X = dataset[:, 0:8]
y = dataset[:, 8]

print("Step 4: Splitting the dataset: Optional")
```

```
print("Step 5: Model(Neural Network) creation")

model = Sequential()

print("Step 5.1: Creating layers and add to the model")

layers12 = Dense(12, input_shape = (8, ), activation = 'relu')
layer3 = Dense(8, activation = 'relu')
layer4 = Dense(1, activation = 'sigmoid')

model.add(layers12)
model.add(layer3)
model.add(layer4)

print("Step 6: Model compilation")

model.compile(
    loss = 'binary_crossentropy',
    optimizer = 'adam',
    metrics = ['accuracy']
)

print("Step 7: Model training")

model.fit(
    X,
    y,
    epochs = 150,
    batch_size = 10,
    verbose = 0

)

# Evaluate the keras model

_, accuracy = model.evaluate(X, y)
print("Accuracy is:", accuracy*100)
```

Output

```
Step 1: Importing libraries
2024-08-16 12:49:00.945172: I tensorflow
erent numerical results due to floating-
e environment variable `TF_ENABLE_ONEDNN
2024-08-16 12:49:02.237748: I tensorflow
erent numerical results due to floating-
e environment variable `TF_ENABLE_ONEDNN
Step 2: Loading the dataset
Step 3: Data preparation
Step 4: Splitting the dataset: Optional
Step 5: Model(Neural Network) creation
Step 5.1: Creating layers and add to the
2024-08-16 12:49:05.673091: I tensorflow
e available CPU instructions in performan
To enable the following instructions: AVX
s.
Step 6: Model compilation
Step 7: Model training
←[1m24/24←[0m ←[32m—————←
Accuracy is: 69.79166865348816
```

10. Model summary

- ✓ By using summary method, we can check the summary of the model.

```
Program      Model summary
Name         demo10.py
Input file   pima-indians-diabetes.csv

print("Topic: First DL Example")
print()

print("Step 1: Importing libraries")

from numpy import loadtxt
from tensorflow.keras.layers import Dense
from tensorflow.keras.models import Sequential

import warnings
warnings.filterwarnings("ignore")

print("Step 2: Loading the dataset")

dataset = loadtxt(
    'pima-indians-diabetes.csv',
    delimiter = ','
)

print("Step 3: Data preparation")

X = dataset[:, 0:8]
y = dataset[:, 8]

print("Step 4: Splitting the dataset: Optional")
```

```
print("Step 5: Model(Neural Network) creation")

model = Sequential()

print("Step 5.1: Creating layers and add to the model")

layers12 = Dense(12, input_shape = (8, ), activation = 'relu')
layer3 = Dense(8, activation = 'relu')
layer4 = Dense(1, activation = 'sigmoid')

model.add(layers12)
model.add(layer3)
model.add(layer4)

print("Step 6: Model compilation")

model.compile(
    loss = 'binary_crossentropy',
    optimizer = 'adam',
    metrics = ['accuracy']
)

print("Step 7: Model training")

model.fit(
    X,
    y,
    epochs = 150,
    batch_size = 10,
    verbose = 0

)

# Model Summary

model.summary()
```


Output

Step 6: Model compilation

Step 7: Model training

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 12)	108
dense_1 (Dense)	(None, 8)	104
dense_2 (Dense)	(None, 1)	9

Total params: 665 (2.60 KB)

Trainable params: 221 (884.00 B)

Non-trainable params: 0 (0.00 B)

Optimizer params: 444 (1.74 KB)

11. Image of the model

- ✓ We can get the image of the model

```
Program      Image of the model
Name         demo11.py
Input file   pima-indians-diabetes.csv

# importing required libraries
from numpy import loadtxt
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense

# load the dataset
dataset = loadtxt('pima-indians-diabetes.csv', delimiter = ',')

# split into input (X) and output (y) variables
X = dataset[:, 0:8]
y = dataset[:, 8]

# define the keras model

model = Sequential()

layer1 = Dense(12, input_shape = (8, ), activation = 'relu')
layer2 = Dense(8, activation = 'relu')
layer3 = Dense(1, activation = 'sigmoid')

model.add(layer1)
model.add(layer2)
model.add(layer3)

# compile the keras model
model.compile(loss = 'binary_crossentropy', optimizer = 'adam',
metrics = ['accuracy'])

# fit the keras model on the dataset
model.fit(X, y, epochs = 150, batch_size = 10, verbose = 0)
```

```
from tensorflow.keras.utils import plot_model
```

```
plot_model(model, to_file = 'model.png', show_shapes = True,  
show_dtype = True, show_layer_names = True, expand_nested =  
True, show_layer_activations = True)
```

Output

