## In [1]:

```
import sys
import pandas as pd
import numpy as np
import sklearn
import keras

print(sys.version)
print(pd.__version__)
print(np.__version__)
print(sklearn.__version__)
print(keras.__version__)
```

Using TensorFlow backend.

```
3.7.6 (default, Jan 8 2020, 20:23:39) [MSC v.1916 64 bit (AMD64)] 0.24.2 1.18.1 0.22.1 2.3.1
```

## In [2]:

## In [3]:

```
df.describe()
```

## Out[3]:

	n_pregnant	glucose_concentration	blood_pressuer (mm Hg)	skin_thickness (mm)	serum_insulin (mu U/ml)	
count	768.000000	768.000000	768.000000	768.000000	768.000000	76
mean	3.845052	120.894531	69.105469	20.536458	79.799479	3
std	3.369578	31.972618	19.355807	15.952218	115.244002	
min	0.000000	0.000000	0.000000	0.000000	0.000000	
25%	1.000000	99.000000	62.000000	0.000000	0.000000	2
50%	3.000000	117.000000	72.000000	23.000000	30.500000	3
75%	6.000000	140.250000	80.000000	32.000000	127.250000	3
max	17.000000	199.000000	122.000000	99.000000	846.000000	6
4						•

# In [4]:

```
df[df['glucose_concentration']==0]
```

### Out[4]:

	n_pregnant	glucose_concentration	blood_pressuer (mm Hg)	skin_thickness (mm)	serum_insulin (mu U/ml)	ВМІ
75	1	0	48	20	0	24.7
182	1	0	74	20	23	27.7
342	1	0	68	35	0	32.0
349	5	0	80	32	0	41.0
502	6	0	68	41	0	39.0
4						•

# In [5]:

```
columns = ['glucose_concentration', 'blood_pressuer (mm Hg)', 'skin_thickness (mm)', 's
erum_insulin (mu U/ml)', 'BMI']
for col in columns:
    df[col].replace(0,np.NaN,inplace=True)
```

# In [6]:

```
# Drop the rows with missing values
df.dropna(inplace=True)

# summarize the number of rows and columns in df
df.describe()
```

## Out[6]:

	n_pregnant	glucose_concentration	blood_pressuer (mm Hg)	skin_thickness (mm)	serum_insulin (mu U/ml)	
count	392.000000	392.000000	392.000000	392.000000	392.000000	39
mean	3.301020	122.627551	70.663265	29.145408	156.056122	3
std	3.211424	30.860781	12.496092	10.516424	118.841690	
min	0.000000	56.000000	24.000000	7.000000	14.000000	1
25%	1.000000	99.000000	62.000000	21.000000	76.750000	2
50%	2.000000	119.000000	70.000000	29.000000	125.500000	3
75%	5.000000	143.000000	78.000000	37.000000	190.000000	3
max	17.000000	198.000000	110.000000	63.000000	846.000000	6
4						•

# In [7]:

```
dataset = df.values
print(dataset.shape)
```

(392, 9)

#### In [8]:

#### In [9]:

```
print(X.shape)
print(Y.shape)
#print(X)
```

(392, 8) (392,)

# In [10]:

```
# Normalize the data using sklearn StandardScaler
from sklearn.preprocessing import StandardScaler
scaler=StandardScaler().fit(X)
print(scaler)
```

StandardScaler(copy=True, with\_mean=True, with\_std=True)

#### In [11]:

```
# Transform and display the training data
X_standardized = scaler.transform(X)
data = pd.DataFrame(X_standardized)
data.describe()
```

#### Out[11]:

	0	1	2	3	4	
count	3.920000e+02	3.920000e+02	3.920000e+02	3.920000e+02	3.920000e+02	3.920000e
mean	-4.021726e-17	3.129583e-17	-4.641624e-16	1.042250e-16	6.485742e-17	1.543550€
std	1.001278e+00	1.001278e+00	1.001278e+00	1.001278e+00	1.001278e+00	1.001278e
min	-1.029213e+00	-2.161731e+00	-3.739001e+00	-2.108484e+00	-1.196867e+00	-2.120941e
25%	-7.174265e-01	-7.665958e-01	-6.941640e-01	-7.755315e-01	-6.681786e-01	-6.676780€
50%	-4.056403e-01	-1.176959e-01	-5.314565e-02	-1.384444e-02	-2.574448e-01	1.621036€
75%	5.297185e-01	6.609841e-01	5.878727e-01	7.478426e-01	2.859877e-01	5.718696€
max	4.271153e+00	2.445459e+00	3.151946e+00	3.223325e+00	5.812990e+00	4.846172e
4						•

## In [12]:

```
#import necessary packages
from sklearn.model_selection import GridSearchCV, KFold
from keras.models import Sequential
from keras.layers import Dense
from keras.wrappers.scikit_learn import KerasClassifier
from keras.optimizers import Adam
```

# In [13]:

```
#Start defining the model

def create_model():
    model=Sequential()
    model.add(Dense(8,input_dim=8,kernel_initializer='normal',activation='relu'))
    model.add(Dense(4,input_dim=8,kernel_initializer='normal',activation='relu'))
    model.add(Dense(1,activation='sigmoid'))

    adam = Adam(lr = 0.01)
    model.compile(loss = 'binary_crossentropy', optimizer = adam, metrics = ['accuracy'])
    return model
```

## In [14]:

```
model = create_model()
print(model.summary())
```

## Model: "sequential\_1"

Layer (type)	Output Shape	Param #
dense_1 (Dense)	(None, 8)	72
dense_2 (Dense)	(None, 4)	36
dense_3 (Dense)	(None, 1)	5
T   1		

Total params: 113 Trainable params: 113 Non-trainable params: 0

None

## In [15]:

```
#define a random seed
seed=6
np.random.seed(seed)
#Start defining the model
def create model():
    model=Sequential()
    model.add(Dense(8,input_dim=8,kernel_initializer='normal',activation='relu'))
    model.add(Dense(4,input_dim=8,kernel_initializer='normal',activation='relu'))
    model.add(Dense(1,activation='sigmoid'))
    adam = Adam(lr = 0.01)
    model.compile(loss = 'binary crossentropy', optimizer = adam, metrics = ['accuracy'
])
    return model
#create model
model=KerasClassifier(build_fn=create_model, verbose=0)
# Define the grid search parameters
batch_size = [10,20,30,40]
epochs = [10, 50, 100]
# make a dictionary of the grid search parameters
param_grid = dict(batch_size=batch_size, epochs=epochs)
# build and fit the GridSearchCV
grid = GridSearchCV(estimator=model, param_grid=param_grid, cv = KFold(random_state=see
d), verbose=10)
grid_result = grid.fit(X_standardized, Y)
# summarize the results
print("Best: {0}, using {1}".format(grid_result.best_score_, grid_result.best_params_))
means = grid_result.cv_results_['mean_test_score']
stds = grid_result.cv_results_['std_test_score']
params = grid_result.cv_results_['params']
for mean, stdev, param in zip(means, stds, params):
    print('{0} ({1}) with: {2}'.format(mean, stdev, param))
```

C:\Users\balsh\anaconda3\lib\site-packages\sklearn\model selection\ split. py:296: FutureWarning: Setting a random state has no effect since shuffle is False. This will raise an error in 0.24. You should leave random state to its default (None), or set shuffle=True. FutureWarning [Parallel(n jobs=1)]: Using backend SequentialBackend with 1 concurrent wo rkers. Fitting 5 folds for each of 12 candidates, totalling 60 fits [CV] batch\_size=10, epochs=10 ...... [CV] ..... batch\_size=10, epochs=10, score=0.722, total= 1.8s [CV] batch\_size=10, epochs=10 ...... [Parallel(n\_jobs=1)]: Done 1 out of 1 | elapsed: 1.7s remaining: 0.0s [CV] ..... batch size=10, epochs=10, score=0.582, total= 1.5s [CV] batch\_size=10, epochs=10 ..... [Parallel(n\_jobs=1)]: Done 2 out of 2 | elapsed: 3.2s remaining: 0.0s [CV] ..... batch\_size=10, epochs=10, score=0.808, total= 1.8s [CV] batch\_size=10, epochs=10 ...... [Parallel(n\_jobs=1)]: Done 3 out of 3 | elapsed: 5.0s remaining: 0.0s [CV] ..... batch\_size=10, epochs=10, score=0.795, total= 1.7s [CV] batch size=10, epochs=10 ..... [Parallel(n jobs=1)]: Done 4 out of 4 | elapsed: 6.8s remaining: 0.0s [CV] ..... batch\_size=10, epochs=10, score=0.846, total= [CV] batch\_size=10, epochs=50 ..... [Parallel(n\_jobs=1)]: Done 5 out of 5 | elapsed: 8.4s remaining: 0.0s [CV] ..... batch\_size=10, epochs=50, score=0.722, total= 4.4s [CV] batch size=10, epochs=50 ..... [Parallel(n jobs=1)]: Done 6 out of 6 | elapsed: 12.8s remaining: 0.0s [CV] ..... batch\_size=10, epochs=50, score=0.709, total= 4.4s [CV] batch size=10, epochs=50 ..... [Parallel(n jobs=1)]: Done 7 out of 7 | elapsed: 17.2s remaining: 0.0s [CV] ..... batch\_size=10, epochs=50, score=0.821, total= 4.5s [CV] batch size=10, epochs=50 ..... [Parallel(n\_jobs=1)]: Done 8 out of 8 | elapsed: 21.7s remaining: 0.0s [CV] ..... batch size=10, epochs=50, score=0.846, total= [CV] batch\_size=10, epochs=50 .....

[Parallel(n\_jobs=1)]: Done 9 out of 9 | elapsed: 26.1s remaining:

0.0s

```
[CV] ..... batch size=10, epochs=50, score=0.859, total= 4.0s
[CV] batch size=10, epochs=100 ......
[CV] ...... batch_size=10, epochs=100, score=0.772, total= 8.4s
[CV] batch_size=10, epochs=100 .....
[CV] ..... batch_size=10, epochs=100, score=0.671, total= 8.2s
[CV] batch_size=10, epochs=100 .....
[CV] ..... batch size=10, epochs=100, score=0.795, total= 7.4s
[CV] batch_size=10, epochs=100 ......
[CV] ..... batch_size=10, epochs=100, score=0.833, total= 7.7s
[CV] batch_size=10, epochs=100 ......
[CV] ..... batch_size=10, epochs=100, score=0.756, total= 7.8s
[CV] batch size=20, epochs=10 .....
[CV] ..... batch size=20, epochs=10, score=0.772, total= 1.3s
[CV] batch_size=20, epochs=10 ......
[CV] ..... batch_size=20, epochs=10, score=0.582, total= 1.1s
[CV] batch_size=20, epochs=10 ......
[CV] ...... batch_size=20, epochs=10, score=0.808, total= 1.2s
[CV] batch_size=20, epochs=10 ......
[CV] ..... batch_size=20, epochs=10, score=0.808, total= 1.1s
[CV] batch size=20, epochs=10 .....
[CV] ..... batch_size=20, epochs=10, score=0.821, total= 1.3s
[CV] batch size=20, epochs=50 ......
[CV] ..... batch_size=20, epochs=50, score=0.759, total= 2.5s
[CV] batch_size=20, epochs=50 .....
[CV] ..... batch_size=20, epochs=50, score=0.620, total= 2.6s
[CV] batch_size=20, epochs=50 .....
[CV] ..... batch_size=20, epochs=50, score=0.833, total= 2.6s
[CV] batch_size=20, epochs=50 .....
[CV] ..... batch_size=20, epochs=50, score=0.872, total= 2.6s
[CV] batch_size=20, epochs=50 .....
[CV] ..... batch size=20, epochs=50, score=0.859, total= 2.5s
[CV] batch_size=20, epochs=100 ......
[CV] ...... batch_size=20, epochs=100, score=0.759, total= 4.4s
[CV] batch_size=20, epochs=100 ......
[CV] ..... batch_size=20, epochs=100, score=0.646, total= 4.2s
[CV] batch_size=20, epochs=100 .....
[CV] ..... batch_size=20, epochs=100, score=0.795, total= 4.5s
[CV] batch_size=20, epochs=100 ......
[CV] ..... batch_size=20, epochs=100, score=0.782, total= 4.4s
[CV] batch size=20, epochs=100 ......
[CV] ...... batch_size=20, epochs=100, score=0.833, total= 4.5s
[CV] batch size=30, epochs=10 ......
[CV] ..... batch size=30, epochs=10, score=0.734, total= 1.1s
[CV] batch size=30, epochs=10 ......
[CV] ..... batch size=30, epochs=10, score=0.608, total= 1.1s
[CV] batch size=30, epochs=10 ......
[CV] ...... batch_size=30, epochs=10, score=0.808, total= 1.2s
[CV] batch size=30, epochs=10 ......
[CV] ..... batch size=30, epochs=10, score=0.808, total= 1.1s
[CV] batch_size=30, epochs=10 ......
[CV] ..... batch size=30, epochs=10, score=0.821, total= 1.0s
[CV] batch_size=30, epochs=50 ......
[CV] ..... batch size=30, epochs=50, score=0.709, total= 2.0s
[CV] batch size=30, epochs=50 .....
[CV] ..... batch size=30, epochs=50, score=0.658, total= 2.0s
[CV] batch size=30, epochs=50 .....
[CV] ..... batch size=30, epochs=50, score=0.859, total= 2.3s
[CV] batch_size=30, epochs=50 ......
[CV] ..... batch_size=30, epochs=50, score=0.846, total= 2.0s
[CV] batch size=30, epochs=50 .....
[CV] ..... batch_size=30, epochs=50, score=0.846, total=
```

[CV]	batch_size=30, epochs=100	
	batch_size=30, epochs=100, score=0.722, total=	
	batch_size=30, epochs=100	
	batch_size=30, epochs=100, score=0.671, total=	3.3s
[CV]	batch_size=30, epochs=100	
	batch_size=30, epochs=100, score=0.821, total=	3.5s
	batch_size=30, epochs=100	
	batch_size=30, epochs=100, score=0.859, total=	3.1s
	batch_size=30, epochs=100	
	batch_size=30, epochs=100, score=0.833, total=	3.4s
	batch_size=40, epochs=10	• • • • •
	batch_size=40, epochs=10, score=0.785, total=	1.0s
	batch_size=40, epochs=10	
	batch_size=40, epochs=10, score=0.620, total=	1.0s
	batch_size=40, epochs=10	
	batch_size=40, epochs=10, score=0.795, total=	1.2s
	batch_size=40, epochs=10	
	batch_size=40, epochs=10, score=0.846, total=	1.0s
	batch_size=40, epochs=10	
	batch_size=40, epochs=10, score=0.859, total=	<b>1.</b> 2s
	batch_size=40, epochs=50	
	batch_size=40, epochs=50, score=0.785, total=	1.9s
	batch_size=40, epochs=50 batch_size=40, epochs=50, score=0.633, total=	1.7s
	batch_size=40, epochs=50	
	batch_size=40, epochs=50, score=0.833, total=	1.7s
	batch_size=40, epochs=50	1./3
	batch_size=40, epochs=50, score=0.821, total=	1.9s
	batch_size=40, epochs=50	
	batch_size=40, epochs=50, score=0.846, total=	1.7s
	batch_size=40, epochs=100	
	batch size=40, epochs=100, score=0.722, total=	2.6s
	batch_size=40, epochs=100	
	batch_size=40, epochs=100, score=0.658, total=	2.7s
	batch_size=40, epochs=100	
	batch_size=40, epochs=100, score=0.769, total=	2.6s
	batch_size=40, epochs=100	
	batch_size=40, epochs=100, score=0.821, total=	2.8s
[cv]	batch_size=40, epochs=100	
		2.75

[Parallel(n\_jobs=1)]: Done 60 out of 60 | elapsed: 2.8min finished

```
Best: 0.7912041544914246, using {'batch_size': 10, 'epochs': 50}
0.7505030870437622 (0.09330364448009798) with: {'batch size': 10, 'epoch
s': 10}
0.7912041544914246 (0.06341564947695214) with: {'batch size': 10, 'epoch
s': 50}
0.7655306696891785 (0.053930915597060584) with: {'batch size': 10, 'epoch
s': 100}
0.7580655574798584 (0.08935860875152117) with: {'batch_size': 20, 'epoch
0.7887698650360108 (0.09280377518135459) with: {'batch_size': 20, 'epoch
s': 50}
0.763063931465149 (0.06344715185877535) with: {'batch size': 20, 'epochs':
0.7555339097976684 (0.08000663660471047) with: {'batch_size': 30, 'epoch
0.783674132823944 (0.08344023121060123) with: {'batch_size': 30, 'epochs':
0.781045114994049 (0.07217075299611107) with: {'batch_size': 30, 'epochs':
100}
0.7810126543045044 (0.08528952648905208) with: {'batch size': 40, 'epoch
s': 10}
0.7835443019866943 (0.078048000929465) with: {'batch size': 40, 'epochs':
0.7656929612159729 (0.07102004888718165) with: {'batch_size': 40, 'epoch
s': 100}
```

#### In [16]:

```
# import necessary packages
from keras.layers import Dropout
# Define a random seed
seed = 6
np.random.seed(seed)
epochs=50
batch_size=10
# Start defining the model
def create_model(learn_rate, dropout_rate):
    # Create model
   model = Sequential()
   model.add(Dense(8, input dim = 8, kernel initializer = 'normal', activation = 'rel
u'))
    model.add(Dropout(dropout rate))
   model.add(Dense(4, input_dim = 8, kernel_initializer = 'normal', activation = 'rel
u'))
   model.add(Dropout(dropout_rate))
    model.add(Dense(1, activation = 'sigmoid'))
    # Compile the model
    adam = Adam(lr = learn_rate)
    model.compile(loss = 'binary_crossentropy', optimizer = adam, metrics = ['accuracy'
1)
    return model
# Create the model
model = KerasClassifier(build_fn = create_model, epochs=epochs, batch_size=batch_size,
verbose = 0) # This comes from the previous best
# define the grid search parameters
learn_rate = [0.0001, 0.001, 0.01, 0.1]
dropout_rate = [0.0, 0.1, 0.2, 0.3]
# make a dictionary of the grid search parameters
param_grid = dict(learn_rate=learn_rate, dropout_rate=dropout_rate)
# build and fit the GridSearchCV
grid = GridSearchCV(estimator=model, param_grid=param_grid, cv = KFold(random_state=see
d), verbose=10)
grid result = grid.fit(X standardized, Y)
# summarize the results
print("Best: {0}, using {1}".format(grid result.best score , grid result.best params ))
means = grid_result.cv_results_['mean_test_score']
stds = grid_result.cv_results_['std_test_score']
params = grid_result.cv_results_['params']
for mean, stdev, param in zip(means, stds, params):
    print('{0} ({1}) with: {2}'.format(mean, stdev, param))
```

Fitting 5 folds for each of 16 candidates, totalling 80 fits [CV] dropout_rate=0.0, learn_rate=0.0001
<pre>[Parallel(n_jobs=1)]: Using backend SequentialBackend with 1 concurrent rkers.</pre>
<pre>[CV] . dropout_rate=0.0, learn_rate=0.0001, score=0.759, total= 4.6s [CV] dropout_rate=0.0, learn_rate=0.0001</pre>
<pre>[Parallel(n_jobs=1)]: Done  1 out of  1   elapsed:  4.5s remaining: 0.0s</pre>
<pre>[CV] . dropout_rate=0.0, learn_rate=0.0001, score=0.620, total= 4.2s [CV] dropout_rate=0.0, learn_rate=0.0001</pre>
<pre>[Parallel(n_jobs=1)]: Done  2 out of  2   elapsed: 8.7s remaining: 0.0s</pre>
<pre>[CV] . dropout_rate=0.0, learn_rate=0.0001, score=0.821, total= 4.2s [CV] dropout_rate=0.0, learn_rate=0.0001</pre>
<pre>[Parallel(n_jobs=1)]: Done  3 out of  3   elapsed: 13.0s remaining: 0.0s</pre>
<pre>[CV] . dropout_rate=0.0, learn_rate=0.0001, score=0.769, total= 4.5s [CV] dropout_rate=0.0, learn_rate=0.0001</pre>
<pre>[Parallel(n_jobs=1)]: Done  4 out of  4   elapsed: 17.5s remaining: 0.0s</pre>
<pre>[CV] . dropout_rate=0.0, learn_rate=0.0001, score=0.769, total= 4.2s [CV] dropout_rate=0.0, learn_rate=0.001</pre>
<pre>[Parallel(n_jobs=1)]: Done 5 out of 5   elapsed: 21.7s remaining: 0.0s</pre>
<pre>[CV] dropout_rate=0.0, learn_rate=0.001, score=0.759, total= 4.3s [CV] dropout_rate=0.0, learn_rate=0.001</pre>
<pre>[Parallel(n_jobs=1)]: Done 6 out of 6   elapsed: 26.0s remaining: 0.0s</pre>
<pre>[CV] dropout_rate=0.0, learn_rate=0.001, score=0.633, total= 4.4s [CV] dropout_rate=0.0, learn_rate=0.001</pre>
<pre>[Parallel(n_jobs=1)]: Done 7 out of 7   elapsed: 30.4s remaining: 0.0s</pre>
<pre>[CV] dropout_rate=0.0, learn_rate=0.001, score=0.808, total= 4.3s [CV] dropout_rate=0.0, learn_rate=0.001</pre>
<pre>[Parallel(n_jobs=1)]: Done 8 out of 8   elapsed: 34.7s remaining: 0.0s</pre>
[CV] dropout_rate=0.0, learn_rate=0.001, score=0.808, total= 4.4s [CV] dropout_rate=0.0, learn_rate=0.001
<pre>[Parallel(n_jobs=1)]: Done 9 out of 9   elapsed: 39.1s remaining: 0.0s</pre>

WO

F C \ / 1	description of a Computation of the Computation of	1	1 -
	dropout_rate=0.0, learn_rate=0.001, score=0.833, tot	ca1=	4.6s
[CV]	dropout_rate=0.0, learn_rate=0.01		• • • •
[CV]	dropout_rate=0.0, learn_rate=0.01, score=0.747, tot	tal=	4.3s
[CV]	dropout_rate=0.0, learn_rate=0.01		
	dropout_rate=0.0, learn_rate=0.01, score=0.671, tot		4.3s
	dropout_rate=0.0, learn_rate=0.01	LUI-	т. ЭЭ
	dropout_rate=0.0, learn_rate=0.01, score=0.769, tot		4.3s
[CV]	dropout_rate=0.0, learn_rate=0.01		
[CV]	dropout_rate=0.0, learn_rate=0.01, score=0.833, tot	tal=	4.3s
[CV]	dropout_rate=0.0, learn_rate=0.01		
	dropout_rate=0.0, learn_rate=0.01, score=0.859, tot		4.3s
			7.73
	dropout_rate=0.0, learn_rate=0.1		
	dropout_rate=0.0, learn_rate=0.1, score=0.734, tot	ca1=	4.6s
[CV]	dropout_rate=0.0, learn_rate=0.1		• • • •
[CV]	dropout_rate=0.0, learn_rate=0.1, score=0.658, tot	tal=	4.2s
[CV]	<pre>dropout_rate=0.0, learn_rate=0.1</pre>		
	dropout_rate=0.0, learn_rate=0.1, score=0.705, tot	al=	4.2s
	dropout_rate=0.0, learn_rate=0.1		
			1 10
	dropout_rate=0.0, learn_rate=0.1, score=0.821, tot		4.4s
	dropout_rate=0.0, learn_rate=0.1		• • • •
[CV]	dropout_rate=0.0, learn_rate=0.1, score=0.795, tot	tal=	4.4s
[CV]	dropout_rate=0.1, learn_rate=0.0001		
[CV]	. dropout_rate=0.1, learn_rate=0.0001, score=0.684, tot	tal=	5.0s
	dropout_rate=0.1, learn_rate=0.0001		
	. dropout_rate=0.1, learn_rate=0.0001, score=0.570, tot		5.3s
	• • • • • • • • • • • • • • • • • • • •		J.J3
	dropout_rate=0.1, learn_rate=0.0001		••••
	. dropout_rate=0.1, learn_rate=0.0001, score=0.718, tot	tal=	5.1s
[CV]	dropout_rate=0.1, learn_rate=0.0001		
[CV]	. dropout_rate=0.1, learn_rate=0.0001, score=0.744, tot	tal=	5.0s
[CV]	dropout_rate=0.1, learn_rate=0.0001		
	. dropout_rate=0.1, learn_rate=0.0001, score=0.679, tot	al=	5.1s
	• • • • • • • • • • • • • • • • • • • •	LUI-	J. <b>1</b> J
- E - E -	dropout_rate=0.1, learn_rate=0.001		
[CV]	. = =	ca1=	5.1s
	dropout_rate=0.1, learn_rate=0.001		• • • •
[CV]	dropout_rate=0.1, learn_rate=0.001, score=0.633, tot	tal=	5.4s
[CV]	<pre>dropout_rate=0.1, learn_rate=0.001</pre>		
	dropout_rate=0.1, learn_rate=0.001, score=0.821, tot		
	dropout_rate=0.1, learn_rate=0.001		
	dropout_rate=0.1, learn_rate=0.001, score=0.833, tot		
	dropout_rate=0.1, learn_rate=0.001		
[CV]	dropout_rate=0.1, learn_rate=0.001, score=0.833, tot	tal=	5.1s
[CV]	dropout_rate=0.1, learn_rate=0.01		
[CV]	dropout rate=0.1, learn rate=0.01, score=0.772, tot	tal=	5.3s
	dropout_rate=0.1, learn_rate=0.01		
	dropout_rate=0.1, learn_rate=0.01, score=0.658, tot		
	dropout_rate=0.1, learn_rate=0.01		
[CV]	dropout_rate=0.1, learn_rate=0.01, score=0.821, tot	tal=	5.1s
[CV]	dropout_rate=0.1, learn_rate=0.01		
[CV]	dropout_rate=0.1, learn_rate=0.01, score=0.885, tot	tal=	5.1s
[CV]	dropout_rate=0.1, learn_rate=0.01		
	dropout_rate=0.1, learn_rate=0.01, score=0.808, tot		
	dropout_rate=0.1, learn_rate=0.1		
	dropout_rate=0.1, learn_rate=0.1, score=0.658, tot		
	dropout_rate=0.1, learn_rate=0.1		
[CV]	dropout_rate=0.1, learn_rate=0.1, score=0.570, tot	tal=	5.0s
[CV]	dropout_rate=0.1, learn_rate=0.1	. <b></b> .	
	dropout_rate=0.1, learn_rate=0.1, score=0.782, tot		
	dropout_rate=0.1, learn_rate=0.1		
	dropout_rate=0.1, learn_rate=0.1, score=0.769, tot		
	dropout_rate=0.1, learn_rate=0.1		
I CV I	dropout rate=0.1, learn rate=0.1, score=0.731, tot	al=	4 95

[CV]	dropout_rate=0.2, learn_rate=0.0001	
[CV]	<pre>. dropout_rate=0.2, learn_rate=0.0001, score=0.759, total=</pre>	4.9s
[CV]	<pre>dropout_rate=0.2, learn_rate=0.0001</pre>	
[CV]	<pre>. dropout_rate=0.2, learn_rate=0.0001, score=0.633, total=</pre>	5.1s
[CV]	dropout_rate=0.2, learn_rate=0.0001	
	. dropout_rate=0.2, learn_rate=0.0001, score=0.705, total=	5.2s
	dropout_rate=0.2, learn_rate=0.0001	
	. dropout_rate=0.2, learn_rate=0.0001, score=0.833, total=	5.3s
	dropout_rate=0.2, learn_rate=0.0001	
	. dropout_rate=0.2, learn_rate=0.0001, score=0.705, total=	4.9s
	dropout_rate=0.2, learn_rate=0.001	
	dropout_rate=0.2, learn_rate=0.001, score=0.759, total=	5.1s
	dropout_rate=0.2, learn_rate=0.001	
	dropout_rate=0.2, learn_rate=0.001, score=0.620, total=	5.2s
	dropout_rate=0.2, learn_rate=0.001	3.23
	dropout_rate=0.2, learn_rate=0.001, score=0.821, total=	5.0s
	dropout_rate=0.2, learn_rate=0.001	3.03
	dropout_rate=0.2, learn_rate=0.001, score=0.859, total=	5.0s
[CV]	-	J.03
[CV]	· · · · · · · · · · · · · · · · · · ·	5.2s
	dropout_rate=0.2, learn_rate=0.01	3.23
	dropout_rate=0.2, learn_rate=0.01, score=0.747, total=	5.5s
	dropout_rate=0.2, learn_rate=0.01	J.J3
	dropout_rate=0.2, learn_rate=0.01, score=0.658, total=	4.9s
	dropout_rate=0.2, learn_rate=0.01	4.53
	dropout_rate=0.2, learn_rate=0.01, score=0.833, total=	5.1s
	dropout_rate=0.2, learn_rate=0.01	
	dropout_rate=0.2, learn_rate=0.01, score=0.833, total=	5.1s
	dropout_rate=0.2, learn_rate=0.01	4.9s
	dropout_rate=0.2, learn_rate=0.01, score=0.795, total=	4.95
	dropout_rate=0.2, learn_rate=0.1	F 0c
	dropout_rate=0.2, learn_rate=0.1, score=0.646, total=	5.0s
	dropout_rate=0.2, learn_rate=0.1	г 1а
	dropout_rate=0.2, learn_rate=0.1, score=0.646, total=	5.1s
	dropout_rate=0.2, learn_rate=0.1	
	dropout_rate=0.2, learn_rate=0.1, score=0.769, total=	
	dropout_rate=0.2, learn_rate=0.1	
	dropout_rate=0.2, learn_rate=0.1, score=0.808, total=	
	dropout_rate=0.2, learn_rate=0.1	
	dropout_rate=0.2, learn_rate=0.1, score=0.808, total=	
	dropout_rate=0.3, learn_rate=0.0001	
	. dropout_rate=0.3, learn_rate=0.0001, score=0.722, total=	
	dropout_rate=0.3, learn_rate=0.0001	
	. dropout_rate=0.3, learn_rate=0.0001, score=0.595, total=	
	dropout_rate=0.3, learn_rate=0.0001	
	. dropout_rate=0.3, learn_rate=0.0001, score=0.705, total=	
	dropout_rate=0.3, learn_rate=0.0001	
	. dropout_rate=0.3, learn_rate=0.0001, score=0.808, total=	
	dropout_rate=0.3, learn_rate=0.0001	
	. dropout_rate=0.3, learn_rate=0.0001, score=0.731, total=	
	dropout_rate=0.3, learn_rate=0.001	
	dropout_rate=0.3, learn_rate=0.001, score=0.734, total=	
	dropout_rate=0.3, learn_rate=0.001	
	dropout_rate=0.3, learn_rate=0.001, score=0.582, total=	
	dropout_rate=0.3, learn_rate=0.001	
	dropout_rate=0.3, learn_rate=0.001, score=0.821, total=	5.15
	dropout_rate=0.3, learn_rate=0.001	
	dropout_rate=0.3, learn_rate=0.001, score=0.808, total=	
	dropout_rate=0.3, learn_rate=0.001	
	dropout_rate=0.3, learn_rate=0.001, score=0.859, total=	
[CV]	dropout_rate=0.3, learn_rate=0.01	• • • • •

```
[CV] ... dropout_rate=0.3, learn_rate=0.01, score=0.747, total= 5.6s
[CV] dropout_rate=0.3, learn_rate=0.01 ......
[CV] ... dropout rate=0.3, learn rate=0.01, score=0.646, total= 5.8s
[CV] dropout_rate=0.3, learn_rate=0.01 .....
[CV] ... dropout rate=0.3, learn rate=0.01, score=0.782, total= 6.3s
[CV] dropout_rate=0.3, learn_rate=0.01 .....
[CV] ... dropout_rate=0.3, learn_rate=0.01, score=0.795, total= 5.3s
[CV] dropout_rate=0.3, learn_rate=0.01 ......
[CV] ... dropout rate=0.3, learn rate=0.01, score=0.821, total= 5.5s
[CV] dropout_rate=0.3, learn_rate=0.1 .....
[CV] .... dropout_rate=0.3, learn_rate=0.1, score=0.646, total= 5.0s
[CV] dropout_rate=0.3, learn_rate=0.1 .....
[CV] .... dropout_rate=0.3, learn_rate=0.1, score=0.595, total= 5.1s
[CV] dropout_rate=0.3, learn_rate=0.1 .....
[CV] .... dropout_rate=0.3, learn_rate=0.1, score=0.795, total= 4.9s
[CV] dropout rate=0.3, learn rate=0.1 ......
[CV] .... dropout_rate=0.3, learn_rate=0.1, score=0.744, total= 4.9s
[CV] dropout_rate=0.3, learn_rate=0.1 .....
[CV] .... dropout_rate=0.3, learn_rate=0.1, score=0.679, total= 4.9s
[Parallel(n jobs=1)]: Done 80 out of 80 | elapsed: 6.6min finished
Best: 0.78864004611969, using {'dropout rate': 0.1, 'learn rate': 0.01}
0.7477442383766174 (0.06724676745381974) with: {'dropout_rate': 0.0, 'lear
n rate': 0.0001}
0.768224585056305 (0.07175017418186987) with: {'dropout_rate': 0.0, 'learn
_rate': 0.001}
0.7758520007133484 (0.06650874261352188) with: {'dropout_rate': 0.0, 'lear
n rate': 0.01}
0.7425835847854614 (0.05898556932094381) with: {'dropout_rate': 0.0, 'lear
n_rate': 0.1}
0.6788380384445191 (0.05945338826029287) with: {'dropout_rate': 0.1, 'lear
n_rate': 0.0001}
0.7759168982505799 (0.07655375483984386) with: {'dropout_rate': 0.1, 'lear
n rate': 0.001}
0.78864004611969 (0.07465831638149004) with: {'dropout rate': 0.1, 'learn
rate': 0.01}
0.7019798755645752 (0.07898891947698417) with: {'dropout_rate': 0.1, 'lear
n_rate': 0.1}
0.7271989464759827 (0.06659103905495732) with: {'dropout rate': 0.2, 'lear
n rate': 0.0001}
0.7810775637626648 (0.08739374146471103) with: {'dropout rate': 0.2, 'lear
n rate': 0.001}
0.7733203530311584 (0.0657499912847478) with: {'dropout rate': 0.2, 'learn
rate': 0.01}
0.7351509213447571 (0.0744789303276814) with: {'dropout rate': 0.2, 'learn
rate': 0.1}
0.7120090842247009 (0.06833877353815908) with: {'dropout rate': 0.3, 'lear
n rate': 0.0001}
0.7607270359992981 (0.09796058824673763) with: {'dropout_rate': 0.3, 'lear
n rate': 0.001}
0.7579681992530822 (0.06100996481663896) with: {'dropout rate': 0.3, 'lear
n rate': 0.01}
0.6916910171508789 (0.07067851744494327) with: {'dropout rate': 0.3, 'lear
n rate': 0.1}
```

## In [17]:

```
# Define a random seed
seed = 6
np.random.seed(seed)
learn rate = 0.01
# Start defining the model
def create_model(activation, init):
    # Create model
    model = Sequential()
    model.add(Dense(8, input_dim = 8, kernel_initializer = init, activation = activatio
n))
    #model.add(Dropout(dropout rate))
    model.add(Dense(4, input dim = 8, kernel initializer = init, activation = activatio
n))
    #model.add(Dropout(dropout rate))
    model.add(Dense(1, activation = 'sigmoid'))
    # Compile the model
    adam = Adam(lr = learn rate)
    model.compile(loss = 'binary_crossentropy', optimizer = adam, metrics = ['accuracy'
])
    return model
# Create the model
model = KerasClassifier(build_fn = create_model, epochs=epochs, batch_size=batch_size,
verbose = 0) # This comes from the previous best
# define the grid search parameters
activations = ['softmax', 'relu', 'tanh', 'linear']
initializers = ['uniform', 'normal', 'zero']
# make a dictionary of the grid search parameters
param_grid = dict(activation = activations, init = initializers)
# build and fit the GridSearchCV
grid = GridSearchCV(estimator=model, param_grid=param_grid, cv = KFold(random_state=see
d), verbose=10)
grid result = grid.fit(X standardized, Y)
# summarize the results
print("Best: {0}, using {1}".format(grid result.best score , grid result.best params ))
means = grid_result.cv_results_['mean_test_score']
stds = grid_result.cv_results_['std_test_score']
params = grid result.cv results ['params']
for mean, stdev, param in zip(means, stds, params):
    print('{0} ({1}) with: {2}'.format(mean, stdev, param))
```

Fitting 5 folds for each of 12 candidates, totalling 60 fits [CV] activation=softmax, init=uniform
[Parallel(n_jobs=1)]: Using backend SequentialBackend with 1 concurrent rkers.
[CV] activation=softmax, init=uniform, score=0.722, total= 4.8s [CV] activation=softmax, init=uniform
<pre>[Parallel(n_jobs=1)]: Done  1 out of  1   elapsed:  4.7s remaining: 0.0s</pre>
<pre>[CV] activation=softmax, init=uniform, score=0.658, total= 5.2s [CV] activation=softmax, init=uniform</pre>
<pre>[Parallel(n_jobs=1)]: Done  2 out of  2   elapsed:  9.9s remaining: 0.0s</pre>
<pre>[CV] activation=softmax, init=uniform, score=0.821, total= 4.4s [CV] activation=softmax, init=uniform</pre>
<pre>[Parallel(n_jobs=1)]: Done  3 out of  3   elapsed: 14.3s remaining: 0.0s</pre>
<pre>[CV] activation=softmax, init=uniform, score=0.833, total= 4.5s [CV] activation=softmax, init=uniform</pre>
<pre>[Parallel(n_jobs=1)]: Done  4 out of  4   elapsed: 18.8s remaining: 0.0s</pre>
<pre>[CV] activation=softmax, init=uniform, score=0.846, total= 4.5s [CV] activation=softmax, init=normal</pre>
<pre>[Parallel(n_jobs=1)]: Done 5 out of 5   elapsed: 23.3s remaining: 0.0s</pre>
[CV] activation=softmax, init=normal, score=0.734, total= 4.6s [CV] activation=softmax, init=normal
<pre>[Parallel(n_jobs=1)]: Done 6 out of 6   elapsed: 28.0s remaining: 0.0s</pre>
[CV] activation=softmax, init=normal, score=0.620, total= 4.5s [CV] activation=softmax, init=normal
<pre>[Parallel(n_jobs=1)]: Done 7 out of 7   elapsed: 32.5s remaining: 0.0s</pre>
<pre>[CV] activation=softmax, init=normal, score=0.795, total= 4.7s [CV] activation=softmax, init=normal</pre>
<pre>[Parallel(n_jobs=1)]: Done  8 out of  8   elapsed:  37.2s remaining: 0.0s</pre>
[CV] activation=softmax, init=normal, score=0.846, total= 4.8s [CV] activation=softmax, init=normal
<pre>[Parallel(n_jobs=1)]: Done 9 out of 9   elapsed: 41.9s remaining: 0.0s</pre>

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	activation=softmax, init=normal,		total=	4.5s
	activation=softmax, init=zero		• • • • • • • • •	• • • • •
[CV]	activation=softmax, init=zero,	score=0.646,	total=	4.4s
[CV]	activation=softmax, init=zero			
[CV]	<pre> activation=softmax, init=zero,</pre>	score=0.570,	total=	4.3s
[CV]	activation=softmax, init=zero			
	activation = softmax, init=zero,		total=	5.0s
	activation=softmax, init=zero			
	activation=softmax, init=zero,			4.7s
	activation=softmax, init=zero			4./3
				4 (
	activation=softmax, init=zero,	-		4.6s
	activation=relu, init=uniform			• • • •
	activation=relu, init=uniform,	-	total=	4.7s
	activation=relu, init=uniform		• • • • • • • • •	
	activation=relu, init=uniform,		total=	4.7s
[CV]	activation=relu, init=uniform			
[CV]	activation=relu, init=uniform,	score=0.782,	total=	4.3s
[CV]	activation=relu, init=uniform			
[CV]	activation=relu, init=uniform,	score=0.808,	total=	4.5s
[cv]	activation=relu, init=uniform			
	activation=relu, init=uniform,		total=	4.4s
	activation=relu, init=normal	-		
	activation=relu, init=normal,		total=	4.4s
	activation=relu, init=normal	-		
	activation=relu, init=normal,			4.5s
				4.53
	activation=relu, init=normal			
	activation=relu, init=normal,		totai=	4.4s
	activation=relu, init=normal			4 2-
	activation=relu, init=normal,	score=0.808,	total=	4.3s
	activation=relu, init=normal		• • • • • • • • •	• • • • •
	activation=relu, init=normal,		total=	4.5s
	activation=relu, init=zero		• • • • • • • • •	
[CV]	activation=relu, init=zero,	score=0.646,	total=	4.8s
	activation=relu, init=zero			
[CV]	activation=relu, init=zero,	score=0.570,	total=	4.4s
[CV]	activation=relu, init=zero			
[CV]	activation=relu, init=zero,	score=0.705,	total=	4.4s
[CV]	activation=relu, init=zero			
	activation=relu, init=zero,			
	activation=relu, init=zero	-		
	activation=relu, init=zero,			
	activation=tanh, init=uniform	-		
	activation=tanh, init=uniform,			
	activation=tanh, init=uniform	-		
	activation=tanh, init=uniform,			
	activation=tanh, init=uniform			
	activation=tanh, init=uniform,			
	activation=tanh, init=uniform			
	activation=tanh, init=uniform,	-		
	activation=tanh, init=uniform			
	activation=tanh, init=uniform,			
	activation=tanh, init=normal			
	<pre> activation=tanh, init=normal,</pre>			
[CV]	activation=tanh, init=normal			
[CV]	activation=tanh, init=normal,	score=0.620,	total=	4.4s
[CV]	activation=tanh, init=normal			
	activation=tanh, init=normal,			
	activation=tanh, init=normal	-		
	activation=tanh, init=normal,			
	activation=tanh, init=normal	-		
	activation=tanh. init=normal.			

[CV]	activation=tanh, init=zero	
[CV]	activation=tanh, init=zero, score=0.646, total=	4.4s
	activation=tanh, init=zero	• • • • •
[CV]	activation=tanh, init=zero, score=0.570, total=	4.5s
	activation=tanh, init=zero	
	activation=tanh, init=zero, score=0.705, total=	4.5s
	activation=tanh, init=zero	• • • • •
	activation=tanh, init=zero, score=0.744, total=	4.6s
	activation=tanh, init=zero	• • • • •
	activation=tanh, init=zero, score=0.679, total=	4.4s
	activation=linear, init=uniform	• • • • •
	activation=linear, init=uniform, score=0.835, total=	4.4s
	activation=linear, init=uniform	• • • • •
	activation=linear, init=uniform, score=0.633, total=	4.5s
	activation=linear, init=uniform	• • • • •
	activation=linear, init=uniform, score=0.821, total=	4.5s
	activation=linear, init=uniform	• • • • •
	activation=linear, init=uniform, score=0.846, total=	4.5s
	activation=linear, init=uniform	• • • • •
	activation=linear, init=uniform, score=0.821, total=	4.3s
	activation=linear, init=normal	
	activation=linear, init=normal, score=0.835, total=	5.1s
	•	• • • • •
	activation=linear, init=normal, score=0.620, total=	4.4s
	activation=linear, init=normal	• • • • •
	activation=linear, init=normal, score=0.821, total=	4.3s
	activation=linear, init=normal	• • • • •
	activation=linear, init=normal, score=0.859, total=	4.4s
	activation=linear, init=normal	• • • • •
	activation=linear, init=normal, score=0.808, total=	4.4s
	activation=linear, init=zero	• • • • •
	activation=linear, init=zero, score=0.646, total=	4.2s
	activation=linear, init=zero	• • • • •
	activation=linear, init=zero, score=0.570, total=	4.4s
	activation=linear, init=zero	• • • • •
	activation=linear, init=zero, score=0.705, total=	4.5s
	activation=linear, init=zero	• • • • •
	activation=linear, init=zero, score=0.744, total=	4.4s
	activation=linear, init=zero	• • • • •
[CV]	activation=linear, init=zero, score=0.679, total=	4.4s

[Parallel(n\_jobs=1)]: Done 60 out of 60 | elapsed: 4.5min finished

```
Best: 0.7911067843437195, using {'activation': 'linear', 'init': 'unifor
m'}
0.7759493708610534 (0.07352347706662563) with: {'activation': 'softmax',
'init': 'uniform'}
0.7708860754966735 (0.08722145913036237) with: {'activation': 'softmax',
'init': 'normal'}
0.6686789989471436 (0.05899773033245815) with: {'activation': 'softmax',
'init': 'zero'}
0.7605647444725037 (0.06994542741212405) with: {'activation': 'relu', 'ini
t': 'uniform'}
0.7476793169975281 (0.06306381621409568) with: {'activation': 'relu', 'ini
t': 'normal'}
0.6686789989471436 (0.05899773033245815) with: {'activation': 'relu', 'ini
t': 'zero'}
0.7632262110710144 (0.09936818298562612) with: {'activation': 'tanh', 'ini
t': 'uniform'}
0.7605972051620483 (0.07558978386106281) with: {'activation': 'tanh', 'ini
t': 'normal'}
0.6686789989471436 (0.05899773033245815) with: {'activation': 'tanh', 'ini
t': 'zero'}
0.7911067843437195 (0.07968826468817176) with: {'activation': 'linear', 'i
nit': 'uniform'}
0.7885751247406005 (0.08587248740892742) with: {'activation': 'linear', 'i
nit': 'normal'}
0.6686789989471436 (0.05899773033245815) with: {'activation': 'linear', 'i
nit': 'zero'}
```

#### In [18]:

```
# Define a random seed
seed = 6
np.random.seed(seed)
init = 'uniform' # taken from previous results
activation = 'linear'
# Start defining the model
def create_model(neuron1, neuron2):
    # Create model
    model = Sequential()
   model.add(Dense(neuron1, input dim = 8, kernel initializer = init, activation = act
ivation))
     model.add(Dropout(dropout rate))
    model.add(Dense(neuron2, input_dim = neuron1, kernel_initializer = init, activation
= activation))
    model.add(Dropout(dropout_rate))
    model.add(Dense(1, activation = 'sigmoid'))
    # Compile the model
    adam = Adam(lr = learn rate)
    model.compile(loss = 'binary_crossentropy', optimizer = adam, metrics = ['accuracy'
])
    return model
# Create the model
model = KerasClassifier(build fn = create model, epochs=50, batch size=10, verbose = 0)
# define the grid search parameters
neuron1 = [4, 8, 16]
neuron2 = [2, 4, 8]
# make a dictionary of the grid search parameters
param_grid = dict(neuron1 = neuron1, neuron2 = neuron2)
# build and fit the GridSearchCV
grid = GridSearchCV(estimator=model, param_grid=param_grid, cv = KFold(random_state=see
d), refit=True, verbose=10) # To retrain with the best parameters found so far
grid result = grid.fit(X standardized, Y)
# summarize the results
print("Best: {0}, using {1}".format(grid_result.best_score_, grid_result.best_params_))
means = grid result.cv results ['mean test score']
stds = grid_result.cv_results_['std_test_score']
params = grid result.cv results ['params']
for mean, stdev, param in zip(means, stds, params):
    print('{0} ({1}) with: {2}'.format(mean, stdev, param))
```

Fitting 5 folds for each of 9 candidates, totalling 45 fits [CV] neuron1=4, neuron2=2
$\label{lem:concurrent} \end{substitute} \begin{substitute}[Parallel(n\_jobs=1)]{:}{:}{:}{:}{:}{:}{:}{:}{:}{:}{:}{:}{:}{$
[CV] neuron1=4, neuron2=2, score=0.835, total= 4.6s [CV] neuron1=4, neuron2=2
<pre>[Parallel(n_jobs=1)]: Done  1 out of  1   elapsed:  4.5s remaining: 0.0s</pre>
[CV] neuron1=4, neuron2=2, score=0.633, total= 4.1s [CV] neuron1=4, neuron2=2
<pre>[Parallel(n_jobs=1)]: Done  2 out of  2   elapsed: 8.6s remaining: 0.0s</pre>
[CV] neuron1=4, neuron2=2, score=0.821, total= 4.5s [CV] neuron1=4, neuron2=2
<pre>[Parallel(n_jobs=1)]: Done  3 out of  3   elapsed: 13.1s remaining: 0.0s</pre>
[CV] neuron1=4, neuron2=2, score=0.846, total= 5.1s [CV] neuron1=4, neuron2=2
<pre>[Parallel(n_jobs=1)]: Done  4 out of  4   elapsed: 18.2s remaining: 0.0s</pre>
[CV] neuron1=4, neuron2=2, score=0.833, total= 4.5s [CV] neuron1=4, neuron2=4
<pre>[Parallel(n_jobs=1)]: Done 5 out of 5   elapsed: 22.7s remaining: 0.0s</pre>
[CV] neuron1=4, neuron2=4, score=0.835, total= 4.5s [CV] neuron1=4, neuron2=4
<pre>[Parallel(n_jobs=1)]: Done 6 out of 6   elapsed: 27.1s remaining: 0.0s</pre>
[CV] neuron1=4, neuron2=4, score=0.608, total= 4.4s [CV] neuron1=4, neuron2=4
<pre>[Parallel(n_jobs=1)]: Done 7 out of 7   elapsed: 31.6s remaining: 0.0s</pre>
[CV] neuron1=4, neuron2=4, score=0.821, total= 4.2s [CV] neuron1=4, neuron2=4
<pre>[Parallel(n_jobs=1)]: Done  8 out of  8   elapsed:  35.8s remaining: 0.0s</pre>
[CV] neuron1=4, neuron2=4, score=0.846, total= 4.3s [CV] neuron1=4, neuron2=4
<pre>[Parallel(n_jobs=1)]: Done 9 out of 9   elapsed: 40.1s remaining: 0.0s</pre>

WO

[CV]	neuron1=4, neuron2=4, score=0.808, tota	l= 4.2s
	neuron1=4, neuron2=8	
[CV]	•	l= 4.3s
	neuron1=4, neuron2=8	
[CV]		l= 4.5s
		1- 4.55
	neuron1=4, neuron2=8	
	neuron1=4, neuron2=8, score=0.821, tota	l= 4.5s
[CV]		• • • • • • • • •
[CV]		ıl= 4.5s
[CV]	neuron1=4, neuron2=8	
[CV]	neuron1=4, neuron2=8, score=0.821, tota	1= 4.4s
[CV]	neuron1=8, neuron2=2	
[cv]		
	neuron1=8, neuron2=2	
[CV]		l= 5.0s
	neuron1=8, neuron2=2	1- 5.03
	·	ıl= 4.4s
[CV]		1= 4.45
[CV]	•	
[CV]		l= 4.4s
[CV]		• • • • • • • • •
	neuron1=8, neuron2=2, score=0.808, tota	ıl= 4.2s
[CV]	neuron1=8, neuron2=4	
[CV]	neuron1=8, neuron2=4, score=0.835, tota	1= 4.3s
[CV]	neuron1=8, neuron2=4	
[CV]	neuron1=8, neuron2=4, score=0.608, tota	l= 4.3s
	neuron1=8, neuron2=4	
[CV]		l= 4.5s
[CV]		
[CV]		ıl= 4.5s
	neuron1=8, neuron2=4	
[CV]		ıl= 4.4s
[CV]		• • • • • • • • •
[CV]		ıl= 4.4s
[CV]	neuron1=8, neuron2=8	
[CV]	neuron1=8, neuron2=8, score=0.620, tota	1= 4.9s
[CV]	neuron1=8, neuron2=8	
[CV]	neuron1=8, neuron2=8, score=0.821, tota	l= 4.6s
	neuron1=8, neuron2=8	
	neuron1=8, neuron2=8, score=0.833, tota	
	neuron1=8, neuron2=8	
	neuron1=16, neuron2=2	
	neuron1=16, neuron2=2, score=0.835, tota	
	neuron1=16, neuron2=2	
	neuron1=16, neuron2=2, score=0.620, tota	
	neuron1=16, neuron2=2	
[CV]	neuron1=16, neuron2=2, score=0.821, tota	1= 4.5s
[CV]	neuron1=16, neuron2=2	
[cv]	neuron1=16, neuron2=2, score=0.833, tota	1= 4.4s
	neuron1=16, neuron2=2	
	neuron1=16, neuron2=2, score=0.833, tota	
	neuron1=16, neuron2=4	
	neuron1=16, neuron2=4, score=0.835, tota	
	neuron1=16, neuron2=4	
	neuron1=16, neuron2=4, score=0.633, tota	
	neuron1=16, neuron2=4	
	neuron1=16, neuron2=4, score=0.821, tota	
	neuron1=16, neuron2=4	
	neuron1=16, neuron2=4, score=0.846, tota	
[CV]	neuron1=16, neuron2=4	
[CV]	neuron1=16. neuron2=4. score=0.833. tota	1= 4.95

```
[CV] neuron1=16, neuron2=8 .....
[CV] ..... neuron1=16, neuron2=8, score=0.835, total= 4.3s
[CV] neuron1=16, neuron2=8 .....
[CV] ..... neuron1=16, neuron2=8, score=0.633, total= 4.6s
[CV] neuron1=16, neuron2=8 .....
[CV] ..... neuron1=16, neuron2=8, score=0.833, total= 4.8s
[CV] neuron1=16, neuron2=8 ......
[CV] ..... neuron1=16, neuron2=8, score=0.833, total= 4.6s
[CV] neuron1=16, neuron2=8 .....
[CV] ..... neuron1=16, neuron2=8, score=0.821, total= 5.2s
[Parallel(n_jobs=1)]: Done 45 out of 45 | elapsed: 3.4min finished
Best: 0.7936708807945252, using {'neuron1': 4, 'neuron2': 2}
0.7936708807945252 (0.08079181748331918) with: {'neuron1': 4, 'neuron2':
2}
0.783479380607605 (0.08890601661969585) with: {'neuron1': 4, 'neuron2': 4}
0.7911067843437195 (0.07968826468817176) with: {'neuron1': 4, 'neuron2':
0.7911067724227905 (0.08009971934827462) with: {'neuron1': 8, 'neuron2':
0.7886075854301453 (0.09087249991485369) with: {'neuron1': 8, 'neuron2':
0.7885751247406005 (0.08432733392428071) with: {'neuron1': 8, 'neuron2':
0.7885751247406005 (0.08432733392428071) with: {'neuron1': 16, 'neuron2':
0.7936708807945252 (0.08079181748331918) with: {'neuron1': 16, 'neuron2':
0.7911067724227905 (0.0792746637362651) with: {'neuron1': 16, 'neuron2':
8}
```

## In [19]:

```
# generate predictions with optimal hyperparameters
y_pred = grid.predict(X_standardized)
```

## In [20]:

```
from sklearn.metrics import classification_report, accuracy_score
print(accuracy_score(Y, y_pred))
print(classification_report(Y, y_pred))
```

#### 0.7857142857142857

	precision	recall	f1-score	support
0	0.81	0.89	0.85	262
1	0.72	0.58	0.64	130
accuracy			0.79	392
macro avg	0.76	0.74	0.75	392
weighted avg	0.78	0.79	0.78	392

# In [21]:

```
# example datapoint
example = df.iloc[1]
print(example)
```

```
n_pregnant
                              0.000
glucose_concentration
                            137.000
blood_pressuer (mm Hg)
                             40.000
skin_thickness (mm)
                             35.000
serum_insulin (mu U/ml)
                           168.000
                             43.100
pedigree_function
                              2.288
age
                             33.000
class
                              1.000
Name: 4, dtype: float64
```

Name. 4, acype. Tioaco

# In [22]:

```
# make a prediction using our optimized deep neural network
prediction = grid.predict(X_standardized[1].reshape(1, -1))
print(prediction)
```

[[1]]