

```
In [2]: # import necessary libraries:
import numpy as np
# importing keras models:
import tensorflow as tf
from keras.models import Sequential
from keras.layers import Dense
import os
import matplotlib.pyplot as plt
```

Using TensorFlow backend.

```
In [3]: # change the working directory
os.chdir("/Users/karimaidrissi/Desktop/DSSA 5104 DL/keras/")
```

```
In [18]: # load the dataset
features_and_targets = np.loadtxt('features_and_targets.csv',delimiter=
',')
# shuffle our dataset by using np.random
np.random.shuffle(features_and_targets)

# columns 0-4 are features
X = features_and_targets[:,0:5]
# columns 5-6 are targets
Y = features_and_targets[:,5:7]

# the random output will remain the same by using seed 10
np.random.seed(10)

# Create a simple model
model = tf.keras.models.Sequential() # intial model with Sequential mode
l type
model.add(tf.keras.layers.Dense(4,input_dim=5,activation="relu")) # full
y connect the 4 hidden layer and 5 input layers by using Dense layer
model.add(tf.keras.layers.Dense(3, activation="relu")) # using ReLu func
tion to fire the model
model.add(tf.keras.layers.Dense(2, activation="sigmoid")) # using Sigmoid
function to activate the last two output layers
# done with the archeticture of our model we will define some paramters
for training our model
# we will compile our model with loss function, optimizer as adam default
t and do the accuracy.
model.compile(loss='mean_squared_error',optimizer="adam", metrics= ["acc
uracy"])
```

```
In [19]: # train the model by passing X and Y with 150 iteration
model.fit(X,Y, epochs=150,verbose=0)

#run and create a history object with model.fit()
history = model.fit(X,Y, epochs=150, verbose=0 )
print(history.history.keys())

# evaluate the model
val_loss, val_accuracy = model.evaluate(X,Y)
print(val_loss, val_accuracy)
# printing the accuracy of our model
print('\n%s: %.2f%%' % ( model.metrics_names[1],val_accuracy*100))
# printing the loss of our model
print('\n%s: %.2f%%' % (model.metrics_names[0], val_loss*100))

# returning the predicting model
predicted_targets = model.predict(X)
predicted_targets

# iterate by 22 times over the predicted target and the observed target
for i in range(22):
    print('Predicted: ',predicted_targets[i,:], 'Observed: ',Y[i,:])
```

```
dict_keys(['loss', 'accuracy'])
22/22 [=====] - 0s 12ms/sample - loss: 0.2130
- accuracy: 0.6364
0.21304577589035034 0.6363636
```

accuracy: 63.64%

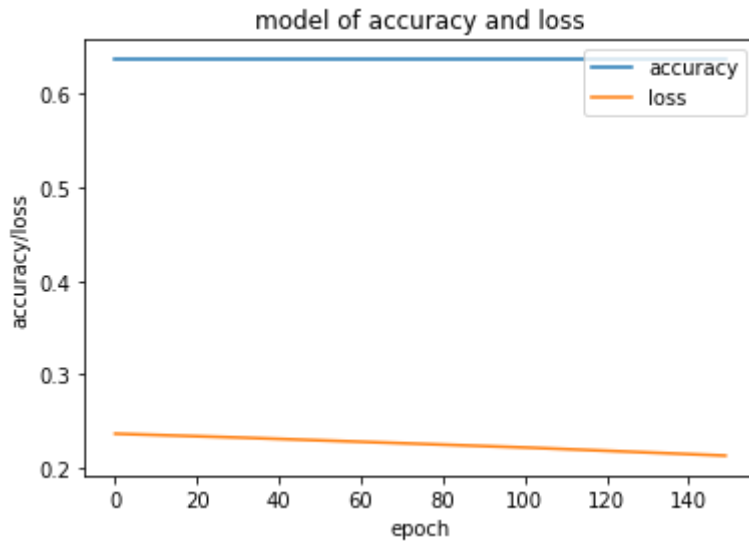
loss: 21.30%

```
Predicted: [0.44608557 0.6543813 ] Observed: [1. 0.]
Predicted: [0.436274 0.5885499] Observed: [0. 1.]
Predicted: [0.3161513 0.68494016] Observed: [0. 1.]
Predicted: [0.3723901 0.51726073] Observed: [0. 1.]
Predicted: [0.36497337 0.5129326 ] Observed: [0. 1.]
Predicted: [0.4271265 0.5481677] Observed: [1. 0.]
Predicted: [0.29423326 0.7019329 ] Observed: [0. 1.]
Predicted: [0.4214809 0.60004795] Observed: [1. 0.]
Predicted: [0.3841725 0.52406037] Observed: [0. 1.]
Predicted: [0.32450667 0.6696159 ] Observed: [0. 1.]
Predicted: [0.42578447 0.5474286 ] Observed: [1. 0.]
Predicted: [0.42790473 0.55164415] Observed: [1. 0.]
Predicted: [0.4271265 0.5481677] Observed: [0. 1.]
Predicted: [0.44002816 0.6047595 ] Observed: [0. 1.]
Predicted: [0.4137374 0.66538984] Observed: [0. 1.]
Predicted: [0.32566786 0.6868797 ] Observed: [0. 1.]
Predicted: [0.4271265 0.5481677] Observed: [1. 0.]
Predicted: [0.4271265 0.5481677] Observed: [1. 0.]
Predicted: [0.444914 0.6254677] Observed: [0. 1.]
Predicted: [0.4271265 0.5481677] Observed: [1. 0.]
Predicted: [0.3134087 0.70009625] Observed: [0. 1.]
Predicted: [0.43858957 0.59857666] Observed: [0. 1.]
```

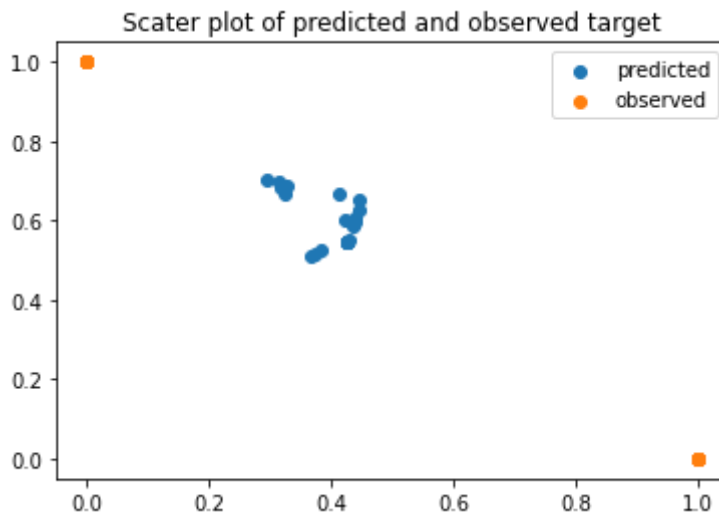
```
In [20]: # list all data in history:
type(history.history)
print(history.history.keys())
# summarize history for accuracy:
print('Accuracy', history.history["accuracy"]) # print history for accuracy
print('Loss', history.history["loss"]) # print history for loss
plt.plot(history.history["accuracy"])
plt.plot(history.history["loss"])
plt.title("model of accuracy and loss")
plt.ylabel("accuracy/loss")
plt.xlabel("epoch")
plt.legend(["accuracy", "loss"], loc="upper right")
plt.show()
```

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```
8960456848145, 0.2202141433954239, 0.22003665566444397, 0.2198573946952
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00071668625, 0.21467962861061096, 0.21449890732765198, 0.21431788802146
912, 0.21413660049438477, 0.21395502984523773, 0.2137732058763504, 0.21
359267830848694, 0.21340960264205933, 0.2132277935743332]
```



```
In [21]: # plotting scastter plot of predicted target and observed target
plt.scatter(predicted_targets[:,0],predicted_targets[:,1])
plt.scatter(Y[:,0],Y[:,1])
plt.title("Scater plot of predicted and observed target")
plt.legend(["predicted","observed"])
plt.show()
```



```
In [22]: # Run this command to see if Tensorflow has detected a GPU to use
from tensorflow.python.client import device_lib
print(device_lib.list_local_devices())

#Run this - if Theano finds a GPU to use it will tell you
#import theano
```

```
[name: "/device:CPU:0"
device_type: "CPU"
memory_limit: 268435456
locality {
}
incarnation: 725552351135439036
, name: "/device:XLA_CPU:0"
device_type: "XLA_CPU"
memory_limit: 17179869184
locality {
}
incarnation: 3332633806451044665
physical_device_desc: "device: XLA_CPU device"
]
```

epochs = 500

```

In [23]: #####
# train the model by passing X and Y with 500 iteration
model.fit(X,Y, epochs=500,verbose=0)

#run and create a history object with model.fit()
history = model.fit(X,Y, epochs=500, verbose=0 )
print(history.history.keys())

# evaluate the model
val_loss, val_accuracy = model.evaluate(X,Y)
print(val_loss, val_accuracy)
# printing the accuracy of our model
print('\n%s: %.2f%%' % ( model.metrics_names[1],val_accuracy*100))
# printing the loss of our model
print('\n%s: %.2f%%' % (model.metrics_names[0], val_loss*100))

# returning the predicting model
predicted_targets = model.predict(X)
predicted_targets

# iterate by 22 times over the predicted target and the observed target
for i in range(22):
    print('Predicted: ',predicted_targets[i,:], 'Observed: ',Y[i,:])

dict_keys(['loss', 'accuracy'])
22/22 [=====] - 0s 75us/sample - loss: 0.0400
- accuracy: 1.0000
0.0399923212826252 1.0

accuracy: 100.00%

loss: 4.00%
Predicted: [0.69049716 0.3352232 ] Observed: [1. 0.]
Predicted: [0.04834585 0.9074958 ] Observed: [0. 1.]
Predicted: [0.02381992 0.9817238 ] Observed: [0. 1.]
Predicted: [0.04701096 0.9079255 ] Observed: [0. 1.]
Predicted: [0.04089212 0.9095444 ] Observed: [0. 1.]
Predicted: [0.69049716 0.3352232 ] Observed: [1. 0.]
Predicted: [0.02380996 0.98165154] Observed: [0. 1.]
Predicted: [0.69049716 0.3352232 ] Observed: [1. 0.]
Predicted: [0.04158007 0.9097821 ] Observed: [0. 1.]
Predicted: [0.01986909 0.98129076] Observed: [0. 1.]
Predicted: [0.69049716 0.3352232 ] Observed: [1. 0.]
Predicted: [0.69049716 0.3352232 ] Observed: [1. 0.]
Predicted: [0.06442876 0.90294194] Observed: [0. 1.]
Predicted: [0.07667875 0.9000406 ] Observed: [0. 1.]
Predicted: [0.0279302 0.97974086] Observed: [0. 1.]
Predicted: [0.02791859 0.979661 ] Observed: [0. 1.]
Predicted: [0.69049716 0.3352232 ] Observed: [1. 0.]
Predicted: [0.69049716 0.3352232 ] Observed: [1. 0.]
Predicted: [0.03821667 0.969257 ] Observed: [0. 1.]
Predicted: [0.69049716 0.3352232 ] Observed: [1. 0.]
Predicted: [0.02044126 0.981122 ] Observed: [0. 1.]
Predicted: [0.07462292 0.9005013 ] Observed: [0. 1.]

```

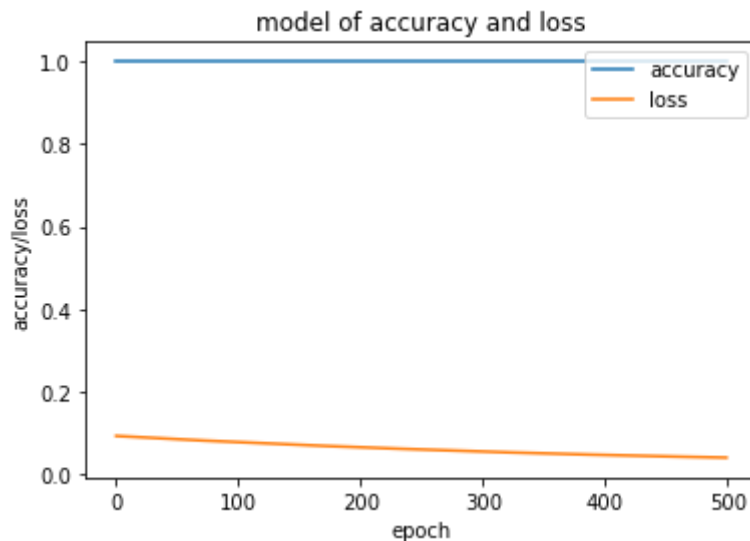
```
In [24]: # list all data in history:
type(history.history)
print(history.history.keys())
# summarize history for accuracy:
print('Accuracy', history.history["accuracy"]) # print history for accuracy
print('Loss', history.history["loss"]) # print history for loss
plt.plot(history.history["accuracy"])
plt.plot(history.history["loss"])
plt.title("model of accuracy and loss")
plt.ylabel("accuracy/loss")
plt.xlabel("epoch")
plt.legend(["accuracy", "loss"], loc="upper right")
plt.show()
```


[illegible]

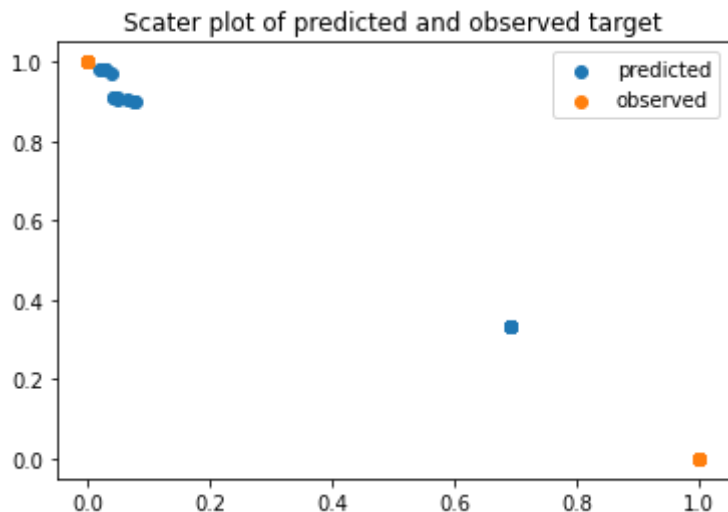
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0.04028739035129547, 0.040228407829999924, 0.0401688814163208, 0.040110
24534702301, 0.04005137458443642]
```



```
In [25]: # plotting scastter plot of predicted target and observed target
plt.scatter(predicted_targets[:,0],predicted_targets[:,1])
plt.scatter(Y[:,0],Y[:,1])
plt.title("Scater plot of predicted and observed target")
plt.legend(["predicted","observed"])
plt.show()
```



```
In [26]: # Run this command to see if Tensorflow has detected a GPU to use
from tensorflow.python.client import device_lib
print(device_lib.list_local_devices())

#Run this - if Theano finds a GPU to use it will tell you
#import theano
```

```
[name: "/device:CPU:0"
device_type: "CPU"
memory_limit: 268435456
locality {
}
incarnation: 5365360192848153725
, name: "/device:XLA_CPU:0"
device_type: "XLA_CPU"
memory_limit: 17179869184
locality {
}
incarnation: 11553298888453344940
physical_device_desc: "device: XLA_CPU device"
]
```

epchos= 1500

```

In [27]: #####
# train the model by passing X and Y with 1500 iteration
model.fit(X,Y, epochs=1500,verbose=0)

#run and create a history object with model.fit()
history = model.fit(X,Y, epochs=1500, verbose=0 )
print(history.history.keys())

# evaluate the model
val_loss, val_accuracy = model.evaluate(X,Y)
print(val_loss, val_accuracy)
# printing the accuracy of our model
print('\n%s: %.2f%%' % ( model.metrics_names[1],val_accuracy*100))
# printing the loss of our model
print('\n%s: %.2f%%' % (model.metrics_names[0], val_loss*100))

# returning the predicting model
predicted_targets = model.predict(X)
predicted_targets

# iterate by 22 times over the predicted target and the observed target
for i in range(22):
    print('Predicted: ',predicted_targets[i,:], 'Observed: ',Y[i,:])

dict_keys(['loss', 'accuracy'])
22/22 [=====] - 0s 76us/sample - loss: 0.0030
- accuracy: 1.0000
0.0030249471310526133 1.0

accuracy: 100.00%

loss: 0.30%
Predicted: [0.91253215 0.09362251] Observed: [1. 0.]
Predicted: [0.00459823 0.9912143 ] Observed: [0. 1.]
Predicted: [0.00686691 0.9945799 ] Observed: [0. 1.]
Predicted: [0.00470164 0.9911451 ] Observed: [0. 1.]
Predicted: [0.00560265 0.99028945] Observed: [0. 1.]
Predicted: [0.91253215 0.09362251] Observed: [1. 0.]
Predicted: [0.00673747 0.9946069 ] Observed: [0. 1.]
Predicted: [0.91253215 0.09362251] Observed: [1. 0.]
Predicted: [0.00546172 0.9904022 ] Observed: [0. 1.]
Predicted: [0.00540533 0.9948873 ] Observed: [0. 1.]
Predicted: [0.91253215 0.09362251] Observed: [1. 0.]
Predicted: [0.91253215 0.09362251] Observed: [1. 0.]
Predicted: [0.01143452 0.9843258 ] Observed: [0. 1.]
Predicted: [0.00786737 0.9878006 ] Observed: [0. 1.]
Predicted: [0.00767107 0.9943975 ] Observed: [0. 1.]
Predicted: [0.00752659 0.9944254 ] Observed: [0. 1.]
Predicted: [0.91253215 0.09362251] Observed: [1. 0.]
Predicted: [0.91253215 0.09362251] Observed: [1. 0.]
Predicted: [0.00643578 0.9948291 ] Observed: [0. 1.]
Predicted: [0.91253215 0.09362251] Observed: [1. 0.]
Predicted: [0.00530329 0.99491274] Observed: [0. 1.]
Predicted: [0.00835794 0.98729926] Observed: [0. 1.]

```

```
In [28]: # list all data in history:
type(history.history)
print(history.history.keys())
# summarize history for accuracy:
print('Accuracy', history.history["accuracy"]) # print history for accuracy
print('Loss', history.history["loss"]) # print history for loss
plt.plot(history.history["accuracy"])
plt.plot(history.history["loss"])
plt.title("model of accuracy and loss")
plt.ylabel("accuracy/loss")
plt.xlabel("epoch")
plt.legend(["accuracy", "loss"], loc="upper right")
plt.show()
```

[illegible]

Loss [0.008613914251327515, 0.008606592193245888, 0.008599899709224701, 0.008593117818236351, 0.008586258627474308, 0.008579332381486893, 0.008572348393499851, 0.008565316908061504, 0.008558609522879124, 0.008551633916795254, 0.008544943295419216, 0.008538170717656612, 0.008531332947313786, 0.00852443091571331, 0.008517485111951828, 0.008510499261319637,

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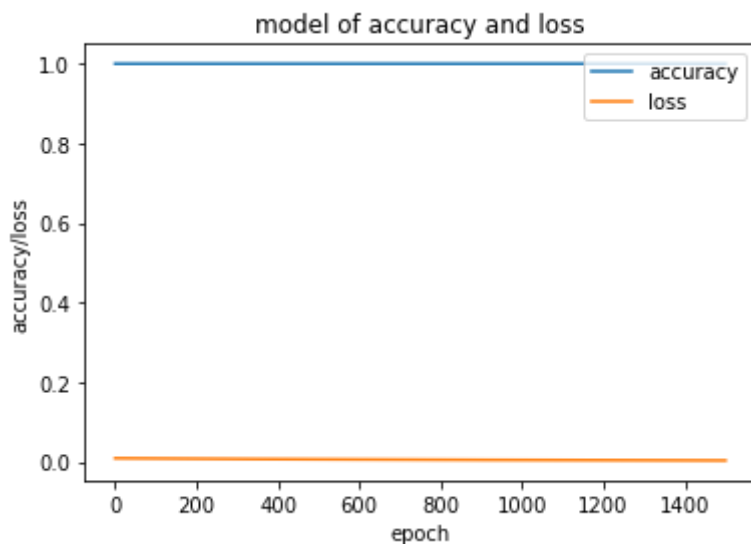
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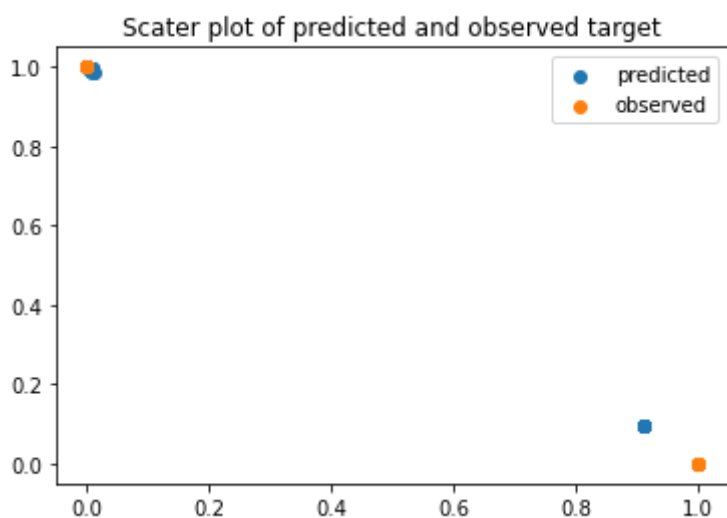
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72709465027, 0.003030716674402356, 0.003028802340850234, 0.003026877995
580435]
```



```
In [29]: # plotting scastter plot of predicted target and observed target
plt.scatter(predicted_targets[:,0],predicted_targets[:,1])
plt.scatter(Y[:,0],Y[:,1])
plt.title("Scater plot of predicted and observed target")
plt.legend(["predicted","observed"])
plt.show()
```



```
In [30]: # Run this command to see if Tensorflow has detected a GPU to use
from tensorflow.python.client import device_lib
print(device_lib.list_local_devices())

#Run this - if Theano finds a GPU to use it will tell you
#import theano
```

```
[name: "/device:CPU:0"
device_type: "CPU"
memory_limit: 268435456
locality {
}
incarnation: 17494523589789408744
, name: "/device:XLA_CPU:0"
device_type: "XLA_CPU"
memory_limit: 17179869184
locality {
}
incarnation: 16852421481777283151
physical_device_desc: "device: XLA_CPU device"
]
```

COMMENT: We conclude by increasing the number of epochs the accuracy result become more accurate and the loss rate goes down.