

Report – Assignment 2

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The assignment is to create a Keras model for the function $(y_1, y_2) = f(x_1, x_2)$ using randomized sets of numbers. The assignment also directs us to train and test the model along with the final report explaining the assignment.

We have implemented this assignment on Spyder (Python). The assignment requires us to use two values of X (i.e. x_1 and x_2) and Y (i.e. y_1 and y_2). To implement this in python, we make use of the dataset for x and y value and we have used keras model.

```
from keras.models import Sequential
import matplotlib.pyplot as plt
from keras.layers import Dense
from keras.optimizers import Adam
import numpy as np

#Training Data
#Training Data
x1 = np.multiply(np.random.random([1000]),10)
x2 = np.multiply(np.random.random([1000]),10)
y1 = np.add(np.add(pow(x1,2),x2),5)
y2 = np.add((np.add(pow(x1,3),x2)),6)
X_train=np.stack((x1,x2)).T
Y_train=np.stack((y1,y2)).T
```

Our model has three-layer, input layer, output layer and one hidden layer. We have two neurons for input layer, two neurons for output layer and four neurons in hidden layer and all the layer are dense. We are using relu activation function for our model.

```
#creating layers for the model
model=Sequential()
model.add(Dense(8,input_shape=(2,),activation='relu'))
model.add(Dense(4,activation='relu'))
model.add(Dense(2))
model.summary()
```

For compiling and training the model we have used model.compile and model.fit. We are using the adam optimizer for our model.

```
#for compiling the model
model.compile(loss='mean_squared_error',optimizer=Adam(lr=0.003), metrics=['accuracy'])
#for training the model
history=model.fit(Y_train,X_train,epochs=10)
```

We are using predict function for predicting the values of y_1 and y_2 and testing the model by randomizing the value of x_1 and x_2 .

```

#testing
#
#Training Data
xt_1 = np.multiply(np.random.random([1000]),10)
xt_2 = np.multiply(np.random.random([1000]),10)
yt_1 = np.add(np.add(pow(x1,2),x2),5)
yt_2 = np.add((np.add(pow(x1,3),x2)),6)

Y_test=np.stack((yt_1,yt_2)).T
print(Y_test)

tes=model.predict(Y_test)
print(tes)

```

After the model is trained we are showing the accuracy for each set and we are plotting the graph for loss and epochs.

```

def plot_loss_accuracy(history):
    fig = plt.figure(figsize=(12, 6))
    ax = fig.add_subplot(1, 2, 1)
    ax.plot(history.history["loss"], 'r-x', label="Train Loss")
    ax.legend()
    ax.set_title('Training loss')
    ax.grid(True)

plot_loss_accuracy(history)

```

Output of Our model

Layer (type)	Output Shape	Param #
=====	=====	=====
dense_36 (Dense)	(None, 8)	24
dense_37 (Dense)	(None, 4)	36
dense_38 (Dense)	(None, 2)	10
=====	=====	=====
Total params: 70		
Trainable params: 70		
Non-trainable params: 0		

Epoch of 10 sets

```
Epoch 1/10
1000/1000 [=====] - 1s 1ms/step - loss: 2404366.2688 - acc:
0.3560
Epoch 2/10
1000/1000 [=====] - 0s 108us/step - loss: 196359.5567 - acc:
0.8900
Epoch 3/10
1000/1000 [=====] - 0s 107us/step - loss: 86094.1333 - acc:
0.9230
Epoch 4/10
1000/1000 [=====] - 0s 94us/step - loss: 51319.2144 - acc:
0.9410
Epoch 5/10
1000/1000 [=====] - 0s 94us/step - loss: 35010.2574 - acc:
0.9490
Epoch 6/10
1000/1000 [=====] - 0s 109us/step - loss: 26399.7962 - acc:
0.9560
Epoch 7/10
1000/1000 [=====] - 0s 109us/step - loss: 21480.1628 - acc:
0.9670
Epoch 8/10
1000/1000 [=====] - 0s 94us/step - loss: 18772.7418 - acc:
0.9700
Epoch 9/10
1000/1000 [=====] - 0s 94us/step - loss: 17209.7596 - acc:
0.9750
Epoch 10/10
1000/1000 [=====] - 0s 94us/step - loss: 16166.7606 - acc:
0.9750
[17 24 10 14 27 29 12 25 10 20] random value of x1 and x2
[40 38 41 59 36 32 46 44 31 42]
[[1889 1311]
 [2020 868]
 [1781 1581]
 [3677 3285]
 [2025 567]
 [1865 183]
 [2260 1972]
 [2561 1311]
 [1061 861]
 [2164 1364]]
                                Y1 and Y2 value of the model
```

The predicted values of the model

```
[[-6.8002507e-02 6.9370264e-01]
 [-6.8002507e-02 6.9370264e-01]
 [-6.8002507e-02 6.9370264e-01]
 [-6.8002507e-02 6.9370264e-01]
 [-6.8002507e-02 6.9370264e-01]
 [ 6.8304817e+01 3.5301712e+00]
 [-6.8002507e-02 6.9370264e-01]
 [-6.8002507e-02 6.9370264e-01]
 [-6.8002507e-02 6.9370264e-01]
 [-6.8002507e-02 6.9370264e-01]]
```

