## MANE 4962 HW 1

Lucas Zhou 662005044

## Problem 1:

```
In [1]: import numpy as np
        x1 = np.array([2, 3, 4, 8, 9])
        x2 = np.array([2, -3, -4, 89])
        features_x1 = len(x1)
        features_x2 = len(x2)
        print('features_x1=', features_x1)
        print('features_x2=', features_x2)
       features x1= 5
       features_x2= 4
In [2]: norm1 x1 = np.linalg.norm(x1, 1)
        norm2 x1 = np.linalg.norm(x1, 2)
        norm_inf_x1 = np.linalg.norm(x1, np.inf)
        norm1_x2 = np.linalg.norm(x2, 1)
        norm2_x2 = np.linalg.norm(x2, 2)
        norm inf x2 = np.linalg.norm(x2, np.inf)
        print("norm1_x1:",norm1_x1)
        print("norm2_x1:", norm2_x1)
        print("norm_inf_x1:", norm_inf_x1)
        print("norm1-x2:",norm1 x2)
        print("norm2-x2:",norm2_x2)
        print("norm inf x2:", norm inf x2)
       norm1_x1: 26.0
       norm2_x1: 13.19090595827292
       norm_inf_x1: 9.0
       norm1-x2: 98.0
       norm2-x2: 89.16277250063504
       norm_inf_x2: 89.0
        Problem 2:
```

Assuming the input and output image are RGB, each has 3 channels.

```
In [3]: feature_dimensions = (1024, 1024, 3)
  output_dimensions = (64, 64, 3)
  length_feature = np.prod(feature_dimensions)
  length_output = np.prod(output_dimensions)
```

```
(a) The length of the feature vector
 In [4]: length_feature
 Out[4]: 3145728
         (b) The length of the output vector
         length_output
 In [5]:
 Out[5]: 12288
         (c) Number of elements in matrix W
         W = 3145728 * 12288
 In [6]:
 In [7]:
 Out[7]: 38654705664
         Number of elements in matrix b
 In [8]: print('elements of b:', length_output)
        elements of b: 12288
         Problem 3:
 In [9]: matrix = np.array([[1, 2], [-1, 0]])
         norm1 = np.linalg.norm(matrix, 1)
         norm2 = np.linalg.norm(matrix, 2)
         norm_inf = np.linalg.norm(matrix, np.inf)
         frobenius_norm = np.linalg.norm(matrix, 'fro')
         1-norm of W
In [10]:
         norm1
Out[10]: 2.0
         2-norm of W
In [11]:
         norm2
Out[11]: 2.2882456112707366
         infinite norm of W
In [12]:
         norm_inf
Out[12]: 3.0
```

model = Sequential()

```
In [13]: frobenius_norm
Out[13]: 2.449489742783178
                          Problem 4:
In [14]: from sklearn import datasets
                          from sklearn.model_selection import train_test_split
                          from sklearn import svm
                          from sklearn.metrics import accuracy_score
                          iris = datasets.load_iris()
                          X = iris.data[:, :3] # only selecting first three features
                          y = iris.target
                          X_train, X_test, y_train, y_test=train_test_split(
                                     Х, у,
                                     test_size=0.40,
                                     train_size=0.60,
                                     random_state=123,
                                      shuffle=True,
                                      stratify=y)
                          clf = svm.SVC()
                          clf.fit(X_train, y_train)
                          preds = clf.predict(X_test)
                          acc = accuracy_score(y_test,clf.predict(X_test) )
                          print('accuracy score :', acc)
                       Problem 5:
In [15]: import tensorflow as tf
                          from numpy import loadtxt
                          from keras.models import Sequential
                          from keras.layers import Dense
                          from keras import optimizers
                          from tensorflow.python.keras.optimizers import *
                          #Load dataset
                          #split into input (X) and output (y)
                          X = np.array([[1], [5], [10], [15], [20], [40], [50], [65], [80], [100]])
                          y = np.array([[1**(1/7)], [5**(1/7)], [10**(1/7)], [15**(1/7)], [20**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40**(1/7)], [40*
                          y = y*1.0
                          #define keras model
```

```
model.add(Dense(6,input_dim=1,activation='relu'))
 model.add(Dense(6,activation='relu'))
 model.add(Dense(6,activation='relu'))
 model.add(Dense(1))
 #compile the keras model
 opt = optimizers.Adam(learning_rate=0.001)
 mse = tf.keras.losses.MeanSquaredError(
     reduction=tf.keras.losses.Reduction.SUM)
 model.compile(loss=mse, optimizer=opt)
 #fit the keras model on the dataset (CPU)
 model.fit(X,y,epochs=2000,batch_size=10, verbose=0)
 model.summary()
 #make class predictions with the model
 predictions = model.predict(X)
 #summarize the first 10 cases
 for i in range(10):
     print('\%s \Rightarrow \%.2f (expected \%.2f)' \% (X[i].tolist(), predictions[i][0], y[i][0]
Model: "sequential"
Layer (type)
                           Output Shape
                                                    Param #
______
 dense (Dense)
                           (None, 6)
                                                    12
dense_1 (Dense)
                           (None, 6)
                                                    42
 dense 2 (Dense)
                           (None, 6)
                                                    42
dense_3 (Dense)
                                                    7
                           (None, 1)
Total params: 103
Non-trainable params: 0
```

Trainable params: 103

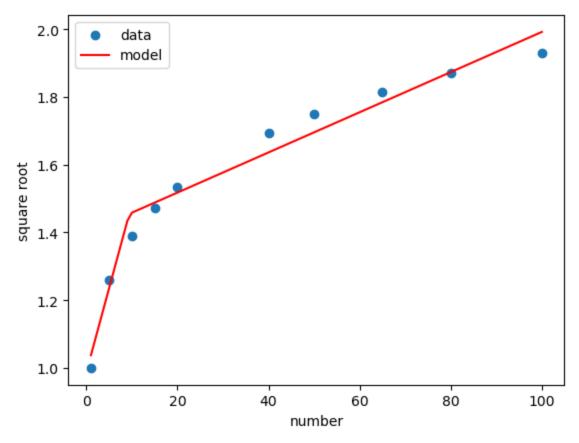
```
1/1 [======] - 0s 85ms/step
[1.0] \Rightarrow 1.04 \text{ (expected 1.00)}
[5.0] \Rightarrow 1.24 \text{ (expected 1.26)}
[10.0] \Rightarrow 1.46 \text{ (expected 1.39)}
[15.0] \Rightarrow 1.49 \text{ (expected 1.47)}
[20.0] => 1.52 (expected 1.53)
[40.0] => 1.64 (expected 1.69)
[50.0] \Rightarrow 1.70 \text{ (expected } 1.75)
[65.0] => 1.78 (expected 1.82)
[80.0] => 1.87 (expected 1.87)
[100.0] => 1.99 (expected 1.93)
```

```
In [16]: import matplotlib.pyplot as plt
         number_grid = np.linspace(1, 100, 100)
```

```
plt.scatter(X,y, label='data')
plt.plot(number_grid,model.predict(np.expand_dims(number_grid,axis=1)) , color='red
plt.xlabel('number')
plt.ylabel('square root')
plt.legend()
```

4/4 [=======] - 0s 2ms/step

Out[16]: <matplotlib.legend.Legend at 0x1f856aa9480>



## Problem 6:

```
In [18]: #define keras model
model = Sequential()
```

```
model.add(Dense(40,input_dim=1,activation='relu'))
model.add(Dense(40,activation='relu'))
model.add(Dense(40,activation='relu'))
model.add(Dense(1))
#compile the keras model
opt = optimizers.Adam(learning_rate=0.001)
mse = tf.keras.losses.MeanSquaredError(
    reduction=tf.keras.losses.Reduction.SUM)
model.compile(loss=mse, optimizer=opt)
print('tarining')
#fit the keras model on the dataset (CPU)
model.fit(X_train,y_train,epochs=2000,batch_size=10, verbose=0)
model.summary()
#make class predictions with the model
predictions = model.predict(X_test)
print('predictions')
for i in range(len(y_test)):
    print('%s => %.2f (expected %.2f)' % (X_test[i].tolist(), predictions[i][0], y_
```

Model: "sequential\_1"

Layer (type)	Output Shape	Param #
dense_4 (Dense)	(None, 40)	80
dense_5 (Dense)	(None, 40)	1640
dense_6 (Dense)	(None, 40)	1640
dense_7 (Dense)	(None, 1)	41

\_\_\_\_\_\_

```
Total params: 3,401
Trainable params: 3,401
Non-trainable params: 0
2/2 [======= ] - 0s 4ms/step
predictions
[6.3] \Rightarrow 4.80 \text{ (expected 4.90)}
[6.8] \Rightarrow 5.34 \text{ (expected 5.50)}
[6.4] \Rightarrow 4.91 \text{ (expected 5.60)}
[5.6] \Rightarrow 3.40 \text{ (expected 4.10)}
[4.9] \Rightarrow 1.59 \text{ (expected 1.40)}
[6.0] \Rightarrow 4.25 \text{ (expected 4.80)}
[6.3] \Rightarrow 4.80 \text{ (expected 4.40)}
[4.4] \Rightarrow 1.16 \text{ (expected 1.30)}
[4.4] \Rightarrow 1.16 \text{ (expected 1.40)}
[5.5] \Rightarrow 2.89 \text{ (expected 4.40)}
[6.9] \Rightarrow 5.44 \text{ (expected 5.10)}
[5.5] \Rightarrow 2.89 \text{ (expected 1.40)}
[5.2] \Rightarrow 1.84 \text{ (expected 3.90)}
[6.5] => 5.02 (expected 5.50)
[7.7] \Rightarrow 6.37 \text{ (expected 6.10)}
[6.5] \Rightarrow 5.02 \text{ (expected 5.80)}
[5.5] \Rightarrow 2.89 \text{ (expected 1.30)}
[4.3] \Rightarrow 1.07 \text{ (expected 1.10)}
[6.1] \Rightarrow 4.46 \text{ (expected 4.70)}
[4.8] \Rightarrow 1.50 \text{ (expected 1.40)}
[5.2] \Rightarrow 1.84 \text{ (expected 1.40)}
[6.3] \Rightarrow 4.80 \text{ (expected 5.10)}
[4.8] \Rightarrow 1.50 \text{ (expected 1.90)}
[6.1] \Rightarrow 4.46 \text{ (expected 4.90)}
[5.1] \Rightarrow 1.76 \text{ (expected 1.60)}
[5.4] \Rightarrow 2.34 \text{ (expected 1.70)}
[5.4] \Rightarrow 2.34 \text{ (expected 1.50)}
[5.6] \Rightarrow 3.40 \text{ (expected 4.90)}
[7.7] \Rightarrow 6.37 \text{ (expected 6.70)}
[5.0] \Rightarrow 1.67 \text{ (expected 1.40)}
[7.4] \Rightarrow 6.02 \text{ (expected 6.10)}
[6.0] \Rightarrow 4.25 \text{ (expected 5.00)}
[4.7] \Rightarrow 1.41 \text{ (expected 1.60)}
[5.1] \Rightarrow 1.76 \text{ (expected 1.40)}
[6.0] \Rightarrow 4.25 \text{ (expected 4.00)}
[5.0] \Rightarrow 1.67 \text{ (expected 3.30)}
```

```
[7.9] \Rightarrow 6.61 \text{ (expected 6.40)}
           [5.4] \Rightarrow 2.34 \text{ (expected 1.70)}
           [5.4] \Rightarrow 2.34 \text{ (expected 1.30)}
           [5.8] \Rightarrow 3.82 \text{ (expected 3.90)}
           [5.0] \Rightarrow 1.67 \text{ (expected 3.50)}
           [5.0] \Rightarrow 1.67 \text{ (expected 1.20)}
           [6.8] \Rightarrow 5.34 \text{ (expected 5.90)}
           [6.7] \Rightarrow 5.23 \text{ (expected 5.20)}
           [5.8] \Rightarrow 3.82 \text{ (expected 5.10)}
           [5.8] \Rightarrow 3.82 \text{ (expected 5.10)}
           [6.3] => 4.80 (expected 5.60)
           [5.5] \Rightarrow 2.89 \text{ (expected 4.00)}
           [5.1] \Rightarrow 1.76 \text{ (expected 1.50)}
           [4.4] \Rightarrow 1.16 \text{ (expected 1.30)}
           [6.5] \Rightarrow 5.02 \text{ (expected 5.10)}
           [5.1] \Rightarrow 1.76 \text{ (expected 1.70)}
           [4.9] \Rightarrow 1.59 \text{ (expected 1.50)}
           [6.7] \Rightarrow 5.23 \text{ (expected 4.70)}
           [6.1] \Rightarrow 4.46 \text{ (expected 4.60)}
           [5.5] \Rightarrow 2.89 \text{ (expected 4.00)}
           [5.7] \Rightarrow 3.61 \text{ (expected 3.50)}
           [5.8] \Rightarrow 3.82 \text{ (expected 5.10)}
           [6.7] \Rightarrow 5.23 \text{ (expected 4.40)}
           [6.4] \Rightarrow 4.91 \text{ (expected 5.30)}
In [19]: import matplotlib.pyplot as plt
            number_grid = np.linspace(4, 8, 30)
             plt.scatter(X_test,y_test, label='test data')
             plt.plot(number_grid,model.predict(np.expand_dims(number_grid,axis=1)) , color='red
             plt.xlabel('sepal length')
             plt.ylabel('petal length')
            plt.legend()
           1/1 [=======] - 0s 18ms/step
Out[19]: <matplotlib.legend.Legend at 0x1f854c756f0>
```

