```
▶ In [69]:
           import pandas as pd
            import tensorflow as tf
            from tensorflow import keras
            from keras.models import Sequential
            from keras.layers import Dense
            import tkinter
            import numpy
            import matplotlib.pyplot as plt
            from sklearn.model selection import train test split
            from keras.wrappers.scikit_learn import KerasRegressor
            from sklearn.metrics import mean squared error
            from sklearn.model selection import cross val score
            from sklearn.model selection import KFold
            from sklearn.preprocessing import StandardScaler
            from sklearn.pipeline import Pipeline
            from sklearn.metrics import r2_score
            # fix random seed for reproducibility
            numpy.random.seed(7)
```

```
▶ In [70]: path = "C:/Users/sbekk/Downloads/housesalesprediction/kc_house_data.csv"
```

```
▶ In [71]: df=pd.read_csv(path)
```

```
#Clean up the unnecessary columns not required for analysis

df=df.drop(['id','date','zipcode','yr_renovated','view','waterfront'],1)

df.head(8)
```

Out[72]:

	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	condition	grade	sqft_above	sqf
0	221900.0	3	1.00	1180	5650	1.0	3	7	1180	
1	538000.0	3	2.25	2570	7242	2.0	3	7	2170	
2	180000.0	2	1.00	770	10000	1.0	3	6	770	
3	604000.0	4	3.00	1960	5000	1.0	5	7	1050	
4	510000.0	3	2.00	1680	8080	1.0	3	8	1680	
5	1225000.0	4	4.50	5420	101930	1.0	3	11	3890	
6	257500.0	3	2.25	1715	6819	2.0	3	7	1715	
7	291850.0	3	1.50	1060	9711	1.0	3	7	1060	

```
▶ In [73]: #put predication column values in y and drop it from X which has other columns for
           y=df['price']
           X=df.drop('price', axis=1)# df=df.values
           # X.head(20)
▶ In [74]: #select 20 percent test and 80 percent training
           X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_s
▶ In [75]: X_train=X_train.values
           print(X train.shape[1])
           X test=X test.values
           #Standardize the data
           meanValue = X train.mean(axis=0)
            stdValue = X train.std(axis=0)
           X train = (X train-meanValue) / stdValue
           X_test=(X_test-meanValue)/stdValue
            print(X train[0])
            # define base model with neuron 130, layers=3, input=14 columns and output=1 colu
            def build ann model():
              model = Sequential([Dense(130, activation=tf.nn.relu,input shape=(X train.shape
                                  Dense(130, activation=tf.nn.relu), Dense(130, activation=tf.
             optimizer = tf.train.RMSPropOptimizer(0.001)
             model.compile(loss='mse',
                            optimizer=optimizer,
                            metrics=['mae'])
              return model
            ann model = build ann model()
              14
              [-0.39003389 -1.44813729 -0.55395019 -0.23023661 -0.91794048 0.9125681
               -0.56106598 -0.86348039 0.46635412 -1.05018286 0.96485795 -0.74469432
               -0.31691025 -0.23245796]
In [76]:
           #Number of rounds(Epochs=500)
           # Store training stats
           history = ann_model.fit(X_train, y_train, epochs=500,
                                validation_split=0.2, verbose=0)
▶ In [77]: # model = Sequential()
           # model.add(Dense(2, input dim=14, activation='relu'))
           # # model.add(Dense(3, activation='sigmoid'))
           # model.add(Dense(1, activation='sigmoid'))
```

Testing set Mean Abs Error: \$75884875.808719

```
y_pred = ann_model.predict(X_test).flatten()
print(y_pred)
mse = mean_squared_error(y_test, y_pred)
print(mse)
```

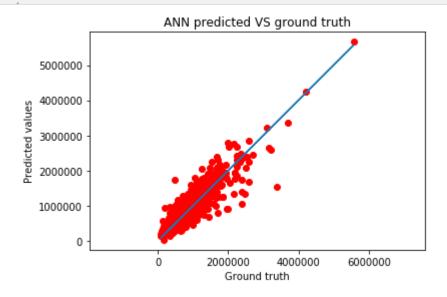
[360247.53 1927065.1 582564.75 ... 381194.1 220061.38 490633.47] 15966229803.899282

```
In [80]: r2 = r2_score(y_test,y_pred)
print(r2)
```

0.8657445624990759

```
▶ In [81]: # model.compile(loss='mse', optimizer='adam', metrics=['accuracy'])
```

```
In [82]: plt.title('ANN predicted VS ground truth')
plt.scatter(y_test, y_pred, color='r')
plt.xlabel('Ground truth')
plt.ylabel('Predicted values')
plt.axis('equal')
plt.xlim(plt.xlim())
plt.ylim(plt.ylim())
_=plt.plot(y_test,y_test)
```



```
▶ In [29]:
           # define base model
            # def baseline_model():
                  # create model
            #
                  model = Sequential()
            #
                  model.add(Dense(13, input dim=14, kernel initializer='normal', activation=
                  model.add(Dense(10, kernel_initializer='normal', activation='relu'))
            #
                  model.add(Dense(8,kernel_initializer='normal', activation='relu'))
            #
                  model.add(Dense(1, kernel_initializer='normal'))
            #
                  # Compile model
                  model.compile(loss='mean_squared_error', optimizer='adam')
            #
            #
                  return model
```

```
▶ In [14]: # seed = 7
                                # numpy.random.seed(seed)
                                # # evaluate model with standardized dataset
                                # estimator = KerasRegressor(build fn=baseline model, epochs=10, batch size=5, ve
                                # type(estimator)
                                # kfold = KFold(n_splits=10, random_state=seed)
                                # results = cross val score(estimator, X, y, cv=kfold)
                                # print("Results: %.2f (%.2f) MSE" % (results.mean(), results.std()))
                                # test predictions = baseline model().predict(X).flatten()
                                # print (test_predictions)
                                # r2 score(y,test predictions)
                                # print(results)
                                # def coeff_determination(y_true, y_pred):
                                                from keras import backend as K
                                                SS res = K.sum(K.square( y true-y pred ))
                                                SS_tot = K.sum(K.square( y_true - K.mean(y_true) ) )
                                                return ( 1 - SS res/(SS tot + K.epsilon()) )
                                # evaluate model with standardized dataset
                                # numpy.random.seed(seed)
                                # estimators = []
                                # estimators.append(('standardize', StandardScaler()))
                                # estimators.append(('mlp', KerasRegressor(build_fn=baseline_model, epochs=50, build_fn=baseline_model, epochs=60, build_fn=baseline_model_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline_fn=baseline
                                # pipeline = Pipeline(estimators)
                                # kfold = KFold(n splits=10, random state=seed)
                                # results = cross_val_score(pipeline, X, y, cv=kfold)
                                # print("Standardized: %.2f (%.2f) MSE" % (results.mean(), results.std()))
```