SCREENSHOTS

```
import pandas as pd
import numpy as np
import tensorflow as tf
import matplotlib.pyplot as plt
from tensorflow.python.framework import ops
from tensorflow.python.framework import tensor_shape
from tensorflow.python.framework import tensor_util
from tensorflow.python.ops import math_ops
from tensorflow.python.ops import random ops
from tensorflow.python.ops import array_ops
from tensorflow.python.layers import utils
from collections import namedtuple
from sklearn.metrics import mean_squared_error
from sklearn.metrics import r2_score
from sklearn.model_selection import KFold
import time
import operator
print(tf.__version__)
```

1.13.1

train=pd.read_csv("train.csv")
test=pd.read_csv("test.csv")

train.head()

y X0 X1 X2 X3 X4 X5 X6 X8 ... X375 X376 X377 X378 X379 X380 X382 X383 X384 X385 0 130.81 d 88.53 d 76.26 d 9 80.62 d 4 13 78.02 az d

5 rows × 378 columns

]: test.head()

ID X0 X1 X2 X3 X4 X5 X6 X8 X10 ... X375 X376 X377 X378 X379 X380 X382 X383 X384 X385 d 0 ... 1 2 b d b 0 ... ai 0 ... d az as а d 0 ... c d 0 ...

5 rows × 377 columns

print(train.shape)
print(test.shape)

(4209, 378) (4209, 377)

train.describe()

	ID	у	X10	X11	X12	X13	X14	X15	X16	
count	4209.000000	4209.000000	4209.000000	4209.0	4209.000000	4209.000000	4209.000000	4209.000000	4209.000000	4209.
mean	4205.960798	100.669318	0.013305	0.0	0.075077	0.057971	0.428130	0.000475	0.002613	0.
std	2437.608688	12.679381	0.114590	0.0	0.263547	0.233716	0.494867	0.021796	0.051061	0.
min	0.000000	72.110000	0.000000	0.0	0.000000	0.000000	0.000000	0.000000	0.000000	0.
25%	2095.000000	90.820000	0.000000	0.0	0.000000	0.000000	0.000000	0.000000	0.000000	0.
50%	4220.000000	99.150000	0.000000	0.0	0.000000	0.000000	0.000000	0.000000	0.000000	0.
75%	6314.000000	109.010000	0.000000	0.0	0.000000	0.000000	1.000000	0.000000	0.000000	0.
max	8417.000000	265.320000	1.000000	0.0	1.000000	1.000000	1.000000	1.000000	1.000000	1.

8 rows × 370 columns

```
[7]: unh_features=[]
      for fea in train:
         if max(train[fea])==min(train[fea]):
             print(fea)
             unh_features.append(fea)
     X11
     X93
     X107
     X233
     X235
     X268
     X289
     X290
     X293
     X297
     X330
     X347
[8]: plt.hist(train.y,bins=300)
     plt.show()
    160
    140
    120
    100
     80
     60
     40
     20
               100
                    125
                               175
                                     200
                                          225
                                                250
                         150
)]: train[train.y >= 170]
          ID
                  y X0 X1 X2 X3 X4 X5 X6 X8 ... X375 X376 X377 X378 X379 X380 X382 X383 X384 X385
    883 1770 265.32 y r ai f d ag
                                           l t ... 0
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 [10]: train=train[train.y < 170]</pre>
        y=train.y
        train=train.drop('y',1)
        df=pd.concat([train,test])
        df=df.drop(unh_features,1)
 [11]: df.shape
 [11]: (8417, 365)
 [12]: dummies=[]
        for col in df:
           if max(df[col]) != 1:
               print(col)
               {\tt dummies.append(col)}
        print(dummies)
        ID
       X0
       X1
        Х2
```

```
ХЗ
      Χ4
      X5
      X6
      ['ID', 'X0', 'X1', 'X2', 'X3', 'X4', 'X5', 'X6', 'X8']
[13]: dummies=dummies[1:]
[14]: for fea in dummies:
          dummy_fea=pd.get_dummies(df[fea], prefix=fea)
          for dummy in dummy_fea:
             df[dummy] = dummy_fea[dummy]
          df = df.drop([fea], 1)
      df = df.drop(['ID'],1)
[15]: df.shape
[15]: (8417, 567)
[16]: df.head()
        X10 X12 X13 X14 X15 X16 X17 X18 X19 X20 ... X8_p X8_q X8_r X8_s X8_t X8_u X8_v X8_w X8_x X8_y
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      5 rows × 567 columns
[17]: trainFinal=df[:len(train)]
      testFinal=df[len(train):]
      yFinal=pd.DataFrame(y)
[18]: def create_weights_biases(num_layers, n_inputs, multiplier, max_nodes):
       '''Use the inputs to create the weights and biases for a network''
       weights = {}
       biases = {}
       for layer in range(1,num_layers):
           if layer == 1:
               weights["h"+str(layer)] = tf.Variable(tf.random_normal([num_features, n_inputs],
                                                                         stddev=np.sqrt(1/num_features)))
               biases["b"+str(layer)] = tf.Variable(tf.random_normal([n_inputs],stddev=0))
               n_previous = n_inputs
           else:
               n_current = int(n_previous * multiplier)
               if n_current >= max_nodes:
                    n_current = max_nodes
               weights["h"+str(layer)] = tf.Variable(tf.random_normal([n_previous, n_current],
                                                                             stddev=np.sqrt(1/n_previous)))
               biases["b"+str(layer)] = tf.Variable(tf.random_normal([n_current], stddev=0))
               n_previous = n_current
```

```
n_current = int(n_previous * multiplier)
           if n_current >= max_nodes:
               n current = max nodes
           weights \verb|"out"| = tf.Variable(tf.random_normal([n\_previous, 1], stddev=np.sqrt(1/n\_previous)))|
           biases["out"] = tf.Variable(tf.random_normal([1],stddev=0))
           return weights, biases
[19]: def network(num_layers, n_inputs, weights, biases, rate, is_training, activation_function):
            '''Add the required number of layers to the network''
           for layer in range(1, num_layers):
               if layer == 1:
                   current layer = eval(activation function + "(tf.matmul(n inputs, weights['h1']) + biases['b1']
                   current_layer = tf.nn.dropout(current_layer, 1-rate)
                   previous_layer = current_layer
               else:
                  current_layer = eval(activation_function + "(tf.matmul(previous_layer,\)
                   weights['h'+str(layer)]) + biases['b'+str(layer)])")
                   current_layer = tf.nn.dropout(current_layer, 1-rate)
                   previous_layer = current_layer
          out_layer = tf.matmul(previous_layer, weights['out']) + biases['out']
          return out_layer
[20]: def model_inputs():
           '''Create placeholders for model's inputs '''
          inputs = tf.placeholder(tf.float32, [None, None], name='inputs')
          targets = tf.placeholder(tf.float32, [None, 1], name='targets')
          learning rate = tf.placeholder(tf.float32, name='learning rate')
          dropout_rate = tf.placeholder(tf.float32, name='dropout_rate')
          is_training = tf.placeholder(tf.bool, name='is_training')
          return inputs, targets, learning_rate, dropout_rate, is_training
[21]: def build_graph(num_layers,n_inputs,weights_multiplier,dropout_rate,learning_rate,max_nodes,activation_fu
           '''Use inputs to build the graph and export the required features for training''
          tf.reset_default_graph()
          inputs, targets, learning_rate, dropout_rate, is_training = model_inputs()
       weights, biases = create_weights_biases(num_layers, n_inputs, weights_multiplier, max_nodes)
       preds = network(num_layers, inputs, weights, biases, dropout_rate, is_training, activation_function)
       with tf.name_scope("cost"):
           cost = tf.losses.mean_squared_error(labels=targets, predictions=preds)
           tf.summary.scalar('cost', cost)
       with tf.name_scope("optimze"):
           optimizer = tf.train.AdamOptimizer(learning_rate).minimize(cost)
       merged = tf.summary.merge_all()
       export_nodes = ['inputs','targets','dropout_rate','is_training','cost','preds','merged',
                        'optimizer','learning_rate']
       Graph = namedtuple('Graph', export_nodes)
       local_dict = locals()
       graph = Graph(*[local_dict[each] for each in export_nodes])
       return graph
```

```
[22]: def train(model, epochs, log_string, learning_rate):
           ''Train the Network and return the average MSE for each iteration of the model'''
          with tf.Session() as sess:
              sess.run(tf.global_variables_initializer())
              testing_loss_summary = []
              iteration = 0
              stop_early = 0
              stop = 10
              learning_rate_decay_threshold = np.random.choice([2,3,4,5])
              original_learning_rate = learning_rate
              print("Training Model: {}".format(log_string))
              train_writer = tf.summary.FileWriter('./logs/1/train/{}'.format(log_string), sess.graph)
              test_writer = tf.summary.FileWriter('./logs/1/test/{}'.format(log_string))
            kf = KFold(n_splits=n_splits, shuffle=True, random_state=2)
            split = 0
            sum_loss_testing = 0
            for train_index, test_index in kf.split(trainFinal):
                x_train = trainFinal.iloc[train_index]
                y_train = yFinal.iloc[train_index]
                x_test = trainFinal.iloc[test_index]
                y_test = yFinal.iloc[test_index]
                training\_check = (len(x\_train)//batch\_size)-1 # Check training progress after this many batch_size)
                testing_check = training_check # Check testing results
                split += 1
                print('Start KFold number {} from {}'.format(split, n_splits))
                for epoch_i in range(1, epochs+1):
                    batch_loss = 0
                    batch_time = 0
                   for batch in range(int(len(x_train)/batch_size)):
                       batch_x = x_train[batch*batch_size:(1+batch)*batch_size]
                       batch_y = y_train[batch*batch_size:(1+batch)*batch_size]
                       start_time = time.time()
                       summary, loss, _ = sess.run([model.merged,
                                                     model.cost,
                                                     model.optimizer],
                                                     {model.inputs: batch_x,
                                                      model.targets: batch_y,
                                                      model.learning_rate: learning_rate,
                                                      model.dropout_rate: dropout_rate,
                                                      model.is_training: True})
                       batch_loss += loss
                       end_time = time.time()
                       batch_time += end_time - start_time
                       train_writer.add_summary(summary, iteration)
```

```
iteration += 1
  if batch % training_check == 0 and batch > 0:
       print('Epoch {:>3}/{} Batch {:>4}/{} - MSE: {:>6.3f}, Seconds: {:>4.2f}'
             .format(epoch_i,
                      epochs,
                      batch,
                      len(x_train) // batch_size,
                      (batch_loss / training_check),
                      batch_time))
       batch_loss = 0
       batch_time = 0
  if batch % testing_check == 0 and batch > 0:
       batch_loss_testing = 0
       batch_time_testing = 0
       \label{eq:forbatch} \textbf{for} \ \ \mathsf{batch} \ \ \ \mathsf{in} \ \ \mathsf{range}(\mathsf{int}(\mathsf{len}(\mathsf{x\_test})/\mathsf{batch\_size})) \colon
           batch_x = x_test[batch*batch_size:(1+batch)*batch_size]
           batch_y = y_test[batch*batch_size:(1+batch)*batch_size]
           start_time_testing = time.time()
             summary, loss = sess.run([model.merged,
                                         model.cost],
                                            {model.inputs: batch_x,
                                             model.targets: batch_y,
                                             model.learning_rate: learning_rate,
                                             model.dropout_rate: 0,
                                             model.is_training: False})
             batch_loss_testing += loss
             end_time_testing = time.time()
             batch_time_testing += end_time_testing - start_time_testing
             test_writer.add_summary(summary, iteration)
         n_batches_testing = batch + 1
         print('Testing MSE: {:>6.3f}, Seconds: {:>4.2f}'
               .format(batch_loss_testing / n_batches_testing,
                        batch_time_testing))
         batch_time_testing = 0
         testing_loss_summary.append(batch_loss_testing)
        if batch_loss_testing <= min(testing_loss_summary):</pre>
             print('New Record!')
            lowest_loss_testing = batch_loss_testing/n_batches_testing
            stop_early = 0 # Reset stop_early if new minimum loss is found
            checkpoint = "./{}.ckpt".format(log_string)
            saver = tf.train.Saver()
            saver.save(sess, checkpoint)
        else:
            print("No Improvement.")
            stop_early += 1 # Increase stop_early if no new minimum loss is found
            if stop_early % learning_rate_decay_threshold == 0:
                 learning_rate *= learning_rate_decay
                 print("New learning rate = ", learning_rate)
            elif stop_early == stop:
                 break
if stop_early == stop:
    print("Stopping training for this fold.")
    print("Lowest MSE =", lowest_loss_testing)
    sum_loss_testing += lowest_loss_testing
```

```
testing_loss_summary = []
                                               learning_rate = original_learning_rate
                                               break
                         average_testing_loss = sum_loss_testing/n_splits
                         print("Stopping training for this iteration.")
                         print("Average MSE for this iteration: ", average_testing_loss)
                  return average_testing_loss
[23]: num_iterations = 25
            results = {}
           for i in range(num_iterations):
                  num features = trainFinal.shape[1]
                  epochs = 50
                  learning_rate = np.random.uniform(0.001, 0.1)
                  learning_rate_decay = np.random.uniform(0.1,0.5)
                  weights_multiplier = np.random.uniform(0.5,2)
                  n_inputs = np.random.randint(int(num_features)*0.1,int(num_features)*2)
          num layers = np.random.choice([2,3,4])
          dropout_rate = np.random.uniform(0,0.3)
          batch_size = np.random.choice([64,128,256])
          max_nodes = np.random.choice([32,64,128,256,512,1024,2048,4096])
          activation_function = np.random.choice(['tf.nn.sigmoid',
                                                                                     'tf.nn.relu',
                                                                                    'tf.nn.elu'])
          print("Starting iteration #",i+1)
          log\_string = 'LR={}, LRD={}, NI={}, NL={}, DR={}, BS={}, MN={}, AF={}'.format(learning\_rate, BR={}, BR={}, BR={}, AF={}'.format(learning\_rate, BR={}, BR={}, BR={}, AF={}, BR={}, AF={}, BR={}, AF={}, AF={}
                                                                                                                                                        learning_rate_decay,
                                                                                                                                                        weights_multiplier,
                                                                                                                                                        n_inputs,
                                                                                                                                                        num layers,
                                                                                                                                                        dropout_rate,
                                                                                                                                                        batch size,
                                                                                                                                                        max_nodes,
                                                                                                                                                        activation_function)
          model = build_graph(num_layers, n_inputs, weights_multiplier,
                                               dropout_rate,learning_rate,max_nodes,activation_function)
          result = train(model, epochs, log_string, learning_rate)
            results[log_string] = result
     Starting iteration # 1
     WARNING:tensorflow:From /opt/anaconda3/lib/python3.7/site-packages/tensorflow/python/framework/op_def_lib
     rary.py:263: colocate_with (from tensorflow.python.framework.ops) is deprecated and will be removed in a
     future version.
     Instructions for updating:
     Colocations handled automatically by placer.
     WARNING:tensorflow:From <ipython-input-19-f1d0724fa693>:7: calling dropout (from tensorflow.python.ops.nn
     ops) with keep prob is deprecated and will be removed in a future version.
     Instructions for updating:
     Please use `rate` instead of `keep_prob`. Rate should be set to `rate = 1 - keep_prob`.
     {\tt WARNING:tensorflow:From /opt/anaconda3/lib/python3.7/site-packages/tensorflow/python/ops/losses/losses\_im}
     pl.py:667: to_float (from tensorflow.python.ops.math_ops) is deprecated and will be removed in a future v
     ersion.
     Instructions for updating:
     Use tf.cast instead.
     Training Model: LR=0.004890773749952225,LRD=0.35431805771808944,WM=1.5946404035599373,NI=756,NL=4,DR=0.13
     858139670331282,BS=128,MN=1024,AF=tf.nn.sigmoid
     Start KFold number 1 from 5
     Epoch 1/50 Batch 25/26 - MSE: 3972.145, Seconds: 1.36
```

early_stop = 0

```
Lowest MSE = 62.24092395489033
       Stopping training for this iteration.
       Average MSE for this iteration: 59.9557729280912
[24]: def find_inputs(model):
            '''Use the log_string from the model to extract the values for all of the model's inputs'''
           learning_rate_start = model.find('LR=') + 3
           learning_rate_end = model.find(',LRD', learning_rate_start)
           learning_rate = float(model[learning_rate_start:learning_rate_end])
           learning_rate_decay_start = model.find('LRD=') + 4
           learning_rate_decay_end = model.find(',WM', learning_rate_decay_start)
           learning_rate_decay = float(model[learning_rate_decay_start:learning_rate_decay_end])
           weights_multiplier_start = model.find('WM=') + 3
           weights_multiplier_end = model.find(',NI', weights_multiplier_start)
           weights_multiplier = float(model[weights_multiplier_start:weights_multiplier_end])
         n_inputs_start = model.find('NI=') + 3
         n_inputs_end = model.find(',NL', n_inputs_start)
         n_inputs = int(model[n_inputs_start:n_inputs_end])
         num_layers_start = model.find('NL=') + 3
         num_layers_end = model.find(',DR', num_layers_start)
         num_layers = int(model[num_layers_start:num_layers_end])
         dropout_rate_start = model.find('DR=') + 3
         dropout rate end = model.find(',BS', dropout rate start)
         dropout_rate = float(model[dropout_rate_start:dropout_rate_end])
         batch size start = model.find('BS=') + 3
         batch_size_end = model.find(',MN', batch_size_start)
         batch_size = int(model[batch_size_start:batch_size_end])
         max\_nodes\_start = model.find('MN=') + 3
         max_nodes_end = model.find(',AF', max_nodes_start)
         max_nodes = int(model[max_nodes_start:max_nodes_end])
          activation_function_start = model.find('AF=') + 3
          activation_function = str(model[activation_function_start:])
          return (learning_rate, learning_rate_decay, weights_multiplier, n_inputs,
                  num_layers, dropout_rate, batch_size, max_nodes, activation_function)
[25]: sorted_results = sorted(results.items(), key=operator.itemgetter(1))
[26]: results_df = pd.DataFrame(columns=["learning_rate",
                                          "learning_rate_decay",
                                          "weights_multiplier",
                                          "n_inputs",
                                          "num_layers"
                                          "dropout_rate",
                                         "batch_size",
                                          "max_nodes",
                                          "activation_function"])
      for result in sorted_results:
          learning_rate, learning_rate_decay, weights_multiplier, n_inputs,\
              num_layers, dropout_rate, batch_size, max_nodes, activation_function = find_inputs(result[0])
```

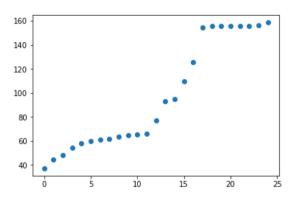
Stopping training for this fold.

```
MSE = result[1]
       new_row = pd.DataFrame([[MSE,
                                 learning_rate,
                                 learning_rate_decay,
                                 {\tt weights\_multiplier},
                                 n_inputs,
                                 num_layers,
                                 dropout_rate,
                                 batch_size,
                                 max_nodes,
                                 activation_function]],
                        columns = ["MSE",
                                    "learning_rate",
                                    "learning_rate_decay",
                                    "weights_multiplier",
                                    "n_inputs",
                                    "num_layers",
                                    "dropout_rate",
                                    "batch_size",
                                    "max_nodes",
                                    "activation_function"])
          results_df = results_df.append(new_row, ignore_index=True,sort=False)
[27]: results_df.head()
```

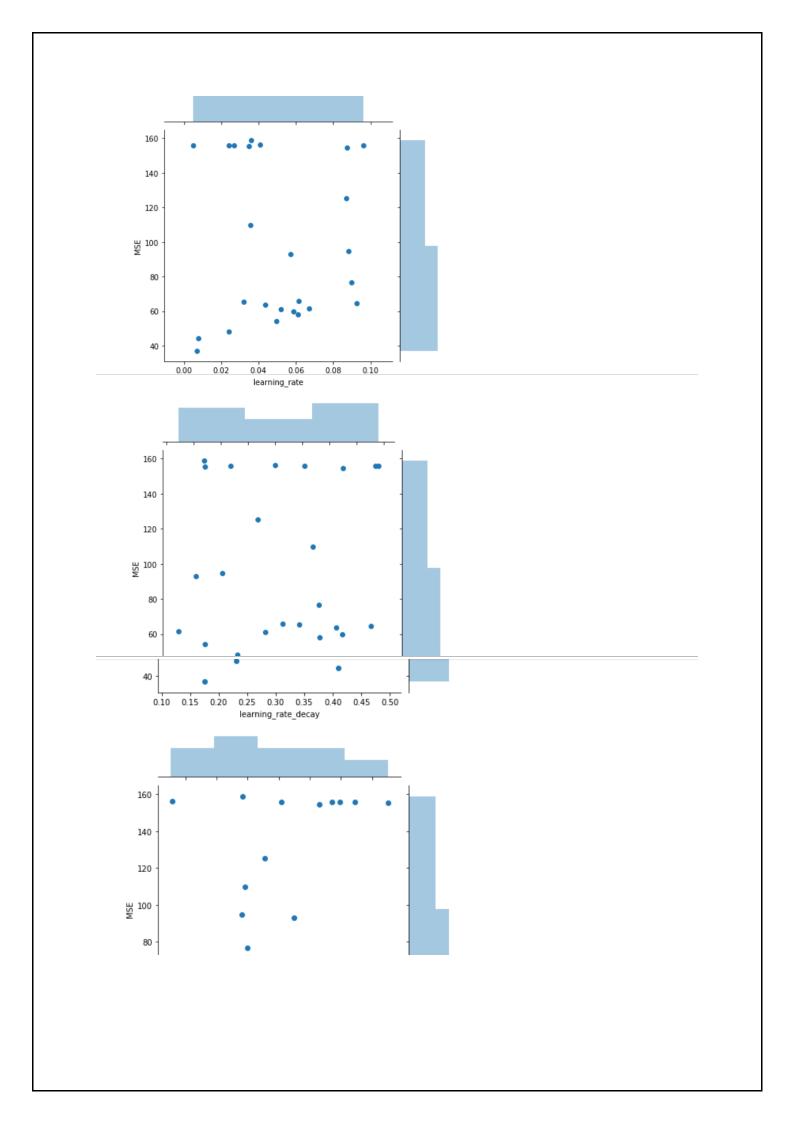
:	learning_rate	learning_rate_decay	weights_multiplier	n_inputs	num_layers	dropout_rate	batch_size	max_nodes	activati
0	0.006848	0.175169	0.505300	1032	2	0.217217	128	1024	
1	0.007817	0.409927	1.509780	271	2	0.133306	256	1024	1
2	0.023933	0.230210	0.643985	295	2	0.136398	256	256	
3	0.049629	0.171783	1.384525	888	2	0.144926	256	2048	1
4	0.061219	0.382154	0.746649	112	3	0.097582	128	128	
4)

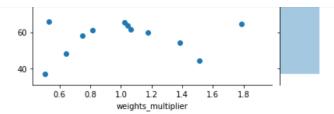
[28]: plt.scatter(results_df.index, results_df.MSE)

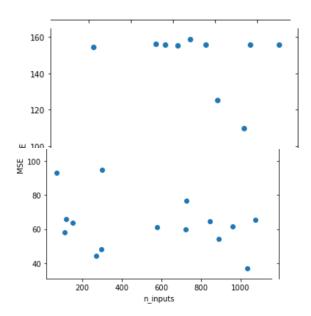
[28]: <matplotlib.collections.PathCollection at 0x7f6714197710>



```
[29]: import seaborn as sns
      for feature in results_df:
          if feature == "MSE":
              continue
          elif feature == "activation_function":
              sns.stripplot(x=feature, y="MSE", data=results_df)
               sns.stripplot(x=feature, y="MSE", data=results_df)
               sns.jointplot(x=feature, y="MSE", data=results_df)
```







```
[30]: best_models = [] # contains the log_strings of the best iterations, to be used for the final predictions
best_R2 = 0 # records the best R2 score
best_predictions = pd.DataFrame([0]*len(trainFinal)) # records the best predictions for each row
current_model = 1 # Used to equally weight the predictions from each iteration
testing_limit = 3 # If 3 consectutive iterations do not improve the best R2, stop predicting

for model, result in sorted_results:
    checkpoint = str(model) + ".ckpt"
```

```
_, _, weights_multiplier, n_inputs, num_layers, _, _, max_nodes, activation_function = find_inputs(mode
model = build_graph(num_layers,n_inputs,weights_multiplier,dropout_rate,
                    learning_rate,max_nodes,activation_function)
# Predict one row at a time
batch_size = 1
with tf.Session() as sess:
   saver = tf.train.Saver()
    {\tt saver.restore}({\tt sess, checkpoint})
   predictions = [] # record the predictions
    for batch in range(int(len(trainFinal)/batch_size)):
       batch_x = trainFinal[batch*batch_size:(1+batch)*batch_size]
        batch_predictions = sess.run([model.preds],
                               {model.inputs: batch_x,
                                model.learning_rate: learning_rate,
                                model.dropout_rate: 0,
                                model.is_training: False})
```

```
for prediction in batch_predictions[0]:
                      predictions.append(prediction)
         predictions = pd.DataFrame(predictions)
         R2 = r2_score(y, predictions)
         print("R2 Score = ", R2)
         # Equally weight each prediction
         combined_predictions = (best_predictions*(current_model-1) + predictions) / current_model
         # Find the r2 score with the new predictions
         new_R2 = r2_score(y, combined_predictions)
         print("New R2 score = ", new_R2)
         if new_R2 >= best_R2:
             best_predictions = combined_predictions
             best_R2 = new_R2
             best_models.append(checkpoint)
             limit = 0
             current_model += 1
             print("Improvement!")
              print()
              print("No improvement.")
               limit += 1
               if limit == testing_limit:
                   print("Stopping predictions.")
                   break
       4
       WARNING:tensorflow:From /opt/anaconda3/lib/python3.7/site-packages/tensorflow/python/training/saver.py:126
       6: checkpoint_exists (from tensorflow.python.training.checkpoint_management) is deprecated and will be rem
       oved in a future version.
       Instructions for updating:
       Use standard file APIs to check for files with this prefix.
       INFO:tensorflow:Restoring parameters from LR=0.0068480379898597865,LRD=0.17516904929685626,WM=0.5053003614
       547112,NI=1032,NL=2,DR=0.21721676724454952,BS=128,MN=1024,AF=tf.nn.relu.ckpt
       R2 Score = 0.6289213546775874
       New R2 score = 0.6289213546775874
       Improvement!
       INFO:tensorflow:Restoring parameters from LR=0.00781732390082656,LRD=0.40992728225770814,WM=1.509779640465
       8745,NI=271,NL=2,DR=0.13330576797679408,BS=256,MN=1024,AF=tf.nn.sigmoid.ckpt
       R2 Score = 0.6660128099903835
      INFO:tensorflow:Restoring parameters from LR=0.0518475775948857,LRD=0.2823241657182445,WM=0.81403872251832
      15,NI=576,NL=2,DR=0.2495012193140341,BS=256,MN=512,AF=tf.nn.sigmoid.ckpt
      R2 Score = 0.6627405909664613
      New R2 score = 0.6978602810391386
      No improvement.
      INFO:tensorflow:Restoring parameters from LR=0.06721761689863513,LRD=0.12307521561981148,WM=1.065323196195
      3413,NI=958,NL=2,DR=0.15110124137972852,BS=128,MN=64,AF=tf.nn.sigmoid.ckpt
      R2 Score = 0.6272154307884952
      New R2 score = 0.6962118421201016
      No improvement.
      INFO:tensorflow:Restoring parameters from LR=0.043469772168736924,LRD=0.41208299100043333,WM=1.04327522493
      63166,NI=150,NL=2,DR=0.06527166637721249,BS=128,MN=128,AF=tf.nn.elu.ckpt
      R2 Score = 0.6586452299697911
      New R2 score = 0.6954826550351285
      No improvement.
      Stopping predictions.
[31]: best_models
```

```
[31]: ['LR=0.0068480379898597865,LRD=0.17516904929685626,WM=0.5053003614547112,NI=1032,NL=2,DR=0.217216767244549
       52,BS=128,MN=1024,AF=tf.nn.relu.ckpt'
        LR=0.00781732390082656,LRD=0.40992728225770814,WM=1.5097796404658745,NI=271,NL=2,DR=0.13330576797679408,
       BS=256, MN=1024, AF=tf.nn.sigmoid.ckpt',
        LR=0.049629404605042365,LRD=0.1717828003094657,WM=1.3845247620511545,NI=888,NL=2,DR=0.14492580384481166,
       BS=256, MN=2048, AF=tf.nn.sigmoid.ckpt',
        LR=0.06121919712054469,LRD=0.3821542267514719,WM=0.7466492797403004,NI=112,NL=3,DR=0.09758179510246075,B
       S=128, MN=128, AF=tf.nn.elu.ckpt',
        LR=0.05890537282055379,LRD=0.42399730663735824,WM=1.1743309126467907,NI=723,NL=2,DR=0.1252452256987556,B
       S=64,MN=2048,AF=tf.nn.elu.ckpt'l
[32]: best_predictions = pd.DataFrame([0]*len(testFinal))
       current_model = 1
       for model in best_models:
           checkpoint = model
            _, _, weights_multiplier, n_inputs, num_layers, _, _, max_nodes, activation_function = find_inputs(mode
           # Remove '.ckpt' from the activation_function string
           activation_function = activation_function[:activation_function.find('.ckpt')]
         model = build_graph(num_layers,n_inputs,weights_multiplier,dropout_rate,
                             learning_rate,max_nodes,activation_function)
         batch_size = 1
         with tf.Session() as sess:
             saver = tf.train.Saver()
             saver.restore(sess, checkpoint)
             predictions = []
             for batch in range(int(len(testFinal)/batch_size)):
                 batch_x = testFinal[batch*batch_size:(1+batch)*batch_size]
                 batch_predictions = sess.run([model.preds],
                                        {model.inputs: batch_x,
                                         model.learning_rate: learning_rate,
                                         model.dropout_rate: 0,
                                         model.is_training: False})
                 for prediction in batch_predictions[0]:
                     predictions.append(prediction)
          predictions = pd.DataFrame(predictions)
          combined_predictions = (best_predictions*(current_model-1) + predictions) / current_model
          best_predictions = combined_predictions
          current model += 1
      INFO:tensorflow:Restoring parameters from LR=0.0068480379898597865,LRD=0.17516904929685626,WM=0.5053003614
      547112,NI=1032,NL=2,DR=0.21721676724454952,BS=128,MN=1024,AF=tf.nn.relu.ckpt
      INFO:tensorflow:Restoring parameters from LR=0.00781732390082656,LRD=0.40992728225770814,WM=1.509779640465
      8745,NI=271,NL=2,DR=0.13330576797679408,BS=256,MN=1024,AF=tf.nn.sigmoid.ckpt
      INFO:tensorflow:Restoring parameters from LR=0.049629404605042365,LRD=0.1717828003094657,WM=1.384524762051
      1545,NI=888,NL=2,DR=0.14492580384481166,BS=256,MN=2048,AF=tf.nn.sigmoid.ckpt
      INFO:tensorflow:Restoring parameters from LR=0.06121919712054469,LRD=0.3821542267514719,WM=0.7466492797403
      004,NI=112,NL=3,DR=0.09758179510246075,BS=128,MN=128,AF=tf.nn.elu.ckpt
      INFO:tensorflow:Restoring parameters from LR=0.05890537282055379,LRD=0.42399730663735824,WM=1.174330912646
      7907,NI=723,NL=2,DR=0.1252452256987556,BS=64,MN=2048,AF=tf.nn.elu.ckpt
[33]: best_predictions['ID'] = test.ID
      best_predictions['y'] = best_predictions[0]
      best_predictions = best_predictions.drop([0],1)
```

```
best_predictions.to_csv("submission.csv", index=False)
[34]: best_predictions.head()
[34]: ID y
     0 1 78.582414
      1 2 96.226047
      2 3 81.614616
      3 4 76.747839
      4 5 105.431487
[35]: best_predictions.describe()
                        у
       ID
[35]:
     count 4209.000000 4209.000000
      mean 4211.039202 100.324548
                      9.867065
       std 2423.078926
      min 1.000000 73.410442
       25% 2115.000000 92.836122
       50% 4202.000000 96.913811
       75% 6310.000000 109.961414
       max 8416.000000 125.187039
[36]: yFinal.describe()
[36]:
      count 4208.000000
      mean 100.630190
       std 12.424146
       min 72.110000
       25% 90.817500
       50% 99.150000
       75% 109.010000
       max 169.910000
```