```
import numpy as np
import pandas as pd
import seaborn as sns
import itertools
import matplotlib.pyplot as plt
import string
import re
import collections
from sklearn import preprocessing

%matplotlib inline

In [2]: # READ DATA
train_df = pd.read_json('train.json.zip')
test_df = pd.read_json('test.json.zip')
```

TRAIN DATA FEATURE ENGINEERING

```
In [3]:
         # convert TARGET to the numeric
         train_df['interest_level'] = train_df['interest_level'].apply(lambda x: 0 if x=='low
                                                                else 1 if x=='medium'
                                                                else 2)
         # REMOVE UNNECESSARY WORDS FROM DESCRIPTION
         train df['description'] = train_df['description'].apply(lambda x: x.replace("<br />"
         train_df['description'] = train_df['description'].apply(lambda x: x.replace("br", ""
         train_df['description'] = train_df['description'].apply(lambda x: x.replace("<a",</pre>
         #basic features
         train_df['rooms'] = train_df['bedrooms'] + train_df['bathrooms']
         # count of photos #
         train_df["num_photos"] = train_df["photos"].apply(len)
         # count of "features" #
         train_df["num_features"] = train_df["features"].apply(len)
         # count of words present in description column #
         train_df["num_description_words"] = train_df["description"].apply(lambda x: len(x.sp
         # description contains email
         regex = r'[\w\.-]+@[\w\.-]+'
         train_df['has_email'] = train_df['description'].apply(lambda x: 1 if re.findall(rege
         # description contains phone
         train df['has phone'] = train df['description'].apply(lambda x:re.sub('['+string.pun
                 .apply(lambda x: [s for s in x if s.isdigit()])\
                 .apply(lambda x: len([s for s in x if len(str(s))==10]))\
                 .apply(lambda x: 1 if x>0 else 0)
         # CONVERT LOWER ALL OF WORDS
         train_df[["features"]] = train_df[["features"]].apply(
             lambda _: [list(map(str.strip, map(str.lower, x))) for x in _])
```

TEST DATA FEATURE ENGINEERING

```
# REMOVE UNNECESSARY WORDS FROM DESCRIPTION
test_df['description'] = test_df['description'].apply(lambda x: x.replace("<br/>",")
```

```
test_df['description'] = test_df['description'].apply(lambda x: x.replace("br", ""))
test_df['description'] = test_df['description'].apply(lambda x: x.replace("<a",</pre>
#basic features
test df['rooms'] = test df['bedrooms'] + test df['bathrooms']
# count of photos #
test_df["num_photos"] = test_df["photos"].apply(len)
# count of "features" #
test_df["num_features"] = test_df["features"].apply(len)
# count of words present in description column #
test_df["num_description_words"] = test_df["description"].apply(lambda x: len(x.spli
# description contains email
regex = r'[\w\.-]+@[\w\.-]+'
test_df['has_email'] = test_df['description'].apply(lambda x: 1 if re.findall(regex,
# description contains phone
test_df['has_phone'] = test_df['description'].apply(lambda x:re.sub('['+string.punct
        .apply(lambda x: [s for s in x if s.isdigit()])\
        .apply(lambda x: len([s for s in x if len(str(s))==10]))\
        .apply(lambda x: 1 if x>0 else 0)
# CONVERT LOWER ALL OF WORDS
test_df[["features"]] = test_df[["features"]].apply(
    lambda _: [list(map(str.strip, map(str.lower, x))) for x in _])
```

MOST FREQUENT FEATURES EXTRACTION

```
In [5]:
         feature value train = train df['features'].tolist()
         feature_value_test = test_df['features'].tolist()
         feature_value_train
         feature_value_test
         feature_lst_train = []
         feature_lst_test = []
         for i in range(len(feature value train)):
             feature lst train += feature value train[i]
         for i in range(len(feature value test)):
             feature lst test += feature value test[i]
         uniq_feature_train = list(set(feature_lst_train))
         uniq_feature_test = list(set(feature_lst_test))
         # see the frequency of each feature
         def most common(lst):
             features = collections.Counter(lst)
             feature_value = features.keys()
             frequency = features.values()
             data = [('feature_value', feature_value),
                     ('frequency', frequency),]
             df = pd.DataFrame.from_dict(dict(data))
             return df.sort_values(by = 'frequency', ascending = False)
         df features train = most common(feature lst train)
         df features test = most common(feature lst test)
```

```
def newColumn(name, df, series):
    feature = pd.Series(0,df.index,name = name)# data : 0
    for row,word in enumerate(series):
        if name in word:
            feature.iloc[row] = 1
    df[name] = feature # feature : series ; value in series : 1 or 0
    return df

# select features based on frequency
facilities = ['elevator', 'cats allowed', 'hardwood floors', 'dogs allowed', 'doorma
for name in facilities:
    train_df = newColumn(name, train_df, train_df['features'])
    test_df = newColumn(name, test_df, test_df['features'])
```

LABEL ENCODING

```
In [6]:
         categorical = ["display_address", "manager_id", "building_id", "street_address"]
         for f in categorical:
                 if train_df[f].dtype=='object':
                     #print(f)
                     lbl = preprocessing.LabelEncoder()
                     lbl.fit(list(train df[f].values) + list(test df[f].values))
                     train_df[f] = lbl.transform(list(train_df[f].values))
                     test_df[f] = lbl.transform(list(test_df[f].values))
In [7]:
         train_df['price'] = np.log10(train_df['price'])
         test_df['price'] = np.log10(test_df['price'])
In [8]:
        train_df['price']
                  3.380211
Out[8]: 4
                  3.579784
        6
                  3.543447
                  3.477121
        10
        15
                  3.446382
                3,447158
        124000
        124002
                3.379306
        124004
                  3,267172
        124008
                  3,622732
                  3.631444
        124009
        Name: price, Length: 49352, dtype: float64
```

DROP UNNECESSARY COLUMNS

```
In [9]: # TRAINING DATASET
    train_df.drop('created', axis=1, inplace=True)
    train_df.drop('description', axis=1, inplace=True)
    train_df.drop('features', axis=1, inplace=True)
    train_df.drop('photos', axis=1, inplace=True)

# TEST DATASET
    test_df.drop('created', axis=1, inplace=True)
    test_df.drop('description', axis=1, inplace=True)
```

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```
test_df.drop('features', axis=1, inplace=True)
test_df.drop('photos', axis=1, inplace=True)
```

```
REGRESSION FOR PRICE
In [10]:
          from sklearn.model_selection import train_test_split
          from sklearn.metrics import make_scorer, mean_absolute_error, mean_squared_error
          from sklearn.preprocessing import StandardScaler
          from sklearn.pipeline import Pipeline
          import xgboost as xgb
          from sklearn.linear_model import LinearRegression
          from sklearn.ensemble import RandomForestRegressor
          from sklearn.svm import SVR
          import optuna
          import math
          from keras import callbacks
          X = train_df.drop(['price'], axis = 1)
          y = train_df.price
          X_train, X_test, y_train, y_test = train_test_split(X, y,
                                                             test_size = .3,
                                                             random_state = 5)
In [11]:
          class Optimizer:
              def __init__(self, metric, trials=100):
                  self.metric = metric
                  self.trials = trials
              def objective(self, trial):
                 model = create model(trial)
                 model.fit(X, y)
                  preds = model.predict(X_test)
```

```
return mean_absolute_error(y_test, preds)
def optimize(self):
   study = optuna.create_study(direction="minimize")
    study.optimize(self.objective, n_trials=self.trials)
    return study
```

```
In [12]:
          from keras.layers import Dense
          from keras.models import Sequential
          def create model(trial):
              model = Sequential()
              model.add(Dense
                               units = trial.suggest_int("units", 32, 256),
                               kernel_initializer=trial.suggest_categorical("kernel_initializer
                               input_dim = X.shape[1],
                               activation= trial.suggest_categorical("activation", ["relu", "si
              model.add(Dense(1))
              model.compile(
                  loss="mean absolute error"
              return model
          optimizer = Optimizer('mae')
```

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keras_study = optimizer.optimize()

print("Number of finished trials: ", len(keras study.trials))

```
print("Best trial:")
keras trial = keras study.best trial
print(" Value: {}".format(keras_trial.value))
print(" Params: ")
for key, value in keras_trial.params.items():
          {}: {}".format(key, value))
[I 2021-09-28 17:53:36,049] A new study created in memory with name: no-name-a331b82
c-4720-4590-9c92-fed25c0a8156
[I 2021-09-28 17:53:39,674] Trial 0 finished with value: 16515.011623708084 and para
meters: {'units': 177, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best i
s trial 0 with value: 16515.011623708084.
[I 2021-09-28 17:53:41,675] Trial 1 finished with value: 90180.31850626413 and param
eters: {'units': 203, 'kernel_initializer': 'glorot_uniform', 'activation': 'relu'}.
Best is trial 0 with value: 16515.011623708084.
[I 2021-09-28 17:53:43,949] Trial 2 finished with value: 80783.38150766604 and param
eters: {'units': 243, 'kernel_initializer': 'glorot_uniform', 'activation': 'silu'}.
Best is trial 0 with value: 16515.011623708084.
[I 2021-09-28 17:53:45,938] Trial 3 finished with value: 9333.99197716002 and parame
ters: {'units': 143, 'kernel_initializer': 'glorot_uniform', 'activation': 'leaky_re
lu'}. Best is trial 3 with value: 9333.99197716002.
[I 2021-09-28 17:53:47,941] Trial 4 finished with value: 75658.22424878449 and param
eters: {'units': 216, 'kernel_initializer': 'glorot_uniform', 'activation': 'leaky_r
elu'}. Best is trial 3 with value: 9333.99197716002.
[I 2021-09-28 17:53:49,956] Trial 5 finished with value: 7424.058743536404 and param
eters: {'units': 214, 'kernel_initializer': 'glorot_uniform', 'activation': 'leaky_r
elu'}. Best is trial 5 with value: 7424.058743536404.
[I 2021-09-28 17:53:52,075] Trial 6 finished with value: 9722.378897348204 and param
eters: {'units': 145, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best is
trial 5 with value: 7424.058743536404.
[I 2021-09-28 17:53:53,991] Trial 7 finished with value: 10187.92243860196 and param
eters: {'units': 83, 'kernel_initializer': 'glorot_uniform', 'activation': 'relu'}.
Best is trial 5 with value: 7424.058743536404.
[I 2021-09-28 17:53:55,911] Trial 8 finished with value: 40956.81029221663 and param
eters: {'units': 68, 'kernel_initializer': 'glorot_uniform', 'activation': 'leaky_re
lu'}. Best is trial 5 with value: 7424.058743536404.
[I 2021-09-28 17:53:57,861] Trial 9 finished with value: 5143.597371840069 and param
eters: {'units': 94, 'kernel_initializer': 'normal', 'activation': 'leaky_relu'}. Be
st is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:53:59,795] Trial 10 finished with value: 7741.619826115889 and para
meters: {'units': 37, 'kernel initializer': 'normal', 'activation': 'leaky relu'}. B
est is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:01,743] Trial 11 finished with value: 5721.640498166707 and para
meters: {'units': 109, 'kernel_initializer': 'normal', 'activation': 'leaky_relu'}.
Best is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:04,035] Trial 12 finished with value: 54769.80959808292 and para
meters: {'units': 109, 'kernel_initializer': 'normal', 'activation': 'leaky_relu'}.
```

```
Best is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:05,980] Trial 13 finished with value: 17666.23294827709 and para
meters: {'units': 114, 'kernel initializer': 'normal', 'activation': 'leaky relu'}.
Best is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:07,933] Trial 14 finished with value: 18301.660984604212 and par
ameters: {'units': 70, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best i
s trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:09,873] Trial 15 finished with value: 26945.909688945576 and par
ameters: {'units': 33, 'kernel_initializer': 'normal', 'activation': 'leaky relu'}.
Best is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:11,832] Trial 16 finished with value: 40552.65903491385 and para
meters: {'units': 114, 'kernel_initializer': 'normal', 'activation': 'leaky_relu'}.
Best is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:13,855] Trial 17 finished with value: 10569.08211923935 and para
meters: {'units': 149, 'kernel_initializer': 'normal', 'activation': 'leaky_relu'}.
Best is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:15,812] Trial 18 finished with value: 11579.790785106967 and par
ameters: {'units': 92, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best i
s trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:17,833] Trial 19 finished with value: 19686.915943272368 and par
ameters: {'units': 53, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best i
s trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:19,835] Trial 20 finished with value: 47414.51819475656 and para
meters: {'units': 168, 'kernel_initializer': 'normal', 'activation': 'leaky_relu'}.
Best is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:21,871] Trial 21 finished with value: 93590.47520912152 and para
meters: {'units': 252, 'kernel_initializer': 'glorot_uniform', 'activation': 'leaky_
relu'}. Best is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:23,842] Trial 22 finished with value: 105114.55650692731 and par
ameters: {'units': 127, 'kernel_initializer': 'glorot_uniform', 'activation': 'leaky
_relu'}. Best is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:25,853] Trial 23 finished with value: 9696.88126831215 and param
eters: {'units': 186, 'kernel_initializer': 'glorot_uniform', 'activation': 'leaky_r
elu'}. Best is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:27,803] Trial 24 finished with value: 19785.04346982452 and para
meters: {'units': 92, 'kernel initializer': 'normal', 'activation': 'leaky relu'}. B
est is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:29,824] Trial 25 finished with value: 41307.75690472844 and para
meters: {'units': 224, 'kernel_initializer': 'normal', 'activation': 'leaky_relu'}.
Best is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:31,799] Trial 26 finished with value: 6426.2445702727155 and par
ameters: { 'units': 131, 'kernel initializer': 'glorot uniform', 'activation': 'leaky
relu'}. Best is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:33,784] Trial 27 finished with value: 43416.01004904071 and para
meters: {'units': 127, 'kernel_initializer': 'normal', 'activation': 'leaky_relu'}.
Best is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:35,717] Trial 28 finished with value: 61678.71735229826 and para
meters: {'units': 101, 'kernel_initializer': 'glorot_uniform', 'activation': 'rel
```

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u'}. Best is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:37,867] Trial 29 finished with value: 14453.937165137655 and par
ameters: {'units': 162, 'kernel initializer': 'normal', 'activation': 'silu'}. Best
is trial 9 with value: 5143.597371840069.
[I 2021-09-28 17:54:40,029] Trial 30 finished with value: 4632.522870892131 and para
meters: {'units': 135, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best i
s trial 30 with value: 4632.522870892131.
[I 2021-09-28 17:54:42,542] Trial 31 finished with value: 10980.277434570655 and par
ameters: {'units': 130, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best
is trial 30 with value: 4632.522870892131.
[I 2021-09-28 17:54:44,574] Trial 32 finished with value: 2928.144715932625 and para
meters: {'units': 75, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best is
trial 32 with value: 2928.144715932625.
[I 2021-09-28 17:54:46,608] Trial 33 finished with value: 6419.799125528503 and para
meters: {'units': 79, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best is
trial 32 with value: 2928.144715932625.
[I 2021-09-28 17:54:48,616] Trial 34 finished with value: 41298.14299345514 and para
meters: {'units': 59, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best is
trial 32 with value: 2928.144715932625.
[I 2021-09-28 17:54:50,682] Trial 35 finished with value: 26940.275950269523 and par
ameters: {'units': 100, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best
is trial 32 with value: 2928.144715932625.
[I 2021-09-28 17:54:52,916] Trial 36 finished with value: 32329.118003190426 and par
ameters: {'units': 189, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best
is trial 32 with value: 2928.144715932625.
[I 2021-09-28 17:54:54,906] Trial 37 finished with value: 2051.6157451483236 and par
ameters: {'units': 52, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best i
s trial 37 with value: 2051.6157451483236.
[I 2021-09-28 17:54:56,873] Trial 38 finished with value: 5329.171856996109 and para
meters: {'units': 46, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best is
trial 37 with value: 2051.6157451483236.
[I 2021-09-28 17:54:58,905] Trial 39 finished with value: 14832.836937347824 and par
ameters: {'units': 83, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best i
s trial 37 with value: 2051.6157451483236.
[I 2021-09-28 17:55:00,956] Trial 40 finished with value: 11058.084726264162 and par
ameters: {'units': 66, 'kernel initializer': 'normal', 'activation': 'silu'}. Best i
s trial 37 with value: 2051.6157451483236.
[I 2021-09-28 17:55:03,018] Trial 41 finished with value: 5299.957124428625 and para
meters: {'units': 45, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best is
trial 37 with value: 2051.6157451483236.
[I 2021-09-28 17:55:05,066] Trial 42 finished with value: 22548.955066453003 and par
ameters: {'units': 45, 'kernel initializer': 'normal', 'activation': 'silu'}. Best i
s trial 37 with value: 2051.6157451483236.
[I 2021-09-28 17:55:07,110] Trial 43 finished with value: 16829.103555201757 and par
ameters: {'units': 59, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best i
s trial 37 with value: 2051.6157451483236.
[I 2021-09-28 17:55:09,142] Trial 44 finished with value: 26058.72848482055 and para
meters: {'units': 77, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best is
```

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trial 37 with value: 2051.6157451483236.
[I 2021-09-28 17:55:11,109] Trial 45 finished with value: 4744.706757013969 and para
meters: {'units': 45, 'kernel initializer': 'normal', 'activation': 'silu'}. Best is
trial 37 with value: 2051.6157451483236.
[I 2021-09-28 17:55:13,124] Trial 46 finished with value: 3023.8700002230544 and par
ameters: {'units': 70, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best i
s trial 37 with value: 2051.6157451483236.
[I 2021-09-28 17:55:15,105] Trial 47 finished with value: 1082.722668145338 and para
meters: {'units': 37, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best is
trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:17,054] Trial 48 finished with value: 2771.9410263718573 and par
ameters: {'units': 36, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best i
s trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:19,426] Trial 49 finished with value: 19559.083900413898 and par
ameters: {'units': 38, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best i
s trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:21,439] Trial 50 finished with value: 3539.899575330315 and para
meters: {'units': 57, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best is
trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:23,383] Trial 51 finished with value: 16250.40572353195 and para
meters: {'units': 32, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best is
trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:25,354] Trial 52 finished with value: 20497.44996769659 and para
meters: {'units': 57, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best is
trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:27,375] Trial 53 finished with value: 15371.60851471764 and para
meters: {'units': 72, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best is
trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:29,379] Trial 54 finished with value: 5020.3144121020405 and par
ameters: {'units': 52, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best i
s trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:31,300] Trial 55 finished with value: 22809.7591546609 and param
eters: {'units': 65, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best is
trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:33,262] Trial 56 finished with value: 13319.44896950012 and para
meters: {'units': 40, 'kernel initializer': 'normal', 'activation': 'silu'}. Best is
trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:35,253] Trial 57 finished with value: 7522.018213229435 and para
meters: {'units': 51, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best is
trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:37,288] Trial 58 finished with value: 5665.0814336141175 and par
ameters: {'units': 63, 'kernel initializer': 'normal', 'activation': 'silu'}. Best i
s trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:39,323] Trial 59 finished with value: 61021.58478819799 and para
meters: {'units': 83, 'kernel_initializer': 'glorot_uniform', 'activation': 'silu'}.
Best is trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:41,238] Trial 60 finished with value: 4092.831026987899 and para
meters: {'units': 32, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best is
```

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trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:43,138] Trial 61 finished with value: 14209.514560250278 and par
ameters: {'units': 32, 'kernel initializer': 'normal', 'activation': 'relu'}. Best i
s trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:45,041] Trial 62 finished with value: 5957.685105244416 and para
meters: {'units': 73, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best is
trial 47 with value: 1082.722668145338.
[I 2021-09-28 17:55:46,964] Trial 63 finished with value: 1044.2942941830224 and par
ameters: {'units': 53, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best i
s trial 63 with value: 1044.2942941830224.
[I 2021-09-28 17:55:48,886] Trial 64 finished with value: 1568.733906626859 and para
meters: {'units': 56, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best is
trial 63 with value: 1044.2942941830224.
[I 2021-09-28 17:55:50,803] Trial 65 finished with value: 32413.610445812927 and par
ameters: {'units': 51, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best i
s trial 63 with value: 1044.2942941830224.
[I 2021-09-28 17:55:52,752] Trial 66 finished with value: 25610.937852538875 and par
ameters: {'units': 90, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best i
s trial 63 with value: 1044.2942941830224.
[I 2021-09-28 17:55:54,639] Trial 67 finished with value: 3410.3496567841744 and par
ameters: {'units': 41, 'kernel_initializer': 'glorot_uniform', 'activation': 'rel
u'}. Best is trial 63 with value: 1044.2942941830224.
[I 2021-09-28 17:55:56,567] Trial 68 finished with value: 4635.826816703268 and para
meters: {'units': 67, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best is
trial 63 with value: 1044.2942941830224.
[I 2021-09-28 17:55:58,499] Trial 69 finished with value: 15543.562669573568 and par
ameters: {'units': 52, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best i
s trial 63 with value: 1044.2942941830224.
[I 2021-09-28 17:56:00,436] Trial 70 finished with value: 1888.8348079191933 and par
ameters: {'units': 75, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best i
s trial 63 with value: 1044.2942941830224.
[I 2021-09-28 17:56:02,820] Trial 71 finished with value: 3448.349923564614 and para
meters: {'units': 63, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best is
trial 63 with value: 1044.2942941830224.
[I 2021-09-28 17:56:04,717] Trial 72 finished with value: 33658.23314232321 and para
meters: {'units': 78, 'kernel initializer': 'normal', 'activation': 'relu'}. Best is
trial 63 with value: 1044.2942941830224.
[I 2021-09-28 17:56:06,636] Trial 73 finished with value: 5696.682805361001 and para
meters: {'units': 87, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best is
trial 63 with value: 1044.2942941830224.
[I 2021-09-28 17:56:08,591] Trial 74 finished with value: 39163.27128353772 and para
meters: {'units': 101, 'kernel initializer': 'normal', 'activation': 'relu'}. Best i
s trial 63 with value: 1044.2942941830224.
[I 2021-09-28 17:56:10,512] Trial 75 finished with value: 24233.888692781005 and par
ameters: {'units': 71, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best i
s trial 63 with value: 1044.2942941830224.
[I 2021-09-28 17:56:12,427] Trial 76 finished with value: 844.9858726442183 and para
meters: {'units': 42, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best is
```

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trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:14,340] Trial 77 finished with value: 1541.7290582079538 and par
ameters: {'units': 41, 'kernel initializer': 'normal', 'activation': 'relu'}. Best i
s trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:16,277] Trial 78 finished with value: 8483.10794473189 and param
eters: {'units': 44, 'kernel_initializer': 'glorot_uniform', 'activation': 'relu'}.
Best is trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:18,271] Trial 79 finished with value: 1540.1521487327038 and par
ameters: {'units': 36, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best i
s trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:20,212] Trial 80 finished with value: 16019.953358552064 and par
ameters: {'units': 49, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best i
s trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:22,122] Trial 81 finished with value: 9527.928814822539 and para
meters: {'units': 39, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best is
trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:24,035] Trial 82 finished with value: 35620.44610788046 and para
meters: {'units': 58, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best is
trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:25,934] Trial 83 finished with value: 21660.368629915865 and par
ameters: {'units': 37, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best i
s trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:27,850] Trial 84 finished with value: 8098.180757526298 and para
meters: {'units': 37, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best is
trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:29,757] Trial 85 finished with value: 5172.574298725549 and para
meters: {'units': 47, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best is
trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:31,691] Trial 86 finished with value: 23359.957063373593 and par
ameters: {'units': 57, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best i
s trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:33,604] Trial 87 finished with value: 16033.852183286865 and par
ameters: {'units': 39, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best i
s trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:35,502] Trial 88 finished with value: 12312.979130937416 and par
ameters: {'units': 44, 'kernel initializer': 'normal', 'activation': 'relu'}. Best i
s trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:37,442] Trial 89 finished with value: 4568.773469425719 and para
meters: {'units': 54, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best is
trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:39,367] Trial 90 finished with value: 23936.22120631501 and para
meters: {'units': 61, 'kernel initializer': 'normal', 'activation': 'relu'}. Best is
trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:41,263] Trial 91 finished with value: 7921.239030455557 and para
meters: {'units': 36, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best is
trial 76 with value: 844.9858726442183.
[I 2021-09-28 17:56:43,687] Trial 92 finished with value: 8220.341535484136 and para
meters: {'units': 47, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best is
```

```
trial 76 with value: 844.9858726442183.
      [I 2021-09-28 17:56:45,584] Trial 93 finished with value: 5019.033403170801 and para
      meters: {'units': 53, 'kernel initializer': 'normal', 'activation': 'relu'}. Best is
      trial 76 with value: 844.9858726442183.
      [I 2021-09-28 17:56:47,556] Trial 94 finished with value: 2125.3077831993783 and par
      ameters: {'units': 43, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best i
      s trial 76 with value: 844.9858726442183.
      [I 2021-09-28 17:56:49,454] Trial 95 finished with value: 17891.298006097823 and par
      ameters: {'units': 43, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best i
      s trial 76 with value: 844.9858726442183.
      [I 2021-09-28 17:56:51,410] Trial 96 finished with value: 9571.66137961033 and param
      eters: {'units': 35, 'kernel_initializer': 'glorot_uniform', 'activation': 'silu'}.
       Best is trial 76 with value: 844.9858726442183.
      [I 2021-09-28 17:56:53,769] Trial 97 finished with value: 30763.805999193533 and par
      ameters: {'units': 225, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best
       is trial 76 with value: 844.9858726442183.
      [I 2021-09-28 17:56:55,723] Trial 98 finished with value: 14620.5028092472 and param
      eters: {'units': 41, 'kernel_initializer': 'normal', 'activation': 'relu'}. Best is
       trial 76 with value: 844.9858726442183.
      [I 2021-09-28 17:56:57,790] Trial 99 finished with value: 20931.296469822406 and par
      ameters: {'units': 49, 'kernel_initializer': 'normal', 'activation': 'silu'}. Best i
      s trial 76 with value: 844.9858726442183.
      Number of finished trials: 100
      Best trial:
        Value: 844.9858726442183
        Params:
         units: 42
         kernel initializer: normal
         activation: relu
In [13]:
       keras_params = keras_study.best_params
       model2 = Sequential()
       model2.add(Dense(**keras_params))
       model2.add(Dense(**keras params))
       model2.add(Dense(**keras params))
       model2.add(Dense(1, activation= keras_params['activation']))
       model2.compile(loss="mean absolute error")
       earlystopping = callbacks.EarlyStopping(monitor ="loss",
                                   mode ="min", patience = 10,
                                   restore_best_weights = True)
       model2.fit(X_train, y_train, validation_data = (X_test, y_test), epochs=100, batch_s
      Epoch 1/100
      3.5171
      Epoch 2/100
      Epoch 3/100
      3.5171
      Epoch 4/100
      3.5171
      Epoch 5/100
```

```
3.5171
 Epoch 6/100
  3.5171
  Epoch 7/100
  Epoch 8/100
  3.5171
  Epoch 9/100
  Epoch 10/100
  3.5171
  Epoch 11/100
  Epoch 12/100
  3.5171
  Epoch 13/100
  Epoch 14/100
  3.5171
  Epoch 15/100
  Epoch 16/100
  3.5171
Out[13]: <keras.callbacks.History at 0x270b0629fd0>
```

In [14]: nr

```
print(model2.get config())
```

{'name': 'sequential_100', 'layers': [{'class_name': 'InputLayer', 'config': {'batch _input_shape': (None, 25), 'dtype': 'float64', 'sparse': False, 'ragged': False, 'na me': 'dense_200_input'}}, {'class_name': 'Dense', 'config': {'name': 'dense_200', 't
rainable': True, 'dtype': 'float32', 'units': 42, 'activation': 'relu', 'use_bias': True, 'kernel_initializer': {'class_name': 'RandomNormal', 'config': {'mean': 0.0, 'stddev': 0.05, 'seed': None}}, 'bias_initializer': {'class_name': 'Zeros', 'confi g': {}}, 'kernel_regularizer': None, 'bias_regularizer': None, 'activity_regularize r': None, 'kernel_constraint': None, 'bias_constraint': None}}, {'class_name': 'Dens e', 'config': {'name': 'dense_201', 'trainable': True, 'dtype': 'float32', 'units': 42, 'activation': 'relu', 'use_bias': True, 'kernel_initializer': {'class_name': 'Ra ndomNormal', 'config': {'mean': 0.0, 'stddev': 0.05, 'seed': None}}, 'bias_initializ er': {'class_name': 'Zeros', 'config': {}}, 'kernel_regularizer': None, 'bias_regula rizer': None, 'activity_regularizer': None, 'kernel_constraint': None, 'bias_constraint': None}}, {'class_name': 'Dense', 'config': {'name': 'dense_202', 'trainable': True, 'dtype': 'float32', 'units': 42, 'activation': 'relu', 'use_bias': True, 'kerne l_initializer': {'class_name': 'RandomNormal', 'config': {'mean': 0.0, 'stddev': 0.0 5, 'seed': None}}, 'bias_initializer': {'class_name': 'Zeros', 'config': {}}, 'kerne l_regularizer': None, 'bias_regularizer': None, 'activity_regularizer': None, 'kerne l_constraint': None, 'bias_constraint': None}}, {'class_name': 'Dense', 'config': {'name': 'dense_203', 'trainable': True, 'dtype': 'float32', 'units': 1, 'activatio n': 'relu', 'use_bias': True, 'kernel_initializer': {'class_name': 'GlorotUniform', 'config': {'seed': None}}, 'bias_initializer': {'class_name': 'Zeros', 'config': {}}, 'kernel_regularizer': None, 'bias_regularizer': None, 'activity_regularizer': N one, 'kernel_constraint': None, 'bias_constraint': None}}]}

BEFORE PREDICTION INVERSE LOG10

```
In [15]: train_df['price'] = 10 ** train_df['price']
```

```
test_df['price'] = 10 ** test_df['price']
In [16]:
         train_df['price']
Out[16]: 4
                   2400.0
                   3800.0
                   3495.0
         10
                   3000.0
         15
                   2795.0
                   2800.0
         124000
         124002
                  2395.0
         124004
                  1850.0
         124008
                  4195.0
                 4280.0
         124009
         Name: price, Length: 49352, dtype: float64
In [17]:
         y_test = 10 ** y_test
In [18]:
          score = model2.evaluate(X_test, y_test, verbose=0)
          print(score)
```

3963.275390625