

In [93]:

```
import pandas as pd
```

In [94]:

```
import numpy as np
```

In [95]:

```
import matplotlib.pyplot as plt
```

In [96]:

```
import datetime
```

In [97]:

```
from sklearn.tree import DecisionTreeRegressor
```

In [98]:

```
from sklearn.metrics import mean_absolute_error
```

In [99]:

```
from sklearn.model_selection import train_test_split
```

In [100]:

```
from sklearn.ensemble import RandomForestRegressor
```

In [101]:

```
from sklearn.impute import SimpleImputer
```

In [102]:

```
from sklearn.model_selection import cross_val_score
```

In [103]:

```
pd.set_option('display.max_columns',100)
```

In [104]:

```
home_data = pd.read_csv('/Users/kabloom/Documents/docs/Data Science/Assignments  
submitted/train.csv')
```

In [105]:

```
home_data.head()
```

Out[105]:

	Id	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	LandContour
0	1	60	RL	65.0	8450	Pave	NaN	Reg	Lvl
1	2	20	RL	80.0	9600	Pave	NaN	Reg	Lvl
2	3	60	RL	68.0	11250	Pave	NaN	IR1	Lvl
3	4	70	RL	60.0	9550	Pave	NaN	IR1	Lvl
4	5	60	RL	84.0	14260	Pave	NaN	IR1	Lvl

In [108]:

```
round(home_data['LotArea'].mean())
```

Out[108]:

10517

In [109]:

```
datetime.datetime.now().year-home_data['YearBuilt'].max()
```

Out[109]:

9

In [110]:

```
home_data['YearBuilt'].min()
```

Out[110]:

1872

In [111]:

```
home_data['YearRemodAdd'].min()
```

Out[111]:

1950

In [112]:

```
home_data['YearRemodAdd'].max()
```

Out[112]:

2010

In [113]:

```
home_data['YrSold'].min()
```

Out[113]:

2006

In [114]:

```
home_data['YrSold'].max()
```

Out[114]:

2010

In [115]:

```
home_data.YrSold.value_counts()
```

Out[115]:

2009	338
2007	329
2006	314
2008	304
2010	175

Name: YrSold, dtype: int64

In [116]:

```
home_data.groupby('YrSold').MoSold.value_counts().head()
```

Out[116]:

YrSold	MoSold	
2006	7	67
	6	48
	5	38
	4	27
	3	25

Name: MoSold, dtype: int64

In [117]:

```
Y=home_data.SalePrice
```

In [118]:

```
Y.head()
```

Out[118]:

0	208500
1	181500
2	223500
3	140000
4	250000

Name: SalePrice, dtype: int64

In [119]:

```
home_features=[ 'LotArea', 'YearBuilt', '1stFlrSF', '2ndFlrSF', 'FullBath', 'BedroomAbvGr', 'TotRmsAbvGrd' ]
```

In [120]:

```
X=home_data[home_features]
```

In [121]:

```
X.head()
```

Out[121]:

	LotArea	YearBuilt	1stFlrSF	2ndFlrSF	FullBath	BedroomAbvGr	TotRmsAbvGrd
0	8450	2003	856	854	2	3	8
1	9600	1976	1262	0	2	3	6
2	11250	2001	920	866	2	3	6
3	9550	1915	961	756	1	3	7
4	14260	2000	1145	1053	2	4	9

In [122]:

```
X.describe()
```

Out[122]:

	LotArea	YearBuilt	1stFlrSF	2ndFlrSF	FullBath	BedroomAbvGr	TotRmsAbvGrd
count	1460.000000	1460.000000	1460.000000	1460.000000	1460.000000	1460.000000	1460.000000
mean	10516.828082	1971.267808	1162.626712	346.992466	1.565068	2.866438	7.000000
std	9981.264932	30.202904	386.587738	436.528436	0.550916	0.815778	1.463319
min	1300.000000	1872.000000	334.000000	0.000000	0.000000	0.000000	4.000000
25%	7553.500000	1954.000000	882.000000	0.000000	1.000000	2.000000	5.000000
50%	9478.500000	1973.000000	1087.000000	0.000000	2.000000	3.000000	6.000000
75%	11601.500000	2000.000000	1391.250000	728.000000	2.000000	3.000000	7.000000
max	215245.000000	2010.000000	4692.000000	2065.000000	3.000000	8.000000	14.000000

In [123]:

```
#Applied Decision Tree Regressor
home_data_model=DecisionTreeRegressor(random_state=1)
train_X,val_X,train_Y,val_Y=train_test_split(X,Y,random_state=1)
home_data_model.fit(train_X,train_Y)
```

Out[123]:

```
DecisionTreeRegressor(criterion='mse', max_depth=None, max_features=
None,
                      max_leaf_nodes=None, min_impurity_decrease=0.0,
                      min_impurity_split=None, min_samples_leaf=1,
                      min_samples_split=2, min_weight_fraction_leaf=0.0,
                      presort=False, random_state=1, splitter='best')
```

In [124]:

```
def get_mae(max_leaf_nodes, train_X, val_X, train_Y, val_Y):
    model=DecisionTreeRegressor(max_leaf_nodes=max_leaf_nodes,random_state=0)
    model.fit(train_X,train_Y)
    preds_val=model.predict(val_X)
    mae=mean_absolute_error(val_Y,preds_val)
    return(mae)
```

In [125]:

```
#Checking error value by tweaking
best_tree_size=0
mean_abs_error=1000000.00
for max_leaf_nodes in [5,50,500,5000]:
    my_mae=get_mae(max_leaf_nodes,train_x,val_x,train_y,val_y)
    print("Max leaf Nodes : %d \t\t Mean Absolute Error : %d" %(max_leaf_nodes,m
y_mae))
    if(mean_abs_error>my_mae):
        mean_abs_error=my_mae
        best_tree_size=max_leaf_nodes
print(best_tree_size)
```

```
-----
-----
NameError                                Traceback (most recent cal
l last)
<ipython-input-125-6736b31caefa> in <module>
      3 mean_abs_error=1000000.00
      4 for max_leaf_nodes in [5,50,500,5000]:
----> 5     my_mae=get_mae(max_leaf_nodes,train_x,val_x,train_y,val_
y)
      6     print("Max leaf Nodes : %d \t\t Mean Absolute Error : %
d" %(max_leaf_nodes,my_mae))
      7     if(mean_abs_error>my_mae):

NameError: name 'train_x' is not defined
```

In [126]:

```
model=DecisionTreeRegressor(max_leaf_nodes=500,random_state=0)
model.fit(train_x,train_y)
preds_val=model.predict(val_x)
mae=mean_absolute_error(val_y,preds_val)
mae
```

```
-----
-----
NameError                                Traceback (most recent call
last)
<ipython-input-126-7aac52cdaf7e> in <module>
      1 model=DecisionTreeRegressor(max_leaf_nodes=500,random_state=
0)
----> 2 model.fit(train_x,train_y)
      3 preds_val=model.predict(val_x)
      4 mae=mean_absolute_error(val_y,preds_val)
      5 mae
```

NameError: name 'train_x' is not defined

In [127]:

```
#Creating Random Forest Model with full training data
full_training_data = pd.read_csv('/Users/kabloom/Documents/docs/Data Science/Ass
ignments submitted/train.csv')
```

In [128]:

```
y_full=full_training_data.SalePrice
```

In [129]:

```
home_features_full=['LotArea','YearBuilt','1stFlrSF','2ndFlrSF','FullBath','Bedr
oomAbvGr','TotRmsAbvGrd']
```

In [130]:

```
x_full=full_training_data[home_features_full]
```

In [131]:

```
f_model=RandomForestRegressor(random_state=1)
f_model.fit(x_full,y_full)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/ensemble/forest.py:24
6: FutureWarning: The default value of n_estimators will change from
10 in version 0.20 to 100 in 0.22.
  "10 in version 0.20 to 100 in 0.22.", FutureWarning)
```

Out[131]:

```
RandomForestRegressor(bootstrap=True, criterion='mse', max_depth=None,
max_features='auto', max_leaf_nodes=None,
min_impurity_decrease=0.0, min_impurity_split=None,
min_samples_leaf=1, min_samples_split=2,
min_weight_fraction_leaf=0.0, n_estimators=10, n_jobs=None,
oob_score=False, random_state=1, verbose=0, warm_start=False)
```

In [132]:

```
full_test_data = pd.read_csv('/Users/kabloom/Documents/docs/Data Science/Assignments submitted/test.csv')
home_features_test=['LotArea', 'YearBuilt', '1stFlrSF', '2ndFlrSF', 'FullBath', 'BedroomAbvGr', 'TotRmsAbvGrd']
x_test=full_test_data[home_features_test]
pred_test=f_model.predict(x_test)
pred_test
```

Out[132]:

```
array([112945. , 149770. , 178100. , ..., 141674.1, 130350. , 228830. ])
```

In [133]:

```
output = pd.DataFrame({'Id':full_test_data.Id, 'SalePrice':pred_test})
```

In [134]:

```
output.to_csv('tripathishivam35HomeData.csv', index=False)
```

In [135]:

```
missing_val_count_by_column=(full_test_data.isnull().sum())
#print(missing_val_count_by_column[missing_val_count_by_column > 0])
```

In [136]:

```
#Get Model Score from Dropping Columns with Missing values
cols_with_missing = [col for col in X_train.columns
                      if X_train[col].isnull().any()]
reduced_X_train = X_train.drop(cols_with_missing, axis=1)
reduced_X_test  = X_test.drop(cols_with_missing, axis=1)
print("Mean Absolute Error from dropping columns with Missing Values:")
print(score_dataset(reduced_X_train, reduced_X_test, y_train, y_test))
```

```
-----
-----
NameError                                Traceback (most recent call
1 last)
<ipython-input-136-f7d17914449f> in <module>
      1 #Get Model Score from Dropping Columns with Missing values
----> 2 cols_with_missing = [col for col in X_train.columns
      3                      if X_train[col].isnull().an
y()]
      4 reduced_X_train = X_train.drop(cols_with_missing, axis=1)
      5 reduced_X_test  = X_test.drop(cols_with_missing, axis=1)

NameError: name 'X_train' is not defined
```

In [137]:

```
#Get Model Score from Imputation
my_imputer = SimpleImputer()
imputed_X_train = my_imputer.fit_transform(X_train)
imputed_X_test = my_imputer.transform(X_test)
print("Mean Absolute Error from Imputation:")
print(score_dataset(imputed_X_train, imputed_X_test, y_train, y_test))
```

```
-----
-----
NameError                                Traceback (most recent call
1 last)
<ipython-input-137-1ee8a805eb5c> in <module>
      1 #Get Model Score from Imputation
      2 my_imputer = SimpleImputer()
----> 3 imputed_X_train = my_imputer.fit_transform(X_train)
      4 imputed_X_test = my_imputer.transform(X_test)
      5 print("Mean Absolute Error from Imputation:")

NameError: name 'X_train' is not defined
```


In [138]:

```
#Get Score from imputation with extra columns showing what was imputed
imputed_X_train_plus = X_train.copy()
imputed_X_test_plus = X_test.copy()

cols_with_missing = (col for col in X_train.columns
                      if X_train[col].isnull().any())
for col in cols_with_missing:
    imputed_X_train_plus[col + '_was_missing'] = imputed_X_train_plus[col].isnull()
    imputed_X_test_plus[col + '_was_missing'] = imputed_X_test_plus[col].isnull()

# Imputation
my_imputer = SimpleImputer()
imputed_X_train_plus = my_imputer.fit_transform(imputed_X_train_plus)
imputed_X_test_plus = my_imputer.transform(imputed_X_test_plus)

print("Mean Absolute Error from Imputation while Track What Was Imputed:")
print(score_dataset(imputed_X_train_plus, imputed_X_test_plus, y_train, y_test))
```

```
-----
-----
NameError                                Traceback (most recent call
last)
<ipython-input-138-8053f0f85c9d> in <module>
      1 #Get Score from imputation with extra columns showing what w
as imputed
----> 2 imputed_X_train_plus = X_train.copy()
      3 imputed_X_test_plus = X_test.copy()
      4
      5 cols_with_missing = (col for col in X_train.columns

NameError: name 'X_train' is not defined
```

In [139]:

```

train_data = pd.read_csv('/Users/kabloom/Documents/docs/Data Science/Assignments
submitted/train.csv')
test_data = pd.read_csv('/Users/kabloom/Documents/docs/Data Science/Assignments
submitted/test.csv')

# Drop houses where the target is missing
train_data.dropna(axis=0, subset=['SalePrice'], inplace=True)

target = train_data.SalePrice

# Since missing values isn't the focus of this tutorial, we use the simplest
# possible approach, which drops these columns.
# For more detail (and a better approach) to missing values, see
# https://www.kaggle.com/dansbecker/handling-missing-values
cols_with_missing = [col for col in train_data.columns
                      if train_data[col].isnull().any()]
candidate_train_predictors = train_data.drop(['Id', 'SalePrice'] + cols_with_mis
sing, axis=1)
candidate_test_predictors = test_data.drop(['Id'] + cols_with_missing, axis=1)

# "cardinality" means the number of unique values in a column.
# We use it as our only way to select categorical columns here. This is convenie
nt, though
# a little arbitrary.
low_cardinality_cols = [cname for cname in candidate_train_predictors.columns if
                        candidate_train_predictors[cname].nunique() < 10
and
                        candidate_train_predictors[cname].dtype == "obje
ct"]
numeric_cols = [cname for cname in candidate_train_predictors.columns if
                 candidate_train_predictors[cname].dtype in ['int
64', 'float64']]
my_cols = low_cardinality_cols + numeric_cols
train_predictors = candidate_train_predictors[my_cols]
test_predictors = candidate_test_predictors[my_cols]

```

In [140]:

```

train_predictors.dtypes.sample(10)
one_hot_encoded_training_predictors = pd.get_dummies(train_predictors)
one_hot_encoded_training_predictors.head()

```

Out[140]:

	MSSubClass	LotArea	OverallQual	OverallCond	YearBuilt	YearRemodAdd	BsmtFinSF1	B
0	60	8450	7	5	2003	2003	706	
1	20	9600	6	8	1976	1976	978	
2	60	11250	7	5	2001	2002	486	
3	70	9550	7	5	1915	1970	216	
4	60	14260	8	5	2000	2000	655	

5 rows × 159 columns

In [141]:

```

from sklearn.model_selection import cross_val_score
from sklearn.ensemble import RandomForestRegressor

def get_mae(X, y):
    # multiple by -1 to make positive MAE score instead of neg value returned as
    # sklearn convention
    return -1 * cross_val_score(RandomForestRegressor(50),
                                X, y,
                                scoring = 'neg_mean_absolute_error').mean()

predictors_without_categoricals = train_predictors.select_dtypes(exclude=['object'])

mae_without_categoricals = get_mae(predictors_without_categoricals, target)

mae_one_hot_encoded = get_mae(one_hot_encoded_training_predictors, target)

print('Mean Absolute Error when Dropping Categoricals: ' + str(int(mae_without_categoricals)))
print('Mean Absolute Error with One-Hot Encoding: ' + str(int(mae_one_hot_encoded)))

```

```

/anaconda3/lib/python3.7/site-packages/sklearn/model_selection/_split.py:2053: FutureWarning: You should specify a value for 'cv' instead of relying on the default value. The default value will change from 3 to 5 in version 0.22.

```

```

warnings.warn(CV_WARNING, FutureWarning)

```

```

/anaconda3/lib/python3.7/site-packages/sklearn/model_selection/_split.py:2053: FutureWarning: You should specify a value for 'cv' instead of relying on the default value. The default value will change from 3 to 5 in version 0.22.

```

```

warnings.warn(CV_WARNING, FutureWarning)

```

```

Mean Absolute Error when Dropping Categoricals: 18631

```

```

Mean Absolute Error with One-Hot Encoding: 18361

```

In [142]:

```

#Applying Gradient Boosting Regressor
bstt = ensm.GradientBoostingRegressor
(loss='ls', criterion='mse', max_depth=4, random_state=0, n_estimators=100, learning_rate=0.1)

```

```

File "<ipython-input-142-9a11800276d3>", line 3
    (loss='ls', criterion='mse', max_depth=4, random_state=0, n_estimators=100, learning_rate=0.1)
    ^

```

```

SyntaxError: invalid syntax

```

In []:

```

# New Tutorail Pandas Housing Data

```

In [150]:

```
train=pd.read_csv('/Users/kabloom/Documents/docs/Data Science/Assignments submitted/train.csv',na_values='#NAME?')
```

In [151]:

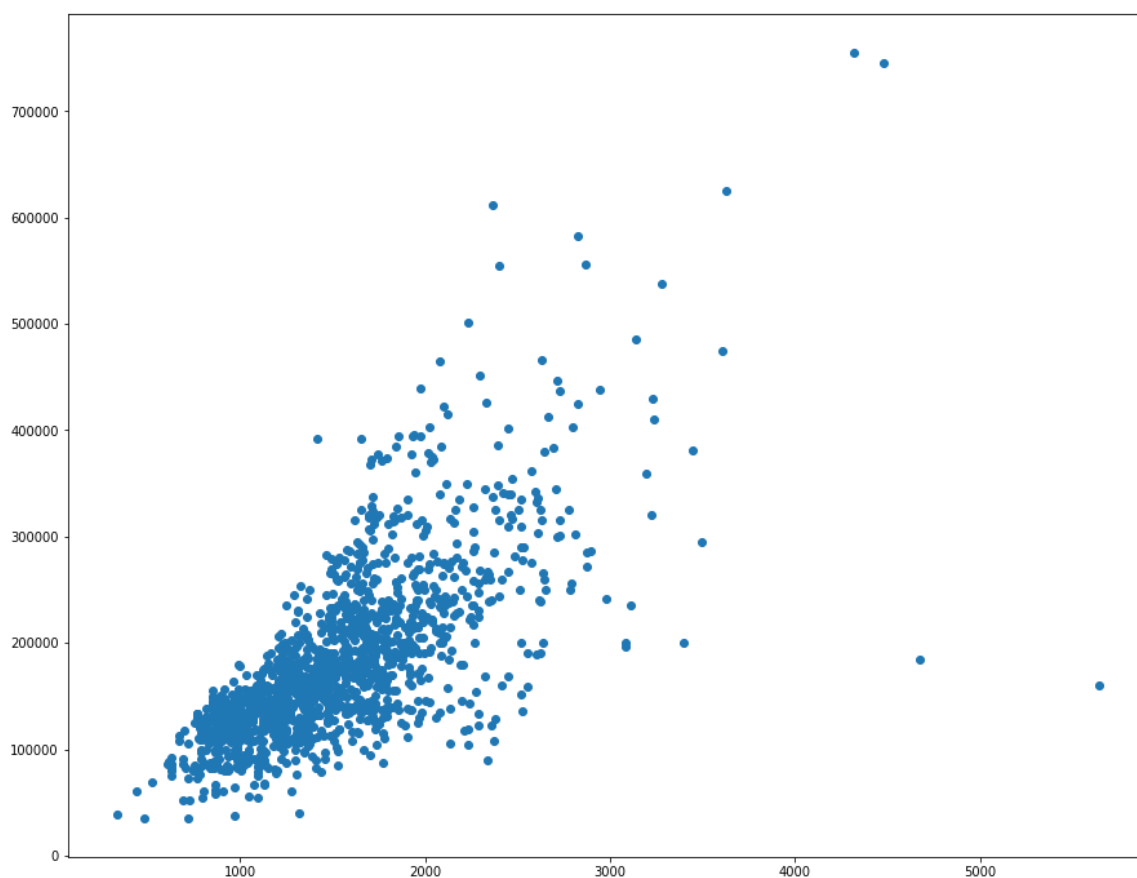
```
test=pd.read_csv('/Users/kabloom/Documents/docs/Data Science/Assignments submitted/test.csv',na_values='#NAME?')
```

In [152]:

```
plt.figure(figsize=(15,12))  
plt.scatter(x=train.GrLivArea,y=train.SalePrice)
```

Out[152]:

<matplotlib.collections.PathCollection at 0x1a1f482b38>



In [153]:

```
train.drop(train[(train['GrLivArea'] >4000) & (train['SalePrice']>600000)].index,  
inplace=True)
```

In [154]:

```
train.shape,test.shape
```

Out[154]:

```
((1458, 81), (1459, 80))
```

In [155]:

```
full=pd.concat([train,test],ignore_index=True,sort=False)
full.drop('Id',axis=1,inplace=True)
full.shape
```

Out[155]:

(2917, 80)

In [156]:

```
missing_values=full.isnull().sum()
```

In [157]:

```
missing_values[missing_values>0].sort_values(ascending=False)
```

Out[157]:

PoolQC	2908
MiscFeature	2812
Alley	2719
Fence	2347
SalePrice	1459
FireplaceQu	1420
LotFrontage	486
GarageYrBlt	159
GarageFinish	159
GarageQual	159
GarageCond	159
GarageType	157
BsmtCond	82
BsmtExposure	82
BsmtQual	81
BsmtFinType2	80
BsmtFinType1	79
MasVnrType	24
MasVnrArea	23
MSZoning	4
BsmtFullBath	2
BsmtHalfBath	2
Functional	2
Utilities	2
BsmtFinSF2	1
BsmtUnfSF	1
BsmtFinSF1	1
TotalBsmtSF	1
SaleType	1
KitchenQual	1
Exterior2nd	1
Exterior1st	1
GarageCars	1
GarageArea	1
Electrical	1

dtype: int64

In [158]:

```
full.groupby(['Neighborhood'])[['LotFrontage']].agg(['mean', 'median', 'count'])
```

Out[158]:

	LotFrontage		
	mean	median	count
Neighborhood			
Blmngtn	46.900000	43.0	20
Blueste	27.300000	24.0	10
BrDale	21.500000	21.0	30
BrkSide	55.789474	51.0	95
ClearCr	88.150000	80.5	20
CollgCr	71.336364	70.0	220
Crawfor	69.951807	70.0	83
Edwards	66.910112	65.0	178
Gilbert	74.207207	64.0	111
IDOTRR	62.241379	60.0	87
MeadowV	25.606061	21.0	33
Mitchel	75.144444	74.0	90
NAmes	75.210667	73.0	375
NPkVill	28.142857	24.0	21
NWAmes	81.517647	80.0	85
NoRidge	90.076923	88.5	52
NridgHt	84.184049	92.0	163
OldTown	61.777293	60.0	229
SWISU	59.068182	60.0	44
Sawyer	74.551020	72.0	98
SawyerW	70.669811	67.0	106
Somerst	64.549383	72.5	162
StoneBr	62.173913	60.0	46
Timber	81.157895	82.0	57
Veenker	72.000000	80.0	16

In [159]:

```
full['LotAreaCut']=pd.qcut(full.LotArea,10)
full.groupby([full['LotAreaCut']])[['LotFrontage']].agg(['mean','median','count'])
```

Out[159]:

	LotFrontage		
	mean	median	count
LotAreaCut			
(1299.999, 4921.8]	35.741036	34.0	251
(4921.8, 7007.2]	55.460674	52.0	267
(7007.2, 7949.0]	62.959839	62.0	249
(7949.0, 8740.4]	67.113725	65.0	255
(8740.4, 9452.0]	69.959184	70.0	245
(9452.0, 10148.8]	73.988235	75.0	255
(10148.8, 11000.0]	73.636364	75.0	253
(11000.0, 12196.8]	83.371681	82.0	226
(12196.8, 14285.8]	84.973684	85.0	228
(14285.8, 215245.0]	93.732673	90.0	202

In [160]:

```
full['LotFrontage']= full.groupby(['LotAreaCut','Neighborhood'])['LotFrontage'].transform(lambda x : x.fillna(x.median()))
full['LotFrontage']= full.groupby(['LotAreaCut'])['LotFrontage'].transform(lambda x : x.fillna(x.median()))
```

In [161]:

```
missing_values = full.isnull().sum()

missing_values[missing_values>0].sort_values(ascending = False)
```

Out[161]:

```
PoolQC          2908
MiscFeature     2812
Alley           2719
Fence           2347
SalePrice       1459
FireplaceQu     1420
GarageYrBlt     159
GarageFinish    159
GarageQual      159
GarageCond      159
GarageType      157
BsmtCond        82
BsmtExposure    82
BsmtQual        81
BsmtFinType2    80
BsmtFinType1    79
MasVnrType      24
MasVnrArea      23
MSZoning         4
BsmtFullBath     2
BsmtHalfBath     2
Functional       2
Utilities        2
BsmtUnfSF        1
BsmtFinSF2       1
TotalBsmtSF      1
BsmtFinSF1       1
SaleType         1
KitchenQual      1
GarageCars       1
Exterior2nd      1
Exterior1st      1
GarageArea       1
Electrical       1
dtype: int64
```

In [162]:

```
columns = ["MasVnrArea", "BsmtUnfSF", "TotalBsmtSF", "GarageCars", "BsmtFinSF2",
           "BsmtFinSF1", "GarageArea"]
for col in columns:
    full[col].fillna(0,inplace= True)
```

In [163]:

```
columns1 = ["PoolQC", "MiscFeature", "Alley", "Fence", "FireplaceQu", "GarageQual",
            "GarageCond", "GarageFinish",
            "GarageYrBlt", "GarageType", "BsmtExposure", "BsmtCond", "BsmtQual", "BsmtFinType2",
            "BsmtFinType1", "MasVnrType"]
for col1 in columns1:
    full[col1].fillna('None',inplace = True)
```


In [164]:

```
columns2 = ["MSZoning", "BsmtFullBath", "BsmtHalfBath", "Utilities", "Functiona
1",
            "Electrical", "KitchenQual", "SaleType", "Exterior1st", "Exterior2nd"
]

for col2 in columns2:
    full[col2].fillna(full[col2].mode()[0], inplace = True)
```

In [165]:

```
full.isnull().sum()[full.isnull().sum()>0]
```

Out[165]:

```
SalePrice      1459
dtype: int64
```

In [166]:

```
numeric_features = full.select_dtypes(include=[np.number])
numeric_features.columns
```

Out[166]:

```
Index(['MSSubClass', 'LotFrontage', 'LotArea', 'OverallQual', 'Overa
llCond',
      'YearBuilt', 'YearRemodAdd', 'MasVnrArea', 'BsmtFinSF1', 'Bsm
tFinSF2',
      'BsmtUnfSF', 'TotalBsmtSF', '1stFlrSF', '2ndFlrSF', 'LowQualFinSF',
      'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath', 'Hal
fBath',
      'BedroomAbvGr', 'KitchenAbvGr', 'TotRmsAbvGrd', 'Fireplaces',
      'GarageCars', 'GarageArea', 'WoodDeckSF', 'OpenPorchSF',
      'EnclosedPorch', '3SsnPorch', 'ScreenPorch', 'PoolArea', 'Mis
cVal',
      'MoSold', 'YrSold', 'SalePrice'],
      dtype='object')
```

In [167]:

```
Numstr = ["MSSubClass", "BsmtFullBath", "BsmtHalfBath", "HalfBath", "BedroomAbvGr",
"KitchenAbvGr", "MoSold",
          "YrSold", "YearBuilt", "YearRemodAdd", "LowQualFinSF", "GarageYrBlt"]

for i in Numstr:
    full[i]=full[i].astype(str)
```

In [168]:

```
full.groupby(['MSSubClass'])[['SalePrice']].agg(['mean', 'median', 'count'])
```

Out[168]:

MSSubClass	SalePrice		
	mean	median	count
120	200779.080460	192000.0	87
150	NaN	NaN	0
160	138647.380952	146000.0	63
180	102300.000000	88500.0	10
190	129613.333333	128250.0	30
20	185224.811567	159250.0	536
30	95829.724638	99900.0	69
40	156125.000000	142500.0	4
45	108591.666667	107500.0	12
50	143302.972222	132000.0	144
60	236513.811448	215000.0	297
70	166772.416667	156000.0	60
75	192437.500000	163500.0	16
80	169736.551724	166500.0	58
85	147810.000000	140750.0	20
90	133541.076923	135980.0	52

In [169]:

```

def map_values():
    full["oMSSubClass"] = full.MSSubClass.map({'180':1,
                                                '30':2, '45':2,
                                                '190':3, '50':3, '90':3,
                                                '85':4, '40':4, '160':4,
                                                '70':5, '20':5, '75':5, '80':5, '150':5,
                                                '120': 6, '60':6})

    full["oMSZoning"] = full.MSZoning.map({'C (all)':1, 'RH':2, 'RM':2, 'RL':3,
    'FV':4})
    full["oNeighborhood"] = full.Neighborhood.map({'MeadowV':1,
                                                    'IDOTRR':2, 'BrDale':2,
                                                    'OldTown':3, 'Edwards':3, 'BrkSid
e':3,
                                                    'Sawyer':4, 'Blueste':4, 'SWISU':
4, 'Names':4,
                                                    'NPkVill':5, 'Mitchel':5,
                                                    'SawyerW':6, 'Gilbert':6, 'NWame
s':6,
                                                    'Blmngtn':7, 'CollgCr':7, 'ClearC
r':7, 'Crawfor':7,
                                                    'Veenker':8, 'Somerst':8, 'Timbe
r':8,
                                                    'StoneBr':9,
                                                    'NoRidge':10, 'NridgHt':10})

    full["oCondition1"] = full.Condition1.map({'Artery':1,
                                                'Feedr':2, 'RRAe':2,
                                                'Norm':3, 'RRAn':3,
                                                'PosN':4, 'RRNe':4,
                                                'PosA':5 , 'RRNn':5})

    full["oBldgType"] = full.BldgType.map({'2fmCon':1, 'Duplex':1, 'Twnhs':1, '1
Fam':2, 'TwnhsE':2})

    full["oHouseStyle"] = full.HouseStyle.map({'1.5Unf':1,
                                                '1.5Fin':2, '2.5Unf':2, 'SFoyer':2,
                                                '1Story':3, 'SLvl':3,
                                                '2Story':4, '2.5Fin':4})

    full["oExterior1st"] = full.Exterior1st.map({'BrkComm':1,
                                                  'AsphShn':2, 'CBlock':2, 'AsbShng':
2,
                                                  'WdShing':3, 'Wd Sdng':3, 'MetalSd'
:3, 'Stucco':3, 'HdBoard':3,
                                                  'BrkFace':4, 'Plywood':4,
                                                  'VinylSd':5,
                                                  'CemntBd':6,
                                                  'Stone':7, 'ImStucc':7})

    full["oMasVnrType"] = full.MasVnrType.map({'BrkCmn':1, 'None':1, 'BrkFace':2
, 'Stone':3})

    full["oExterQual"] = full.ExterQual.map({'Fa':1, 'TA':2, 'Gd':3, 'Ex':4})

    full["oFoundation"] = full.Foundation.map({'Slab':1,
                                                'BrkTil':2, 'CBlock':2, 'Stone':2,
                                                'Wood':3, 'PConc':4})

```

```

full["oBsmtQual"] = full.BsmtQual.map({'Fa':2, 'None':1, 'TA':3, 'Gd':4, 'Ex':5})

full["oBsmtExposure"] = full.BsmtExposure.map({'None':1, 'No':2, 'Av':3, 'Min':3, 'Gd':4})

full["oHeating"] = full.Heating.map({'Floor':1, 'Grav':1, 'Wall':2, 'OthW':3, 'GasW':4, 'GasA':5})

full["oHeatingQC"] = full.HeatingQC.map({'Po':1, 'Fa':2, 'TA':3, 'Gd':4, 'Ex':5})

full["oKitchenQual"] = full.KitchenQual.map({'Fa':1, 'TA':2, 'Gd':3, 'Ex':4})

full["oFunctional"] = full.Functional.map({'Maj2':1, 'Maj1':2, 'Min1':2, 'Min2':2, 'Mod':2, 'Sev':2, 'Typ':3})

full["oFireplaceQu"] = full.FireplaceQu.map({'None':1, 'Po':1, 'Fa':2, 'TA':3, 'Gd':4, 'Ex':5})

full["oGarageType"] = full.GarageType.map({'CarPort':1, 'None':1, 'Detchd':2, '2Types':3, 'Basement':3, 'Attchd':4, 'BuiltIn':5})

full["oGarageFinish"] = full.GarageFinish.map({'None':1, 'Unf':2, 'RFn':3, 'Fin':4})

full["oPavedDrive"] = full.PavedDrive.map({'N':1, 'P':2, 'Y':3})

full["oSaleType"] = full.SaleType.map({'COD':1, 'ConLD':1, 'ConLI':1, 'ConLW':1, 'Oth':1, 'WD':1, 'CWD':2, 'Con':3, 'New':3})

full["oSaleCondition"] = full.SaleCondition.map({'AdjLand':1, 'Abnorml':2, 'Alloca':2, 'Family':2, 'Normal':3, 'Partial':4})

return "Done!"

```

In [170]:

```
map_values()
```

Out[170]:

```
'Done!'
```

In [171]:

```

full.drop("LotAreaCut",axis=1,inplace=True)

full.drop(['SalePrice'],axis=1,inplace=True)

```

In [172]:

```
full.shape
```

Out[172]:

```
(2917, 101)
```

In [173]:

```
full[['YearBuilt', 'YearRemodAdd', 'GarageYrBlt']].head()
```

Out[173]:

	YearBuilt	YearRemodAdd	GarageYrBlt
0	2003	2003	2003.0
1	1976	1976	1976.0
2	2001	2002	2001.0
3	1915	1970	1998.0
4	2000	2000	2000.0

In [174]:

```
from sklearn.base import BaseEstimator, clone, TransformerMixin, RegressorMixin
```

In [175]:

```
class labenc(BaseEstimator, TransformerMixin):
    def __init__(self):
        pass
    def fit(self, X, y=None):
        return self
    def transform(self, X):
        label = LabelEncoder()
        X['YearBuilt'] = label.fit_transform(X['YearBuilt'])
        X['YearRemodAdd'] = label.fit_transform(X['YearRemodAdd'])
        X['GarageYrBlt'] = label.fit_transform(X['GarageYrBlt'])
        return X
```

In [176]:

```
class skewness(BaseEstimator, TransformerMixin):
    def __init__(self, skew=0.5):
        self.skew = skew
    def fit(self, X, y=None):
        return self
    def transform(self, X):
        X_numeric = X.select_dtypes(exclude=["object"])
        skewness = X_numeric.apply(lambda x: skew(x))
        skewness_features = skewness[abs(skewness) >= self.skew].index
        X[skewness_features] = np.log1p(X[skewness_features])
        return X
```

In [177]:

```
from sklearn.pipeline import make_pipeline, Pipeline
```

In [178]:

```
class dummies(BaseEstimator,TransformerMixin):
    def __init__(self):
        pass
    def fit(self,X,y=None):
        return self
    def transform(self,X):
        X = pd.get_dummies(X)
        return X
```

In [179]:

```
pipeline = Pipeline([('labenc',labenc()),('skewness',skewness(skew =1)),('dummies',dummies())])
```

In [180]:

```
from sklearn.preprocessing import LabelEncoder,Imputer,OneHotEncoder,RobustScaler,StandardScaler,Imputer
from scipy.stats import skew
```

In [181]:

```
full_copy = full.copy()
data_pipeline = pipeline.fit_transform(full_copy)
```

In [182]:

```
data_pipeline.shape
```

Out[182]:

```
(2917, 406)
```

In [183]:

```
data_pipeline.head()
```

Out[183]:

	LotFrontage	LotArea	OverallQual	OverallCond	YearBuilt	YearRemodAdd	MasVnrArea	...
0	4.189655	9.042040	7	5	110	53	5.283204	...
1	4.394449	9.169623	6	8	83	26	0.000000	...
2	4.234107	9.328212	7	5	108	52	5.093750	...
3	4.110874	9.164401	7	5	25	20	0.000000	...
4	4.442651	9.565284	8	5	107	50	5.860786	...

5 rows × 406 columns

In [184]:

```
robust_scaler = RobustScaler()
```

In [185]:

```
n_train = train.shape[0]
n_train
```

Out[185]:

1458

In [186]:

```
X= data_pipeline[:n_train]
y = train.SalePrice
test_X = data_pipeline[n_train:]
X.shape,y.shape,test_X.shape
```

Out[186]:

((1458, 406), (1458,), (1459, 406))

In [187]:

```
X_scaled = robust_scaler.fit(X).transform(X)
y_log = np.log(train.SalePrice)
test_X_scaled = robust_scaler.transform(test_X)
```

In [188]:

```
X_scaled.shape,y_log.shape,test_X.shape
```

Out[188]:

((1458, 406), (1458,), (1459, 406))

In [189]:

```
from sklearn.linear_model import LinearRegression,BayesianRidge,ElasticNet,Lasso
,SGDRegressor,Ridge
```

In [190]:

```
lasso = Lasso(alpha = 0.001)
lasso.fit(X_scaled,y_log)
y_pred_lasso = lasso.predict(test_X_scaled)
```

In [191]:

```
FI_lasso = pd.DataFrame({"Feature Importance":lasso.coef_, index=data_pipeline.  
columns})  
FI_lasso.sort_values("Feature Importance",ascending=False)
```


Out[191]:

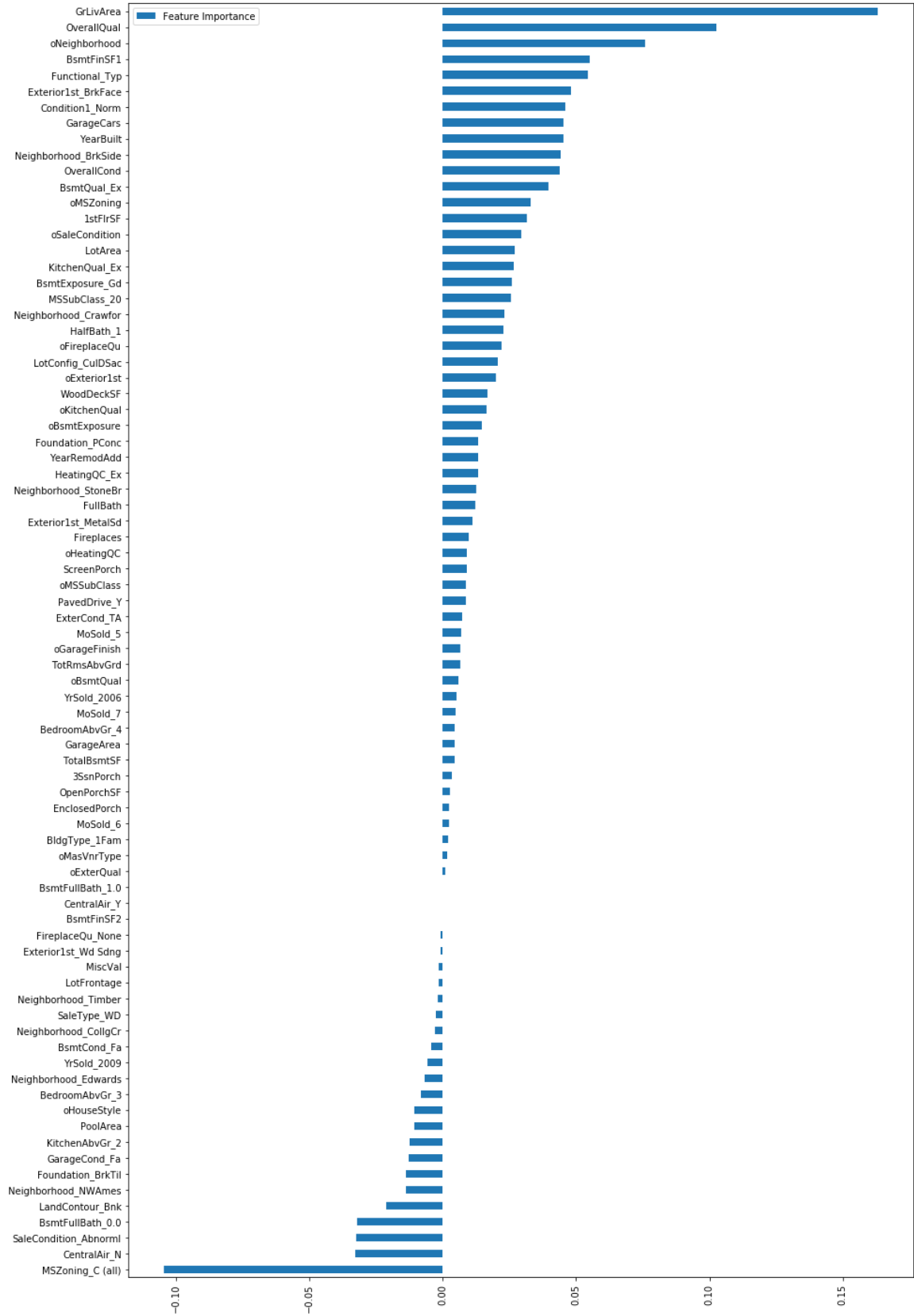
Feature Importance	
GrLivArea	0.163010
OverallQual	0.102665
oNeighborhood	0.076002
BsmtFinSF1	0.055243
Functional_Typ	0.054297
Exterior1st_BrkFace	0.048122
Condition1_Norm	0.045894
GarageCars	0.045438
YearBuilt	0.045378
Neighborhood_BrkSide	0.044335
OverallCond	0.043769
BsmtQual_Ex	0.039762
oMSZoning	0.032991
1stFlrSF	0.031554
oSaleCondition	0.029414
LotArea	0.026918
KitchenQual_Ex	0.026612
BsmtExposure_Gd	0.025852
MSSubClass_20	0.025700
Neighborhood_Crawfor	0.023158
HalfBath_1	0.022972
oFireplaceQu	0.022057
LotConfig_CulDSac	0.020708
oExterior1st	0.020122
WoodDeckSF	0.016964
oKitchenQual	0.016402
oBsmtExposure	0.014838
Foundation_PConc	0.013472
YearRemodAdd	0.013453
HeatingQC_Ex	0.013290
...	...
BsmtQual_Gd	-0.000000
RoofMatl_Membran	0.000000
RoofMatl_CompShg	0.000000
RoofMatl_ClyTile	-0.000000
RoofStyle_Shed	0.000000

Feature Importance	
RoofStyle_Mansard	0.000000
ExterCond_Po	-0.000000
BsmtFinSF2	-0.000072
FireplaceQu_None	-0.000635
Exterior1st_Wd Sdng	-0.000671
MiscVal	-0.001562
LotFrontage	-0.001595
Neighborhood_Timber	-0.001829
SaleType_WD	-0.002551
Neighborhood_CollgCr	-0.002945
BsmtCond_Fa	-0.004276
YrSold_2009	-0.005709
Neighborhood_Edwards	-0.006648
BedroomAbvGr_3	-0.008201
oHouseStyle	-0.010532
PoolArea	-0.010542
KitchenAbvGr_2	-0.012274
GarageCond_Fa	-0.012713
Foundation_BrkTil	-0.013611
Neighborhood_NWAmes	-0.013618
LandContour_Bnk	-0.021147
BsmtFullBath_0.0	-0.032013
SaleCondition_Abnorml	-0.032362
CentralAir_N	-0.032854
MSZoning_C (all)	-0.104515

406 rows × 1 columns

In [192]:

```
FI_lasso[FI_lasso["Feature Importance"]!=0].sort_values("Feature Importance").plot(kind="barh",figsize=(15,25))  
plt.xticks(rotation=90)  
plt.show()
```



In [193]:

```
class add_feature(BaseEstimator, TransformerMixin):
    def __init__(self, additional=1):
        self.additional = additional

    def fit(self, X, y=None):
        return self

    def transform(self, X):
        if self.additional==1:
            X["TotalHouse"] = X["TotalBsmtSF"] + X["1stFlrSF"] + X["2ndFlrSF"]
            X["TotalArea"] = X["TotalBsmtSF"] + X["1stFlrSF"] + X["2ndFlrSF"] +
X["GarageArea"]

            else:
                X["TotalHouse"] = X["TotalBsmtSF"] + X["1stFlrSF"] + X["2ndFlrSF"]
                X["TotalArea"] = X["TotalBsmtSF"] + X["1stFlrSF"] + X["2ndFlrSF"] +
X["GarageArea"]

                X["+_TotalHouse_OverallQual"] = X["TotalHouse"] * X["OverallQual"]
                X["+_GrLivArea_OverallQual"] = X["GrLivArea"] * X["OverallQual"]
                X["+_oMSZoning_TotalHouse"] = X["oMSZoning"] * X["TotalHouse"]
                X["+_oMSZoning_OverallQual"] = X["oMSZoning"] + X["OverallQual"]
                X["+_oMSZoning_YearBuilt"] = X["oMSZoning"] + X["YearBuilt"]
                X["+_oNeighborhood_TotalHouse"] = X["oNeighborhood"] * X["TotalHous
e"]
                X["+_oNeighborhood_OverallQual"] = X["oNeighborhood"] + X["OverallQu
al"]

                X["+_oNeighborhood_YearBuilt"] = X["oNeighborhood"] + X["YearBuilt"]
                X["+_BsmtFinSF1_OverallQual"] = X["BsmtFinSF1"] * X["OverallQual"]

                X["-_oFunctional_TotalHouse"] = X["oFunctional"] * X["TotalHouse"]
                X["-_oFunctional_OverallQual"] = X["oFunctional"] + X["OverallQual"]
                X["-_LotArea_OverallQual"] = X["LotArea"] * X["OverallQual"]
                X["-_TotalHouse_LotArea"] = X["TotalHouse"] + X["LotArea"]
                X["-_oCondition1_TotalHouse"] = X["oCondition1"] * X["TotalHouse"]
                X["-_oCondition1_OverallQual"] = X["oCondition1"] + X["OverallQual"]

                X["Bsmt"] = X["BsmtFinSF1"] + X["BsmtFinSF2"] + X["BsmtUnfSF"]
                X["Rooms"] = X["FullBath"]+X["TotRmsAbvGrd"]
                X["PorchArea"] = X["OpenPorchSF"]+X["EnclosedPorch"]+X["3SsnPorch"]+
X["ScreenPorch"]
                X["TotalPlace"] = X["TotalBsmtSF"] + X["1stFlrSF"] + X["2ndFlrSF"] +
X["GarageArea"] + X["OpenPorchSF"]+X["EnclosedPorch"]+X["3SsnPorch"]+X["ScreenPo
rch"]

        return X
```

In [194]:

```

pipeline = Pipeline([('labenc',labenc()),('add_feature', add_feature(additional=
2)),
                    ('skewness',skewness(skew =1)),('dummies',dummies())])

full_pipe = pipeline.fit_transform(full)
full_pipe.shape

```

Out[194]:

(2917, 427)

In [195]:

```

n_train=train.shape[0]
X = full_pipe[:n_train]
test_X = full_pipe[n_train:]
y= train.SalePrice

X_scaled = robust_scaler.fit(X).transform(X)
y_log = np.log(train.SalePrice)
test_X_scaled = robust_scaler.transform(test_X)

```

In [196]:

```
print(X_scaled.shape)
```

(1458, 427)

In [197]:

```
from sklearn.decomposition import PCA,KernelPCA
```

In [198]:

```

pca = PCA(0.95)

#pca = PCA(n_components = 426)
X_scaled = pca.fit_transform(X_scaled)
#X_scaled = pca.inverse_transform(lower_dimension_pca)
var1 = np.round(pca.explained_variance_ratio_*100, decimals = 1)
var1

```

Out[198]:

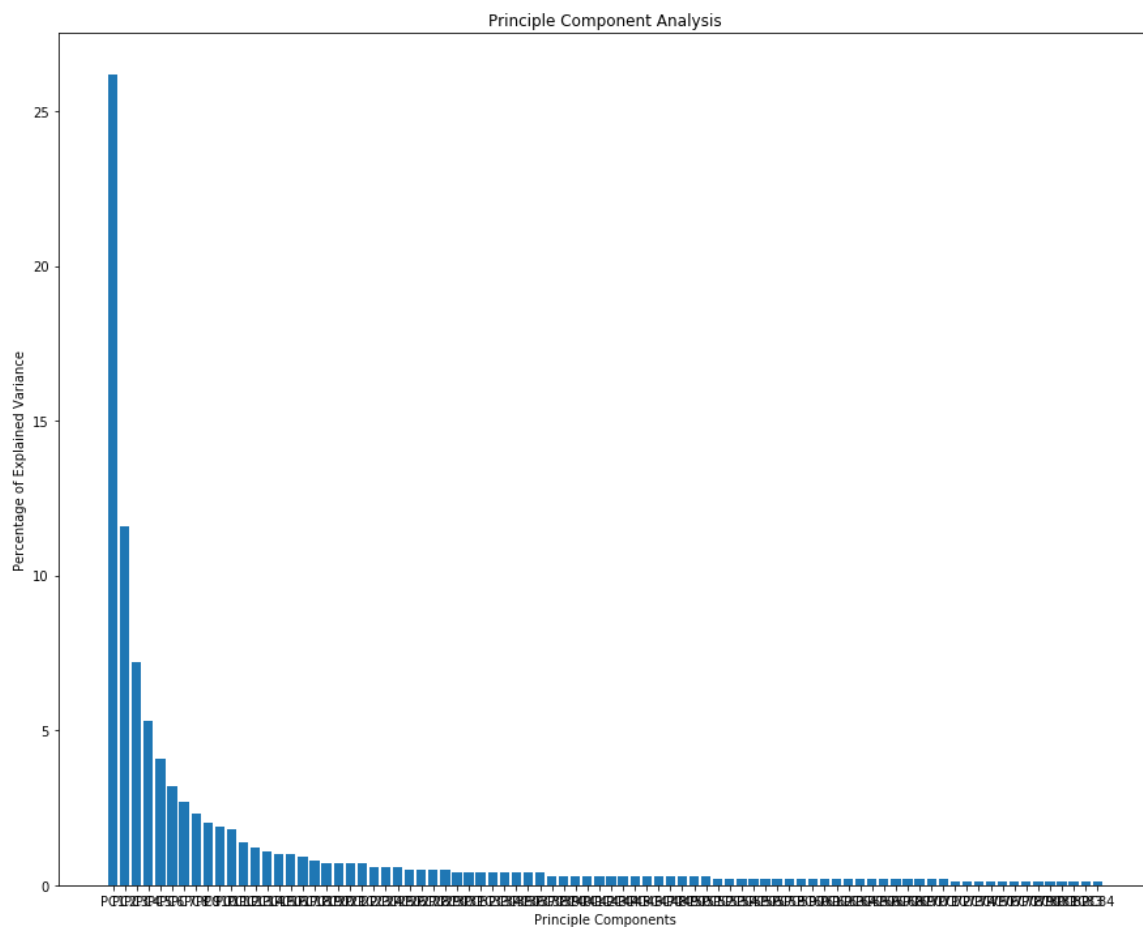
```

array([[26.2, 11.6,  7.2,  5.3,  4.1,  3.2,  2.7,  2.3,  2. ,  1.9,
  1.8,
        1.4,  1.2,  1.1,  1. ,  1. ,  0.9,  0.8,  0.7,  0.7,  0.7,
  0.7,
        0.6,  0.6,  0.6,  0.5,  0.5,  0.5,  0.5,  0.4,  0.4,  0.4,
  0.4,
        0.4,  0.4,  0.4,  0.4,  0.3,  0.3,  0.3,  0.3,  0.3,  0.3,
  0.3,
        0.3,  0.3,  0.3,  0.3,  0.3,  0.3,  0.3,  0.2,  0.2,  0.2,
  0.2,
        0.2,  0.2,  0.2,  0.2,  0.2,  0.2,  0.2,  0.2,  0.2,  0.2,
  0.2,
        0.2,  0.2,  0.2,  0.2,  0.2,  0.1,  0.1,  0.1,  0.1,  0.1,
  0.1,
        0.1,  0.1,  0.1,  0.1,  0.1,  0.1,  0.1,  0.1])

```

In [199]:

```
label =['PC' + str(x) for x in range(1,len(var1)+1)]  
plt.figure(figsize=(15,12))  
plt.bar(x=range(1,len(var1)+1), height = var1 ,tick_label = label)  
  
plt.ylabel("Percentage of Explained Variance")  
plt.xlabel("Principle Components")  
plt.title("Principle Component Analysis")  
plt.show()
```



In [200]:

```
test_X_scaled = pca.fit_transform(test_X_scaled)
var = np.round(pca.explained_variance_ratio_*100, decimals = 1)
var
```

Out[200]:

```
array([[27.3, 12.1,  6.9,  5.7,  4. ,  3.3,  2.8,  2.3,  2.1,  2. ,
        1.7,
        1.3,  1.1,  1. ,  1. ,  0.9,  0.8,  0.8,  0.7,  0.7,  0.6,
        0.6,
        0.5,  0.5,  0.5,  0.5,  0.5,  0.5,  0.4,  0.4,  0.4,  0.4,
        0.4,
        0.4,  0.4,  0.3,  0.3,  0.3,  0.3,  0.3,  0.3,  0.3,  0.3,
        0.3,
        0.3,  0.3,  0.3,  0.3,  0.2,  0.2,  0.2,  0.2,  0.2,  0.2,
        0.2,
        0.2,  0.2,  0.2,  0.2,  0.2,  0.2,  0.2,  0.2,  0.2,  0.2,
        0.2,
        0.2,  0.2,  0.2,  0.2,  0.1,  0.1,  0.1,  0.1,  0.1,  0.1,
        0.1,
        0.1,  0.1,  0.1,  0.1,  0.1,  0.1])
```

In [201]:

```
X_scaled.shape, test_X_scaled.shape
```

Out[201]:

```
((1458, 84), (1459, 83))
```

In [202]:

```
# Define Root Mean Square Error
def rmse_cv(model,X,y):
    rmse = np.sqrt(-cross_val_score(model,X,y,scoring="neg_mean_squared_error",c
v=5))
    return rmse
```

In [203]:

```
from sklearn.ensemble import ExtraTreesRegressor,GradientBoostingRegressor,Rando
mForestRegressor,VotingClassifier
from sklearn.svm import LinearSVR,SVR
from sklearn.kernel_ridge import KernelRidge
from xgboost import XGBRegressor
```


In [204]:

```
models = [LinearRegression(),
           Ridge(),
           Lasso(alpha=0.01,max_iter=10000),
           RandomForestRegressor(),
           GradientBoostingRegressor(),
           SVR(),
           LinearSVR(),
           ElasticNet(alpha = 0.001,max_iter=10000),
           SGDRegressor(max_iter=1000, tol = 1e-3),
           BayesianRidge(),
           KernelRidge(alpha=0.6,kernel='polynomial',degree = 2,coef0=2.5),
           ExtraTreesRegressor(),
           XGBRegressor()]

names = ['LR','Ridge','Lasso','RF','GBR','SVR','LSVR','ENet','SGDR','BayRidge',
         'Kernel','XTreeR','XGBR']
```

In [205]:

```
for model,name in zip(models,names):  
    score = rmse_cv(model,X_scaled,y_log)  
    print("{}: {:.6f}, {:.4f}".format(name,score.mean(),score.std()))
```

LR: 0.127207, 0.012788
Ridge: 0.127144, 0.012795
Lasso: 0.130874, 0.011881

```
/anaconda3/lib/python3.7/site-packages/sklearn/ensemble/forest.py:24
6: FutureWarning: The default value of n_estimators will change from
10 in version 0.20 to 100 in 0.22.
    "10 in version 0.20 to 100 in 0.22.", FutureWarning)
/anaconda3/lib/python3.7/site-packages/sklearn/ensemble/forest.py:24
6: FutureWarning: The default value of n_estimators will change from
10 in version 0.20 to 100 in 0.22.
    "10 in version 0.20 to 100 in 0.22.", FutureWarning)
/anaconda3/lib/python3.7/site-packages/sklearn/ensemble/forest.py:24
6: FutureWarning: The default value of n_estimators will change from
10 in version 0.20 to 100 in 0.22.
    "10 in version 0.20 to 100 in 0.22.", FutureWarning)
/anaconda3/lib/python3.7/site-packages/sklearn/ensemble/forest.py:24
6: FutureWarning: The default value of n_estimators will change from
10 in version 0.20 to 100 in 0.22.
    "10 in version 0.20 to 100 in 0.22.", FutureWarning)
/anaconda3/lib/python3.7/site-packages/sklearn/ensemble/forest.py:24
6: FutureWarning: The default value of n_estimators will change from
10 in version 0.20 to 100 in 0.22.
    "10 in version 0.20 to 100 in 0.22.", FutureWarning)
```

RF: 0.141926, 0.008570
GBR: 0.129264, 0.009111

```
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:196: Futu
reWarning: The default value of gamma will change from 'auto' to 'sc
ale' in version 0.22 to account better for unscaled features. Set ga
mma explicitly to 'auto' or 'scale' to avoid this warning.
    "avoid this warning.", FutureWarning)
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:196: Futu
reWarning: The default value of gamma will change from 'auto' to 'sc
ale' in version 0.22 to account better for unscaled features. Set ga
mma explicitly to 'auto' or 'scale' to avoid this warning.
    "avoid this warning.", FutureWarning)
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:196: Futu
reWarning: The default value of gamma will change from 'auto' to 'sc
ale' in version 0.22 to account better for unscaled features. Set ga
mma explicitly to 'auto' or 'scale' to avoid this warning.
    "avoid this warning.", FutureWarning)
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:196: Futu
reWarning: The default value of gamma will change from 'auto' to 'sc
ale' in version 0.22 to account better for unscaled features. Set ga
mma explicitly to 'auto' or 'scale' to avoid this warning.
    "avoid this warning.", FutureWarning)
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:196: Futu
reWarning: The default value of gamma will change from 'auto' to 'sc
ale' in version 0.22 to account better for unscaled features. Set ga
mma explicitly to 'auto' or 'scale' to avoid this warning.
    "avoid this warning.", FutureWarning)
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:196: Futu
reWarning: The default value of gamma will change from 'auto' to 'sc
ale' in version 0.22 to account better for unscaled features. Set ga
mma explicitly to 'auto' or 'scale' to avoid this warning.
    "avoid this warning.", FutureWarning)
```

SVR: 0.129020, 0.012752

```
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:922: Conv  
ergenceWarning: Liblinear failed to converge, increase the number of  
iterations.
```

```
"the number of iterations.", ConvergenceWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:922: Conv  
ergenceWarning: Liblinear failed to converge, increase the number of  
iterations.
```

```
"the number of iterations.", ConvergenceWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:922: Conv  
ergenceWarning: Liblinear failed to converge, increase the number of  
iterations.
```

```
"the number of iterations.", ConvergenceWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:922: Conv  
ergenceWarning: Liblinear failed to converge, increase the number of  
iterations.
```

```
"the number of iterations.", ConvergenceWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:922: Conv  
ergenceWarning: Liblinear failed to converge, increase the number of  
iterations.
```

```
"the number of iterations.", ConvergenceWarning)
```

```
LSVR: 0.130379, 0.012712
```

```
ENet: 0.125751, 0.013029
```

```
SGDR: 0.135695, 0.014213
```

```
BayRidge: 0.125546, 0.013013
```

```
Kernel: 0.117835, 0.008849
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/ensemble/forest.py:24  
6: FutureWarning: The default value of n_estimators will change from  
10 in version 0.20 to 100 in 0.22.
```

```
"10 in version 0.20 to 100 in 0.22.", FutureWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/ensemble/forest.py:24  
6: FutureWarning: The default value of n_estimators will change from  
10 in version 0.20 to 100 in 0.22.
```

```
"10 in version 0.20 to 100 in 0.22.", FutureWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/ensemble/forest.py:24  
6: FutureWarning: The default value of n_estimators will change from  
10 in version 0.20 to 100 in 0.22.
```

```
"10 in version 0.20 to 100 in 0.22.", FutureWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/ensemble/forest.py:24  
6: FutureWarning: The default value of n_estimators will change from  
10 in version 0.20 to 100 in 0.22.
```

```
"10 in version 0.20 to 100 in 0.22.", FutureWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/ensemble/forest.py:24  
6: FutureWarning: The default value of n_estimators will change from  
10 in version 0.20 to 100 in 0.22.
```

```
"10 in version 0.20 to 100 in 0.22.", FutureWarning)
```

```
XTreeR: 0.142756, 0.010216
```

```

/anaconda3/lib/python3.7/site-packages/xgboost/core.py:587: FutureWarning: Series.base is deprecated and will be removed in a future version
    if getattr(data, 'base', None) is not None and \
/anaconda3/lib/python3.7/site-packages/xgboost/core.py:587: FutureWarning: Series.base is deprecated and will be removed in a future version
    if getattr(data, 'base', None) is not None and \
/anaconda3/lib/python3.7/site-packages/xgboost/core.py:587: FutureWarning: Series.base is deprecated and will be removed in a future version
    if getattr(data, 'base', None) is not None and \
/anaconda3/lib/python3.7/site-packages/xgboost/core.py:587: FutureWarning: Series.base is deprecated and will be removed in a future version
    if getattr(data, 'base', None) is not None and \
XGBR: 0.129692, 0.009626

```

In [208]:

```

from sklearn.model_selection import cross_val_score, KFold, GridSearchCV, RandomizedSearchCV, StratifiedKFold, train_test_split
class grid():
    def __init__(self, model):
        self.model = model
    def grid_get(self, X, y, param_grid):
        grid_search = GridSearchCV(self.model, param_grid, cv=5, scoring='neg_mean_squared_error')
        grid_search.fit(X, y)
        print(grid_search.best_params_, np.sqrt(-grid_search.best_score_))
        grid_search.cv_results_['mean_test_score'] = np.sqrt(-grid_search.cv_results_['mean_test_score'])
        print(pd.DataFrame(grid_search.cv_results_)[['params', 'mean_test_score', 'std_test_score']])

class random():
    def __init__(self, model):
        self.model = model
    def random_get(self, X, y, param_grid):
        random_search = RandomizedSearchCV(self.model, param_grid, random_state=0, n_iter=10, scoring='neg_mean_squared_error')
        random_search.fit(X, y)
        print(random_search.best_params_, np.sqrt(-random_search.best_score_))
        random_search.cv_results_['mean_test_score'] = np.sqrt(-random_search.cv_results_['mean_test_score'])
        print(pd.DataFrame(random_search.cv_results_)[['params', 'mean_test_score', 'std_test_score']])

```

In [209]:

```
grid(Lasso()).grid_get(X_scaled,y_log,{ 'alpha':[0.01,0.001,0.0001,0.0002,0.0003,  
0.0004,0.0005,0.0006,0.0007,0.0009],  
                                     'max_iter':[10000]})
```

```
{'alpha': 0.001, 'max_iter': 10000} 0.12590997865928494
                                params  mean_test_score  std_test_s
core
0      {'alpha': 0.01, 'max_iter': 10000}          0.131410          0.00
3112
1      {'alpha': 0.001, 'max_iter': 10000}          0.125910          0.00
3366
2      {'alpha': 0.0001, 'max_iter': 10000}          0.127476          0.00
3325
3      {'alpha': 0.0002, 'max_iter': 10000}          0.127158          0.00
3325
4      {'alpha': 0.0003, 'max_iter': 10000}          0.126879          0.00
3330
5      {'alpha': 0.0004, 'max_iter': 10000}          0.126638          0.00
3335
6      {'alpha': 0.0005, 'max_iter': 10000}          0.126438          0.00
3337
7      {'alpha': 0.0006, 'max_iter': 10000}          0.126278          0.00
3340
8      {'alpha': 0.0007, 'max_iter': 10000}          0.126149          0.00
3342
9      {'alpha': 0.0009, 'max_iter': 10000}          0.125958          0.00
3357
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split0_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split1_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split2_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split3_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split4_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('mean_train_score'), which will not be available by default any more in 0.21. If
you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('std_train_score'), which will not be available by default any more in 0.21. If
you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
```

In [210]:

```
random(Lasso()).random_get(X_scaled,y_log,{'alpha':[0.01,0.001,0.0001,0.0002,0.0003,0.0004,0.0005,0.0006,0.0007,0.0009]}))
```

```
{'alpha': 0.001} 0.12589791575759118
      params  mean_test_score  std_test_score
0  {'alpha': 0.01}          0.131757          0.002214
1  {'alpha': 0.001}          0.125898          0.002171
2  {'alpha': 0.0001}          0.127435          0.002091
3  {'alpha': 0.0002}          0.127098          0.002098
4  {'alpha': 0.0003}          0.126817          0.002108
5  {'alpha': 0.0004}          0.126582          0.002120
6  {'alpha': 0.0005}          0.126390          0.002131
7  {'alpha': 0.0006}          0.126233          0.002143
8  {'alpha': 0.0007}          0.126120          0.002152
9  {'alpha': 0.0009}          0.125958          0.002167
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/model_selection/_split.py:2053: FutureWarning: You should specify a value for 'cv' instead of relying on the default value. The default value will change from 3 to 5 in version 0.22.
```

```
warnings.warn(CV_WARNING, FutureWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:125: FutureWarning: You are accessing a training score ('split0_train_score'), which will not be available by default any more in 0.21.
```

```
If you need training scores, please set return_train_score=True
```

```
warnings.warn(*warn_args, **warn_kwargs)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:125: FutureWarning: You are accessing a training score ('split1_train_score'), which will not be available by default any more in 0.21.
```

```
If you need training scores, please set return_train_score=True
```

```
warnings.warn(*warn_args, **warn_kwargs)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:125: FutureWarning: You are accessing a training score ('split2_train_score'), which will not be available by default any more in 0.21.
```

```
If you need training scores, please set return_train_score=True
```

```
warnings.warn(*warn_args, **warn_kwargs)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:125: FutureWarning: You are accessing a training score ('mean_train_score'), which will not be available by default any more in 0.21. If
```

```
you need training scores, please set return_train_score=True
```

```
warnings.warn(*warn_args, **warn_kwargs)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:125: FutureWarning: You are accessing a training score ('std_train_score'), which will not be available by default any more in 0.21. If
```

```
you need training scores, please set return_train_score=True
```

```
warnings.warn(*warn_args, **warn_kwargs)
```


In [211]:

```
grid(Ridge()).grid_get(X_scaled,y_log,  
                        {'alpha':[10,20,25,30,35,40,45,50,55,57,60,65,70,75,80,10  
0], 'max_iter':[10000]})
```

```

{'alpha': 100, 'max_iter': 10000} 0.12547326886690283
                                params  mean_test_score  std_test_sco
re
0      {'alpha': 10, 'max_iter': 10000}          0.127292          0.0033
28
1      {'alpha': 20, 'max_iter': 10000}          0.126863          0.0033
27
2      {'alpha': 25, 'max_iter': 10000}          0.126684          0.0033
26
3      {'alpha': 30, 'max_iter': 10000}          0.126524          0.0033
26
4      {'alpha': 35, 'max_iter': 10000}          0.126382          0.0033
25
5      {'alpha': 40, 'max_iter': 10000}          0.126255          0.0033
24
6      {'alpha': 45, 'max_iter': 10000}          0.126141          0.0033
23
7      {'alpha': 50, 'max_iter': 10000}          0.126039          0.0033
22
8      {'alpha': 55, 'max_iter': 10000}          0.125948          0.0033
20
9      {'alpha': 57, 'max_iter': 10000}          0.125914          0.0033
20
10     {'alpha': 60, 'max_iter': 10000}          0.125867          0.0033
19
11     {'alpha': 65, 'max_iter': 10000}          0.125794          0.0033
18
12     {'alpha': 70, 'max_iter': 10000}          0.125730          0.0033
17
13     {'alpha': 75, 'max_iter': 10000}          0.125672          0.0033
15
14     {'alpha': 80, 'max_iter': 10000}          0.125621          0.0033
14
15     {'alpha': 100, 'max_iter': 10000}         0.125473          0.0033
09

```

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split0_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split1_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split2_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split3_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split4_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('mean_train_score'), which will not be available by default any more in 0.21. If
you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('std_train_score'), which will not be available by default any more in 0.21. If
you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
```

In [212]:

```
random(Ridge()).random_get(X_scaled,y_log,
                           {'alpha':[10,20,25,30,35,40,45,50,55,57,60,65,70,75,80,100],
                           'max_iter':[10000]})
```

```
{'max_iter': 10000, 'alpha': 80} 0.12564921489535272
```

	params	mean_test_score	std_test_score
0	{'max_iter': 10000, 'alpha': 20}	0.126761	0.002110
1	{'max_iter': 10000, 'alpha': 45}	0.126066	0.002120
2	{'max_iter': 10000, 'alpha': 55}	0.125899	0.002122
3	{'max_iter': 10000, 'alpha': 57}	0.125871	0.002122
4	{'max_iter': 10000, 'alpha': 75}	0.125684	0.002125
5	{'max_iter': 10000, 'alpha': 35}	0.126288	0.002117
6	{'max_iter': 10000, 'alpha': 25}	0.126581	0.002113
7	{'max_iter': 10000, 'alpha': 80}	0.125649	0.002125
8	{'max_iter': 10000, 'alpha': 60}	0.125832	0.002123
9	{'max_iter': 10000, 'alpha': 50}	0.125977	0.002121

/anaconda3/lib/python3.7/site-packages/sklearn/model_selection/_split.py:2053: FutureWarning: You should specify a value for 'cv' instead of relying on the default value. The default value will change from 3 to 5 in version 0.22.

warnings.warn(CV_WARNING, FutureWarning)

/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:

125: FutureWarning: You are accessing a training score ('split0_train_score'), which will not be available by default any more in 0.21.

If you need training scores, please set return_train_score=True

warnings.warn(*warn_args, **warn_kwargs)

/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:

125: FutureWarning: You are accessing a training score ('split1_train_score'), which will not be available by default any more in 0.21.

If you need training scores, please set return_train_score=True

warnings.warn(*warn_args, **warn_kwargs)

/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:

125: FutureWarning: You are accessing a training score ('split2_train_score'), which will not be available by default any more in 0.21.

If you need training scores, please set return_train_score=True

warnings.warn(*warn_args, **warn_kwargs)

/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:

125: FutureWarning: You are accessing a training score ('mean_train_score'), which will not be available by default any more in 0.21. If

you need training scores, please set return_train_score=True

warnings.warn(*warn_args, **warn_kwargs)

/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:

125: FutureWarning: You are accessing a training score ('std_train_score'), which will not be available by default any more in 0.21. If

you need training scores, please set return_train_score=True

warnings.warn(*warn_args, **warn_kwargs)

In [213]:

```
param_grid = {'C':[0.05,0.1,0.15,0.2],"epsilon":[0.0015,0.002,0.0025,0.025,0.0004]}  
grid(LinearSVR()).grid_get(X_scaled,y_log,param_grid)
```

[illegible]

```
"the number of iterations.", ConvergenceWarning)
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:922: Conv
vergenceWarning: Liblinear failed to converge, increase the number of
```

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[illegible]

```
"the number of iterations.", ConvergenceWarning)
```

```
{'C': 0.15, 'epsilon': 0.002} 0.12588870949551445
```

	params	mean_test_score	std_test_score
0	{'C': 0.05, 'epsilon': 0.0015}	0.127670	0.003651
1	{'C': 0.05, 'epsilon': 0.002}	0.127768	0.003645
2	{'C': 0.05, 'epsilon': 0.0025}	0.127802	0.003675
3	{'C': 0.05, 'epsilon': 0.025}	0.127952	0.003708
4	{'C': 0.05, 'epsilon': 0.0004}	0.127856	0.003632
5	{'C': 0.1, 'epsilon': 0.0015}	0.126160	0.003905
6	{'C': 0.1, 'epsilon': 0.002}	0.126080	0.003903
7	{'C': 0.1, 'epsilon': 0.0025}	0.126175	0.003968
8	{'C': 0.1, 'epsilon': 0.025}	0.126043	0.003983
9	{'C': 0.1, 'epsilon': 0.0004}	0.126119	0.003879
10	{'C': 0.15, 'epsilon': 0.0015}	0.126073	0.004039
11	{'C': 0.15, 'epsilon': 0.002}	0.125889	0.003944
12	{'C': 0.15, 'epsilon': 0.0025}	0.125965	0.003929
13	{'C': 0.15, 'epsilon': 0.025}	0.126161	0.004169
14	{'C': 0.15, 'epsilon': 0.0004}	0.126175	0.003935
15	{'C': 0.2, 'epsilon': 0.0015}	0.126375	0.004057
16	{'C': 0.2, 'epsilon': 0.002}	0.126274	0.004095
17	{'C': 0.2, 'epsilon': 0.0025}	0.126556	0.004186
18	{'C': 0.2, 'epsilon': 0.025}	0.126254	0.003922
19	{'C': 0.2, 'epsilon': 0.0004}	0.126732	0.004116

```
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:922: Conv  
ergenceWarning: Liblinear failed to converge, increase the number of  
iterations.
```

```
    "the number of iterations.", ConvergenceWarning)  
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:  
125: FutureWarning: You are accessing a training score ('split0_train_score'), which will not be available by default any more in 0.21.  
If you need training scores, please set return_train_score=True  
    warnings.warn(*warn_args, **warn_kwargs)  
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:  
125: FutureWarning: You are accessing a training score ('split1_train_score'), which will not be available by default any more in 0.21.  
If you need training scores, please set return_train_score=True  
    warnings.warn(*warn_args, **warn_kwargs)  
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:  
125: FutureWarning: You are accessing a training score ('split2_train_score'), which will not be available by default any more in 0.21.  
If you need training scores, please set return_train_score=True  
    warnings.warn(*warn_args, **warn_kwargs)  
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:  
125: FutureWarning: You are accessing a training score ('split3_train_score'), which will not be available by default any more in 0.21.  
If you need training scores, please set return_train_score=True  
    warnings.warn(*warn_args, **warn_kwargs)  
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:  
125: FutureWarning: You are accessing a training score ('split4_train_score'), which will not be available by default any more in 0.21.  
If you need training scores, please set return_train_score=True  
    warnings.warn(*warn_args, **warn_kwargs)  
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:  
125: FutureWarning: You are accessing a training score ('mean_train_score'), which will not be available by default any more in 0.21. If  
you need training scores, please set return_train_score=True  
    warnings.warn(*warn_args, **warn_kwargs)  
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:  
125: FutureWarning: You are accessing a training score ('std_train_score'), which will not be available by default any more in 0.21. If  
you need training scores, please set return_train_score=True  
    warnings.warn(*warn_args, **warn_kwargs)
```

In [214]:

```
param_grid = {'C':[0.05,0.1,0.15,0.2],"epsilon":[0.0015,0.002,0.0025,0.025,0.0004]}  
random(LinearSVR()).random_get(X_scaled,y_log,param_grid)
```

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```
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:922: Conv
ergenceWarning: Liblinear failed to converge, increase the number of
iterations.
```

```
"the number of iterations.", ConvergenceWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:922: Conv
ergenceWarning: Liblinear failed to converge, increase the number of
iterations.
```

```
"the number of iterations.", ConvergenceWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:922: Conv
ergenceWarning: Liblinear failed to converge, increase the number of
iterations.
```

```
"the number of iterations.", ConvergenceWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:922: Conv
ergenceWarning: Liblinear failed to converge, increase the number of
iterations.
```

```
"the number of iterations.", ConvergenceWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:922: Conv
ergenceWarning: Liblinear failed to converge, increase the number of
iterations.
```

```
"the number of iterations.", ConvergenceWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:922: Conv
ergenceWarning: Liblinear failed to converge, increase the number of
iterations.
```

```
"the number of iterations.", ConvergenceWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:922: Conv
ergenceWarning: Liblinear failed to converge, increase the number of
iterations.
```

```
"the number of iterations.", ConvergenceWarning)
```

```
{'epsilon': 0.0004, 'C': 0.2} 0.12678993862511065
```

	params	mean_test_score	std_test_score
0	{'epsilon': 0.025, 'C': 0.2}	0.127104	0.002853
1	{'epsilon': 0.002, 'C': 0.05}	0.130269	0.002116
2	{'epsilon': 0.0004, 'C': 0.2}	0.126790	0.002844
3	{'epsilon': 0.025, 'C': 0.1}	0.127116	0.002618
4	{'epsilon': 0.0015, 'C': 0.15}	0.127591	0.002796
5	{'epsilon': 0.0025, 'C': 0.2}	0.127116	0.002822
6	{'epsilon': 0.002, 'C': 0.1}	0.127559	0.002652
7	{'epsilon': 0.025, 'C': 0.15}	0.127252	0.002806
8	{'epsilon': 0.0004, 'C': 0.05}	0.130315	0.002087
9	{'epsilon': 0.0025, 'C': 0.05}	0.130144	0.002181

```
/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:922: Conv  
ergenceWarning: Liblinear failed to converge, increase the number of  
iterations.
```

```
    "the number of iterations.", ConvergenceWarning)  
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:  
125: FutureWarning: You are accessing a training score ('split0_train_score'), which will not be available by default any more in 0.21.  
If you need training scores, please set return_train_score=True  
    warnings.warn(*warn_args, **warn_kwargs)  
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:  
125: FutureWarning: You are accessing a training score ('split1_train_score'), which will not be available by default any more in 0.21.  
If you need training scores, please set return_train_score=True  
    warnings.warn(*warn_args, **warn_kwargs)  
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:  
125: FutureWarning: You are accessing a training score ('split2_train_score'), which will not be available by default any more in 0.21.  
If you need training scores, please set return_train_score=True  
    warnings.warn(*warn_args, **warn_kwargs)  
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:  
125: FutureWarning: You are accessing a training score ('mean_train_score'), which will not be available by default any more in 0.21. If  
you need training scores, please set return_train_score=True  
    warnings.warn(*warn_args, **warn_kwargs)  
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:  
125: FutureWarning: You are accessing a training score ('std_train_score'), which will not be available by default any more in 0.21. If  
you need training scores, please set return_train_score=True  
    warnings.warn(*warn_args, **warn_kwargs)
```


In [215]:

```
param_grid = {'C':[11,13,15], 'kernel':['rbf'], "gamma":[0.0003,0.0004], "epsilon":  
[0.008,0.009]}  
grid(SVR()).grid_get(X_scaled,y_log,param_grid)
```

```
{'C': 15, 'epsilon': 0.008, 'gamma': 0.0004, 'kernel': 'rbf'} 0.1243
937886301839
```

	params	mean_test_sco
re \		
0	{'C': 11, 'epsilon': 0.008, 'gamma': 0.0003, '...	0.1252
97		
1	{'C': 11, 'epsilon': 0.008, 'gamma': 0.0004, '...	0.1245
72		
2	{'C': 11, 'epsilon': 0.009, 'gamma': 0.0003, '...	0.1252
71		
3	{'C': 11, 'epsilon': 0.009, 'gamma': 0.0004, '...	0.1246
03		
4	{'C': 13, 'epsilon': 0.008, 'gamma': 0.0003, '...	0.1252
03		
5	{'C': 13, 'epsilon': 0.008, 'gamma': 0.0004, '...	0.1244
76		
6	{'C': 13, 'epsilon': 0.009, 'gamma': 0.0003, '...	0.1251
89		
7	{'C': 13, 'epsilon': 0.009, 'gamma': 0.0004, '...	0.1245
55		
8	{'C': 15, 'epsilon': 0.008, 'gamma': 0.0003, '...	0.1251
28		
9	{'C': 15, 'epsilon': 0.008, 'gamma': 0.0004, '...	0.1243
94		
10	{'C': 15, 'epsilon': 0.009, 'gamma': 0.0003, '...	0.1251
04		
11	{'C': 15, 'epsilon': 0.009, 'gamma': 0.0004, '...	0.1244
51		

	std_test_score
0	0.004451
1	0.004298
2	0.004439
3	0.004311
4	0.004440
5	0.004315
6	0.004435
7	0.004330
8	0.004453
9	0.004325
10	0.004449
11	0.004334

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split0_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split1_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split2_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split3_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split4_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('mean_train_score'), which will not be available by default any more in 0.21. If
you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('std_train_score'), which will not be available by default any more in 0.21. If
you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
```

In [216]:

```
param_grid = {'C':[11,13,15], 'kernel':['rbf'], "gamma":[0.0003,0.0004], "epsilon":  
[0.008,0.009]}  
random(SVR()).random_get(X_scaled,y_log,param_grid)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/model_selection/_split.py:2053: FutureWarning: You should specify a value for 'cv' instead of relying on the default value. The default value will change from 3 to 5 in version 0.22.
```

```
warnings.warn(CV_WARNING, FutureWarning)
```

```
{'kernel': 'rbf', 'gamma': 0.0004, 'epsilon': 0.008, 'C': 15} 0.1244596963550175
```

	params	mean_test_score
0	{'kernel': 'rbf', 'gamma': 0.0003, 'epsilon': ...}	0.12541
1	{'kernel': 'rbf', 'gamma': 0.0004, 'epsilon': ...}	0.12448
2	{'kernel': 'rbf', 'gamma': 0.0003, 'epsilon': ...}	0.12536
3	{'kernel': 'rbf', 'gamma': 0.0003, 'epsilon': ...}	0.12534
4	{'kernel': 'rbf', 'gamma': 0.0003, 'epsilon': ...}	0.12538
5	{'kernel': 'rbf', 'gamma': 0.0003, 'epsilon': ...}	0.12531
6	{'kernel': 'rbf', 'gamma': 0.0004, 'epsilon': ...}	0.12470
7	{'kernel': 'rbf', 'gamma': 0.0004, 'epsilon': ...}	0.12455
8	{'kernel': 'rbf', 'gamma': 0.0004, 'epsilon': ...}	0.12446
9	{'kernel': 'rbf', 'gamma': 0.0004, 'epsilon': ...}	0.12469

	std_test_score
0	0.003041
1	0.003002
2	0.003049
3	0.003053
4	0.003033
5	0.003051
6	0.003007
7	0.002981
8	0.002996
9	0.002994

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split0_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split1_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split2_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('mean_train_score'), which will not be available by default any more in 0.21. If
you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('std_train_score'), which will not be available by default any more in 0.21. If
you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
```

In [217]:

```
param_grid={'alpha':[0.1,0.15,0.2,0.25], 'kernel':['polynomial'], 'degree':[3],  
'coef0':[0.6,0.7,0.75,0.8,0.9]}  
grid(KernelRidge()).grid_get(X_scaled,y_log,param_grid)
```

```
{'alpha': 0.25, 'coef0': 0.9, 'degree': 3, 'kernel': 'polynomial'}
0.13498125343336792
```

	params	mean_test_sco
re \		
0	{'alpha': 0.1, 'coef0': 0.6, 'degree': 3, 'ker...	0.1645
14		
1	{'alpha': 0.1, 'coef0': 0.7, 'degree': 3, 'ker...	0.1505
85		
2	{'alpha': 0.1, 'coef0': 0.75, 'degree': 3, 'ke...	0.1465
20		
3	{'alpha': 0.1, 'coef0': 0.8, 'degree': 3, 'ker...	0.1436
30		
4	{'alpha': 0.1, 'coef0': 0.9, 'degree': 3, 'ker...	0.1400
97		
5	{'alpha': 0.15, 'coef0': 0.6, 'degree': 3, 'ke...	0.1651
89		
6	{'alpha': 0.15, 'coef0': 0.7, 'degree': 3, 'ke...	0.1495
47		
7	{'alpha': 0.15, 'coef0': 0.75, 'degree': 3, 'k...	0.1449
39		
8	{'alpha': 0.15, 'coef0': 0.8, 'degree': 3, 'ke...	0.1416
44		
9	{'alpha': 0.15, 'coef0': 0.9, 'degree': 3, 'ke...	0.1375
76		
10	{'alpha': 0.2, 'coef0': 0.6, 'degree': 3, 'ker...	0.1665
89		
11	{'alpha': 0.2, 'coef0': 0.7, 'degree': 3, 'ker...	0.1493
69		
12	{'alpha': 0.2, 'coef0': 0.75, 'degree': 3, 'ke...	0.1442
56		
13	{'alpha': 0.2, 'coef0': 0.8, 'degree': 3, 'ker...	0.1405
79		
14	{'alpha': 0.2, 'coef0': 0.9, 'degree': 3, 'ker...	0.1360
00		
15	{'alpha': 0.25, 'coef0': 0.6, 'degree': 3, 'ke...	0.1683
76		
16	{'alpha': 0.25, 'coef0': 0.7, 'degree': 3, 'ke...	0.1496
79		
17	{'alpha': 0.25, 'coef0': 0.75, 'degree': 3, 'k...	0.1440
88		
18	{'alpha': 0.25, 'coef0': 0.8, 'degree': 3, 'ke...	0.1400
49		
19	{'alpha': 0.25, 'coef0': 0.9, 'degree': 3, 'ke...	0.1349
81		

std_test_score

0	0.005877
1	0.005173
2	0.004967
3	0.004819
4	0.004635
5	0.005621
6	0.004905
7	0.004689
8	0.004533
9	0.004335
10	0.005428
11	0.004715
12	0.004496
13	0.004335
14	0.004128

15	0.005272
16	0.004570
17	0.004352
18	0.004189
19	0.003977

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split0_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split1_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split2_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split3_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split4_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('mean_train_score'), which will not be available by default any more in 0.21. If
you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('std_train_score'), which will not be available by default any more in 0.21. If
you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
```

In [218]:

```
param_grid={'alpha':[0.1,0.15,0.2,0.25], 'kernel':['polynomial'], 'degree':[3],  
'coef0':[0.6,0.7,0.75,0.8,0.9]}  
random(KernelRidge()).random_get(X_scaled,y_log,param_grid)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/model_selection/_split.py:2053: FutureWarning: You should specify a value for 'cv' instead of relying on the default value. The default value will change from 3 to 5 in version 0.22.
```

```
warnings.warn(CV_WARNING, FutureWarning)
```

```
{'kernel': 'polynomial', 'degree': 3, 'coef0': 0.9, 'alpha': 0.25}
0.13788763231762488
```

	params	mean_test_score
0	{'kernel': 'polynomial', 'degree': 3, 'coef0': ...}	0.14466
1	{'kernel': 'polynomial', 'degree': 3, 'coef0': ...}	0.15531
2	{'kernel': 'polynomial', 'degree': 3, 'coef0': ...}	0.13788
3	{'kernel': 'polynomial', 'degree': 3, 'coef0': ...}	0.14490
4	{'kernel': 'polynomial', 'degree': 3, 'coef0': ...}	0.17794
5	{'kernel': 'polynomial', 'degree': 3, 'coef0': ...}	0.14995
6	{'kernel': 'polynomial', 'degree': 3, 'coef0': ...}	0.15543
7	{'kernel': 'polynomial', 'degree': 3, 'coef0': ...}	0.14459
8	{'kernel': 'polynomial', 'degree': 3, 'coef0': ...}	0.14095
9	{'kernel': 'polynomial', 'degree': 3, 'coef0': ...}	0.14984

	std_test_score
0	0.005907
1	0.006844
2	0.005288
3	0.005875
4	0.008736
5	0.006353
6	0.006886
7	0.005893
8	0.005272
9	0.006279

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split0_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split1_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split2_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('mean_train_score'), which will not be available by default any more in 0.21. If
you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('std_train_score'), which will not be available by default any more in 0.21. If
you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
```

In [219]:

```
grid(ElasticNet()).grid_get(X_scaled,y_log,{ 'alpha':[0.004,0.005,0.006,0.0065],  
                                              'l1_ratio':[0.065,0.06,0.070,0.075],  
                                              'max_iter':[10000]})
```

```
{'alpha': 0.0065, 'l1_ratio': 0.075, 'max_iter': 10000} 0.1262414912
449528
```

	params	mean_test_sco
re \		
0	{'alpha': 0.004, 'l1_ratio': 0.065, 'max_iter'...	0.1268
01		
1	{'alpha': 0.004, 'l1_ratio': 0.06, 'max_iter':...	0.1268
52		
2	{'alpha': 0.004, 'l1_ratio': 0.07, 'max_iter':...	0.1267
51		
3	{'alpha': 0.004, 'l1_ratio': 0.075, 'max_iter'...	0.1267
03		
4	{'alpha': 0.005, 'l1_ratio': 0.065, 'max_iter'...	0.1266
03		
5	{'alpha': 0.005, 'l1_ratio': 0.06, 'max_iter':...	0.1266
59		
6	{'alpha': 0.005, 'l1_ratio': 0.07, 'max_iter':...	0.1265
49		
7	{'alpha': 0.005, 'l1_ratio': 0.075, 'max_iter'...	0.1264
97		
8	{'alpha': 0.006, 'l1_ratio': 0.065, 'max_iter'...	0.1264
28		
9	{'alpha': 0.006, 'l1_ratio': 0.06, 'max_iter':...	0.1264
88		
10	{'alpha': 0.006, 'l1_ratio': 0.07, 'max_iter':...	0.1263
71		
11	{'alpha': 0.006, 'l1_ratio': 0.075, 'max_iter'...	0.1263
19		
12	{'alpha': 0.0065, 'l1_ratio': 0.065, 'max_iter'...	0.1263
49		
13	{'alpha': 0.0065, 'l1_ratio': 0.06, 'max_iter'...	0.1264
09		
14	{'alpha': 0.0065, 'l1_ratio': 0.07, 'max_iter'...	0.1262
93		
15	{'alpha': 0.0065, 'l1_ratio': 0.075, 'max_iter'...	0.1262
41		

	std_test_score
0	0.003329
1	0.003328
2	0.003331
3	0.003332
4	0.003334
5	0.003332
6	0.003335
7	0.003336
8	0.003337
9	0.003336
10	0.003338
11	0.003338
12	0.003338
13	0.003337
14	0.003338
15	0.003339

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split0_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split1_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split2_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split3_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split4_train_score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('mean_train_score'), which will not be available by default any more in 0.21. If
you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('std_train_score'), which will not be available by default any more in 0.21. If
you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
```

In [220]:

```
random(ElasticNet()).random_get(X_scaled,y_log,{'alpha':[0.004,0.005,0.006,0.0065],  
                                                'l1_ratio':[0.065,0.06,0.070,0.075],  
                                                'max_iter':[10000]})
```



```
{'max_iter': 10000, 'l1_ratio': 0.07, 'alpha': 0.0065} 0.12626750848
365328
```

	params	mean_test_scor
e \		
0	{'max_iter': 10000, 'l1_ratio': 0.06, 'alpha':...	0.12680
2		
1	{'max_iter': 10000, 'l1_ratio': 0.07, 'alpha':...	0.12650
4		
2	{'max_iter': 10000, 'l1_ratio': 0.065, 'alpha'...	0.12639
3		
3	{'max_iter': 10000, 'l1_ratio': 0.06, 'alpha':...	0.12644
8		
4	{'max_iter': 10000, 'l1_ratio': 0.06, 'alpha':...	0.12637
6		
5	{'max_iter': 10000, 'l1_ratio': 0.065, 'alpha'...	0.12655
9		
6	{'max_iter': 10000, 'l1_ratio': 0.07, 'alpha':...	0.12670
4		
7	{'max_iter': 10000, 'l1_ratio': 0.07, 'alpha':...	0.12626
8		
8	{'max_iter': 10000, 'l1_ratio': 0.07, 'alpha':...	0.12634
1		
9	{'max_iter': 10000, 'l1_ratio': 0.075, 'alpha'...	0.12645
6		

	std_test_score
0	0.002106
1	0.002119
2	0.002124
3	0.002120
4	0.002124
5	0.002116
6	0.002110
7	0.002131
8	0.002127
9	0.002121

```
/anaconda3/lib/python3.7/site-packages/sklearn/model_selection/_split.py:2053: FutureWarning: You should specify a value for 'cv' instead of relying on the default value. The default value will change from 3 to 5 in version 0.22.
```

```
warnings.warn(CV_WARNING, FutureWarning)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:125: FutureWarning: You are accessing a training score ('split0_train_score'), which will not be available by default any more in 0.21. If you need training scores, please set return_train_score=True
```

```
warnings.warn(*warn_args, **warn_kwargs)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:125: FutureWarning: You are accessing a training score ('split1_train_score'), which will not be available by default any more in 0.21. If you need training scores, please set return_train_score=True
```

```
warnings.warn(*warn_args, **warn_kwargs)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:125: FutureWarning: You are accessing a training score ('split2_train_score'), which will not be available by default any more in 0.21. If you need training scores, please set return_train_score=True
```

```
warnings.warn(*warn_args, **warn_kwargs)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:125: FutureWarning: You are accessing a training score ('mean_train_score'), which will not be available by default any more in 0.21. If you need training scores, please set return_train_score=True
```

```
warnings.warn(*warn_args, **warn_kwargs)
```

```
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:125: FutureWarning: You are accessing a training score ('std_train_score'), which will not be available by default any more in 0.21. If you need training scores, please set return_train_score=True
```

```
warnings.warn(*warn_args, **warn_kwargs)
```

In [221]:

```
class AverageWeight(BaseEstimator, RegressorMixin):
    def __init__(self, model, weight):
        self.model = model
        self.weight = weight

    def fit(self, X, y):
        self.models_ = [clone(x) for x in self.model]
        for model in self.models_:
            model.fit(X, y)
        return self

    def predict(self, X):
        w = list()
        pred = np.array([model.predict(X) for model in self.models_])
        # for every data point, single model prediction times weight, then add them together
        for data in range(pred.shape[1]):
            single = [pred[model, data]*weight for model, weight in zip(range(pred.shape[0]), self.weight)]
            w.append(np.sum(single))
        return w
```

In [222]:

```
lasso = Lasso(alpha= 0.0005, max_iter= 10000)
ridge = Ridge(alpha=64, max_iter= 10000)
svr = SVR(C = 13, epsilon= 0.009, gamma = 0.0004, kernel = 'rbf')
ker = KernelRidge(alpha=0.25 ,kernel='polynomial',degree=3 , coef0=0.9)
ela = ElasticNet(alpha=0.0065,l1_ratio=0.065,max_iter=10000)
bay = BayesianRidge()
```

In [223]:

```
lasso = Lasso(alpha= 0.0005, max_iter= 10000)
ridge = Ridge(alpha=45, max_iter= 10000)
svr = SVR(C = 0.2, epsilon= 0.025, gamma = 0.0004, kernel = 'rbf')
ker = KernelRidge(alpha=0.15 ,kernel='polynomial',degree=3 , coef0=0.9)
ela = ElasticNet(alpha=0.0065,l1_ratio=0.075,max_iter=10000)
bay = BayesianRidge()
```

In [224]:

```
# Lasso          0.111296607965
# Ridge          0.110201749615
# SVR            0.108231959163
# Kernel Ridge CV 0.108256222016
# Elastic CV     0.111127583451
# BayRidge:      0.110577
#0.107705581748
#0.107767677927
# assign weights based on their gridsearch score
w1 = 0.047
w2 = 0.2
w3 = 0.25
w4 = 0.3
w5 = 0.003
w6 = 0.2

weight_avg = AverageWeight(model = [lasso,ridge,svr,ker,ela,bay],weight=[w1,w2,w
3,w4,w5,w6])
score = rmse_cv(weight_avg,X_scaled,y_log)
print(score.mean())
```

0.11900337535219294

In [225]:

```
weight_avg = AverageWeight(model = [svr,ker],weight=[0.50,0.50])
score = rmse_cv(weight_avg,X_scaled,y_log)
print(score.mean())
```

0.12121529805160784

In [226]:

```

evc = VotingClassifier( estimators= [('lasso',lasso),('ridge',ridge),('svr',svr
),
                                ('ker',ker),('ela',ela),('bay',bay)], votin
g = 'hard')
evc.fit(X_scaled,y_log)

```

Out[226]:

```

VotingClassifier(estimators=[('lasso', Lasso(alpha=0.0005, copy_X=True,
fit_intercept=True, max_iter=10000,
normalize=False, positive=False, precompute=False, random_state=None,
selection='cyclic', tol=0.0001, warm_start=False)), ('ridge', Ridge(alpha=45, copy_X=True, fit_intercept=True, max_iter=10000,
normalize=False, lambda_1=1e-06, lambda_2=1e-06, n_iter=300,
normalize=False, tol=0.001, verbose=False))],
flatten_transform=None, n_jobs=None, voting='hard', weights=None)

```

In [227]:

```

class stacking(BaseEstimator, RegressorMixin, TransformerMixin):
    def __init__(self, mod, meta_model):
        self.mod = mod
        self.meta_model = meta_model
        self.kf = KFold(n_splits=5, random_state=42, shuffle=True)

    def fit(self, X, y):
        self.saved_model = [list() for i in self.mod]
        oof_train = np.zeros((X.shape[0], len(self.mod)))

        for i, model in enumerate(self.mod):
            for train_index, val_index in self.kf.split(X, y):
                renew_model = clone(model)
                renew_model.fit(X[train_index], y[train_index])
                self.saved_model[i].append(renew_model)
                oof_train[val_index, i] = renew_model.predict(X[val_index])

        self.meta_model.fit(oof_train, y)
        return self

    def predict(self, X):
        whole_test = np.column_stack([np.column_stack(model.predict(X) for model
in single_model).mean(axis=1) for single_model in self.saved_model])
        return self.meta_model.predict(whole_test)

    def get_oof(self, X, y, test_X):
        oof = np.zeros((X.shape[0], len(self.mod)))
        test_single = np.zeros((test_X.shape[0], 5))
        test_mean = np.zeros((test_X.shape[0], len(self.mod)))
        for i, model in enumerate(self.mod):
            for j, (train_index, val_index) in enumerate(self.kf.split(X, y)):
                clone_model = clone(model)
                clone_model.fit(X[train_index], y[train_index])
                oof[val_index, i] = clone_model.predict(X[val_index])
                test_single[:, j] = clone_model.predict(test_X)
            test_mean[:, i] = test_single.mean(axis=1)
        return oof, test_mean

```

In [228]:

```

X_scaled_imputed = Imputer().fit_transform(X_scaled)
y_log_imputed = Imputer().fit_transform(y_log.values.reshape(-1, 1)).ravel()

```

```

/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
58: DeprecationWarning: Class Imputer is deprecated; Imputer was depre
cated in version 0.20 and will be removed in 0.22. Import impute.S
impleImputer from sklearn instead.
    warnings.warn(msg, category=DeprecationWarning)
/anaconda3/lib/python3.7/site-packages/sklearn/utils/deprecation.py:
58: DeprecationWarning: Class Imputer is deprecated; Imputer was depre
cated in version 0.20 and will be removed in 0.22. Import impute.S
impleImputer from sklearn instead.
    warnings.warn(msg, category=DeprecationWarning)

```

In [229]:

```
X_scaled_imputed.shape,y_log_imputed.shape,test_X_scaled.shape
```

Out[229]:

```
((1458, 84), (1458,), (1459, 83))
```

In [230]:

```
stack_model = stacking(mod=[lasso,ridge,svr,ker,ela,bay],meta_model=ker)
```

In [231]:

```
score = rmse_cv(stack_model,X_scaled_imputed,y_log_imputed)
```

In [232]:

```
print(score.mean())
```

```
0.11991883371044768
```

In [233]:

```
X_train_stack,X_test_stack = stack_model.get_oof(X_scaled_imputed,y_log_imputed,
test_X_scaled)
```

```

-----
-----
ValueError                                Traceback (most recent call
last)
<ipython-input-233-de4390529faf> in <module>
----> 1 X_train_stack,X_test_stack = stack_model.get_oof(X_scaled_im
puted,y_log_imputed,test_X_scaled)

<ipython-input-227-2fdd33de90dc> in get_oof(self, X, y, test_X)
    33         clone_model.fit(X[train_index],y[train_index
])
    34         oof[val_index,i] = clone_model.predict(X[val
_index])
--> 35         test_single[:,j] = clone_model.predict(test_
X)
    36         test_mean[:,i] = test_single.mean(axis=1)
    37         return oof, test_mean

/anaconda3/lib/python3.7/site-packages/sklearn/linear_model/base.py
in predict(self, X)
    211         Returns predicted values.
    212         """
--> 213         return self._decision_function(X)
    214
    215     _preprocess_data = staticmethod(_preprocess_data)

/anaconda3/lib/python3.7/site-packages/sklearn/linear_model/coordina
te_descent.py in _decision_function(self, X)
    803         dense_output=True) + sel
f.intercept_
    804         else:
--> 805         return super(ElasticNet, self)._decision_functio
n(X)
    806
    807

/anaconda3/lib/python3.7/site-packages/sklearn/linear_model/base.py
in _decision_function(self, X)
    196         X = check_array(X, accept_sparse=['csr', 'csc', 'co
o'])
    197         return safe_sparse_dot(X, self.coef_.T,
--> 198                             dense_output=True) + self.int
ercept_
    199
    200     def predict(self, X):

/anaconda3/lib/python3.7/site-packages/sklearn/utils/extmath.py in s
afe_sparse_dot(a, b, dense_output)
    171         return ret
    172     else:
--> 173         return np.dot(a, b)
    174
    175

ValueError: shapes (1459,83) and (84,) not aligned: 83 (dim 1) != 84
(dim 0)

```


In [234]:

```
X_train_stack.shape,X_test_stack.shape,X_scaled_imputed.shape,y_log_imputed.shap
e
```

```
-----
-----
NameError                                Traceback (most recent call
last)
<ipython-input-234-8a9493e35d12> in <module>
----> 1 X_train_stack.shape,X_test_stack.shape,X_scaled_imputed.shap
e,y_log_imputed.shape
```

NameError: name 'X_train_stack' is not defined

In [235]:

```
X_train_add = np.hstack((X_scaled_imputed,X_train_stack))
X_test_add = np.hstack((test_X_scaled,X_test_stack))
X_train_add.shape,X_test_add.shape
```

```
-----
-----
NameError                                Traceback (most recent call
last)
<ipython-input-235-7fefb47b0469> in <module>
----> 1 X_train_add = np.hstack((X_scaled_imputed,X_train_stack))
      2 X_test_add = np.hstack((test_X_scaled,X_test_stack))
      3 X_train_add.shape,X_test_add.shape
```

NameError: name 'X_train_stack' is not defined

In [236]:

```
score = rmse_cv(stack_model,X_train_add,y_log_imputed)
print(score.mean())
```

```
-----
-----
NameError                                Traceback (most recent call
last)
<ipython-input-236-93c8db269170> in <module>
----> 1 score = rmse_cv(stack_model,X_train_add,y_log_imputed)
      2 print(score.mean())
```

NameError: name 'X_train_add' is not defined

In [237]:

```
stack_model = stacking(mod=[lasso,ridge,svr,ker,ela,bay],meta_model=ker)
stack_model.fit(X_scaled_imputed,y_log_imputed)
```

Out[237]:

```
stacking(meta_model=KernelRidge(alpha=0.15, coef0=0.9, degree=3, gamma=None, kernel='polynomial',
    kernel_params=None),
    mod=[Lasso(alpha=0.0005, copy_X=True, fit_intercept=True, max_iter=10000,
    normalize=False, positive=False, precompute=False, random_state=None,
    selection='cyclic', tol=0.0001, warm_start=False), Ridge(alpha=4
5, copy_X=True, fit_intercept=True, max_iter=10000,
    normalize=False, random_state=None, lambda_1=1e-06, lambda_2=
1e-06, n_iter=300,
    normalize=False, tol=0.001, verbose=False)])
```

In [238]:

```
print (stack_model.score(X_scaled_imputed,y_log_imputed))
```

0.9442012768202671

In [239]:

```
p_pred = np.expml(stack_model.predict(test_X_scaled))  
p_pred
```

```

-----
-----
ValueError                                Traceback (most recent call
last)
<ipython-input-239-2d2db05ea870> in <module>
----> 1 p_pred = np.expml(stack_model.predict(test_X_scaled))
      2 p_pred

<ipython-input-227-2fdd33de90dc> in predict(self, X)
      21     def predict(self,X):
      22         whole_test = np.column_stack([np.column_stack(model.
predict(X) for model in single_model).mean(axis=1)
----> 23                                     for single_model in se
lf.saved_model])
      24         return self.meta_model.predict(whole_test)
      25

<ipython-input-227-2fdd33de90dc> in <listcomp>(.0)
      21     def predict(self,X):
      22         whole_test = np.column_stack([np.column_stack(model.
predict(X) for model in single_model).mean(axis=1)
----> 23                                     for single_model in se
lf.saved_model])
      24         return self.meta_model.predict(whole_test)
      25

/anaconda3/lib/python3.7/site-packages/numpy/lib/shape_base.py in co
lumn_stack(tup)
      587     """
      588     arrays = []
--> 589     for v in tup:
      590         arr = array(v, copy=False, subok=True)
      591         if arr.ndim < 2:

<ipython-input-227-2fdd33de90dc> in <genexpr>(.0)
      20
      21     def predict(self,X):
----> 22         whole_test = np.column_stack([np.column_stack(model.
predict(X) for model in single_model).mean(axis=1)
      23                                     for single_model in se
lf.saved_model])
      24         return self.meta_model.predict(whole_test)

/anaconda3/lib/python3.7/site-packages/sklearn/linear_model/base.py
in predict(self, X)
      211         Returns predicted values.
      212         """
--> 213         return self._decision_function(X)
      214
      215         _preprocess_data = staticmethod(_preprocess_data)

/anaconda3/lib/python3.7/site-packages/sklearn/linear_model/coordina
te_descent.py in _decision_function(self, X)
      803         dense_output=True) + sel
f.intercept_
      804     else:
--> 805         return super(ElasticNet, self)._decision_functio
n(X)
      806
      807

```

```

/anaconda3/lib/python3.7/site-packages/sklearn/linear_model/base.py
in _decision_function(self, X)
196         X = check_array(X, accept_sparse=['csr', 'csc', 'co
o'])
197         return safe_sparse_dot(X, self.coef_.T,
--> 198                             dense_output=True) + self.int
ercept_
199
200     def predict(self, X):

/anaconda3/lib/python3.7/site-packages/sklearn/utils/extmath.py in s
afe_sparse_dot(a, b, dense_output)
171         return ret
172     else:
--> 173         return np.dot(a, b)
174
175

```

ValueError: shapes (1459,83) and (84,) not aligned: 83 (dim 1) != 84 (dim 0)

In [240]:

```

final_result = pd.DataFrame({'Id':test.Id,'SalePrice':p_pred})
final_result.to_csv("final_result_submission.csv",index= False)

```

```

-----
-----
NameError                                Traceback (most recent cal
l last)
<ipython-input-240-4673ae51e8e0> in <module>
----> 1 final_result = pd.DataFrame({'Id':test.Id,'SalePrice':p_pred
})
      2 final_result.to_csv("final_result_submission.csv",index= Fal
se)

```

NameError: name 'p_pred' is not defined

In []: