```
In [1]: import pandas as pd
        import numpy as np
        import seaborn as sns
        import matplotlib
        import xgboost as xgb
        from sklearn.preprocessing import StandardScaler
        import matplotlib.pyplot as plt
        from scipy.stats import skew
        from scipy.stats.stats import pearsonr
        %config InlineBackend.figure_format = 'png' #set 'png' here when working
        %matplotlib inline
        from sklearn.model_selection import KFold, cross_val_score
```

In [38]: train = pd.read\_csv(r'train.csv') test = pd.read\_csv(r'test.csv')

In [39]: train

Out[39]:

		Id	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	Lan
	0	1	60	RL	65.0	8450	Pave	NaN	Reg	
	1	2	20	RL	80.0	9600	Pave	NaN	Reg	
	2	3	60	RL	68.0	11250	Pave	NaN	IR1	
	3	4	70	RL	60.0	9550	Pave	NaN	IR1	
	4	5	60	RL	84.0	14260	Pave	NaN	IR1	
	•••									
	1455	1456	60	RL	62.0	7917	Pave	NaN	Reg	
	1456	1457	20	RL	85.0	13175	Pave	NaN	Reg	
	1457	1458	70	RL	66.0	9042	Pave	NaN	Reg	
	1458	1459	20	RL	68.0	9717	Pave	NaN	Reg	
	1459	1460	20	RL	75.0	9937	Pave	NaN	Reg	

1460 rows × 81 columns

In [40]: test

Out[40]:		Id	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	Lan
	0	1461	20	RH	80.0	11622	Pave	NaN	Reg	
	1	1462	20	RL	81.0	14267	Pave	NaN	IR1	
	2	1463	60	RL	74.0	13830	Pave	NaN	IR1	
	3	1464	60	RL	78.0	9978	Pave	NaN	IR1	
	4	1465	120	RL	43.0	5005	Pave	NaN	IR1	
	•••									
	1454	2915	160	RM	21.0	1936	Pave	NaN	Reg	
	1455	2916	160	RM	21.0	1894	Pave	NaN	Reg	
	1456	2917	20	RL	160.0	20000	Pave	NaN	Reg	
	1457	2918	85	RL	62.0	10441	Pave	NaN	Reg	
	1458	2919	60	RL	74.0	9627	Pave	NaN	Reg	

1459 rows × 80 columns

In [41]: all\_data = pd.concat((train.loc[:,'MSSubClass':'SaleCondition'],test.loc[

In [42]: all\_data

Out[42]: MSSubClass MSZoning LotFrontage LotArea Street Alley LotShape LandContc 0 60 RL 65.0 8450 Pave NaN Reg 80.0 9600 20 RL Pave NaN Reg 2 RL68.0 60 11250 Pave NaN IR1 3 70 RL60.0 9550 Pave NaN IR1 4 60 RL84.0 IR1 14260 Pave NaN 21.0 1454 160 RM1936 Pave NaN Reg 1455 160 RM 21.0 1894 Pave NaN Reg 1456 20000 20 RL160.0 Pave NaN Reg 1457 RL62.0 10441 Pave 85 NaN Reg 1458 60 RL 74.0 9627 Pave NaN Reg

2919 rows × 79 columns

```
In [43]: # model regression . fit
# model lasso . fit
```

```
In [44]: #features more normal
           matplotlib.rcParams['figure.figsize'] = (12.0, 6.0)
           prices = pd.DataFrame({"price":train["SalePrice"], "log(price + 1)":np.lo
           prices.hist()
Out[44]: array([[<AxesSubplot:title={'center':'price'}>,
                    <AxesSubplot:title={'center':'log(price + 1)'}>]], dtype=object)
                              price
                                                                         log(price + 1)
           700
                                                         400
           600
           500
                                                         300
           400
                                                         200
           300
           200
                                                         100
           100
                100000200000300000400000500000600000700000
                                                            10.5
                                                                  11.0
                                                                       11.5
                                                                            12.0
                                                                                 12.5
                                                                                       13.0
                                                                                            13.5
```

```
In [45]: train["SalePrice"] = np.log1p(train["SalePrice"])
```

```
In [46]: train["SalePrice"] #define target
```

```
Out[46]: 0
                  12.247699
          1
                  12.109016
          2
                  12.317171
          3
                  11.849405
          4
                  12,429220
                     . . .
          1455
                  12.072547
          1456
                  12.254868
                  12.493133
          1457
          1458
                  11.864469
          1459
                  11.901590
          Name: SalePrice, Length: 1460, dtype: float64
```

```
In [47]: numeric_feats = all_data.dtypes[all_data.dtypes != "object"].index
```

In [48]: numeric\_feats

```
Out[48]: Index(['MSSubClass', 'LotFrontage', 'LotArea', 'OverallQual', 'OverallCon
         d',
                 'YearBuilt', 'YearRemodAdd', 'MasVnrArea', 'BsmtFinSF1', 'BsmtFinS
         F2',
                'BsmtUnfSF', 'TotalBsmtSF', '1stFlrSF', '2ndFlrSF', 'LowQualFinSF'
                 'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath', 'HalfBath
                 'BedroomAbvGr', 'KitchenAbvGr', 'TotRmsAbvGrd', 'Fireplaces',
                 'GarageYrBlt', 'GarageCars', 'GarageArea', 'WoodDeckSF', 'OpenPorc
         hSF',
                'EnclosedPorch', '3SsnPorch', 'ScreenPorch', 'PoolArea', 'MiscVal'
                 'MoSold', 'YrSold'],
               dtype='object')
         skewed_feats = train[numeric_feats].apply(lambda x: skew(x.dropna())) #co
In [49]:
         skewed_feats = skewed_feats[skewed_feats > 0.75]
         skewed_feats = skewed_feats.index
         all_data[skewed_feats] = np.log1p(all_data[skewed_feats])
In [50]: all data[skewed feats]
Out [50]:
```

:		MSSubClass	LotFrontage	LotArea	MasVnrArea	BsmtFinSF1	BsmtFinSF2	Bsmtl
	0	4.110874	4.189655	9.042040	5.283204	6.561031	0.0	5.01
	1	3.044522	4.394449	9.169623	0.000000	6.886532	0.0	5.65
	2	4.110874	4.234107	9.328212	5.093750	6.188264	0.0	6.07
	3	4.262680	4.110874	9.164401	0.000000	5.379897	0.0	6.29
	4	4.110874	4.442651	9.565284	5.860786	6.486161	0.0	6.19
1	454	5.081404	3.091042	7.568896	0.000000	0.000000	0.0	6.30
1	455	5.081404	3.091042	7.546974	0.000000	5.533389	0.0	5.68
1	456	3.044522	5.081404	9.903538	0.000000	7.110696	0.0	0.00
1	457	4.454347	4.143135	9.253591	0.000000	5.823046	0.0	6.35
1	458	4.110874	4.317488	9.172431	4.553877	6.632002	0.0	5.47

2919 rows × 21 columns

```
In [51]: all_data = pd.get_dummies(all_data)
In [52]: #filling NA's with the mean of the column:
    all_data = all_data.fillna(all_data.mean())
```

```
In [53]: #creating matrices for sklearn:
X_train = all_data[:train.shape[0]]
X_test = all_data[train.shape[0]:]
y = train.SalePrice
```

In [54]: all\_data

Out[54]:		MSSubClass	LotFrontage	LotArea	OverallQual	OverallCond	YearBuilt	YearRem
	0	4.110874	4.189655	9.042040	7	5	2003	
	1	3.044522	4.394449	9.169623	6	8	1976	
	2	4.110874	4.234107	9.328212	7	5	2001	
	3	4.262680	4.110874	9.164401	7	5	1915	
	4	4.110874	4.442651	9.565284	8	5	2000	
	•••							
	1454	5.081404	3.091042	7.568896	4	7	1970	
	1455	5.081404	3.091042	7.546974	4	5	1970	
	1456	3.044522	5.081404	9.903538	5	7	1960	
	1457	4.454347	4.143135	9.253591	5	5	1992	
	1458	4.110874	4.317488	9.172431	7	5	1993	

2919 rows × 288 columns

# RMSE Value - Ridge

```
In [59]: cv_ridge
```

Out [59]: [0.13777538277187865]

## **RMSE Value - Lasso**

```
In [60]: model_lasso = LassoCV(alphas = [0.1]).fit(X_train, y)
In [61]: rmse_cv(model_lasso).mean()
Out[61]: 0.20921930047608214
```

## Obseerved that ridge performed better for alpha

```
Optimizing Alpha
In [62]: # For Ridge
         alphas=np.arange(5, 15, 1).tolist() #running on alpha from 5 to 15 at a 1
         cv_ridge = [rmse_cv(Ridge(alpha =alpha)).mean()
                     for alpha in alphas]
In [63]: min_index=cv_ridge.index(min(cv_ridge)) #min cv see for index
         cv_ridge[min_index] #this is minimum cv score we could achieve
Out[63]: 0.12733734668670776
In [64]: alphas[min_index] #min CV achieved at alpha = 10
Out[64]: 10
In [65]: # For Lasso
         alphas=np.arange(0, 0.001, 0.0001).tolist() #running on alpha from 0 to 0
         lasso cv=[]
         alpha_curr=[]
         for i in alphas:
             model_lasso_cv = rmse_cv(LassoCV(alphas=[i]).fit(X_train, y)).mean()
             alpha curr.append(i)
             lasso cv.append(model lasso cv)
         /Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
         r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
         alpha=0 may lead to unexpected results and is discouraged.
           model = cd_fast.enet_coordinate_descent_gram(
         /Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
         r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
         converge. You might want to increase the number of iterations. Duality ga
         p: 4.859170575181508, tolerance: 0.018912592760396085
           model = cd_fast.enet_coordinate_descent_gram(
         /Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
         r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
```

alpha=0 may lead to unexpected results and is discouraged.

```
model = cd fast.enet coordinate descent gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 4.257882622560601, tolerance: 0.01800219138548883
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r model/ coordinate descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
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/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 4.095357955603021, tolerance: 0.018373605848561597
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
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/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 4.829918807482244, tolerance: 0.019008081403702633
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 4.703140526782022, tolerance: 0.01881061188370518
 model = cd fast.enet coordinate descent gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:1714: UserWarning: With alpha=0, this algo
rithm does not converge well. You are advised to use the LinearRegression
estimator
 model.fit(X, y)
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:647: UserWarning: Coordinate descent with
no regularization may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:647: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations, check the
scale of the features or consider increasing regularisation. Duality gap:
6.153e+00, tolerance: 2.328e-02 Linear regression models with null weight
for the l1 regularization term are more efficiently fitted using one of t
he solvers implemented in sklearn.linear_model.Ridge/RidgeCV instead.
 model = cd fast.enet coordinate descent(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
```

```
model = cd fast.enet coordinate descent gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.4079296266881443, tolerance: 0.014608206095799353
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
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 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.3428077749517797, tolerance: 0.014601004078052359
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.4383502763372746, tolerance: 0.015531757542978097
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
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/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r model/ coordinate descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.827884824674669, tolerance: 0.015448874096282393
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
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/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.662900988799123, tolerance: 0.015453647710308875
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:1714: UserWarning: With alpha=0, this algo
rithm does not converge well. You are advised to use the LinearRegression
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/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:647: UserWarning: Coordinate descent with
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/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:647: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations, check the
```

```
scale of the features or consider increasing regularisation. Duality gap:
4.859e+00, tolerance: 1.891e-02 Linear regression models with null weight
for the l1 regularization term are more efficiently fitted using one of t
he solvers implemented in sklearn.linear_model.Ridge/RidgeCV instead.
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r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.2535264274607414, tolerance: 0.014421032323759136
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
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/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 2.9705961284435602, tolerance: 0.013878761882709149
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
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r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.086940408869907, tolerance: 0.014621128652612829
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
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/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.3486242125725028, tolerance: 0.01453514652569633
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
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/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.0780428069976438, tolerance: 0.014544304735976838
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:1714: UserWarning: With alpha=0, this algo
rithm does not converge well. You are advised to use the LinearRegression
estimator
```

```
model.fit(X, y)
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:647: UserWarning: Coordinate descent with
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/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:647: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations, check the
scale of the features or consider increasing regularisation. Duality gap:
4.258e+00, tolerance: 1.800e-02 Linear regression models with null weight
for the l1 regularization term are more efficiently fitted using one of t
he solvers implemented in sklearn.linear_model.Ridge/RidgeCV instead.
 model = cd_fast.enet_coordinate_descent(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
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 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.078947772285062, tolerance: 0.014791078401076713
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd fast.enet coordinate descent gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.01427064223833, tolerance: 0.014455328041034693
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 2.952829479429994, tolerance: 0.014416323383021182
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.0683216215797984, tolerance: 0.014911027581501767
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
```

```
r model/ coordinate descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 2.977869365490676, tolerance: 0.014914232105412309
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:1714: UserWarning: With alpha=0, this algo
rithm does not converge well. You are advised to use the LinearRegression
estimator
 model.fit(X, y)
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:647: UserWarning: Coordinate descent with
no regularization may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:647: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations, check the
scale of the features or consider increasing regularisation. Duality gap:
4.095e+00, tolerance: 1.837e-02 Linear regression models with null weight
for the l1 regularization term are more efficiently fitted using one of t
he solvers implemented in sklearn.linear_model.Ridge/RidgeCV instead.
 model = cd_fast.enet_coordinate_descent(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r model/ coordinate descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.662546611421817, tolerance: 0.01542865087978492
 model = cd fast.enet coordinate descent gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd fast.enet coordinate descent gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.7217488118484425, tolerance: 0.015089979013565014
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.1000647992446773, tolerance: 0.014491631155835854
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
```

```
converge. You might want to increase the number of iterations. Duality ga
p: 3.2917687071829107, tolerance: 0.01546918576152043
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r model/ coordinate descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.715077253629545, tolerance: 0.01555239160345648
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:1714: UserWarning: With alpha=0, this algo
rithm does not converge well. You are advised to use the LinearRegression
estimator
 model.fit(X, y)
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:647: UserWarning: Coordinate descent with
no regularization may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:647: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations, check the
scale of the features or consider increasing regularisation. Duality gap:
4.830e+00, tolerance: 1.901e-02 Linear regression models with null weight
for the l1 regularization term are more efficiently fitted using one of t
he solvers implemented in sklearn.linear_model.Ridge/RidgeCV instead.
 model = cd_fast.enet_coordinate_descent(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.5948797841396924, tolerance: 0.015228754219030277
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.507123236427276, tolerance: 0.014892726351484208
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
```

p: 2.8620210884815833, tolerance: 0.014286517661843702

```
model = cd fast.enet coordinate descent gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.3364463600519514, tolerance: 0.01528597962708133
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: UserWarning: Coordinate descent with
alpha=0 may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 3.5593462547941215, tolerance: 0.015536468891600251
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:1714: UserWarning: With alpha=0, this algo
rithm does not converge well. You are advised to use the LinearRegression
estimator
 model.fit(X, y)
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:647: UserWarning: Coordinate descent with
no regularization may lead to unexpected results and is discouraged.
 model = cd_fast.enet_coordinate_descent(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:647: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations, check the
scale of the features or consider increasing regularisation. Duality gap:
4.703e+00, tolerance: 1.881e-02 Linear regression models with null weight
for the l1 regularization term are more efficiently fitted using one of t
he solvers implemented in sklearn.linear_model.Ridge/RidgeCV instead.
 model = cd_fast.enet_coordinate_descent(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 0.5392151380338897, tolerance: 0.01800219138548883
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:647: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations, check the
scale of the features or consider increasing regularisation. Duality gap:
9.122e-01, tolerance: 2.328e-02
 model = cd_fast.enet_coordinate_descent(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
r_model/_coordinate_descent.py:633: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations. Duality ga
p: 0.769086169657208, tolerance: 0.015531757542978097
 model = cd_fast.enet_coordinate_descent_gram(
/Users/vipulsahni/opt/anaconda3/lib/python3.9/site-packages/sklearn/linea
```

```
r_model/_coordinate_descent.py:647: ConvergenceWarning: Objective did not
converge. You might want to increase the number of iterations, check the
scale of the features or consider increasing regularisation. Duality gap:
5.392e-01, tolerance: 1.800e-02
  model = cd_fast.enet_coordinate_descent(
```

In [66]: min\_index=lasso\_cv.index(min(lasso\_cv)) #min cv see for index
lasso\_cv[min\_index] #this is minimum cv score we could achieve with lasso

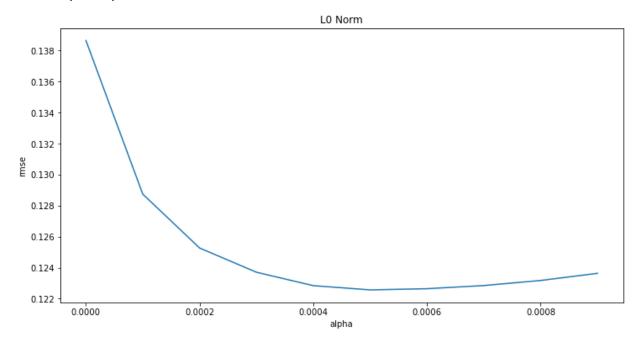
Out[66]: 0.12256735885048128

In [67]: alpha\_curr[min\_index] #min CV achieved at alpha = 0.0005

Out[67]: 0.0005

```
In [68]: #L0 Norm of the coefficients that lasso produces
  lasso_cv = pd.Series(lasso_cv, index = alphas)
  lasso_cv.plot(title = "L0 Norm")
  plt.xlabel("alpha")
  plt.ylabel("rmse")
```

Out[68]: Text(0, 0.5, 'rmse')



For Ridge, we were able to optimize alpha = 10, to get RMSE CV score of 0.127

For Lasso, alpha = 0.0005 to get RMSE CV score of 0.1225 - lower than Ridge

```
In [29]: #Now checking predictions - Lasso
model_lasso = LassoCV(alphas = [0.0005]).fit(X_train, y)
model_lasso.predict(X_test)
```

```
Out[29]: array([11.69490559, 11.92823243, 12.10183291, ..., 12.03779924,
                11.68640318, 12.33887195])
In [30]:
         #Now checking predictions - Ridge
         model_ridge = Ridge(alpha=0.1).fit(X_train, y) #0.1 because submitting to
In [31]: pred_ridge=model_ridge.predict(X_test)
In [32]: pred ridge df=pd.DataFrame(np.expm1(pred ridge))
In [33]: X_test['SalePrice']=pred_ridge_df[0]
         X test['Id'] = X test.index
         X test['Id'] += 1461
         /var/folders/6p/fvv3r7rj335gs5y3bm1f750m0000gn/T/ipykernel_18452/25514226
         27.py:1: SettingWithCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row_indexer,col_indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas-do
         cs/stable/user quide/indexing.html#returning-a-view-versus-a-copy
           X_test['SalePrice']=pred_ridge_df[0]
         /var/folders/6p/fvv3r7rj335gs5y3bm1f750m0000gn/T/ipykernel_18452/25514226
         27.py:2: SettingWithCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row_indexer,col_indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas-do
         cs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
           X_test['Id'] = X_test.index
         /var/folders/6p/fvv3r7rj335gs5y3bm1f750m0000gn/T/ipykernel_18452/25514226
         27.py:3: SettingWithCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row_indexer,col_indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas-do
         cs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
           X_test['Id'] += 1461
```

#### Submission score=0.135

```
In [34]: X_test[['Id','SalePrice']].to_csv('submission.csv') #0.135
```

### Add features to ridge regression | Ensembling and Stacking

```
In [69]: model_ridge = Ridge(alpha=10).fit(X_train, y)
    model_lasso = LassoCV(alphas = [0.0005]).fit(X_train, y)
    X_train['Lasso_val'] = model_lasso.predict(X_train)
    X_train['Ridge_val'] = model_ridge.predict(X_train.loc[:,'MSSubClass':'Sa
```

/var/folders/6p/fvv3r7rj335gs5y3bm1f750m0000gn/T/ipykernel\_18452/33504323
49.py:3: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy

X\_train['Lasso\_val'] = model\_lasso.predict(X\_train)

/var/folders/6p/fvv3r7rj335gs5y3bm1f750m0000gn/T/ipykernel\_18452/33504323
49.py:4: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy

X\_train['Ridge\_val'] = model\_ridge.predict(X\_train.loc[:,'MSSubClass':'
SaleCondition\_Partial'])

In [70]: X\_train

Out[70]:		MSSubClass	LotFrontage	LotArea	OverallQual	OverallCond	YearBuilt	YearRem
	0	4.110874	4.189655	9.042040	7	5	2003	
	1	3.044522	4.394449	9.169623	6	8	1976	
	2	4.110874	4.234107	9.328212	7	5	2001	
	3	4.262680	4.110874	9.164401	7	5	1915	
	4	4.110874	4.442651	9.565284	8	5	2000	
	•••							
	1455	4.110874	4.143135	8.976894	6	5	1999	
	1456	3.044522	4.454347	9.486152	6	6	1978	
	1457	4.262680	4.204693	9.109746	7	9	1941	
	1458	3.044522	4.234107	9.181735	5	6	1950	
	1459	3.044522	4.330733	9.204121	5	6	1965	

1460 rows × 290 columns

In [165... X\_train.head()

Out[165]:		MSSubClass	LotFrontage	LotArea	OverallQual	OverallCond	YearBuilt	YearRemod
	0	4.110874	4.189655	9.042040	7	5	2003	2
	1	3.044522	4.394449	9.169623	6	8	1976	1
	2	4.110874	4.234107	9.328212	7	5	2001	2
	3	4.262680	4.110874	9.164401	7	5	1915	1
	4	4.110874	4.442651	9.565284	8	5	2000	2

5 rows × 290 columns

```
In [168... # Check Ridge train predictions
model_ridge = Ridge(alpha=10).fit(X_train, y)
In [175... rmse_cv(model_ridge).mean() #RMSE is improved, earlier it was 0.127
```

Out[175]: 0.12267213549556055

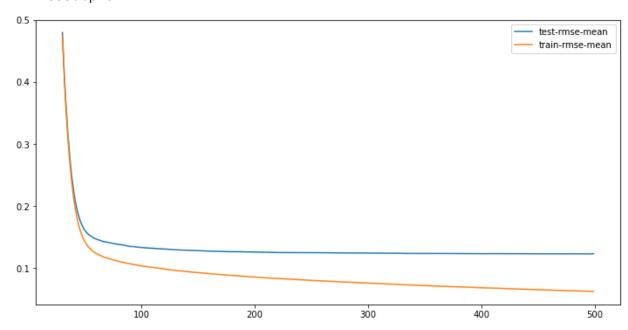
### **XG** Boost

```
In [198... dtrain = xgb.DMatrix(X_train.loc[:,'MSSubClass':'SaleCondition_Partial'],
    dtest = xgb.DMatrix(X_test)

    params = {"max_depth":2, "eta":0.1}
    model = xgb.cv(params, dtrain, num_boost_round=500, early_stopping_round

In [199... model.loc[30:,["test-rmse-mean", "train-rmse-mean"]].plot()
```

#### Out[199]: <AxesSubplot:>



```
model_xgb = xgb.XGBRegressor(n_estimators=360, max_depth=2, learning_rate
          model xgb.fit(X train.loc[:,'MSSubClass':'SaleCondition Partial'], y)
Out[200]: XGBRegressor(base_score=0.5, booster='gbtree', callbacks=None,
                        colsample_bylevel=1, colsample_bynode=1, colsample_bytree=1
                        early_stopping_rounds=None, enable_categorical=False,
                        eval metric=None, gamma=0, gpu id=-1, grow policy='depthwis
          e',
                        importance type=None, interaction constraints='',
                        learning_rate=0.1, max_bin=256, max_cat_to_onehot=4,
                        max_delta_step=0, max_depth=2, max_leaves=0, min_child_weig
          ht=1,
                        missing=nan, monotone_constraints='()', n_estimators=360, n
          jobs=0,
                        num_parallel_tree=1, predictor='auto', random_state=0, reg_
          alpha=0,
                        reg_lambda=1, ...)
         model_lasso = LassoCV(alphas = [0.0005]).fit(X_train.loc[:,'MSSubClass':'
In [204...
          xgb_preds = np.expm1(model_xgb.predict(X_test))
          lasso_preds = np.expm1(model_lasso.predict(X_test))
          predictions = pd.DataFrame({"xgb":xgb_preds, "lasso":lasso_preds})
In [205...
          predictions.plot(x = "xgb", y = "lasso", kind = "scatter")
Out[205]: <AxesSubplot:xlabel='xgb', ylabel='lasso'>
           600000
           500000
           400000
           300000
           200000
```

Using only xgb\_preds, model is overfitting. On training RMSE is  $\sim$ 0.1, but on test it is 0.13

300000

400000

```
In [212... rmse_cv(model_xgb).mean() #RMSE is improved, earlier it was 0.127
```

200000

Out[212]: 0.10913110265345634

100000

100000

500000

```
In [206... preds = 0.7*lasso_preds + 0.3*xgb_preds
In [214... solution = pd.DataFrame({"id":test.Id, "SalePrice":preds})
# solution.to_csv("ridge_sol.csv", index = False)
In [215... solution.to_csv('submission.csv') #Score: 0.12299
```

## Improve upon this XGboost model

### **Feature Engineering**

```
In [217... | all_data['GarageYrBltn'] = abs(all_data['YrSold'] - all_data['GarageYrBlt
         all data['YearRemodAddn'] = abs(all data['YrSold'] - all data['YearRemodA
         all data['YearBuiltn'] = abs(all data['YrSold'] - all data['YearBuilt'])
         all data['SF'] = all data['1stFlrSF']+all data['2ndFlrSF']+all data['Tota
         /var/folders/6p/fvv3r7rj335gs5y3bm1f750m0000gn/T/ipykernel_86807/35110614
         70.py:1: PerformanceWarning: DataFrame is highly fragmented. This is usu
         ally the result of calling `frame.insert` many times, which has poor perf
         ormance. Consider joining all columns at once using pd.concat(axis=1) in
         stead. To get a de-fragmented frame, use `newframe = frame.copy()`
           all_data['GarageYrBltn'] = abs(all_data['YrSold'] - all_data['GarageYrB
         lt'])
         /var/folders/6p/fvv3r7rj335gs5y3bm1f750m0000gn/T/ipykernel_86807/35110614
         70.py:2: PerformanceWarning: DataFrame is highly fragmented. This is usu
         ally the result of calling `frame.insert` many times, which has poor perf
                   Consider joining all columns at once using pd.concat(axis=1) in
         stead. To get a de-fragmented frame, use `newframe = frame.copy()`
           all_data['YearRemodAddn'] = abs(all_data['YrSold'] - all_data['YearRemo
         dAdd'])
         /var/folders/6p/fvv3r7rj335qs5v3bm1f750m0000qn/T/ipykernel 86807/35110614
         70.py:3: PerformanceWarning: DataFrame is highly fragmented. This is usu
         ally the result of calling `frame.insert` many times, which has poor perf
         ormance. Consider joining all columns at once using pd.concat(axis=1) in
         stead. To get a de-fragmented frame, use `newframe = frame.copy()`
           all_data['YearBuiltn'] = abs(all_data['YrSold'] - all_data['YearBuilt']
         /var/folders/6p/fvv3r7rj335gs5y3bm1f750m0000gn/T/ipykernel_86807/35110614
         70.py:4: PerformanceWarning: DataFrame is highly fragmented. This is usu
         ally the result of calling `frame.insert` many times, which has poor perf
         ormance. Consider joining all columns at once using pd.concat(axis=1) in
         stead. To get a de-fragmented frame, use `newframe = frame.copy()`
           all_data['SF'] = all_data['1stFlrSF']+all_data['2ndFlrSF']+all_data['To
         talBsmtSF']+all_data['GrLivArea']+all_data['HalfBath']+all_data['FullBath
         ']
```

In [218... all\_data

Out[218]:		MSSubClass	LotFrontage	LotArea	OverallQual	OverallCond	YearBuilt	YearRer
	0	4.110874	4.189655	9.042040	7	5	2003	
	1	3.044522	4.394449	9.169623	6	8	1976	
	2	4.110874	4.234107	9.328212	7	5	2001	
	3	4.262680	4.110874	9.164401	7	5	1915	
	4	4.110874	4.442651	9.565284	8	5	2000	
	•••							
	1454	5.081404	3.091042	7.568896	4	7	1970	
	1455	5.081404	3.091042	7.546974	4	5	1970	
	1456	3.044522	5.081404	9.903538	5	7	1960	
	1457	4.454347	4.143135	9.253591	5	5	1992	
	1458	4.110874	4.317488	9.172431	7	5	1993	

2919 rows × 292 columns

In [224	all_data						
Out[224]:	MSSubClass	LotFrontage	LotArea	OverallQual	OverallCond	YearBuilt	Year

	MSSubClass	LotFrontage	LotArea	OverallQual	OverallCond	YearBuilt	YearR
0	0.419418	-0.020358	-0.103719	0.646183	-0.507284	1.046258	
1	-1.120845	0.619103	0.146544	-0.063185	2.188279	0.154764	-
2	0.419418	0.118440	0.457629	0.646183	-0.507284	0.980221	
3	0.638691	-0.266348	0.136301	0.646183	-0.507284	-1.859351	
4	0.419418	0.769612	0.922662	1.355551	-0.507284	0.947203	
•••							
1454	1.821276	-3.450727	-2.993401	-1.481920	1.289758	-0.043346	
1455	1.821276	-3.450727	-3.036401	-1.481920	-0.507284	-0.043346	
1456	-1.120845	2.764091	1.586172	-0.772552	1.289758	-0.373528	
1457	0.915540	-0.165615	0.311255	-0.772552	-0.507284	0.683057	
1458	0.419418	0.378796	0.152052	0.646183	-0.507284	0.716075	

2919 rows × 292 columns

```
In [225... df_train = all_data.iloc[:1460,:]
    df_test = all_data.iloc[1460:,:]
    final = pd.concat([df_train,y], axis = 1)
```

### In [226... final

Out[226]:		MSSubClass	LotFrontage	LotArea	OverallQual	OverallCond	YearBuilt	YearR
	0	0.419418	-0.020358	-0.103719	0.646183	-0.507284	1.046258	
	1	-1.120845	0.619103	0.146544	-0.063185	2.188279	0.154764	-
	2	0.419418	0.118440	0.457629	0.646183	-0.507284	0.980221	
	3	0.638691	-0.266348	0.136301	0.646183	-0.507284	-1.859351	-
	4	0.419418	0.769612	0.922662	1.355551	-0.507284	0.947203	
	•••							
	1455	0.419418	-0.165615	-0.231508	-0.063185	-0.507284	0.914184	
	1456	-1.120845	0.806133	0.767440	-0.063185	0.391237	0.220801	
	1457	0.638691	0.026597	0.029092	0.646183	3.086800	-1.000876	
	1458	-1.120845	0.118440	0.170303	-0.772552	0.391237	-0.703711	

1460 rows × 293 columns

-1.120845

1459

```
In [231... from sklearn.linear_model import BayesianRidge, HuberRegressor, Ridge, Or
from sklearn.ensemble import GradientBoostingRegressor
from catboost import CatBoostRegressor
```

0.214215 -0.772552

0.391237 -0.208437

0.420154

```
In [270... baseline_model = GradientBoostingRegressor()
   baseline_model.fit(df_train, y)
```

#### Out[270]: GradientBoostingRegressor()

```
In [287... br_params = {
          'n_iter': 304,
          'tol': 0.16864712769300896,
          'alpha_1': 5.589616542154059e-07,
          'alpha_2': 9.799343618469923,
          'lambda_1': 1.7735725582463822,
          'lambda_2': 3.616928181181732e-06
}

ridge_params = {
          'alpha': 10
}
```

```
In [295... models = {'gbr':GradientBoostingRegressor(),
                    'br':BayesianRidge(**br params),
                    'ridge':Ridge(**ridge params),
                    'catboost':CatBoostRegressor(loss_function='RMSE',n_estimators=
In [274... for name, model in models.items():
             model.fit(df_train, y)
In [275... results = {}
         kf = KFold(n_splits=10)
         for name, model in models.items():
              result = np.exp(np.sqrt(-cross_val_score(model, df_train, y, scoring=
              results[name] = result
In [276... for name, result in results.items():
             print("-----\n" + name)
             print(np.mean(result))
             print(np.std(result))
         gbr
         1.1334937793062636
         0.02041514019070494
         br
         1.136720968542448
         0.02708918872877562
         ridge
         1.1377247226437417
         0.025916296541081737
         catboost
         1.1237199881446363
         0.02114096999987419
In [291... y_pred = (
             0.0 * np.exp(models['gbr'].predict(df_test)) +
             0.0 * np.exp(models['br'].predict(df test)) +
             0.0 * np.exp(models['ridge'].predict(df_test))+
             1 * np.exp(models['catboost'].predict(df_test)))
In [292... solution = pd.DataFrame({"id":test.Id, "SalePrice":y pred})
         # solution.to_csv("ridge_sol.csv", index = False)
In [293... solution.to_csv('submission.csv',index=False) #Score: 0.12299
In [294... solution
```

Out[294]:		id	SalePrice
	0	1461	125953.825089
	1	1462	161424.964322
	2	1463	187908.591855
	3	1464	196804.694788
	4	1465	183519.623195
	•••		
	1454	2915	80918.325435
	1455	2916	78242.653140
	1456	2917	160394.307425
	1457	2918	113555.787541
	1458	2919	224401.442934

1459 rows × 2 columns

Using ensemble model with some feature engineering gives us better result at public leaderboard 0.124, than by using just xgboost 0.132