

```
In [3]: 1 import pandas as pd
2 import numpy as np
3 import matplotlib.pyplot as plt
4 import seaborn as sns
5
6 #prep
7 from sklearn.model_selection import train_test_split, GridSearchCV, RandomizedSearchCV
8 from sklearn.preprocessing import LabelEncoder, StandardScaler, MaxAbsScaler
9
10 #models
11 from sklearn.linear_model import LogisticRegression, LogisticRegressionCV
12 from sklearn.tree import DecisionTreeRegressor
13 from sklearn.ensemble import RandomForestRegressor
14
15 #validation libraries
16 #from sklearn.cross_validation import KFold, StratifiedKFold
17 from IPython.display import display
18 from sklearn import metrics
19
```

```
In [4]: 1 df_train = pd.read_csv('train.csv')
2 #df_test = pd.read_csv('test.csv')
```

```
In [5]: 1 #frames = [df_train, df_test]
2 #result = pd.concat(frames, sort = True)
```

```
In [6]: 1 df_train.head()
```

```
Out[6]:
```

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

5 rows x 81 columns

```
In [7]: 1 df_train['MSZoning'].value_counts() # categorical values
```

```
Out[7]: RL      1151
RM        218
FV         65
RH         16
C (all)    10
Name: MSZoning, dtype: int64
```

```
In [70]: 1 #plt.scatter_plot()
```

```
In [9]: 1 df_train['MSZoning'].isnull()
```

```
Out[9]: 0      False
        1      False
        2      False
        3      False
        4      False
        5      False
        6      False
        7      False
        8      False
        9      False
       10      False
       11      False
       12      False
       13      False
       14      False
       15      False
       16      False
       17      False
       18      False
       19      False
       20      False
       21      False
       22      False
       23      False
       24      False
       25      False
       26      False
       27      False
       28      False
       29      False
       ...
     1430      False
     1431      False
     1432      False
     1433      False
     1434      False
     1435      False
     1436      False
     1437      False
     1438      False
     1439      False
     1440      False
     1441      False
     1442      False
     1443      False
     1444      False
     1445      False
     1446      False
     1447      False
     1448      False
     1449      False
     1450      False
     1451      False
     1452      False
     1453      False
```

```
1454    False
1455    False
1456    False
1457    False
1458    False
1459    False
Name: MSZoning, Length: 1460, dtype: bool
```

```
In [10]: 1 df_train.shape
```

```
Out[10]: (1460, 81)
```

```
In [11]: 1 df_train.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1460 entries, 0 to 1459
Data columns (total 81 columns):
Id                1460 non-null int64
MSSubClass        1460 non-null int64
MSZoning          1460 non-null object
LotFrontage       1201 non-null float64
LotArea           1460 non-null int64
Street            1460 non-null object
Alley             91 non-null object
LotShape          1460 non-null object
LandContour       1460 non-null object
Utilities         1460 non-null object
LotConfig         1460 non-null object
LandSlope         1460 non-null object
Neighborhood      1460 non-null object
Condition1        1460 non-null object
Condition2        1460 non-null object
BldgType          1460 non-null object
HouseStyle        1460 non-null object
OverallQual       1460 non-null int64
OverallCond       1460 non-null int64
YearBuilt         1460 non-null int64
YearRemodAdd      1460 non-null int64
RoofStyle         1460 non-null object
RoofMatl          1460 non-null object
Exterior1st       1460 non-null object
Exterior2nd       1460 non-null object
MasVnrType        1452 non-null object
MasVnrArea        1452 non-null float64
ExterQual         1460 non-null object
ExterCond         1460 non-null object
Foundation        1460 non-null object
BsmtQual          1423 non-null object
BsmtCond          1423 non-null object
BsmtExposure      1422 non-null object
BsmtFinType1      1423 non-null object
BsmtFinSF1        1460 non-null int64
BsmtFinType2      1422 non-null object
BsmtFinSF2        1460 non-null int64
BsmtUnfSF         1460 non-null int64
TotalBsmtSF       1460 non-null int64
Heating           1460 non-null object
HeatingQC         1460 non-null object
CentralAir        1460 non-null object
Electrical        1459 non-null object
1stFlrSF          1460 non-null int64
2ndFlrSF          1460 non-null int64
LowQualFinSF      1460 non-null int64
GrLivArea         1460 non-null int64
BsmtFullBath      1460 non-null int64
BsmtHalfBath      1460 non-null int64
FullBath          1460 non-null int64
HalfBath          1460 non-null int64
BedroomAbvGr      1460 non-null int64
```

```
KitchenAbvGr      1460 non-null int64
KitchenQual       1460 non-null object
TotRmsAbvGrd      1460 non-null int64
Functional        1460 non-null object
Fireplaces        1460 non-null int64
FireplaceQu       770 non-null object
GarageType        1379 non-null object
GarageYrBltd      1379 non-null float64
GarageFinish      1379 non-null object
GarageCars        1460 non-null int64
GarageArea        1460 non-null int64
GarageQual        1379 non-null object
GarageCond        1379 non-null object
PavedDrive        1460 non-null object
WoodDeckSF        1460 non-null int64
OpenPorchSF       1460 non-null int64
EnclosedPorch     1460 non-null int64
3SsnPorch         1460 non-null int64
ScreenPorch       1460 non-null int64
PoolArea          1460 non-null int64
PoolQC            7 non-null object
Fence             281 non-null object
MiscFeature       54 non-null object
MiscVal           1460 non-null int64
MoSold            1460 non-null int64
YrSold            1460 non-null int64
SaleType          1460 non-null object
SaleCondition     1460 non-null object
SalePrice         1460 non-null int64
dtypes: float64(3), int64(35), object(43)
memory usage: 924.0+ KB
```

```
In [12]: 1 df_train.columns
```

```
Out[12]: Index(['Id', 'MSSubClass', 'MSZoning', 'LotFrontage', 'LotArea', 'Street',
               'Alley', 'LotShape', 'LandContour', 'Utilities', 'LotConfig',
               'LandSlope', 'Neighborhood', 'Condition1', 'Condition2', 'BldgType',
               'HouseStyle', 'OverallQual', 'OverallCond', 'YearBuilt', 'YearRemodAdd',
               'RoofStyle', 'RoofMatl', 'Exterior1st', 'Exterior2nd', 'MasVnrType',
               'MasVnrArea', 'ExterQual', 'ExterCond', 'Foundation', 'BsmtQual',
               'BsmtCond', 'BsmtExposure', 'BsmtFinType1', 'BsmtFinSF1',
               'BsmtFinType2', 'BsmtFinSF2', 'BsmtUnfSF', 'TotalBsmtSF', 'Heating',
               'HeatingQC', 'CentralAir', 'Electrical', '1stFlrSF', '2ndFlrSF',
               'LowQualFinSF', 'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath',
               'HalfBath', 'BedroomAbvGr', 'KitchenAbvGr', 'KitchenQual',
               'TotRmsAbvGrd', 'Functional', 'Fireplaces', 'FireplaceQu', 'GarageType',
               'GarageYrBlt', 'GarageFinish', 'GarageCars', 'GarageArea', 'GarageQual',
               'GarageCond', 'PavedDrive', 'WoodDeckSF', 'OpenPorchSF',
               'EnclosedPorch', '3SsnPorch', 'ScreenPorch', 'PoolArea', 'PoolQC',
               'Fence', 'MiscFeature', 'MiscVal', 'MoSold', 'YrSold', 'SaleType',
               'SaleCondition', 'SalePrice'],
              dtype='object')
```

```
In [13]: 1 len(df_train.columns)
```

```
Out[13]: 81
```

```
In [14]: 1 for x,y in zip (df_train.dtypes.index, df_train.dtypes):
          2     print(" {} || {} " .format(x,y))
```

```
Id || int64
MSSubClass || int64
MSZoning || object
LotFrontage || float64
LotArea || int64
Street || object
Alley || object
LotShape || object
LandContour || object
Utilities || object
LotConfig || object
LandSlope || object
Neighborhood || object
Condition1 || object
Condition2 || object
BldgType || object
HouseStyle || object
OverallQual || int64
OverallCond || int64
YearBuilt || int64
YearRemodAdd || int64
RoofStyle || object
RoofMatl || object
Exterior1st || object
Exterior2nd || object
MasVnrType || object
MasVnrArea || float64
ExterQual || object
ExterCond || object
Foundation || object
BsmtQual || object
BsmtCond || object
BsmtExposure || object
BsmtFinType1 || object
BsmtFinSF1 || int64
BsmtFinType2 || object
BsmtFinSF2 || int64
BsmtUnfSF || int64
TotalBsmtSF || int64
Heating || object
HeatingQC || object
CentralAir || object
Electrical || object
1stFlrSF || int64
2ndFlrSF || int64
LowQualFinSF || int64
GrLivArea || int64
BsmtFullBath || int64
BsmtHalfBath || int64
FullBath || int64
HalfBath || int64
BedroomAbvGr || int64
KitchenAbvGr || int64
KitchenQual || object
```

```
TotRmsAbvGrd || int64
Functional || object
Fireplaces || int64
FireplaceQu || object
GarageType || object
GarageYrBlt || float64
GarageFinish || object
GarageCars || int64
GarageArea || int64
GarageQual || object
GarageCond || object
PavedDrive || object
WoodDeckSF || int64
OpenPorchSF || int64
EnclosedPorch || int64
3SsnPorch || int64
ScreenPorch || int64
PoolArea || int64
PoolQC || object
Fence || object
MiscFeature || object
MiscVal || int64
MoSold || int64
YrSold || int64
SaleType || object
SaleCondition || object
SalePrice || int64
```



```
In [15]: 1 df_train.isna().sum().sort_values(ascending=False) / len(df_train) * 100
```

```
Out[15]: PoolQC          99.520548
MiscFeature      96.301370
Alley            93.767123
Fence            80.753425
FireplaceQu      47.260274
LotFrontage      17.739726
GarageCond        5.547945
GarageType        5.547945
GarageYrBlt       5.547945
GarageFinish      5.547945
GarageQual        5.547945
BsmtExposure      2.602740
BsmtFinType2      2.602740
BsmtFinType1      2.534247
BsmtCond          2.534247
BsmtQual          2.534247
MasVnrArea        0.547945
MasVnrType        0.547945
Electrical        0.068493
Utilities         0.000000
YearRemodAdd      0.000000
MSSubClass        0.000000
Foundation        0.000000
ExterCond         0.000000
ExterQual         0.000000
Exterior2nd       0.000000
Exterior1st       0.000000
RoofMatl          0.000000
RoofStyle         0.000000
YearBuilt         0.000000
...
GarageArea        0.000000
PavedDrive        0.000000
WoodDeckSF        0.000000
OpenPorchSF       0.000000
3SsnPorch         0.000000
BsmtUnfSF         0.000000
ScreenPorch       0.000000
PoolArea          0.000000
MiscVal           0.000000
MoSold            0.000000
YrSold            0.000000
SaleType          0.000000
Functional        0.000000
TotRmsAbvGrd      0.000000
KitchenQual       0.000000
KitchenAbvGr      0.000000
BedroomAbvGr      0.000000
HalfBath          0.000000
FullBath          0.000000
BsmtHalfBath      0.000000
BsmtFullBath      0.000000
GrLivArea         0.000000
LowQualFinSF      0.000000
2ndFlrSF          0.000000
```

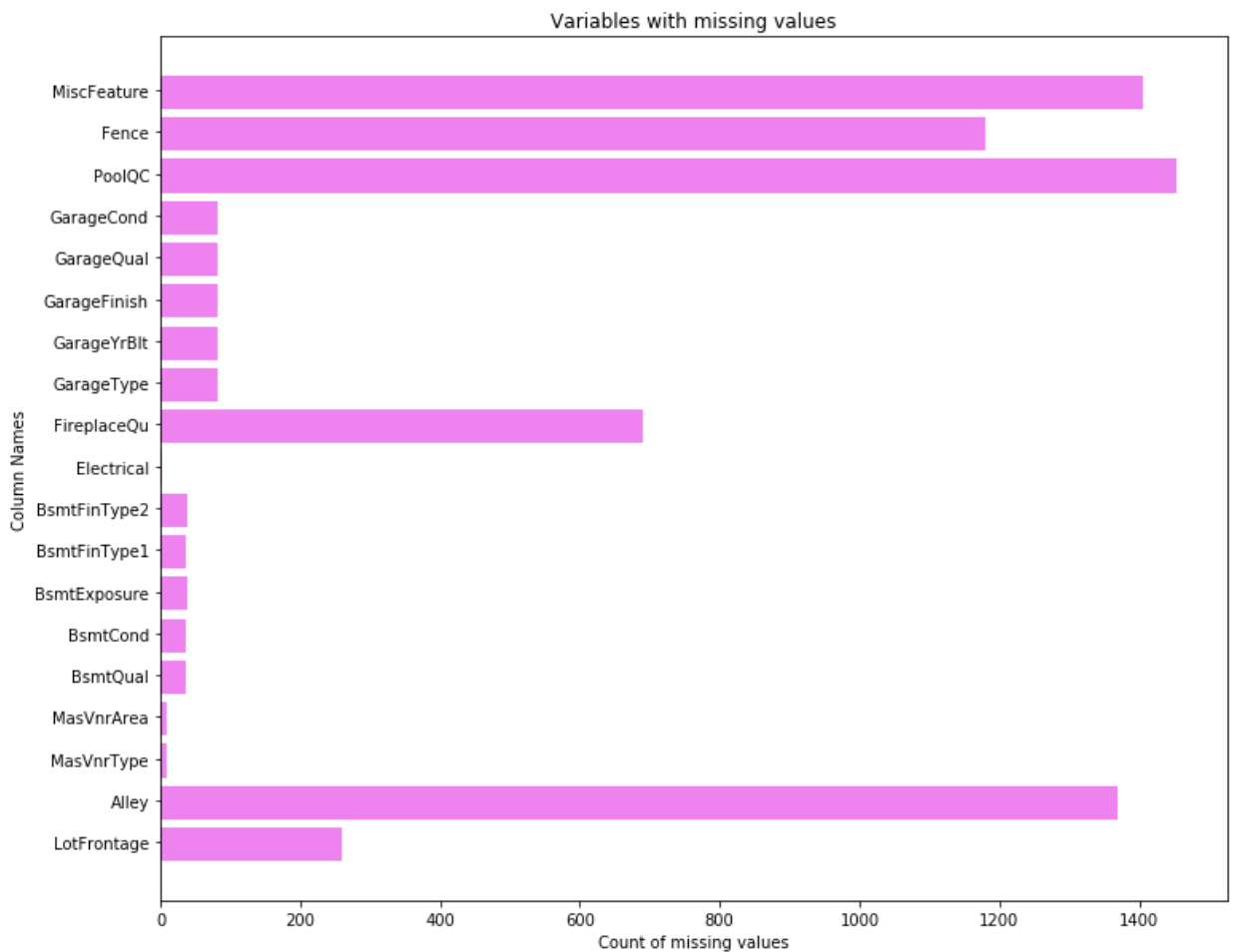
```
1stFlrSF      0.000000
CentralAir    0.000000
SaleCondition  0.000000
Heating       0.000000
TotalBsmtSF   0.000000
Id            0.000000
Length: 81, dtype: float64
```

```
In [44]: 1 #null_columns.shape
```

```

In [17]: 1 null_columns=df_train.columns[df_train.isnull().any()]
2 df_train[null_columns].isnull().sum()
3
4 labels = []
5 values = []
6 for col in null_columns:
7     labels.append(col)
8     values.append(df_train[col].isnull().sum())
9 ind = np.arange(len(labels))
10 width = 0.1
11 fig, ax = plt.subplots(figsize=(12,10))
12 rects = ax.barh(ind, np.array(values), color='violet')
13 ax.set_yticks(ind+((width)/2.))
14 ax.set_yticklabels(labels, rotation='horizontal')
15 ax.set_xlabel("Count of missing values")
16 ax.set_ylabel("Column Names")
17 ax.set_title("Variables with missing values");
18

```

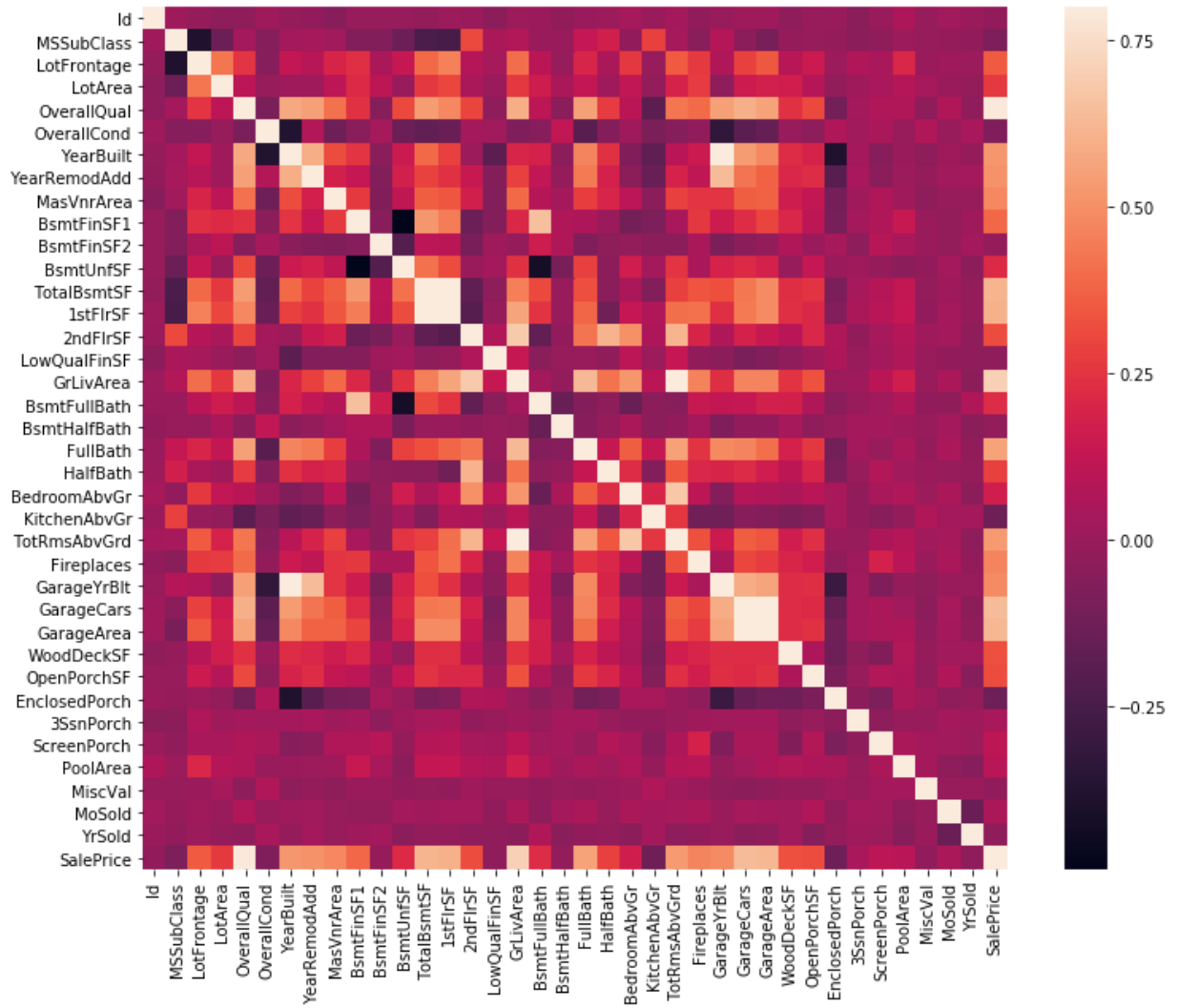


```

In [18]: 1 #df_train#.isna().sum().sort_values(ascending=False) / len(df_train) *

```

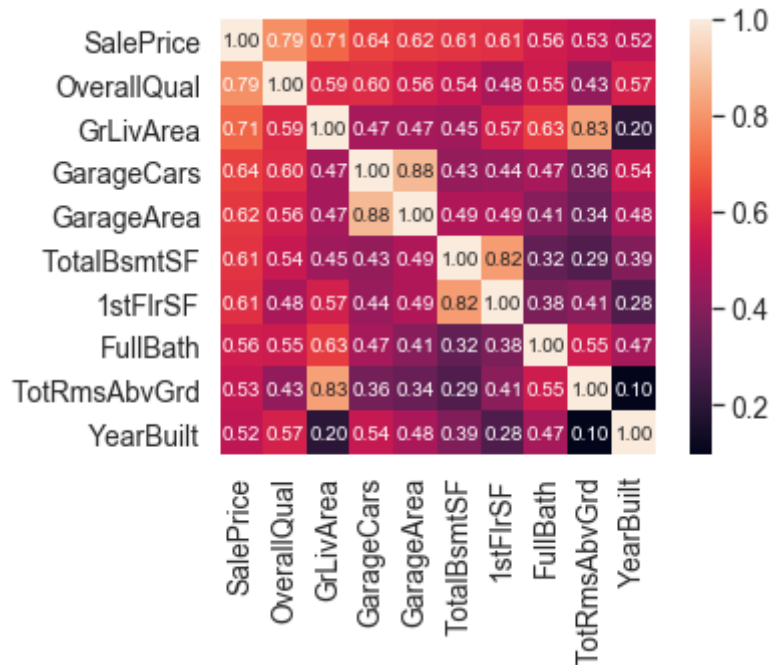
```
In [19]: 1 corrmat = df_train.corr()
2 f, ax = plt.subplots(figsize=(13, 10))
3 sns.heatmap(corrmat, vmax=.8, square=True);
4
```



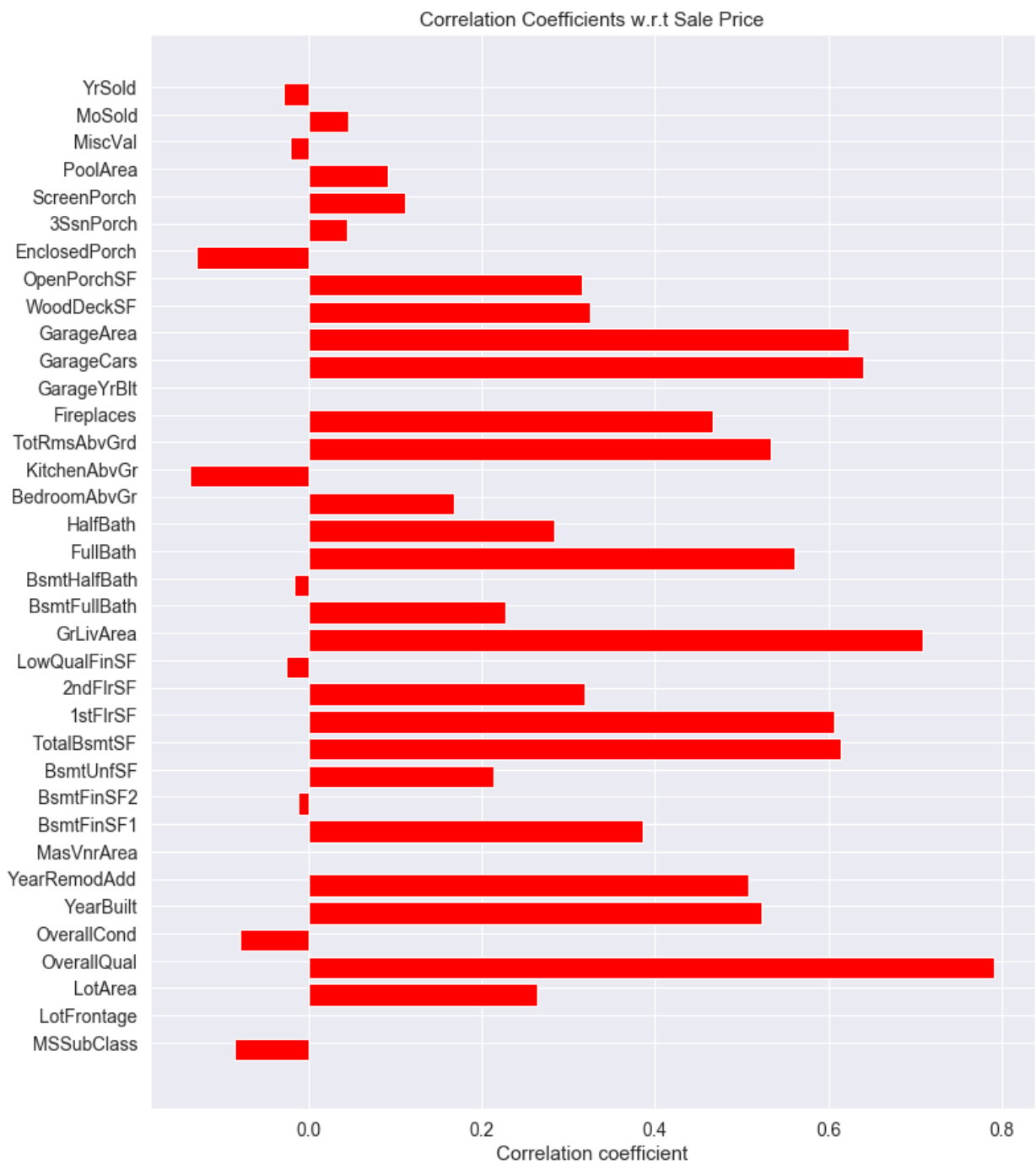
```

In [20]: 1 #saleprice correlation matrix
2 k = 10 #number of variables for heatmap
3 cols = corrmat.nlargest(k, 'SalePrice')['SalePrice'].index
4 cm = np.corrcoef(df_train[cols].values.T)
5 sns.set(font_scale=1.25)
6 hm = sns.heatmap(cm, cbar=True, annot=True, square=True, fmt='.2f', anr
7 plt.show()

```



```
In [21]: 1  #plotting correlations
2  num_feat=df_train.columns[df_train.dtypes!=object]
3  num_feat=num_feat[1:-1]
4  labels = []
5  values = []
6  for col in num_feat:
7      labels.append(col)
8      values.append(np.corrcoef(df_train[col].values, df_train.SalePrice.
9
10 ind = np.arange(len(labels))
11 width = 0.5
12 fig, ax = plt.subplots(figsize=(12,15))
13 rects = ax.barh(ind, np.array(values), color='red')
14 ax.set_yticks(ind+((width)/2.))
15 ax.set_yticklabels(labels, rotation='horizontal')
16 ax.set_xlabel("Correlation coefficient")
17 ax.set_title("Correlation Coefficients w.r.t Sale Price");
```



```
In [22]: 1 correlations=df_train.corr()
2         attrs = correlations.iloc[:-1,:-1] # all except target
3
4         threshold = 0.5
5         important_corrs = (attrs[abs(attrs) > threshold][attrs != 1.0]) \
6             .unstack().dropna().to_dict()
7
8         unique_important_corrs = pd.DataFrame(
9             list(set([(tuple(sorted(key)), important_corrs[key]) \
10                 for key in important_corrs])),
11             columns=['Attribute Pair', 'Correlation'])
12
13         # sorted by absolute value
14         unique_important_corrs = unique_important_corrs.ix[
15             abs(unique_important_corrs['Correlation']).argsort()[::-1]]
16
17         unique_important_corrs
```

/Users/shradhitsuBudhi/anaconda3/envs/python37charm/lib/python3.7/site-packages/ipykernel_launcher.py:14: DeprecationWarning:

.ix is deprecated. Please use
.loc for label based indexing or
.iloc for positional indexing

See the documentation here:

<http://pandas.pydata.org/pandas-docs/stable/indexing.html#ix-indexer-is-deprecated> (<http://pandas.pydata.org/pandas-docs/stable/indexing.html#ix-indexer-is-deprecated>)

Out [22]:

	Attribute Pair	Correlation
25	(GarageArea, GarageCars)	0.882475
4	(GarageYrBlt, YearBuilt)	0.825667
11	(GrLivArea, TotRmsAbvGrd)	0.825489
5	(1stFlrSF, TotalBsmtSF)	0.819530
16	(2ndFlrSF, GrLivArea)	0.687501
27	(BedroomAbvGr, TotRmsAbvGrd)	0.676620
13	(BsmtFinSF1, BsmtFullBath)	0.649212
0	(GarageYrBlt, YearRemodAdd)	0.642277
21	(FullBath, GrLivArea)	0.630012
12	(2ndFlrSF, TotRmsAbvGrd)	0.616423
1	(2ndFlrSF, HalfBath)	0.609707
9	(GarageCars, OverallQual)	0.600671
19	(GrLivArea, OverallQual)	0.593007
26	(YearBuilt, YearRemodAdd)	0.592855
24	(GarageCars, GarageYrBlt)	0.588920

	Attribute Pair	Correlation
20	(OverallQual, YearBuilt)	0.572323
23	(1stFlrSF, GrLivArea)	0.566024
7	(GarageArea, GarageYrBlt)	0.564567
15	(GarageArea, OverallQual)	0.562022
17	(FullBath, TotRmsAbvGrd)	0.554784
6	(OverallQual, YearRemodAdd)	0.550684
14	(FullBath, OverallQual)	0.550600
2	(GarageYrBlt, OverallQual)	0.547766
22	(GarageCars, YearBuilt)	0.537850
8	(OverallQual, TotalBsmtSF)	0.537808
10	(BsmtFinSF1, TotalBsmtSF)	0.522396
18	(BedroomAbvGr, GrLivArea)	0.521270
3	(2ndFlrSF, BedroomAbvGr)	0.502901

```
In [23]: 1 df_train['LotArea_norm'] = df_train['LotArea']
2
3 ss = StandardScaler()
4 mas = MaxAbsScaler()
5 qs = QuantileTransformer()
6
7 df_train['LotArea_norm'] = ss.fit_transform(df_train[['LotArea']])
8 df_train['LotArea_mas'] = mas.fit_transform(df_train[['LotArea']])
9 df_train['LotArea_qs'] = qs.fit_transform(df_train[['LotArea']])
10
11
12 df_train[['LotArea_norm', 'LotArea_mas', 'LotArea_qs', 'LotArea']].head(5)
```

```
Out[23]:
```

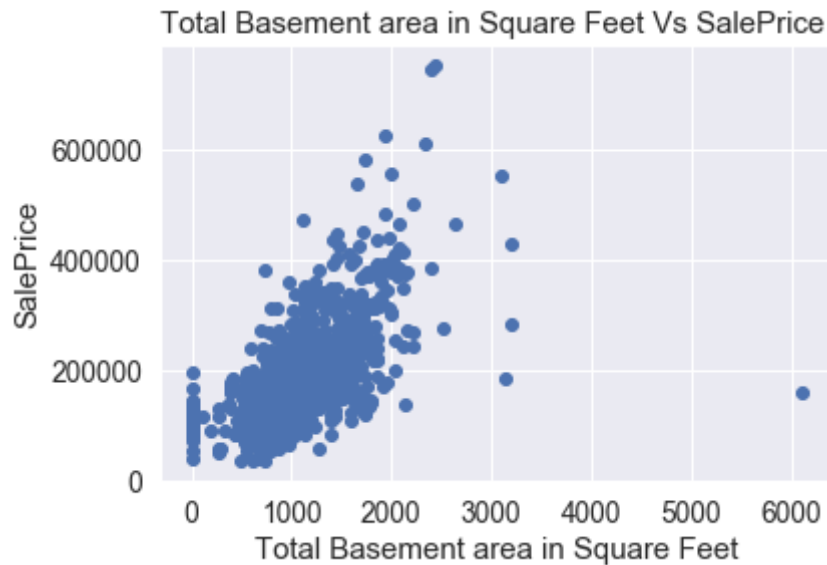
	LotArea_norm	LotArea_mas	LotArea_qs	LotArea
0	-0.207142	0.039258	0.349349	8450
1	-0.091886	0.044600	0.525025	9600
2	0.073480	0.052266	0.720220	11250
3	-0.096897	0.044368	0.509667	9550
4	0.375148	0.066250	0.897123	14260

```
In [24]: 1 plt.figure(figsize=(8, 6))
2         sns.distplot(df_train['SalePrice'], color="r", kde=False, )
3         plt.title("Distribution of Sale Price")
4         plt.ylabel("Number of Occurences")
5
6         plt.xlabel("Sale Price");
```



```
In [25]: 1 sns.distplot?
```

```
In [26]: 1 plt.scatter(df_train["TotalBsmtSF"],df_train["SalePrice"])
2 plt.title("Total Basement area in Square Feet Vs SalePrice ")
3 plt.ylabel("SalePrice")
4 plt.xlabel("Total Basement area in Square Feet");
```



```
In [27]: 1 #sns.factorplot("Fireplaces","SalePrice",data=df_train);
2
```

```
In [28]: 1 #GarageArea has got some outliers lets remove them.
2 upperlimit = np.percentile(df_train.GarageArea.values, 99.5)
3 df_train['GarageArea'].ix[df_train['GarageArea']>upperlimit] = upperlin
4
5 plt.scatter(df_train.GarageArea, df_train["SalePrice"].values,color='violet')
6 plt.title("Garage Area Vs SalePrice ")
7 plt.ylabel("SalePrice")
8 plt.xlabel("Garage Area in sq feet");
```

/Users/shradhitsuBudhi/anaconda3/envs/python37charm/lib/python3.7/site-packages/ipykernel_launcher.py:3: DeprecationWarning:
.ix is deprecated. Please use
.loc for label based indexing or
.iloc for positional indexing

See the documentation here:

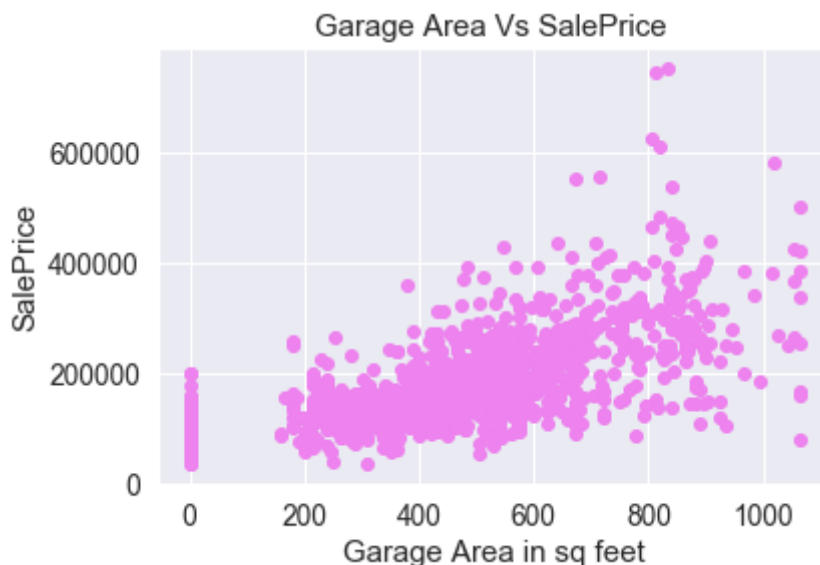
<http://pandas.pydata.org/pandas-docs/stable/indexing.html#ix-indexer-is-deprecated> (<http://pandas.pydata.org/pandas-docs/stable/indexing.html#ix-indexer-is-deprecated>)

This is separate from the ipykernel package so we can avoid doing imports until

/Users/shradhitsuBudhi/anaconda3/envs/python37charm/lib/python3.7/site-packages/pandas/core/indexing.py:190: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame

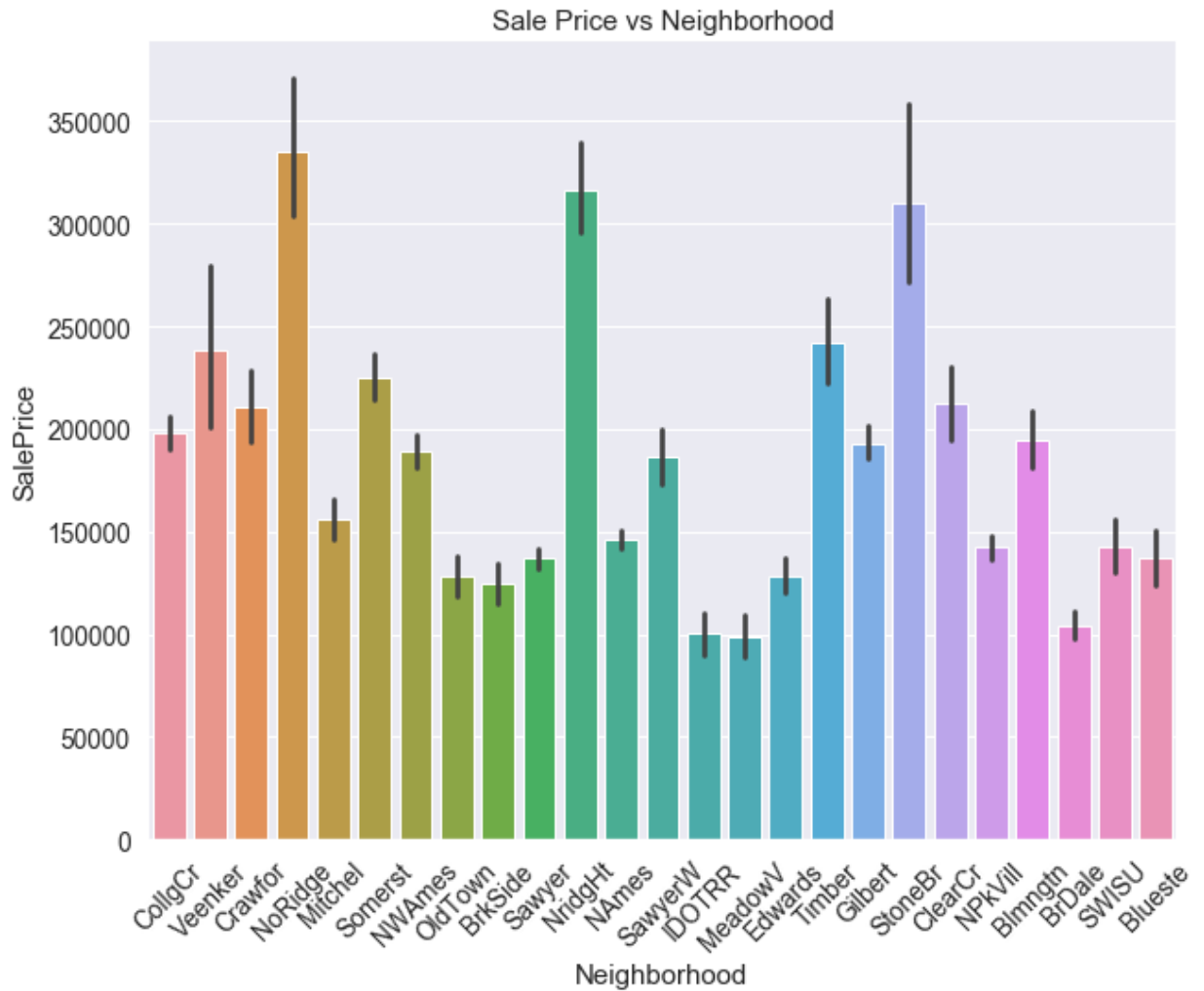
See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy> (<http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy>)

self._setitem_with_indexer(indexer, value)



```
In [29]: 1 upperlimit = np.percentile(df_train.GarageArea.values, 99.5)
2 #upperlimit
```

```
In [30]: 1 plt.figure(figsize=(10, 8))
2 plt.xticks(rotation=45)
3 sns.barplot(df_train["Neighborhood"], df_train["SalePrice"])
4 plt.title("Sale Price vs Neighborhood");
```



```
In [42]: 1 df_train.set_index('Id', inplace=True)
2 #df.set_index('month')
```

In [43]: 1 df_train

Out[43]:

	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	LandContour	Utilities
Id									
1	60	RL	65.0	8450	Pave	NaN	Reg	Lvl	AllPul
2	20	RL	80.0	9600	Pave	NaN	Reg	Lvl	AllPul
3	60	RL	68.0	11250	Pave	NaN	IR1	Lvl	AllPul
4	70	RL	60.0	9550	Pave	NaN	IR1	Lvl	AllPul
5	60	RL	84.0	14260	Pave	NaN	IR1	Lvl	AllPul
6	50	RL	85.0	14115	Pave	NaN	IR1	Lvl	AllPul
7	20	RL	75.0	10084	Pave	NaN	Reg	Lvl	AllPul
8	60	RL	NaN	10382	Pave	NaN	IR1	Lvl	AllPul
9	50	RM	51.0	6120	Pave	NaN	Reg	Lvl	AllPul
10	190	RL	50.0	7420	Pave	NaN	Reg	Lvl	AllPul
11	20	RL	70.0	11200	Pave	NaN	Reg	Lvl	AllPul
12	60	RL	85.0	11924	Pave	NaN	IR1	Lvl	AllPul
13	20	RL	NaN	12968	Pave	NaN	IR2	Lvl	AllPul
14	20	RL	91.0	10652	Pave	NaN	IR1	Lvl	AllPul
15	20	RL	NaN	10920	Pave	NaN	IR1	Lvl	AllPul
16	45	RM	51.0	6120	Pave	NaN	Reg	Lvl	AllPul
17	20	RL	NaN	11241	Pave	NaN	IR1	Lvl	AllPul
18	90	RL	72.0	10791	Pave	NaN	Reg	Lvl	AllPul
19	20	RL	66.0	13695	Pave	NaN	Reg	Lvl	AllPul
20	20	RL	70.0	7560	Pave	NaN	Reg	Lvl	AllPul
21	60	RL	101.0	14215	Pave	NaN	IR1	Lvl	AllPul
22	45	RM	57.0	7449	Pave	Grvl	Reg	Bnk	AllPul
23	20	RL	75.0	9742	Pave	NaN	Reg	Lvl	AllPul
24	120	RM	44.0	4224	Pave	NaN	Reg	Lvl	AllPul
25	20	RL	NaN	8246	Pave	NaN	IR1	Lvl	AllPul
26	20	RL	110.0	14230	Pave	NaN	Reg	Lvl	AllPul
27	20	RL	60.0	7200	Pave	NaN	Reg	Lvl	AllPul
28	20	RL	98.0	11478	Pave	NaN	Reg	Lvl	AllPul
29	20	RL	47.0	16321	Pave	NaN	IR1	Lvl	AllPul
30	30	RM	60.0	6324	Pave	NaN	IR1	Lvl	AllPul
...
1431	60	RL	60.0	21930	Pave	NaN	IR3	Lvl	AllPul

	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	LandContour	Utilities
Id									
1432	120	RL	NaN	4928	Pave	NaN	IR1	Lvl	AllPul
1433	30	RL	60.0	10800	Pave	Grvl	Reg	Lvl	AllPul
1434	60	RL	93.0	10261	Pave	NaN	IR1	Lvl	AllPul
1435	20	RL	80.0	17400	Pave	NaN	Reg	Low	AllPul
1436	20	RL	80.0	8400	Pave	NaN	Reg	Lvl	AllPul
1437	20	RL	60.0	9000	Pave	NaN	Reg	Lvl	AllPul
1438	20	RL	96.0	12444	Pave	NaN	Reg	Lvl	AllPul
1439	20	RM	90.0	7407	Pave	NaN	Reg	Lvl	AllPul
1440	60	RL	80.0	11584	Pave	NaN	Reg	Lvl	AllPul
1441	70	RL	79.0	11526	Pave	NaN	IR1	Bnk	AllPul
1442	120	RM	NaN	4426	Pave	NaN	Reg	Lvl	AllPul
1443	60	FV	85.0	11003	Pave	NaN	Reg	Lvl	AllPul
1444	30	RL	NaN	8854	Pave	NaN	Reg	Lvl	AllPul
1445	20	RL	63.0	8500	Pave	NaN	Reg	Lvl	AllPul
1446	85	RL	70.0	8400	Pave	NaN	Reg	Lvl	AllPul
1447	20	RL	NaN	26142	Pave	NaN	IR1	Lvl	AllPul
1448	60	RL	80.0	10000	Pave	NaN	Reg	Lvl	AllPul
1449	50	RL	70.0	11767	Pave	NaN	Reg	Lvl	AllPul
1450	180	RM	21.0	1533	Pave	NaN	Reg	Lvl	AllPul
1451	90	RL	60.0	9000	Pave	NaN	Reg	Lvl	AllPul
1452	20	RL	78.0	9262	Pave	NaN	Reg	Lvl	AllPul
1453	180	RM	35.0	3675	Pave	NaN	Reg	Lvl	AllPul
1454	20	RL	90.0	17217	Pave	NaN	Reg	Lvl	AllPul
1455	20	FV	62.0	7500	Pave	Pave	Reg	Lvl	AllPul
1456	60	RL	62.0	7917	Pave	NaN	Reg	Lvl	AllPul
1457	20	RL	85.0	13175	Pave	NaN	Reg	Lvl	AllPul
1458	70	RL	66.0	9042	Pave	NaN	Reg	Lvl	AllPul
1459	20	RL	68.0	9717	Pave	NaN	Reg	Lvl	AllPul
1460	20	RL	75.0	9937	Pave	NaN	Reg	Lvl	AllPul

1460 rows × 83 columns

- PoolQC 99.520548

- MiscFeature 96.301370
- Alley 93.767123
- Fence 80.753425
- FireplaceQu 47.260274
- LotFrontage 17.739726
- GarageCond 5.547945
- GarageType 5.547945
- GarageYrBlt 5.547945
- GarageFinish 5.547945
- GarageQual 5.547945
- BsmtExposure 2.602740
- BsmtFinType2 2.602740
- BsmtFinType1 2.534247
- BsmtCond 2.534247
- BsmtQual 2.534247
- MasVnrArea 0.547945
- MasVnrType 0.547945
- Electrical 0.068493

```
In [51]: 1 sum(df_train['FireplaceQu'].isna())
```

```
Out[51]: 690
```

```
In [52]: 1 len(df_train['FireplaceQu'].isna())
```

```
Out[52]: 1460
```

```
In [53]: 1 df_train['FireplaceQu'].value_counts()
```

```
Out[53]: Gd      380
         TA      313
         Fa       33
         Ex       24
         Po       20
         Name: FireplaceQu, dtype: int64
```

```
In [55]: 1 from sklearn.preprocessing import LabelEncoder
         2 le = LabelEncoder()
```

```
In [61]: 1 #le.fit(df_train['FireplaceQu'])
         2 df['Alley'].fillna(value=df['MSZoning'].value_counts().index[0],inplace=True)
         3
```



```
In [58]: 1 #df_train['FireplaceQu']
```

```
Out[58]: Id
1      NaN
2      TA
3      TA
4      Gd
5      TA
6      NaN
7      Gd
8      TA
9      TA
10     TA
11     NaN
12     Gd
13     NaN
14     Gd
15     Fa
16     NaN
17     TA
18     NaN
19     NaN
20     NaN
21     Gd
22     Gd
23     Gd
24     TA
25     TA
26     Gd
27     NaN
28     Gd
29     Gd
30     NaN
...
1431   Gd
1432   NaN
1433   NaN
1434   TA
1435   Gd
1436   Gd
1437   NaN
1438   Gd
1439   NaN
1440   TA
1441   Gd
1442   TA
1443   Ex
1444   Gd
1445   NaN
1446   NaN
1447   NaN
1448   TA
1449   NaN
1450   NaN
1451   NaN
1452   Gd
1453   NaN
```

```
1454    NaN
1455    NaN
1456     TA
1457     TA
1458     Gd
1459    NaN
1460    NaN
Name: FireplaceQu, Length: 1460, dtype: object
```

```
In [63]: 1 len(df_train.columns)
```

```
Out[63]: 83
```

```
In [68]: 1 for x in df_train.dtypes:  
        2     print(x)
```

```
int64  
object  
float64  
int64  
object  
object  
object  
object  
object  
object  
object  
object  
object  
object  
object  
object  
int64  
int64  
int64  
int64  
object  
object  
object  
object  
object  
float64  
object  
object  
object  
object  
object  
object  
object  
int64  
object  
int64  
int64  
int64  
object  
object  
object  
object  
int64  
int64  
int64  
int64  
int64  
int64  
int64  
int64  
int64  
int64  
int64  
object  
int64
```

```
object
int64
object
object
float64
object
int64
float64
object
object
object
int64
int64
int64
int64
int64
int64
object
object
object
int64
int64
int64
object
object
int64
float64
float64
float64
```

In []:

1