Feature Engineering

September 3, 2022

```
import pandas as pd
[2]:
     import numpy as np
     import matplotlib.pyplot as plt
     import seaborn as sns
    train=pd.read_csv("PEP1.csv")
[5]:
     train.describe(include='all')
[5]:
                        Id
                             MSSubClass MSZoning
                                                     LotFrontage
                                                                          LotArea Street
              1460.000000
                            1460.000000
                                              1460
                                                     1201.000000
                                                                      1460.000000
                                                                                     1460
     count
                                                 5
                                                                                        2
     unique
                       NaN
                                     NaN
                                                              NaN
                                                                              NaN
     top
                       NaN
                                     NaN
                                                RL
                                                              NaN
                                                                              NaN
                                                                                     Pave
     freq
                       NaN
                                     NaN
                                              1151
                                                              NaN
                                                                              NaN
                                                                                     1454
                              56.897260
               730.500000
                                               NaN
                                                       70.049958
                                                                    10516.828082
                                                                                      NaN
     mean
                                                       24.284752
     std
               421.610009
                              42.300571
                                               NaN
                                                                     9981.264932
                                                                                      NaN
     min
                 1.000000
                               20.000000
                                               NaN
                                                       21.000000
                                                                      1300.000000
                                                                                      NaN
     25%
               365.750000
                               20.000000
                                               NaN
                                                       59.000000
                                                                     7553.500000
                                                                                      NaN
     50%
               730.500000
                              50.000000
                                                       69.000000
                                               NaN
                                                                     9478.500000
                                                                                      NaN
     75%
              1095.250000
                              70.000000
                                               NaN
                                                       80.000000
                                                                    11601.500000
                                                                                      NaN
              1460.000000
                             190.000000
                                               NaN
                                                      313.000000
                                                                   215245.000000
                                                                                      NaN
     max
             Alley LotShape LandContour Utilities
                                                              PoolArea PoolQC
                                                                                Fence
                91
                        1460
                                     1460
                                                1460
                                                          1460.000000
                                                                             7
                                                                                   281
     count
                 2
                           4
                                         4
                                                    2
                                                                             3
                                                                                     4
     unique
                                                                   NaN
     top
              Grvl
                         Reg
                                      Lvl
                                              AllPub
                                                                   NaN
                                                                            Gd
                                                                                MnPrv
     freq
                50
                         925
                                     1311
                                                1459
                                                                   NaN
                                                                             3
                                                                                   157
     mean
               NaN
                         NaN
                                      NaN
                                                 NaN
                                                              2.758904
                                                                           NaN
                                                                                   NaN
     std
               NaN
                         NaN
                                      NaN
                                                 NaN
                                                             40.177307
                                                                           NaN
                                                                                   NaN
     min
               NaN
                         NaN
                                      NaN
                                                 NaN
                                                              0.000000
                                                                           NaN
                                                                                   NaN
     25%
               NaN
                         NaN
                                      NaN
                                                                           NaN
                                                 NaN
                                                              0.000000
                                                                                   NaN
     50%
                                                                           NaN
               NaN
                         NaN
                                      NaN
                                                 NaN
                                                              0.000000
                                                                                   NaN
     75%
               NaN
                         NaN
                                      NaN
                                                              0.000000
                                                                           NaN
                                                                                   NaN
                                                 {\tt NaN}
               NaN
     max
                         NaN
                                      NaN
                                                 NaN
                                                           738.000000
                                                                           NaN
                                                                                   NaN
             MiscFeature
                                 MiscVal
                                                MoSold
                                                                       SaleType
                                                               YrSold
     count
                       54
                            1460.000000
                                           1460.000000
                                                         1460.000000
                                                                            1460
```

unique	4	NaN	NaN	NaN	9
top	Shed	NaN	NaN	NaN	WD
freq	49	NaN	NaN	NaN	1267
mean	NaN	43.489041	6.321918	2007.815753	NaN
std	NaN	496.123024	2.703626	1.328095	NaN
min	NaN	0.000000	1.000000	2006.000000	NaN
25%	NaN	0.000000	5.000000	2007.000000	NaN
50%	NaN	0.000000	6.000000	2008.000000	NaN
75%	NaN	0.000000	8.000000	2009.000000	NaN
max	NaN	15500.000000	12.000000	2010.000000	NaN

	SaleCondition	SalePrice
count	1460	1460.000000
unique	6	NaN
top	Normal	NaN
freq	1198	NaN
mean	NaN	180921.195890
std	NaN	79442.502883
min	NaN	34900.000000
25%	NaN	129975.000000
50%	NaN	163000.000000
75%	NaN	214000.000000
max	NaN	755000.000000

[11 rows x 81 columns]

- [6]: train.shape
- [6]: (1460, 81)
- [7]: train.isnull().sum().sort_values(ascending= False).head(20)
- [7]: PoolQC 1453 MiscFeature 1406 Alley 1369 Fence 1179 FireplaceQu 690 ${\tt LotFrontage}$ 259 GarageCond 81 GarageType 81 GarageYrBlt 81 GarageFinish 81 GarageQual 81 BsmtExposure 38 BsmtFinType238 BsmtFinType1 37 BsmtCond 37

BsmtQual 37
MasVnrArea 8
MasVnrType 8
Electrical 1
Utilities 0
dtype: int64

[9]: train.drop(['PoolQC','MiscFeature','Alley','Fence','LotFrontage'],axis=1)

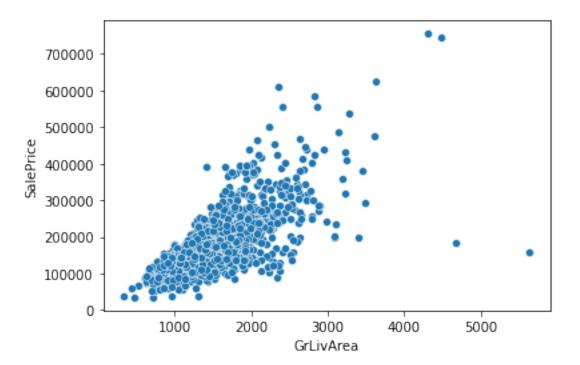
		_										_				
[9]:		Id	MSSub	Class	MSZon	ing	LotA	rea	Street	LotSh	nape Lan	dContou	ır \			
	0	1		60		RL		450	Pave		Reg	L_{λ}				
	1	2		20		RL	9	600	Pave		Reg	Lv	<i>7</i> 1			
	2	3		60		RL	11	250	Pave		IR1	Lv	<i>7</i> 1			
	3	4		70		RL	9	550	Pave		IR1	Lv	<i>7</i> 1			
	4	5		60		RL	14	260	Pave		IR1	Lv	<i>7</i> 1			
	 1455	 1456		60	•	 RL	7	917	 Pave	•••	Reg	Lv	₇ 7			
	1456	1457		20		RL		175	Pave		Reg	Lv				
	1457	1458		70		RL		042	Pave		Reg	Lv				
	1458	1459		20		RL		717	Pave		Reg	Lv				
	1459	1460		20		RL		937	Pave		Reg	L				
			_			101		_					_		,	
	^	Utiliti			_	_	•		ClosedPo		BSsnPorc		enPor		\	
	0	AllF		Insid			tl			0		0		0		
	1	AllF			R2		tl			0		0		0		
	2	AllF		Insid			tl			0		0		0		
	3	AllF		Corne			tl			272		0		0		
	4	AllF	'ub	FI	32	G	tl			0		0		0		
					•••			•	••		•••	•		_		
	1455	AllF		Insid			tl			0		0		0		
	1456	AllF		Insid			tl			0		0		0		
	1457	AllF		Insid			tl			0		0		0		
	1458	AllF		Insid			tl			112		0		0		
	1459	AllF	ub	Insid	ie	G	tl			0		0		0		
		PoolAre	a Mis	cVal	MoSol	d Y	rSold	Sa	aleType	Sale	eConditi	on Sale	ePric	е		
	0		0	0	:	2	2008		WD		Norm	al 2	20850	0		
	1		0	0		5	2007		WD		Norm	al 1	L8150	0		
	2		0	0	!	9	2008		WD		Norm	al 2	22350	0		
	3		0	0	:	2	2006		WD		Abnor	ml 1	L4000	0		
	4		0	0	1:	2	2008		WD		Norm	al 2	25000	0		
	•••		•••		•••		•••			••	•••					
	1455		0	0	;	8	2007		WD		Norm	al 1	17500	0		
	1456		0	0	:	2	2010		WD		Norm	al 2	21000	0		
	1457		0	2500		5	2010		WD		Norm	al 2	26650	0		
	1458		0	0		4	2010		WD		Norm	ıal 1	L4212	5		
	1459		0	0		6	2008		WD		Norm	ıal 1	L4750	0		

[1460 rows x 76 columns]

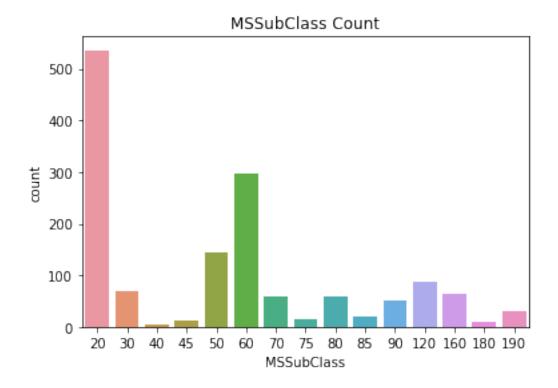
```
[10]: print(train.shape)
sns.scatterplot(y="SalePrice", x="GrLivArea", data=train)

(1460, 81)
```

[10]: <AxesSubplot:xlabel='GrLivArea', ylabel='SalePrice'>



[12]: Text(0.5, 1.0, 'MSSubClass Count')

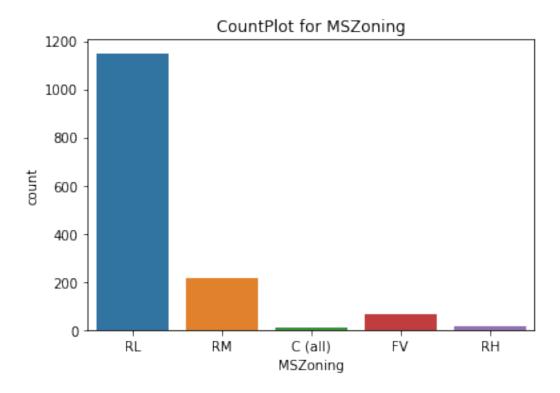


```
[14]: cols = ['MSSubClass', 'MSZoning', 'LotFrontage', 'LotArea', 'Street',
             'Alley', 'LotShape', 'LandContour', 'Utilities', 'LotConfig',
             'LandSlope', 'Neighborhood', 'Condition1', 'Condition2', 'BldgType',
             'HouseStyle', 'OverallQual', 'OverallCond', 'YearBuilt', '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']
      features=[]
      obj_col=[]
      for col in cols:
          if (train[col].dtype) == object :
              obj_col.append(col)
      obj_col
```

```
KeyError
                                               Traceback (most recent call_
→last)
      /usr/local/lib/python3.7/site-packages/pandas/core/indexes/base.py in_
2897
                      try:
  -> 2898
                         return self._engine.get_loc(casted_key)
     2899
                      except KeyError as err:
      pandas/_libs/index.pyx in pandas._libs.index.IndexEngine.get_loc()
      pandas/_libs/index.pyx in pandas._libs.index.IndexEngine.get_loc()
      pandas/_libs/hashtable_class_helper.pxi in pandas._libs.hashtable.
→PyObjectHashTable.get_item()
      pandas/_libs/hashtable_class_helper.pxi in pandas._libs.hashtable.
→PyObjectHashTable.get_item()
      KeyError: 'KitchenAbvGr'
  The above exception was the direct cause of the following exception:
                                               Traceback (most recent call_
      KeyError
→last)
      <ipython-input-14-8c467f101b6d> in <module>
       19 obj_col=[]
       20 for col in cols:
  ---> 21 if (train[col].dtype) == object :
                 obj_col.append(col)
       23 obj_col
      /usr/local/lib/python3.7/site-packages/pandas/core/frame.py in_
→__getitem__(self, key)
```

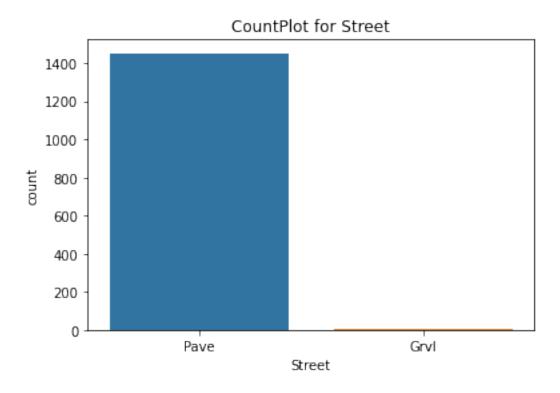
```
2904
                             if self.columns.nlevels > 1:
            2905
                                  return self._getitem_multilevel(key)
         -> 2906
                             indexer = self.columns.get_loc(key)
            2907
                             if is_integer(indexer):
            2908
                                  indexer = [indexer]
             /usr/local/lib/python3.7/site-packages/pandas/core/indexes/base.py in ⊔

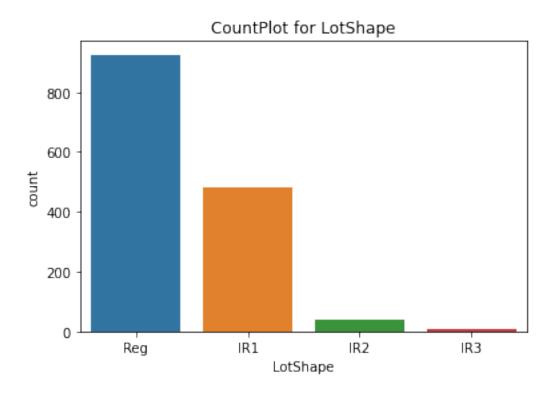
→get_loc(self, key, method, tolerance)
            2898
                                 return self._engine.get_loc(casted_key)
            2899
                             except KeyError as err:
         -> 2900
                                  raise KeyError(key) from err
            2901
            2902
                         if tolerance is not None:
             KeyError: 'KitchenAbvGr'
[15]: def plotting(col):
          sns.countplot(x=col,data=train)
          plt.title('CountPlot for '+ col)
          print('No. of missing values are : ')
          print(train[col].isnull().sum())
[16]: plotting('MSZoning')
     No. of missing values are :
     0
```



[17]: plotting('Street')

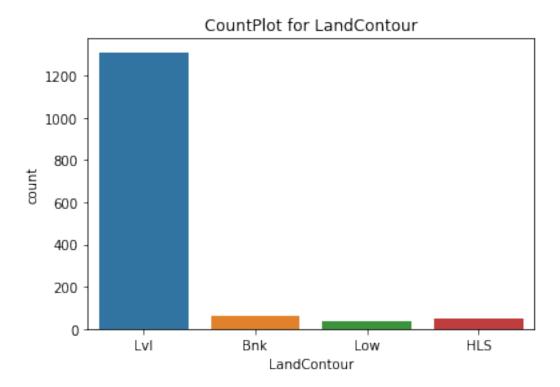
No. of missing values are : 0





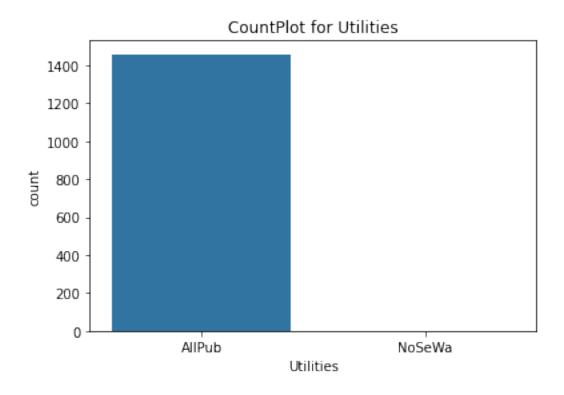
[19]: plotting('LandContour')

No. of missing values are : $\ensuremath{\text{0}}$

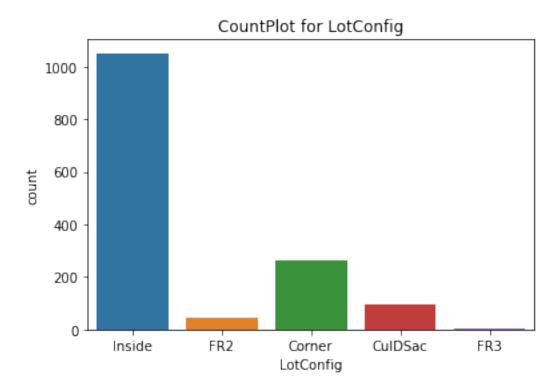


[20]: plotting('Utilities')

No. of missing values are : 0

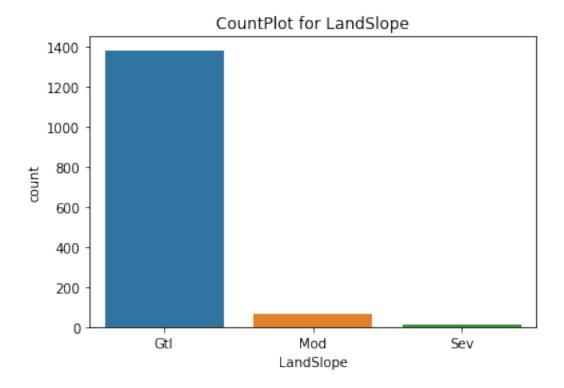


```
[21]: plotting('LotConfig')
      features.append('LotConfig')
      train.LotConfig=train.LotConfig.replace({'Inside':1,'Corner':2,'CulDSac':
      →3,'FR2':4,'FR3':5})
      test.LotConfig=test.LotConfig.replace({'Inside':1,'Corner':2,'CulDSac':3,'FR2':
       \hookrightarrow4,'FR3':5})
     No. of missing values are :
     0
             NameError
                                                         Traceback (most recent call_
      →last)
              <ipython-input-21-d7bd811a34f4> in <module>
                4 train.LotConfig=train.LotConfig.replace({'Inside':1,'Corner':
      →2, 'CulDSac':3, 'FR2':4, 'FR3':5})
         ----> 5 test.LotConfig=test.LotConfig.replace({'Inside':1,'Corner':
      →2, 'CulDSac':3, 'FR2':4, 'FR3':5})
```



[22]: plotting('LandSlope')

No. of missing values are : $\mathbf{0}$



```
[23]: train_neighbor=train.copy()
      print(train_neighbor[['Neighborhood', 'SalePrice']].

¬groupby(['Neighborhood'],as_index=True).mean())
      plotting('Neighborhood')
      #This looks promising as Price varies from area to area and it can be seen_
      →through the mean of the prices of selected Neighborhood.
      features.append('Neighborhood')
      train.Neighborhood=train.Neighborhood.replace({'MeadowV':1,
                                                      'IDOTRR':2, 'BrDale':2,
                                                      'OldTown':3, 'Edwards':3, u

¬'BrkSide':3,
                                                      'Sawyer':4, 'Blueste':4, 'SWISU':
       \hookrightarrow4, 'NAmes':4,
                                                      'NPkVill':5, 'Mitchel':5,
                                                      'SawyerW':6, 'Gilbert':6, L
       →'NWAmes':6,
                                                      'Blmngtn':7, 'CollgCr':7, ⊔
       'Veenker':8, 'Somerst':8, u

¬'Timber':8,
                                                      'StoneBr':9,
                                                      'NoRidge':10, 'NridgHt':10})
      test.Neighborhood=test.Neighborhood.replace({'MeadowV':1,
```

```
'IDOTRR':2, 'BrDale':2,
'OldTown':3, 'Edwards':3,

→'BrkSide':3,

'Sawyer':4, 'Blueste':4, 'SWISU':

→4, 'NAmes':4,

'NPkVill':5, 'Mitchel':5,
'SawyerW':6, 'Gilbert':6,

→'NWAmes':6,

'Blmngtn':7, 'CollgCr':7,

→'ClearCr':7, 'Crawfor':7,

'Veenker':8, 'Somerst':8,

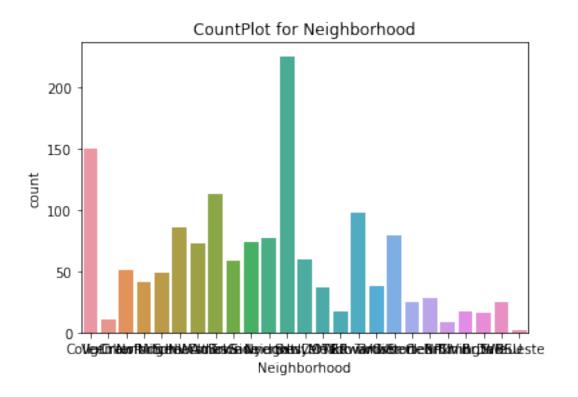
→'Timber':8,

'StoneBr':9,
'NoRidge':10, 'NridgHt':10})
```

SalePrice

Neighborhood							
Blmngtn	194870.882353						
Blueste	137500.000000						
BrDale	104493.750000						
BrkSide	124834.051724						
ClearCr	212565.428571						
CollgCr	197965.773333						
Crawfor	210624.725490						
Edwards	127318.571429						
Gilbert	192854.506329						
IDOTRR	100123.783784						
MeadowV	98576.470588						
Mitchel	156270.122449						
NPkVill	142694.444444						
NWAmes	189050.068493						
NoRidge	335295.317073						
NridgHt	316270.623377						
OldTown	128225.300885						
SWISU	142591.360000						
Sawyer	136793.135135						
SawyerW	186555.796610						
Somerst	225379.837209						
StoneBr	310499.000000						
Timber	242247.447368						
Veenker	238772.727273						
mes 145847.080000							
No. of missing values are :							
0							

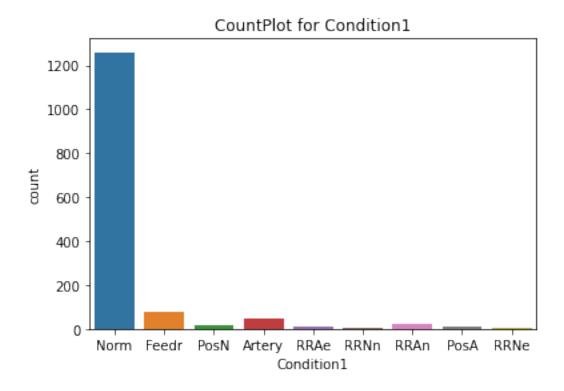
→-----



```
[24]: plotting('Condition1')

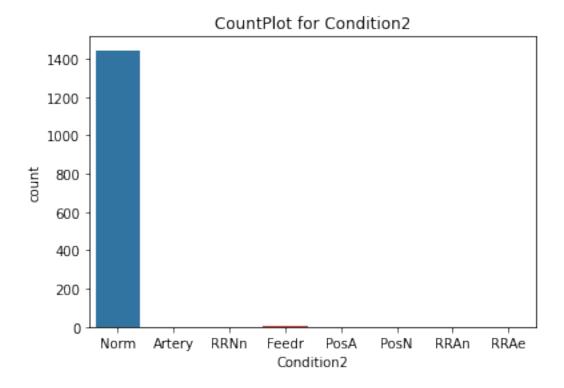
No. of missing values are :
```

0

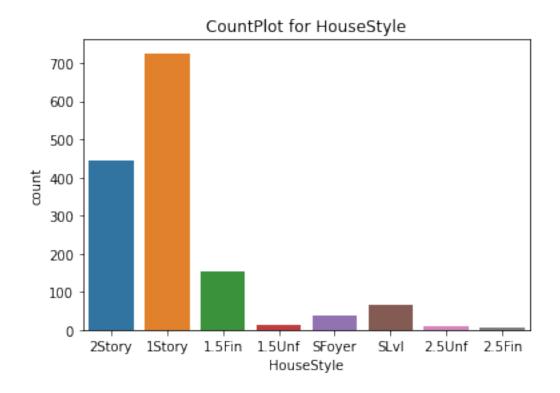


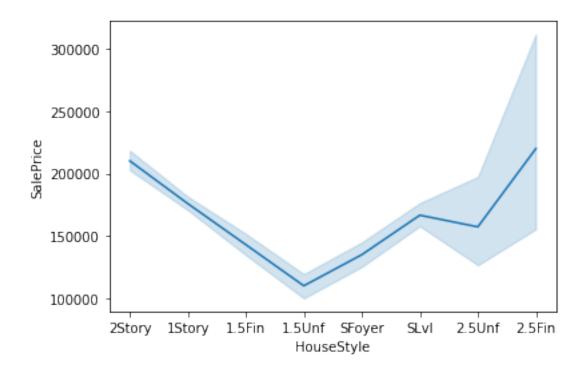
[25]: plotting('Condition2')

No. of missing values are : 0



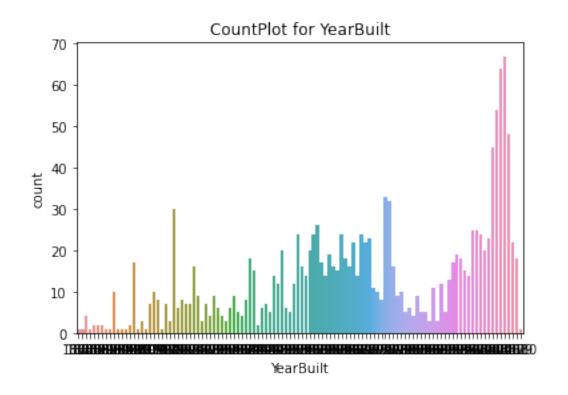
No. of missing values are : $\ensuremath{\text{0}}$

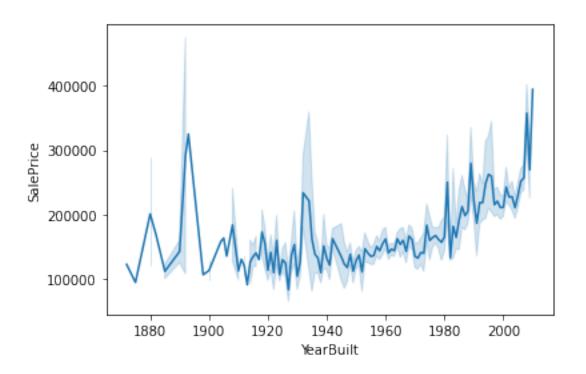




```
[27]: plotting('YearBuilt')
  plt.show()
  sns.lineplot(y='SalePrice',x='YearBuilt',data=train)
  # It is clearly visible. So it should be part of our features variable.
  features.append('YearBuilt')
```

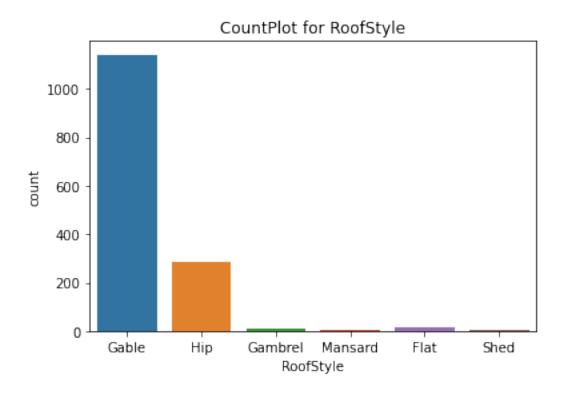
No. of missing values are : \circ



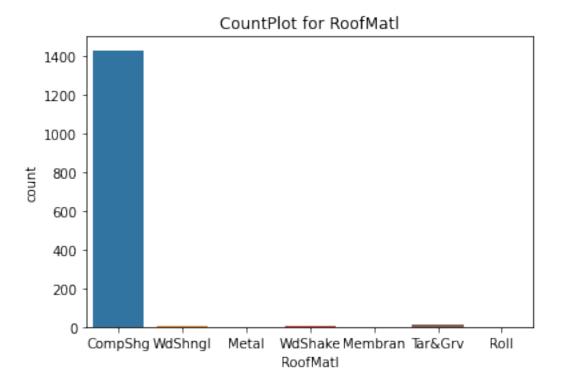


```
[28]: #Roof Style : Type of roof
plotting('RoofStyle')
plt.show()
#RoofMatl: Roof material
plotting('RoofMatl')
```

No. of missing values are : Ω



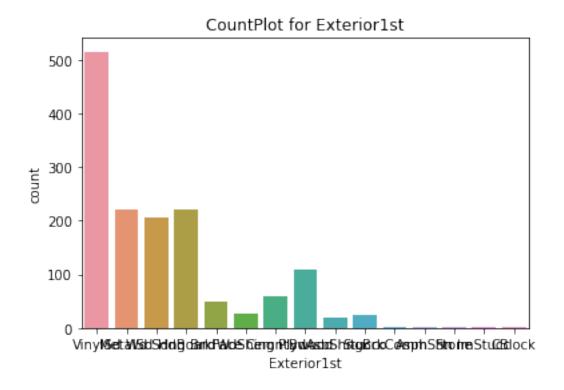
No. of missing values are : 0



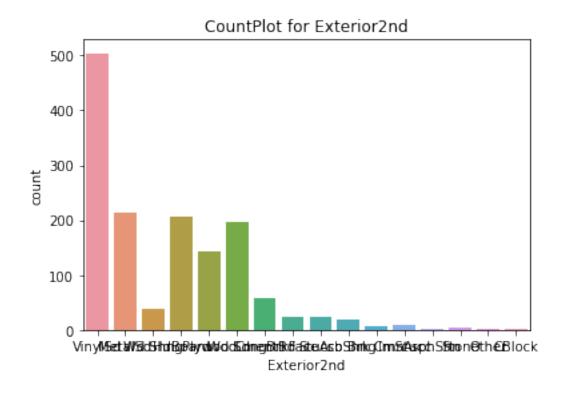
```
[29]: plotting('Exterior1st')
     plt.show()
     #Exterior2nd: Exterior covering on house (if more than one material)
     plotting('Exterior2nd')
     plt.show()
     train.Exterior1st=train.Exterior1st.replace({'BrkComm':1,
                                             'AsphShn':2, 'CBlock':2, 'AsbShng':
      -2,
                                             'WdShing':3, 'Wd Sdng':3, u
      'BrkFace':4, 'Plywood':4,
                                             'VinylSd':5,
                                             'CemntBd':6,
                                             'Stone':7, 'ImStucc':7})
     test.Exterior1st=test.Exterior1st.replace({'BrkComm':1,
                                             'AsphShn':2, 'CBlock':2, 'AsbShng':
      →2,
                                             'WdShing':3, 'Wd Sdng':3,
      'BrkFace':4, 'Plywood':4,
                                             'VinylSd':5,
```

```
'CemntBd':6,
'Stone':7, 'ImStucc':7})
features.append('Exterior1st')
```

No. of missing values are : \circ



No. of missing values are : $\ensuremath{\text{0}}$

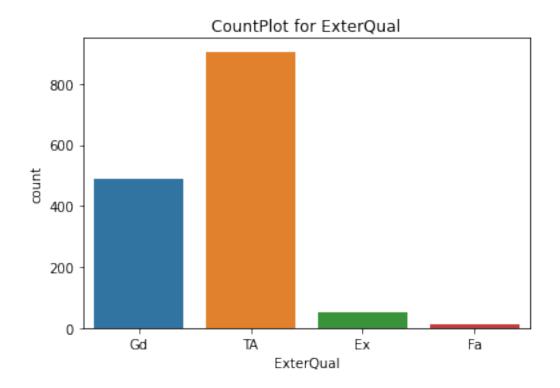


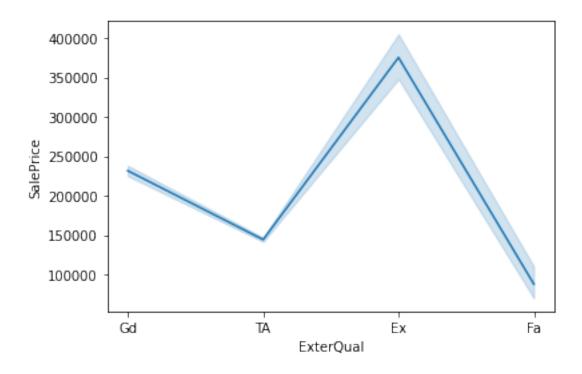
sns.lineplot(y='SalePrice',x='ExterQual',data=train)

plt.show()

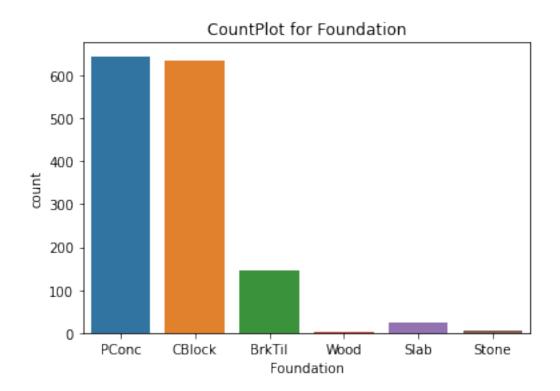
```
features.append('ExterQual')
train.ExterQual = train.ExterQual.replace({'Fa':1, 'TA':2, 'Gd':3, 'Ex':4})
test.ExterQual = test.ExterQual.replace({'Fa':1, 'TA':2, 'Gd':3, 'Ex':4})
```

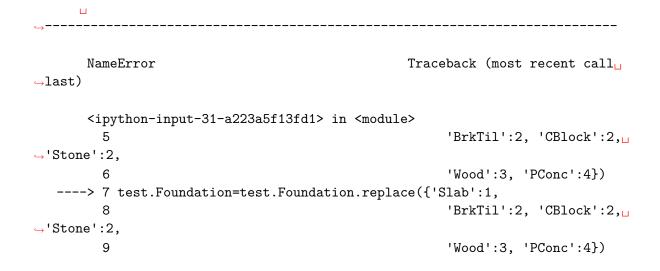
No. of missing values are : $\ensuremath{\text{0}}$

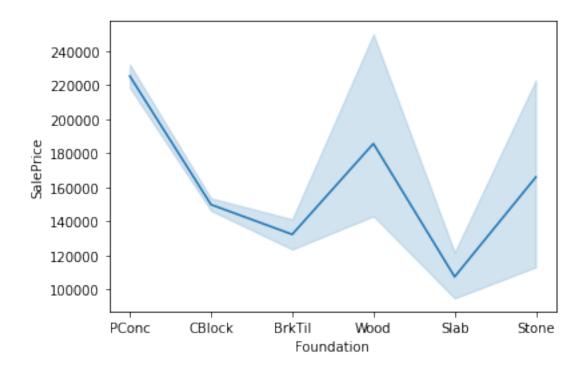




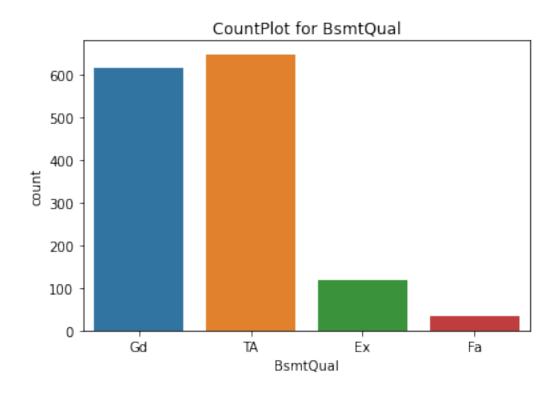
No. of missing values are : \circ







```
[32]: plotting('BsmtQual')
      train.BsmtQual=train.BsmtQual.fillna('None').replace({'Fa':2, 'None':1, 'TA':3,_
       \hookrightarrow 'Gd':4, 'Ex':5})
      test.BsmtQual=test.BsmtQual.fillna('None').replace({'Fa':2, 'None':1, 'TA':3, L
       \hookrightarrow 'Gd':4, 'Ex':5})
      features.append('BsmtQual')
     No. of missing values are :
     37
              NameError
                                                            Traceback (most recent call⊔
       →last)
              <ipython-input-32-fa674785bf39> in <module>
                1 plotting('BsmtQual')
                2 train.BsmtQual=train.BsmtQual.fillna('None').replace({'Fa':2, 'None':
       \rightarrow 1, 'TA':3, 'Gd':4, 'Ex':5})
          ----> 3 test.BsmtQual=test.BsmtQual.fillna('None').replace({'Fa':2, 'None':
      \rightarrow 1, 'TA':3, 'Gd':4, 'Ex':5})
                4 features.append('BsmtQual')
```



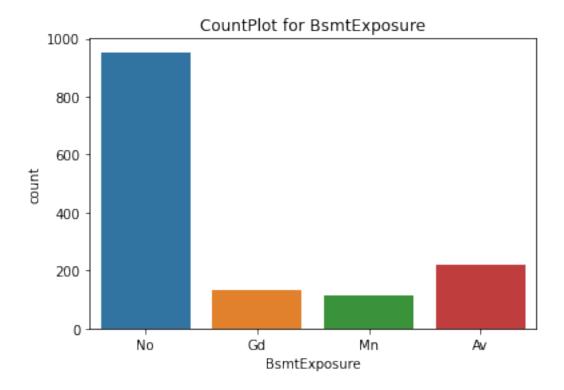
```
2 train.BsmtExposure=train.BsmtExposure.fillna('None').replace({'None':

→1, 'No':2, 'Av':3, 'Mn':3, 'Gd':4})

----> 3 test.BsmtExposure=test.BsmtExposure.fillna('None').replace({'None':

→1, 'No':2, 'Av':3, 'Mn':3, 'Gd':4})

4 features.append('BsmtExposure')
```



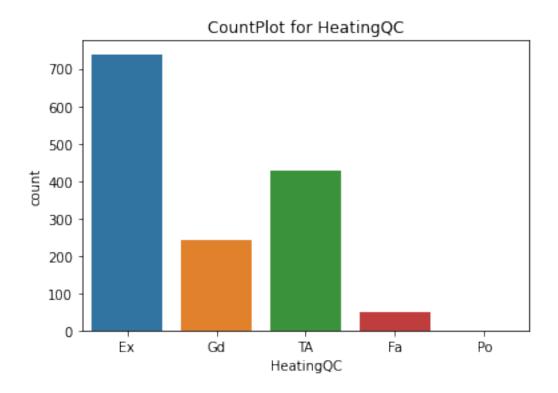
```
plotting('HeatingQC')
train.HeatingQC.replace({'Po':1, 'Fa':2, 'TA':3, 'Gd':4, 'Ex':5},inplace=True)
test.HeatingQC.replace({'Po':1, 'Fa':2, 'TA':3, 'Gd':4, 'Ex':5},inplace=True)
features.append('HeatingQC')

No. of missing values are :
0

NameError

NameError

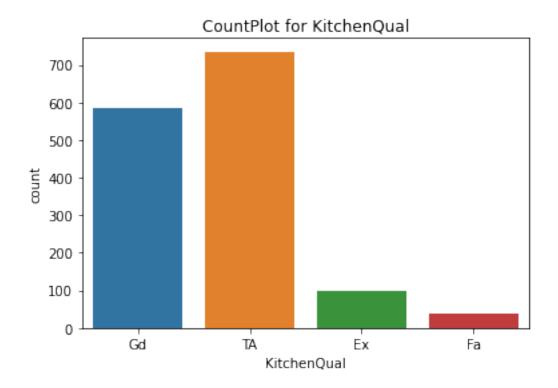
Traceback (most recent call_u → last)
```



```
[35]: plotting('KitchenQual')
train.KitchenQual.replace({'Fa':1, 'TA':2, 'Gd':3, 'Ex':4},inplace=True)
test.KitchenQual.replace({'Fa':1, 'TA':2, 'Gd':3, 'Ex':4},inplace=True)
features.append('KitchenQual')
```

No. of missing values are : $\ensuremath{\text{0}}$

⊔



```
[36]: plotting('FireplaceQu')
train.FireplaceQu.fillna('None',inplace=True)
train.FireplaceQu.replace({'None':1, 'Po':1, 'Fa':2, 'TA':3, 'Gd':4, 'Ex':
→5},inplace=True)
test.FireplaceQu.fillna('None',inplace=True)
test.FireplaceQu.replace({'None':1, 'Po':1, 'Fa':2, 'TA':3, 'Gd':4, 'Ex':
→5},inplace=True)
```

```
features.append('FireplaceQu')

No. of missing values are :
690

NameError Traceback (most recent callust)

<ipython-input-36-3bd8e3a474b3> in <module>
2 train.FireplaceQu.fillna('None',inplace=True)
3 train.FireplaceQu.replace({'None':1, 'Po':1, 'Fa':2, 'TA':3, 'Gd':4, use 'Ex':5},inplace=True)

----> 4 test.FireplaceQu.fillna('None',inplace=True)

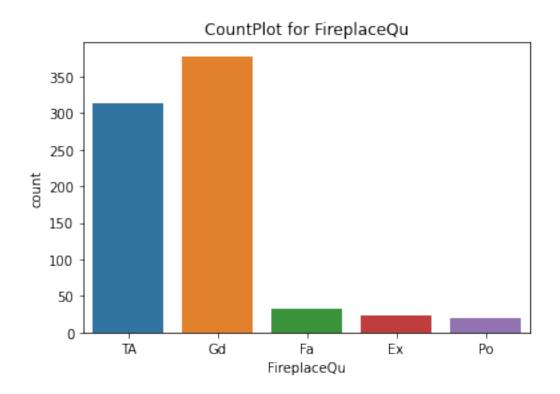
----> 4 test.FireplaceQu.fillna('None',inplace=True)
```

5 test.FireplaceQu.replace({'None':1, 'Po':1, 'Fa':2, 'TA':3, 'Gd':4, |

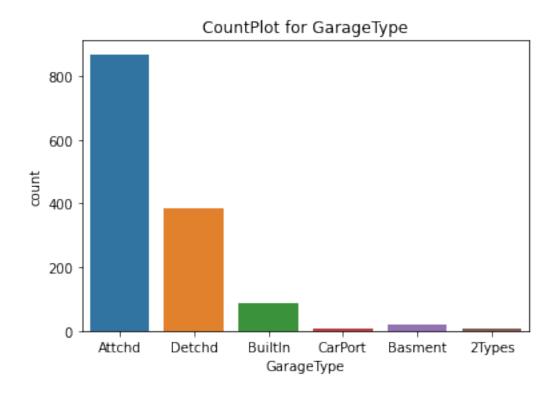
NameError: name 'test' is not defined

6 features.append('FireplaceQu')

→'Ex':5},inplace=True)



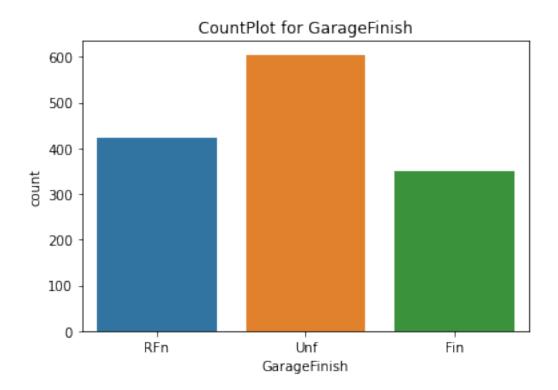
```
[37]: plotting('GarageType')
      train.GarageType.fillna('None',inplace=True)
      test.GarageType.fillna('None',inplace=True)
      train.GarageType.replace({'CarPort':1, 'None':1,
                                                  'Detchd':2,
                                                  '2Types':3, 'Basment':3,
                                                  'Attchd':4, 'BuiltIn':
      →5},inplace=True)
      test.GarageType.replace({'CarPort':1, 'None':1,
                                                  'Detchd':2,
                                                  '2Types':3, 'Basment':3,
                                                  'Attchd':4, 'BuiltIn':
      →5},inplace=True)
      features.append('GarageType')
     No. of missing values are :
     81
             NameError
                                                        Traceback (most recent call⊔
      →last)
             <ipython-input-37-8da765d6e288> in <module>
               1 plotting('GarageType')
               2 train.GarageType.fillna('None',inplace=True)
         ----> 3 test.GarageType.fillna('None',inplace=True)
               4 train.GarageType.replace({'CarPort':1, 'None':1,
               5
                                                             'Detchd':2,
             NameError: name 'test' is not defined
```



```
[38]: plotting('GarageFinish')
      train.GarageFinish.fillna('None',inplace=True)
      test.GarageFinish.fillna('None',inplace=True)
      train.GarageFinish.replace({'None':1, 'Unf':2, 'RFn':3, 'Fin':4},inplace=True)
      test.GarageFinish.replace({'None':1, 'Unf':2, 'RFn':3, 'Fin':4},inplace=True)
      features.append('GarageFinish')
     No. of missing values are :
     81
             NameError
                                                        Traceback (most recent call_
      →last)
             <ipython-input-38-0f0163201bfb> in <module>
               1 plotting('GarageFinish')
               2 train.GarageFinish.fillna('None',inplace=True)
         ----> 3 test.GarageFinish.fillna('None',inplace=True)
               4 train.GarageFinish.replace({'None':1, 'Unf':2, 'RFn':3, 'Fin':
      →4},inplace=True)
```

```
5 test.GarageFinish.replace({'None':1, 'Unf':2, 'RFn':3, 'Fin': \rightarrow4},inplace=True)
```

NameError: name 'test' is not defined



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

NameError

Traceback (most recent call_

NameError: name 'test' is not defined

```
[40]: full.isnull().sum().sort_values(ascending=False)
             NameError
                                                         Traceback (most recent call_
      →last)
             <ipython-input-40-02d623efd052> in <module>
         ----> 1 full.isnull().sum().sort_values(ascending=False)
             NameError: name 'full' is not defined
[41]: train.groupby(['Neighborhood'])[['LotFrontage']].agg(['mean', 'median', 'count'])
                   LotFrontage
[41]:
                          mean median count
      Neighborhood
                     27.800000
                                 21.0
                                          15
      2
                     49.400000
                                 50.0
                                          50
      3
                     62.440000
                                 60.0
                                         250
                     68.164384
                                 66.0
                                         73
      4
      5
                     63.930233
                                 69.0
                                         43
                     77.409722
                                 76.0
                                         144
      6
      7
                     70.726804
                                 70.0
                                         194
      8
                     68.400000
                                 75.0
                                         115
      9
                     62.700000
                                         20
                                 61.5
      10
                     84.908257
                                  90.0
                                         109
                     76.462366
      mes
                                 73.0
                                         186
[43]: train['LotAreaCut'] = pd.qcut(train.LotArea,10)
      train.groupby([train['LotAreaCut']])[['LotFrontage']].
       →agg(['mean', 'median', 'count'])
[43]:
                          LotFrontage
                                 mean median count
      LotAreaCut
      (1299.999, 5000.0]
                            36.691057
                                         35.0
                                                123
      (5000.0, 7071.2]
                            55.984252
                                         52.0
                                                127
      (7071.2, 8063.1]
                            63.869919
                                         61.0
                                                123
```

```
(8063.1, 8788.8]
                          67.143939
                                     65.0
                                            132
     (8788.8, 9475.0]
                                     70.0
                                            123
                          70.544715
     (9475.0, 10193.0]
                          75.195312 76.0
                                            128
     (10193.0, 11049.9]
                          73.780488
                                     75.0
                                            123
     (11049.9, 12194.6]
                          84.649123 84.5
                                            114
     (12194.6, 14340.9]
                          84.733945
                                     85.0
                                            109
     (14340.9, 215245.0]
                          93.628866
                                     90.0
                                             97
[44]: train['LotFrontage'] = train.
      →groupby(['LotAreaCut','Neighborhood'])['LotFrontage'].transform(lambda x : x.
      \rightarrowfillna(x.median()))
[47]: columns = ["MasVnrArea", "BsmtUnfSF", "TotalBsmtSF", "GarageCars", [
      →"BsmtFinSF2", "BsmtFinSF1", "GarageArea"]
     for col in columns:
         train[col].fillna(0,inplace= True)
[48]: columns1 = ["PoolQC" , "MiscFeature", "Alley", "Fence", "FireplaceQu", __
      "GarageYrBlt", "GarageType", "BsmtExposure", "BsmtCond", "BsmtQual", "
      → "BsmtFinType2", "BsmtFinType1", "MasVnrType"]
     for col1 in columns1:
         train[col1].fillna('None',inplace = True)
[49]: columns2 = ["MSZoning", "BsmtFullBath", "BsmtHalfBath", "Utilities", [
      "Electrical", "KitchenQual", "SaleType", "Exterior1st", __
      for col2 in columns2:
         train[col2].fillna(full[col2].mode()[0],inplace = True)
            NameError
                                                    Traceback (most recent call_
      →last)
            <ipython-input-49-768063900bd8> in <module>
              4 for col2 in columns2:
                   train[col2].fillna(full[col2].mode()[0],inplace = True)
            NameError: name 'full' is not defined
```

```
[50]: train.drop("LotAreaCut",axis=1,inplace=True)
[51]: features
[51]: ['LotShape',
       'LotConfig',
       'Neighborhood',
       'HouseStyle',
       'YearBuilt',
       'ExterQual']
[52]: cols_num =
       → ["MSSubClass", "BsmtFullBath", "BsmtHalfBath", "HalfBath", "BedroomAbvGr", "KitchenAbvGr", "MoSol
                "YrSold", "YearBuilt", "YearRemodAdd", "LowQualFinSF", "GarageYrBlt"]
      for i in cols_num :
          train[i]=train[i].astype(str)
                                                        Traceback (most recent call_
             KeyError
      →last)
             /usr/local/lib/python3.7/site-packages/pandas/core/indexes/base.py in_
      →get_loc(self, key, method, tolerance)
            2897
                              try:
         -> 2898
                                  return self._engine.get_loc(casted_key)
            2899
                              except KeyError as err:
             pandas/_libs/index.pyx in pandas._libs.index.IndexEngine.get_loc()
             pandas/_libs/index.pyx in pandas._libs.index.IndexEngine.get_loc()
             pandas/_libs/hashtable_class_helper.pxi in pandas._libs.hashtable.
      →PyObjectHashTable.get_item()
             pandas/_libs/hashtable_class_helper.pxi in pandas._libs.hashtable.
      →PyObjectHashTable.get_item()
             KeyError: 'KitchenAbvGr'
```

The above exception was the direct cause of the following exception:

Traceback (most recent call_

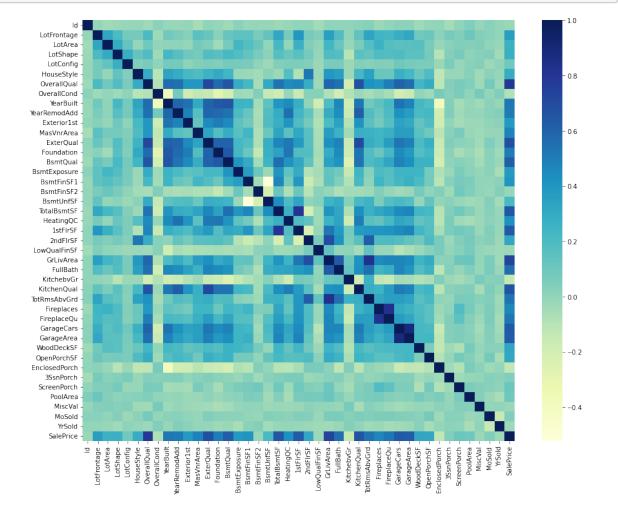
KeyError

```
→last)
             <ipython-input-52-48c69d59c9e4> in <module>
               4 for i in cols_num :
                     train[i]=train[i].astype(str)
             /usr/local/lib/python3.7/site-packages/pandas/core/frame.py in_
      →__getitem__(self, key)
                              if self.columns.nlevels > 1:
            2904
            2905
                                  return self._getitem_multilevel(key)
         -> 2906
                              indexer = self.columns.get_loc(key)
                              if is_integer(indexer):
            2907
                                  indexer = [indexer]
            2908
             /usr/local/lib/python3.7/site-packages/pandas/core/indexes/base.py in ____
      →get_loc(self, key, method, tolerance)
            2898
                                  return self._engine.get_loc(casted_key)
            2899
                              except KeyError as err:
         -> 2900
                                  raise KeyError(key) from err
            2901
            2902
                          if tolerance is not None:
             KeyError: 'KitchenAbvGr'
[53]: features
[53]: ['LotShape',
       'LotConfig',
       'Neighborhood',
       'HouseStyle',
       'YearBuilt',
       'ExterQual']
[54]: train.isnull().sum(0).sort_values(ascending=False)
```

[54]: Electrical 1 SalePrice 0 Foundation 0 RoofMatl 0 0 Exterior1st Functiol 0 Fireplaces 0 FireplaceQu 0 GarageType 0 0 Ιd

Length: 81, dtype: int64

```
[55]: #Checking correlation via heatmap:
plt.figure(figsize=(15,12))
sns.heatmap(train.corr(),cmap="YlGnBu")
plt.show()
```



```
[56]: core=train.corr()
    core_des=core.sort_values(by='SalePrice',ascending=False)
    core_des
```

[56]:		Id	LotFrontage	LotArea	LotShape	LotConfig	\
	SalePrice	-0.021673	0.361292	0.268179	0.270477	0.092499	
	OverallQual	-0.030213	0.206488	0.091392	0.190817	0.064424	
	GrLivArea	0.003138	0.353064	0.232927	0.192468	0.060986	
	ExterQual	0.003413	0.122734	0.040373	0.180443	0.051259	
	KitchenQual	0.013709	0.140458	0.055988	0.151122	0.034402	
	TotalBsmtSF	-0.025908	0.313779	0.223914	0.172605	0.042197	
	GarageCars	0.016863	0.281073	0.152484	0.194461	0.097318	
	1stFlrSF	0.003921	0.394293	0.269234	0.165273	0.048999	
	GarageArea	0.014316	0.317903	0.163680	0.161068	0.068097	
	BsmtQual	-0.033854	0.121623	0.067061	0.217596	0.083792	
	FullBath	0.005750	0.187091	0.119679	0.181745	0.060946	
	TotRmsAbvGrd	0.025227	0.339434	0.175425	0.126653	0.026261	
	FireplaceQu	0.003231	0.229629	0.177234	0.182126	0.030781	
	YearBuilt	-0.013463	0.091459	0.007462	0.226674	0.083040	
	YearRemodAdd	-0.022694	0.056648	0.007500	0.172670	0.075750	
	Foundation	-0.007603	0.023025	-0.037182	0.165594	0.047767	
	MasVnrArea	-0.053952	0.152038	0.083715	0.076833	0.042114	
	Fireplaces	-0.023122	0.239627	0.261176	0.192602	0.041528	
	HeatingQC	-0.015546	0.070610	-0.001521	0.117117	0.050443	
	BsmtFinSF1	-0.015012	0.146506	0.175320	0.127540	0.064786	
	Exterior1st	-0.001654	0.027434	-0.008814	0.100561	0.009628	
	BsmtExposure	0.015689	0.149516	0.229513	0.182028	0.069130	
	LotFrontage	-0.010181	1.000000	0.323789	0.204337	0.074351	
	WoodDeckSF	-0.030262	0.092277	0.168621	0.159403	0.056044	
	OpenPorchSF	-0.002115	0.105699	0.061861	0.079791	0.059458	
	2ndFlrSF	0.005248	0.070666	0.041026	0.084127	0.031950	
	HouseStyle	-0.007124	0.077233	-0.006970	0.127657	0.044283	
	LotShape	-0.028067	0.204337	0.303146	1.000000	0.276815	
	LotArea	-0.037689	0.323789	1.000000	0.303146	0.131158	
	${\tt BsmtUnfSF}$	-0.007499	0.127331	-0.003251	0.014493	-0.027797	
	ScreenPorch	0.001491	0.042183	0.045288	0.066678	-0.016082	
	PoolArea	0.048476	0.106658	0.036491	0.011531	0.024744	
	LotConfig	-0.015053	0.074351	0.131158	0.276815	1.000000	
	MoSold	0.023511	0.022187	0.005744	0.031968	-0.023353	
	3SsnPorch	-0.046600	0.067284	0.021362	0.034485	0.058946	
	BsmtFinSF2	-0.005802	0.064977	0.114297	0.061950	0.010835	
	MiscVal	-0.006195	0.035609	0.039072	0.029176	0.031796	
	Id	1.000000	-0.010181	-0.037689	-0.028067	-0.015053	
	${\tt LowQualFinSF}$	-0.044191	0.038046	0.005536	-0.007576	-0.024001	
	YrSold	0.000378	0.000349	-0.013699	-0.037661	0.011064	
	OverallCond	0.012925	-0.029367	-0.002723	-0.032042	0.012180	
	EnclosedPorch	0.003103	0.017687	-0.016505	-0.094430	-0.021226	

	HouseStyle	OverallQual	OverallCond	YearBuilt	YearRemodAdd	\
SalePrice	0.292263	0.795774	-0.077948	0.523608	0.507717	•
OverallQual	0.295311	1.000000	-0.090442	0.571368	0.549827	
GrLivArea	0.387912	0.589414	-0.077423	0.194397	0.286758	
ExterQual	0.252170	0.723286	-0.137724	0.597356	0.586675	
KitchenQual	0.224432	0.670627	-0.025207	0.528822	0.624563	
TotalBsmtSF	-0.028472	0.538595	-0.174931	0.400266	0.294941	
GarageCars	0.272117	0.600741	-0.185275	0.537301	0.419882	
1stFlrSF	-0.086386	0.469092	-0.144717	0.281253	0.238769	
GarageArea	0.221161	0.557230	-0.150503	0.477998	0.370000	
BsmtQual	0.241591	0.658657	-0.200155	0.656603	0.546599	
FullBath	0.354523	0.548420	-0.193473	0.467146	0.437927	
TotRmsAbvGrd	0.346882	0.420621	-0.055782	0.091220	0.188418	
FireplaceQu	0.113930	0.497400	-0.055271	0.219047	0.199877	
YearBuilt	0.335103	0.571368	-0.375572	1.000000	0.592105	
YearRemodAdd	0.293811	0.549827	0.074615	0.592105	1.000000	
Foundation	0.307361	0.577360	-0.279920	0.644117	0.573413	
MasVnrArea	0.196764	0.398785	-0.124261	0.308722	0.172590	
Fireplaces	0.116101	0.392472	-0.022406	0.144982	0.109898	
HeatingQC	0.175592	0.456379	-0.013482	0.448075	0.549420	
BsmtFinSF1	-0.024982	0.222467	-0.042987	0.249818	0.123039	
Exterior1st	0.217661	0.396564	-0.172392	0.545668	0.480334	
BsmtExposure	-0.020630	0.295445	-0.031653	0.254469	0.207194	
LotFrontage	0.077233	0.206488	-0.029367	0.091459	0.056648	
WoodDeckSF	0.116652	0.236753	-0.002675	0.223700	0.204591	
OpenPorchSF	0.194471	0.297186	-0.029800	0.184116	0.222790	
2ndFlrSF	0.529462	0.290052	0.030504	0.006968	0.137423	
HouseStyle	1.000000	0.295311	-0.103506	0.335103	0.293811	
LotShape	0.127657	0.190817	-0.032042	0.226674	0.172670	
LotArea	-0.006970	0.091392	-0.002723	0.007462	0.007500	
BsmtUnfSF	0.013420	0.308990	-0.136723	0.148836	0.180958	
ScreenPorch	-0.014635	0.066298	0.054632	-0.049972	-0.038360	
PoolArea	0.027652	0.044758	0.002321	-0.005171	-0.003426	
LotConfig	0.044283	0.064424	0.012180	0.083040	0.075750	
MoSold	0.015159	0.072531	-0.003735	0.012975	0.022020	
3SsnPorch	-0.011804	0.030991	0.025426	0.031578	0.045504	
BsmtFinSF2	-0.042940	-0.058324	0.040034	-0.048683	-0.067377	
MiscVal	-0.033895	-0.031238	0.068728	-0.034273	-0.010160	
Id	-0.007124	-0.030213	0.012925	-0.013463	-0.022694	
LowQualFinSF	-0.028556	-0.030129	0.025414	-0.183770	-0.062289	
YrSold	-0.008630	-0.026575	0.043795	-0.013250	0.036145	
OverallCond	-0.103506	-0.090442	1.000000	-0.375572	0.074615	
EnclosedPorch	-0.121171	-0.113176	0.070120	-0.387093	-0.193551	
KitchebvGr	-0.024264	-0.184077	-0.087170	-0.174630	-0.149406	

```
WoodDeckSF
                                OpenPorchSF
                                                              3SsnPorch \
                                              EnclosedPorch
                                                                0.044568
SalePrice
                     0.324758
                                   0.321142
                                                   -0.128646
OverallQual
                     0.236753
                                   0.297186
                                                  -0.113176
                                                                0.030991
                •••
GrLivArea
                     0.247120
                                   0.302569
                                                    0.012976
                                                                0.022520
                •••
ExterQual
                     0.205586
                                   0.279707
                                                  -0.149025
                                                                0.037582
KitchenQual
                     0.219112
                                   0.249783
                                                                0.022769
                                                   -0.105126
TotalBsmtSF
                     0.234476
                                   0.215140
                                                   -0.096773
                                                                0.040921
GarageCars
                     0.225462
                                   0.209968
                                                  -0.151110
                                                                0.035957
                                                  -0.064247
1stFlrSF
                     0.235439
                                   0.178859
                                                                0.059521
GarageArea
                     0.222482
                                   0.228246
                                                   -0.121194
                                                                0.035880
                •••
BsmtQual
                     0.262482
                                   0.260816
                                                  -0.203033
                                                                0.010975
                •••
FullBath
                                   0.252678
                                                   -0.114558
                     0.186147
                                                                0.035715
TotRmsAbvGrd
                     0.163218
                                   0.220052
                                                    0.005700
                                                               -0.006235
                •••
FireplaceQu
                     0.211295
                                   0.156796
                                                  -0.042915
                                                                0.037962
YearBuilt
                     0.223700
                                                   -0.387093
                                   0.184116
                                                                0.031578
YearRemodAdd
                     0.204591
                                   0.222790
                                                   -0.193551
                                                                0.045504
Foundation
                     0.176950
                                   0.228337
                                                   -0.167918
                                                                0.017907
                •••
MasVnrArea
                     0.156812
                                   0.101698
                                                   -0.109126
                                                                0.019933
                     0.198294
Fireplaces
                                   0.160924
                                                   -0.023879
                                                                0.011657
HeatingQC
                     0.129869
                                   0.172574
                                                   -0.101368
                                                                0.059826
                •••
BsmtFinSF1
                     0.204993
                                   0.072717
                                                   -0.103853
                                                                0.029177
Exterior1st
                     0.126352
                                   0.195974
                                                   -0.175786
                                                                0.001081
                                                              -0.017925
BsmtExposure
                     0.244759
                                   0.083108
                                                   -0.076415
LotFrontage
                     0.092277
                                   0.105699
                                                    0.017687
                                                                0.067284
                •••
WoodDeckSF
                     1.000000
                                   0.053686
                                                  -0.125616
                                                              -0.032642
                •••
OpenPorchSF
                     0.053686
                                   1.000000
                                                   -0.092212
                                                              -0.005191
                •••
                                   0.197938
2ndFlrSF
                     0.089902
                                                    0.063211
                                                              -0.024107
                •••
HouseStyle
                     0.116652
                                   0.194471
                                                   -0.121171
                                                              -0.011804
                •••
LotShape
                     0.159403
                                   0.079791
                                                   -0.094430
                                                                0.034485
                     0.168621
                                   0.061861
                                                   -0.016505
                                                                0.021362
LotArea
BsmtUnfSF
                    -0.005609
                                   0.128917
                                                   -0.002422
                                                                0.020806
                •••
ScreenPorch
                                   0.077150
                    -0.073884
                                                   -0.083009
                                                               -0.031481
PoolArea
                     0.069100
                                   0.032920
                                                    0.060101
                                                               -0.007417
LotConfig
                     0.056044
                                   0.059458
                                                   -0.021226
                                                                0.058946
                •••
MoSold
                     0.021509
                                   0.072416
                                                   -0.029097
                                                                0.029483
                •••
3SsnPorch
                    -0.032642
                                  -0.005191
                                                  -0.037366
                                                                1.000000
                                                              -0.030041
BsmtFinSF2
                     0.068312
                                   0.004974
                                                    0.036406
MiscVal
                    -0.009445
                                  -0.018308
                                                    0.018319
                                                                0.000340
                •••
Ιd
                    -0.030262
                                  -0.002115
                                                               -0.046600
                                                    0.003103
                •••
LowQualFinSF
                    -0.025305
                                   0.019292
                                                    0.061028
                                                              -0.004315
YrSold
                     0.022579
                                  -0.056537
                                                   -0.010035
                                                                0.018610
OverallCond
                •••
                    -0.002675
                                  -0.029800
                                                    0.070120
                                                                0.025426
EnclosedPorch
                    -0.125616
                                  -0.092212
                                                    1.000000
                                                              -0.037366
KitchebvGr
                    -0.089916
                                  -0.069810
                                                    0.037213
                                                              -0.024635
                ScreenPorch
                              PoolArea
                                          MiscVal
                                                      MoSold
                                                                YrSold
                                                                         SalePrice
```

0.046124 -0.028882

1.000000

0.099490 -0.021203

SalePrice

0.111415

```
OverallQual
                 0.066298
                           0.044758 -0.031238
                                               0.072531 -0.026575
                                                                    0.795774
GrLivArea
                 0.107768
                           0.115999 -0.001637
                                               0.057296 -0.035971
                                                                    0.734968
ExterQual
                 0.013799
                           0.012827 -0.039184
                                               0.044280 -0.005866
                                                                    0.686756
KitchenQual
                 0.031731
                           0.046646 -0.054012
                                               0.052985 0.002788
                                                                    0.662236
TotalBsmtSF
                 0.092574
                           0.036089 -0.018514
                                               0.025775 -0.014880
                                                                    0.651153
GarageCars
                 0.050909
                           0.019446 -0.043004
                                               0.039520 -0.038494
                                                                    0.641047
1stFlrSF
                           0.062756 -0.021045
                                               0.040558 -0.012810
                 0.095042
                                                                    0.631530
GarageArea
                 0.053015
                           0.026642 -0.027236
                                               0.032470 -0.027209
                                                                    0.629217
BsmtQual
                 0.030462
                           0.012592 -0.043146
                                               0.016445 -0.017597
                                                                    0.624758
FullBath
                           0.045656 -0.014122
                                               0.054756 -0.018697
                 -0.007496
                                                                    0.562165
TotRmsAbvGrd
                 0.060920
                           0.059728 0.025297
                                               0.039221 -0.033901
                                                                    0.537769
FireplaceQu
                 0.147088 0.032945 -0.024111
                                               0.054510 -0.050337
                                                                    0.527256
YearBuilt
                 -0.049972 -0.005171 -0.034273
                                               0.012975 -0.013250
                                                                    0.523608
YearRemodAdd
                 -0.038360 -0.003426 -0.010160
                                               0.022020 0.036145
                                                                    0.507717
Foundation
                 -0.053332 -0.016876 -0.064806
                                               0.021040 -0.033946
                                                                    0.502768
MasVnrArea
                 0.064258 - 0.020350 - 0.029346 - 0.005006 - 0.007215
                                                                    0.477810
Fireplaces
                 0.051142 -0.024315
                                                                    0.469862
HeatingQC
                 -0.034613 -0.057745 -0.026780
                                               0.015935 -0.009365
                                                                    0.428024
BsmtFinSF1
                 0.068389
                           0.052981 0.004741 -0.004281
                                                        0.015767
                                                                    0.409384
Exterior1st
                 -0.049561 -0.025228 -0.024777
                                               0.004053
                                                         0.014829
                                                                    0.383349
BsmtExposure
                 0.057929
                           0.013197 -0.005497
                                               0.018078 -0.063932
                                                                    0.382907
LotFrontage
                           0.106658 0.035609
                                               0.022187
                 0.042183
                                                         0.000349
                                                                    0.361292
WoodDeckSF
                 -0.073884 0.069100 -0.009445
                                               0.021509
                                                         0.022579
                                                                    0.324758
OpenPorchSF
                 0.077150
                           0.032920 -0.018308
                                               0.072416 -0.056537
                                                                    0.321142
2ndFlrSF
                           0.074258 0.016497
                                               0.034665 -0.027775
                                                                    0.320532
                 0.041504
HouseStyle
                 -0.014635
                           0.027652 -0.033895
                                               0.015159 -0.008630
                                                                    0.292263
                 0.066678
LotShape
                           0.011531 0.029176
                                               0.031968 -0.037661
                                                                    0.270477
                                                                    0.268179
LotArea
                 0.045288
                           0.036491 0.039072
                                               0.005744 -0.013699
BsmtUnfSF
                 -0.012494 -0.034936 -0.023813
                                               0.033997 -0.040953
                                                                    0.214460
ScreenPorch
                           0.056293 0.031915
                                               0.023151 0.010608
                  1.000000
                                                                    0.111415
                           1.000000
                                     0.031965 -0.018601 -0.064030
PoolArea
                 0.056293
                                                                    0.099490
LotConfig
                 -0.016082
                           0.024744
                                     0.031796 -0.023353
                                                         0.011064
                                                                    0.092499
MoSold
                 0.023151 -0.018601 -0.006544
                                               1.000000 -0.145267
                                                                    0.046124
3SsnPorch
                 -0.031481 -0.007417
                                     0.000340
                                               0.029483
                                                         0.018610
                                                                    0.044568
BsmtFinSF2
                 0.088774
                           0.046350
                                     0.004905 -0.015362
                                                         0.031618
                                                                   -0.011422
MiscVal
                 0.031915
                           0.031965
                                     1.000000 -0.006544
                                                         0.004878
                                                                   -0.021203
Τd
                 0.001491
                           0.048476 -0.006195 0.023511
                                                                   -0.021673
                                                         0.000378
                 0.026756
                           0.066428 -0.003807 -0.022268 -0.028964
                                                                   -0.025625
LowQualFinSF
YrSold
                 0.010608 -0.064030 0.004878 -0.145267
                                                         1.000000
                                                                   -0.028882
OverallCond
                 0.054632
                           0.002321
                                     0.068728 -0.003735
                                                         0.043795
                                                                   -0.077948
EnclosedPorch
                 -0.083009
                           0.060101
                                     0.018319 -0.029097 -0.010035
                                                                   -0.128646
KitchebvGr
                 -0.051696 -0.013481 0.062318 0.026554 0.031623
                                                                   -0.135946
```

[43 rows x 43 columns]

```
[57]: core_des_top=core_des[(core_des['SalePrice']>0.5) & (core_des['SalePrice'] !=1)] core_des_top.index
```

```
[57]: Index(['OverallQual', 'GrLivArea', 'ExterQual', 'KitchenQual', 'TotalBsmtSF',
             'GarageCars', '1stFlrSF', 'GarageArea', 'BsmtQual', 'FullBath',
             'TotRmsAbvGrd', 'FireplaceQu', 'YearBuilt', 'YearRemodAdd',
             'Foundation'],
            dtype='object')
[58]: features.append(['OverallQual', 'GrLivArea', 'Neighborhood', 'ExterQual', __
       'TotalBsmtSF', 'GarageCars', '1stFlrSF', 'GarageArea', 'BsmtQual',
             'FullBath', 'GarageFinish', 'TotRmsAbvGrd', 'FireplaceQu',
             'Foundation'])
[59]: core_des_low=core_des[(core_des['SalePrice']<0)]
      core_des_low.index
[59]: Index(['BsmtFinSF2', 'MiscVal', 'Id', 'LowQualFinSF', 'YrSold', 'OverallCond',
             'EnclosedPorch', 'KitchebvGr'],
            dtype='object')
[60]: features.append(['BsmtFinSF2', 'MiscVal', 'Id', 'OverallCond', 'EnclosedPorch'])
[61]: features=['LotShape',
       'LotConfig',
       'Neighborhood',
       'HouseStyle',
       'Exterior1st',
       'ExterQual',
       'Foundation',
       'BsmtQual',
       'BsmtExposure',
       'HeatingQC',
       'KitchenQual',
       'FireplaceQu',
       'GarageType',
       'GarageFinish',
       'OverallQual',
        'GrLivArea',
        'Neighborhood',
        'ExterQual',
        'KitchenQual',
        'TotalBsmtSF',
        'GarageCars',
        '1stFlrSF',
        'GarageArea',
        'BsmtQual',
        'FullBath',
        'GarageFinish',
```

```
'TotRmsAbvGrd',
         'FireplaceQu',
         'Foundation',
        'BsmtFinSF2', 'MiscVal', 'Id', 'OverallCond', 'EnclosedPorch', 'SalePrice']
[62]: train_d=train.copy()
      train_d=train_d[features]
[63]: train_d
[63]:
             LotShape LotConfig Neighborhood HouseStyle Exterior1st ExterQual \
      0
                    1
                                               7
                                                            4
                                                                           5
                                                                                       3
                                 1
                                                            3
                                                                           3
                                                                                       2
      1
                    1
                                 4
                                               8
      2
                    2
                                               7
                                                            4
                                                                           5
                                                                                       3
                                 1
      3
                    2
                                 2
                                               7
                                                            4
                                                                           3
                                                                                       2
                    2
      4
                                 4
                                              10
                                                            4
                                                                           5
                                                                                       3
      1455
                                                                           5
                                                                                       2
                     1
                                 1
                                               6
                                                            4
      1456
                                                                                       2
                    1
                                 1
                                               6
                                                            3
                                                                           4
      1457
                     1
                                 1
                                               7
                                                            4
                                                                           6
                                                                                       4
      1458
                                                            3
                                                                           3
                                                                                       2
                     1
                                 1
                                             mes
      1459
                     1
                                 1
                                               3
                                                            3
                                                                           3
                                                                                       3
                         BsmtQual
                                                                   GarageFinish \
            Foundation
                                    BsmtExposure
                                                   HeatingQC
      0
                      4
                                 4
                                                            5
                                                                             RFn
      1
                      2
                                 4
                                                4
                                                            5
                                                                             R.Fn
                                                3
                                                            5
      2
                      4
                                 4
                                                                             RFn
      3
                      2
                                 3
                                                2
                                                            4
                                                                             Unf
      4
                      4
                                                3
                                                            5
                                 4
                                                                             RFn
      1455
                      4
                                 4
                                                2
                                                            5
                                                                             RFn
                      2
                                                2
                                                            3
      1456
                                 4
                                                                             Unf
                                                2
      1457
                      2
                                 3
                                                            5
                                                                             RFn
                      2
                                                3
      1458
                                 3
                                                            4
                                                                             Unf
      1459
                                 3
                                                2
                                                                             Fin
             TotRmsAbvGrd FireplaceQu Foundation BsmtFinSF2
                                                                   MiscVal
                                                                               Ιd
      0
                         8
                                      1
                                                  4
                                                                0
                                                                          0
                                                                                1
                         6
                                      3
                                                  2
                                                                0
                                                                                2
      1
                                                                          0
      2
                         6
                                      3
                                                  4
                                                                0
                                                                          0
                                                                                3
      3
                         7
                                      4
                                                  2
                                                                0
                                                                                4
                                                                          0
      4
                         9
                                      3
                                                  4
                                                                0
                                                                          0
                                                                                5
      1455
                         7
                                      3
                                                  4
                                                                0
                                                                          0
                                                                             1456
      1456
                         7
                                      3
                                                  2
                                                                             1457
                                                              163
                                                                          0
      1457
                         9
                                      4
                                                  2
                                                                             1458
                                                                       2500
                                                  2
                         5
      1458
                                                            1029
                                                                          0
                                                                             1459
```

```
OverallCond EnclosedPorch
                                         SalePrice
      0
                      5
                                            208500
      1
                      8
                                      0
                                            181500
                      5
      2
                                      0
                                            223500
      3
                      5
                                    272
                                            140000
      4
                      5
                                      0
                                            250000
                      5
                                            175000
      1455
                                      0
                      6
      1456
                                      0
                                            210000
      1457
                      9
                                      0
                                            266500
      1458
                      6
                                    112
                                            142125
      1459
                      6
                                      0
                                            147500
      [1458 rows x 35 columns]
[65]: train_d=pd.get_dummies(train_d)
[66]: train_d.columns
[66]: Index(['LotShape', 'LotConfig', 'HouseStyle', 'Exterior1st', 'ExterQual',
             'Foundation', 'BsmtQual', 'BsmtExposure', 'HeatingQC', 'KitchenQual',
             'FireplaceQu', 'OverallQual', 'GrLivArea', 'ExterQual', 'KitchenQual',
             'TotalBsmtSF', 'GarageCars', '1stFlrSF', 'GarageArea', 'BsmtQual',
             'FullBath', 'TotRmsAbvGrd', 'FireplaceQu', 'Foundation', 'BsmtFinSF2',
             'MiscVal', 'Id', 'OverallCond', 'EnclosedPorch', 'SalePrice',
             'Neighborhood_1', 'Neighborhood_2', 'Neighborhood_3', 'Neighborhood_4',
             'Neighborhood_5', 'Neighborhood_6', 'Neighborhood_7', 'Neighborhood_8',
             'Neighborhood_9', 'Neighborhood_10', 'Neighborhood_mes',
             'GarageType_2Types', 'GarageType_Attchd', 'GarageType_Basment',
             'GarageType_BuiltIn', 'GarageType_CarPort', 'GarageType_Detchd',
             'GarageType_None', 'GarageFinish_Fin', 'GarageFinish_None',
             'GarageFinish_RFn', 'GarageFinish_Unf', 'Neighborhood_1',
             'Neighborhood_2', 'Neighborhood_3', 'Neighborhood_4', 'Neighborhood_5',
             'Neighborhood_6', 'Neighborhood_7', 'Neighborhood_8', 'Neighborhood_9',
             'Neighborhood_10', 'Neighborhood_mes', 'GarageFinish_Fin',
             'GarageFinish_None', 'GarageFinish_RFn', 'GarageFinish_Unf'],
            dtype='object')
[67]: for col in train_d.columns:
          print(col,)
     LotShape
     LotConfig
     HouseStyle
     Exterior1st
```

1459

6

1

2

290

0 1460

ExterQual

Foundation

BsmtQual

BsmtExposure

HeatingQC

KitchenQual

FireplaceQu

OverallQual

 ${\tt GrLivArea}$

ExterQual

KitchenQual

 ${\tt TotalBsmtSF}$

GarageCars

1stFlrSF

GarageArea

BsmtQual

FullBath

 ${\tt TotRmsAbvGrd}$

FireplaceQu

Foundation

BsmtFinSF2

MiscVal

Ιd

OverallCond

 ${\tt EnclosedPorch}$

SalePrice

Neighborhood_1

Neighborhood_2

Neighborhood_3

Neighborhood_4

 ${\tt Neighborhood_5}$

Neighborhood_6

Neighborhood_7

Neighborhood_8

Neighborhood 9

Neighborhood_10

Neighborhood_mes

GarageType_2Types

GarageType_Attchd

GarageType_Basment

GarageType_BuiltIn

GarageType_CarPort

GarageType_Detchd

GarageType_None

GarageFinish_Fin

GarageFinish_None

GarageFinish_RFn

GarageFinish_Unf

```
Neighborhood 1
     Neighborhood_2
     Neighborhood_3
     Neighborhood 4
     Neighborhood 5
     Neighborhood 6
     Neighborhood 7
     Neighborhood 8
     Neighborhood 9
     Neighborhood_10
     Neighborhood_mes
     GarageFinish_Fin
     GarageFinish_None
     GarageFinish_RFn
     GarageFinish_Unf
[68]: from sklearn.linear_model import LinearRegression
      from sklearn.model_selection import train_test_split
      predictors=train_d.drop(['Id', 'SalePrice'], axis=1)
      target=train.SalePrice
      x_train,x_cv,y_train,y_cv=train_test_split(predictors,target,test_size=0.
       →35,random_state=0)
[69]: from sklearn.model_selection import KFold,cross_val_score
      logreg = LinearRegression()
      logreg.fit(x_train, y_train)
[69]: LinearRegression()
[70]: logreg.predict(x_cv)
[70]: array([144139.43782158, 152744.003098 , 43963.37022548, 127465.94601343,
             269474.27559947, 120459.03952188, 208047.63481301, 117075.56596074,
             139363.55870788, 141787.84145595, 200122.06941272, 101730.18720464,
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                              79123.65915782, 165009.0824828 , 389122.19903686,
             217246.115668 ,
             303928.97626247, 106442.9142123 , 121999.08365686, 107320.6988759 ,
             105894.79780088, 225869.4817535 , 201594.31318564, 100798.1195574 ,
             140659.93627001, 396144.61414194, 174525.60224797, 221849.04495508,
             144402.43231512, 240588.32413429, 87665.39512821, 113179.88204427,
             244369.69716486, 280094.39038557, 94168.41635413, 68620.57308695,
             129230.2181316 , 246004.26136834, 127420.15010849, 117961.18099309,
             184893.35462445, 176174.65556824, 315266.00303225, 327251.94531695,
             422163.98599134, 261506.8430623 , 247736.76714408, 98545.3165135 ,
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             307280.356098 , 83519.49957952, 195376.21498372, 126644.05136689,
             178446.22793128, 125911.58601471, 125170.75335233, 267354.49417125,
```

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248035.79653976, 226585.44026249, 247006.78802771, 227329.84475522,
213483.28690774, 129677.14535453, 268924.79587656, 228703.04988093,
210169.65185134, 140090.01812955, 135344.0279052 , 139194.34722826,
147080.38203471, 95399.2430098, 287521.86415949, 143295.71991319,
128640.3272776 , 290851.08193852, 356965.4888421 , 94336.38001244,
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[71]: from sklearn.metrics import accuracy_score
            y_pred=logreg.predict(x_cv)
            logreg.score(x_cv,y_cv)
[71]: 0.8637143770467263
[72]: features train=['LotShape', 'LotConfig', 'Neighborhood', 'HouseStyle', 'LotConfig', 'Neighborhood', 'N
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