

Project_Feature Engineering

November 21, 2022

```
[1]: import warnings
warnings.filterwarnings('ignore')
```

```
[2]: # Import essential libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
from sklearn.impute import SimpleImputer
from sklearn.base import TransformerMixin
```

```
[3]: # import dataset
data = pd.read_csv('PEP1.csv')
data.head()
```

```
[3]:
```

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

	LandContour	Utilities	...	PoolArea	PoolQC	Fence	MiscFeature	MiscVal	MoSold	\
0	Lvl	AllPub	...	0	NaN	NaN	NaN	0	2	
1	Lvl	AllPub	...	0	NaN	NaN	NaN	0	5	
2	Lvl	AllPub	...	0	NaN	NaN	NaN	0	9	
3	Lvl	AllPub	...	0	NaN	NaN	NaN	0	2	
4	Lvl	AllPub	...	0	NaN	NaN	NaN	0	12	

	YrSold	SaleType	SaleCondition	SalePrice
0	2008	WD	Normal	208500
1	2007	WD	Normal	181500
2	2008	WD	Normal	223500
3	2006	WD	Abnorml	140000
4	2008	WD	Normal	250000

[5 rows x 81 columns]

1 1. Understand the dataset:

- Identify the shape of the dataset
- Identify variables with null values
- Identify variables with unique values

```
[4]: # a. Identify the shape of the dataset
print('Shape of the dataset :',data.shape)
```

Shape of the dataset : (1460, 81)

```
[5]: # b. Identify variables with null values

data_null = pd.DataFrame(data[data.columns[data.isna().any()]].isna().sum())
data_null['percentage'] = data.isna().sum() / data.shape[0]*100
data_null = data_null.reset_index()
data_null = data_null.rename(columns={'index': 'Variables',0:'Count'})
data_null = data_null.sort_values(by='Count',ascending=False)
data_null
```

```
[5]:
```

	Variables	Count	percentage
16	PoolQC	1453	99.520548
18	MiscFeature	1406	96.301370
1	Alley	1369	93.767123
17	Fence	1179	80.753425
10	FireplaceQu	690	47.260274
0	LotFrontage	259	17.739726
11	GarageType	81	5.547945
12	GarageYrBlt	81	5.547945
13	GarageFinish	81	5.547945
14	GarageQual	81	5.547945
15	GarageCond	81	5.547945
6	BsmtExposure	38	2.602740
8	BsmtFinType2	38	2.602740
7	BsmtFinType1	37	2.534247
5	BsmtCond	37	2.534247
4	BsmtQual	37	2.534247
3	MasVnrArea	8	0.547945
2	MasVnrType	8	0.547945
9	Electrical	1	0.068493

```
[6]: # c. Identify variables with unique values
```

```
for i in data.columns:
```

```
print({i : data[i].unique()})
```

```
{'Id': array([ 1, 2, 3, ..., 1458, 1459, 1460], dtype=int64)}
{'MSSubClass': array([ 60, 20, 70, 50, 190, 45, 90, 120, 30, 85, 80,
160, 75,
180, 40], dtype=int64)}
{'MSZoning': array(['RL', 'RM', 'C (all)', 'FV', 'RH'], dtype=object)}
{'LotFrontage': array([ 65., 80., 68., 60., 84., 85., 75., nan, 51.,
50., 70.,
91., 72., 66., 101., 57., 44., 110., 98., 47., 108., 112.,
74., 115., 61., 48., 33., 52., 100., 24., 89., 63., 76.,
81., 95., 69., 21., 32., 78., 121., 122., 40., 105., 73.,
77., 64., 94., 34., 90., 55., 88., 82., 71., 120., 107.,
92., 134., 62., 86., 141., 97., 54., 41., 79., 174., 99.,
67., 83., 43., 103., 93., 30., 129., 140., 35., 37., 118.,
87., 116., 150., 111., 49., 96., 59., 36., 56., 102., 58.,
38., 109., 130., 53., 137., 45., 106., 104., 42., 39., 144.,
114., 128., 149., 313., 168., 182., 138., 160., 152., 124., 153.,
46.] )}
{'LotArea': array([ 8450, 9600, 11250, ..., 17217, 13175, 9717], dtype=int64)}
{'Street': array(['Pave', 'Grv1'], dtype=object)}
{'Alley': array([nan, 'Grv1', 'Pave'], dtype=object)}
{'LotShape': array(['Reg', 'IR1', 'IR2', 'IR3'], dtype=object)}
{'LandContour': array(['Lvl', 'Bnk', 'Low', 'HLS'], dtype=object)}
{'Utilities': array(['AllPub', 'NoSeWa'], dtype=object)}
{'LotConfig': array(['Inside', 'FR2', 'Corner', 'CulDSac', 'FR3'],
dtype=object)}
{'LandSlope': array(['Gtl', 'Mod', 'Sev'], dtype=object)}
{'Neighborhood': array(['CollgCr', 'Veenker', 'Crawfor', 'NoRidge', 'Mitchel',
'Somerst',
'NWAmes', 'OldTown', 'BrkSide', 'Sawyer', 'NridgHt', 'mes',
'SawyerW', 'IDOTRR', 'MeadowV', 'Edwards', 'Timber', 'Gilbert',
'StoneBr', 'ClearCr', 'NPkVill', 'Blmngtn', 'BrDale', 'SWISU',
'Blueste'], dtype=object)}
{'Condition1': array(['Norm', 'Feedr', 'PosN', 'Artery', 'RRAe', 'RRNn', 'RRAn',
'PosA',
'RRNe'], dtype=object)}
{'Condition2': array(['Norm', 'Artery', 'RRNn', 'Feedr', 'PosN', 'PosA', 'RRAn',
'RRAe'],
dtype=object)}
{'BldgType': array(['1Fam', '2fmCon', 'Duplex', 'TwnhsE', 'Twnhs'],
dtype=object)}
{'HouseStyle': array(['2Story', '1Story', '1.5Fin', '1.5Unf', 'SFoyer', 'SLvl',
'2.5Unf',
'2.5Fin'], dtype=object)}
{'OverallQual': array([ 7, 6, 8, 5, 9, 4, 10, 3, 1, 2], dtype=int64)}
{'OverallCond': array([5, 8, 6, 7, 4, 2, 3, 9, 1], dtype=int64)}
```

```

{'YearBuilt': array([2003, 1976, 2001, 1915, 2000, 1993, 2004, 1973, 1931, 1939,
1965,
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2007, 1951, 1957, 1927, 1920, 1966, 1959, 1994, 1954, 1953, 1955,
1983, 1975, 1997, 1934, 1963, 1981, 1964, 1999, 1972, 1921, 1945,
1982, 1998, 1956, 1948, 1910, 1995, 1991, 2009, 1950, 1961, 1977,
1985, 1979, 1885, 1919, 1990, 1969, 1935, 1988, 1971, 1952, 1936,
1923, 1924, 1984, 1926, 1940, 1941, 1987, 1986, 2008, 1908, 1892,
1916, 1932, 1918, 1912, 1947, 1925, 1900, 1980, 1989, 1992, 1949,
1880, 1928, 1978, 1922, 1996, 2010, 1946, 1913, 1937, 1942, 1938,
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1872, 1905], dtype=int64)}
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1962, 2007, 1960, 2001, 1967, 2004, 2008, 1997, 1959, 1990, 1955,
1983, 1980, 1966, 1963, 1987, 1964, 1972, 1996, 1998, 1989, 1953,
1956, 1968, 1981, 1992, 2009, 1982, 1961, 1993, 1999, 1985, 1979,
1977, 1969, 1958, 1991, 1971, 1952, 1975, 2010, 1984, 1986, 1994,
1988, 1954, 1957, 1951, 1978, 1974], dtype=int64)}
{'RoofStyle': array(['Gable', 'Hip', 'Gambrel', 'Mansard', 'Flat', 'Shed'],
dtype=object)}
{'RoofMatl': array(['CompShg', 'WdShngl', 'Metal', 'WdShake', 'Membran',
'Tar&Grv',
'Roll', 'ClyTile'], dtype=object)}
{'Exterior1st': array(['VinylSd', 'MetalSd', 'Wd Sdng', 'HdBoard', 'BrkFace',
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'CemntBd', 'Plywood', 'AsbShng', 'Stucco', 'BrkComm', 'AsphShn',
'Stone', 'ImStucc', 'CBlock'], dtype=object)}
{'Exterior2nd': array(['VinylSd', 'MetalSd', 'Wd Shng', 'HdBoard', 'Plywood',
'Wd Sdng',
'CmentBd', 'BrkFace', 'Stucco', 'AsbShng', 'Brk Cmn', 'ImStucc',
'AsphShn', 'Stone', 'Other', 'CBlock'], dtype=object)}
{'MasVnrType': array(['BrkFace', 'None', 'Stone', 'BrkCmn', nan], dtype=object)}
{'MasVnrArea': array([1.960e+02, 0.000e+00, 1.620e+02, 3.500e+02, 1.860e+02,
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2.860e+02, 3.060e+02, 2.120e+02, 1.800e+02, 3.800e+02, 2.810e+02,
6.400e+02, 2.000e+02, 2.460e+02, 1.320e+02, 6.500e+02, 1.010e+02,
4.120e+02, 2.720e+02, 4.560e+02, 1.031e+03, 1.780e+02, 5.730e+02,
3.440e+02, 2.870e+02, 1.670e+02, 1.115e+03, 4.000e+01, 1.040e+02,
5.760e+02, 4.430e+02, 4.680e+02, 6.600e+01, 2.200e+01, 2.840e+02,
7.600e+01, 2.030e+02, 6.800e+01, 1.830e+02, 4.800e+01, 2.800e+01,
3.360e+02, 6.000e+02, 7.680e+02, 4.800e+02, 2.200e+02, 1.840e+02,
1.129e+03, 1.160e+02, 1.350e+02, 2.660e+02, 8.500e+01, 3.090e+02,
1.360e+02, 2.880e+02, 7.000e+01, 3.200e+02, 5.000e+01, 1.200e+02,
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6.530e+02, 1.120e+02, 4.910e+02, 2.680e+02, 7.480e+02, 9.800e+01,
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1.540e+02, 4.250e+02, 4.500e+01, 1.378e+03, 3.370e+02, 1.490e+02,
1.430e+02, 5.100e+01, 1.710e+02, 2.340e+02, 6.300e+01, 7.660e+02,
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7.310e+02, 4.480e+02, 2.940e+02, 3.100e+02, 2.370e+02, 4.260e+02,
9.600e+01, 4.380e+02, 1.940e+02, 1.190e+02]])}
{'ExterQual': array(['Gd', 'TA', 'Ex', 'Fa'], dtype=object)}
{'ExterCond': array(['TA', 'Gd', 'Fa', 'Po', 'Ex'], dtype=object)}
{'Foundation': array(['PConc', 'CBlock', 'BrkTil', 'Wood', 'Slab', 'Stone'],
dtype=object)}
{'BsmtQual': array(['Gd', 'TA', 'Ex', nan, 'Fa'], dtype=object)}
{'BsmtCond': array(['TA', 'Gd', nan, 'Fa', 'Po'], dtype=object)}
{'BsmtExposure': array(['No', 'Gd', 'Mn', 'Av', nan], dtype=object)}

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{'BsmtFinType1': array(['GLQ', 'ALQ', 'Unf', 'Rec', 'BLQ', nan, 'LwQ'],
dtype=object)}
{'BsmtFinSF1': array([ 706,  978,  486,  216,  655,  732, 1369,  859,    0,
851,  906,
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dtype=int64)}
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175, 820, 1474, 264, 479, 147, 232, 380, 544, 294, 258,
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 402000, 423000, 230500, 173500, 103600, 257500, 372500, 159434,
 285000, 227875, 148800, 392000, 194700, 755000, 335000, 108480,
 141500, 89000, 123500, 138500, 196000, 312500, 361919, 213000,

```

55000, 302000, 254000, 179540, 52000, 102776, 189000, 130500,
159500, 341000, 103000, 236500, 131400, 93500, 239900, 299800,
236000, 265979, 260400, 275500, 158900, 179400, 215200, 337000,
264132, 216837, 538000, 134900, 102000, 395000, 221500, 175900,
187100, 161500, 233000, 107900, 160200, 146800, 269790, 143500,
485000, 582933, 227680, 135500, 159950, 144500, 55993, 157900,
224900, 271000, 224000, 183000, 139500, 232600, 147400, 237000,
139950, 174900, 133500, 189950, 250580, 248900, 169000, 200500,
66500, 303477, 132250, 328900, 122900, 154500, 118858, 142953,
611657, 125500, 255000, 154300, 173733, 75000, 35311, 238000,
176500, 145900, 169990, 193000, 117500, 184900, 253000, 239799,
244400, 150900, 197500, 172000, 116500, 214900, 178900, 37900,
99500, 182000, 167500, 85500, 178400, 336000, 159895, 255900,
117000, 395192, 195000, 197000, 348000, 173900, 337500, 121600,
206000, 232000, 136905, 119200, 227000, 203000, 213490, 194000,
287000, 293077, 310000, 119750, 84000, 315500, 262280, 278000,
139600, 556581, 84900, 176485, 200141, 185850, 328000, 167900,
151400, 91500, 138800, 155900, 83500, 252000, 92900, 176432,
274725, 134500, 184100, 133700, 118400, 212900, 163900, 259000,
239500, 94000, 424870, 174500, 116900, 201800, 218000, 235128,
108959, 233170, 245350, 625000, 171900, 154900, 392500, 745000,
186700, 104900, 262000, 219210, 116050, 271900, 229456, 80500,
137900, 367294, 101800, 138887, 265900, 248328, 465000, 186500,
169900, 171750, 294000, 165400, 301500, 99900, 128900, 183900,
378500, 381000, 185750, 68400, 150500, 281000, 333168, 206900,
295493, 111000, 156500, 72500, 52500, 155835, 108500, 283463,
410000, 156932, 144152, 216000, 274300, 466500, 58500, 237500,
377500, 246578, 281213, 137450, 193879, 282922, 257000, 223000,
274970, 182900, 192140, 143750, 64500, 394617, 149700, 149300,
121000, 179600, 92000, 287090, 266500, 142125, 147500],
dtype=int64)}

```

2. 2. Generate a separate dataset for numerical and categorical variables

```

[7]: data_categorical = data.select_dtypes(exclude=np.number)
data_numerical = data.select_dtypes(include=np.number)

```

```

[8]: data_categorical.head()

```

```

[8]:  MSZoning Street Alley LotShape LandContour Utilities LotConfig LandSlope \
0      RL   Pave   NaN      Reg          Lvl    AllPub    Inside    Gtl
1      RL   Pave   NaN      Reg          Lvl    AllPub      FR2    Gtl
2      RL   Pave   NaN      IR1          Lvl    AllPub    Inside    Gtl
3      RL   Pave   NaN      IR1          Lvl    AllPub    Corner    Gtl
4      RL   Pave   NaN      IR1          Lvl    AllPub      FR2    Gtl

```

	Neighborhood	Condition1	...	GarageType	GarageFinish	GarageQual	GarageCond	\
0	CollgCr	Norm	...	Attchd	RFn	TA	TA	
1	Veenker	Feedr	...	Attchd	RFn	TA	TA	
2	CollgCr	Norm	...	Attchd	RFn	TA	TA	
3	Crawfor	Norm	...	Detchd	Unf	TA	TA	
4	NoRidge	Norm	...	Attchd	RFn	TA	TA	

	PavedDrive	PoolQC	Fence	MiscFeature	SaleType	SaleCondition
0	Y	NaN	NaN	NaN	WD	Normal
1	Y	NaN	NaN	NaN	WD	Normal
2	Y	NaN	NaN	NaN	WD	Normal
3	Y	NaN	NaN	NaN	WD	Abnorml
4	Y	NaN	NaN	NaN	WD	Normal

[5 rows x 43 columns]

```
[9]: data_numerical.head()
```

```
[9]:
```

	Id	MSSubClass	LotFrontage	LotArea	OverallQual	OverallCond	YearBuilt	\
0	1	60	65.0	8450	7	5	2003	
1	2	20	80.0	9600	6	8	1976	
2	3	60	68.0	11250	7	5	2001	
3	4	70	60.0	9550	7	5	1915	
4	5	60	84.0	14260	8	5	2000	

	YearRemodAdd	MasVnrArea	BsmtFinSF1	...	WoodDeckSF	OpenPorchSF	\
0	2003	196.0	706	...	0	61	
1	1976	0.0	978	...	298	0	
2	2002	162.0	486	...	0	42	
3	1970	0.0	216	...	0	35	
4	2000	350.0	655	...	192	84	

	EnclosedPorch	3SsnPorch	ScreenPorch	PoolArea	MiscVal	MoSold	YrSold	\
0	0	0	0	0	0	2	2008	
1	0	0	0	0	0	5	2007	
2	0	0	0	0	0	9	2008	
3	272	0	0	0	0	2	2006	
4	0	0	0	0	0	12	2008	

	SalePrice
0	208500
1	181500
2	223500
3	140000
4	250000

[5 rows x 38 columns]

3. EDA of numerical variables:

- Missing value treatment
- Identify the skewness and distribution
- Identify significant variables using a correlation matrix
- Pair plot for distribution and density

```
[10]: # a. Missing value treatment

class SimpleImputerLookup(SimpleImputer):
    def __init__(self, fill_values=None, **kwargs):
        kwargs["strategy"] = "constant"
        super().__init__(**kwargs)
        self.fill_values = fill_values
    def fit(self, X):
        super().fit(X)
        if isinstance(self.fill_values, dict) and hasattr(X, "columns"):
            self.statistics_ = np.array([self.fill_values[c] for c in X.
→columns])
        else:
            self.statistics_ = np.array(self.fill_values)
        return self

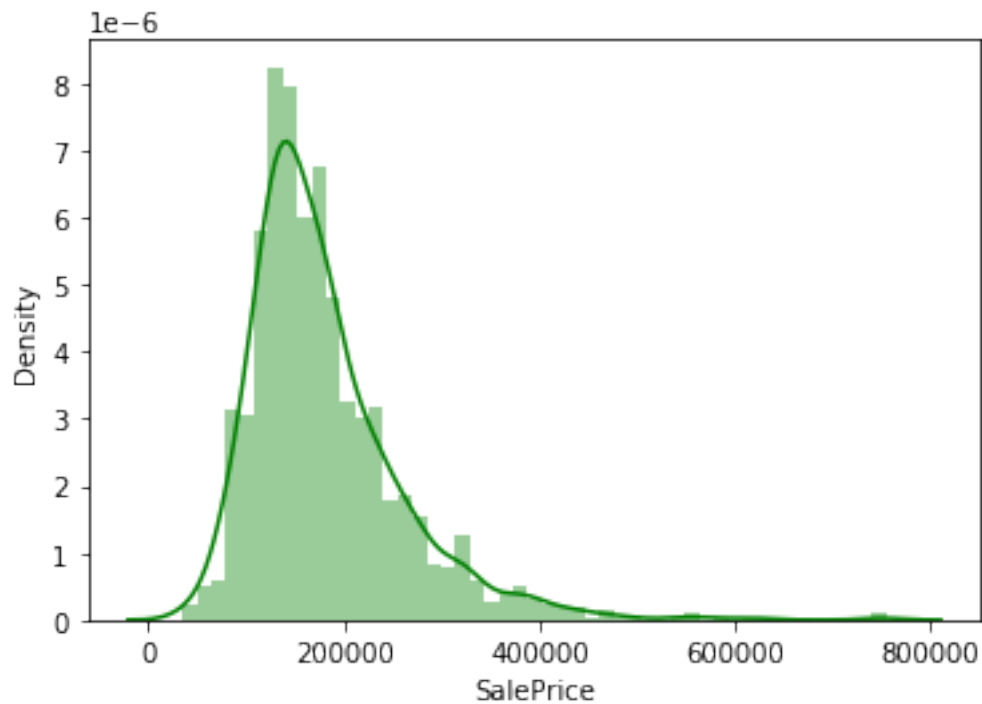
numeric_cols = [data.select_dtypes(exclude=['object']).columns]
impute_numeric_data = []
for col in numeric_cols:
    imp = SimpleImputerLookup(fill_values=data_numeric[col].mean())
    impute_numeric_data = pd.DataFrame(imp.fit_transform(data_numeric))
    impute_numeric_data.columns = data_numeric.columns
    impute_numeric_data.append(impute_numeric_data)

[11]: print('Before implementation of Simple Imputer on numeric data :',data.isna().
→sum().any())
print('After implementation of Simple Imputer on numeric data :
→',impute_numeric_data.isna().sum().any())
```

Before implementation of Simple Imputer on numeric data : True

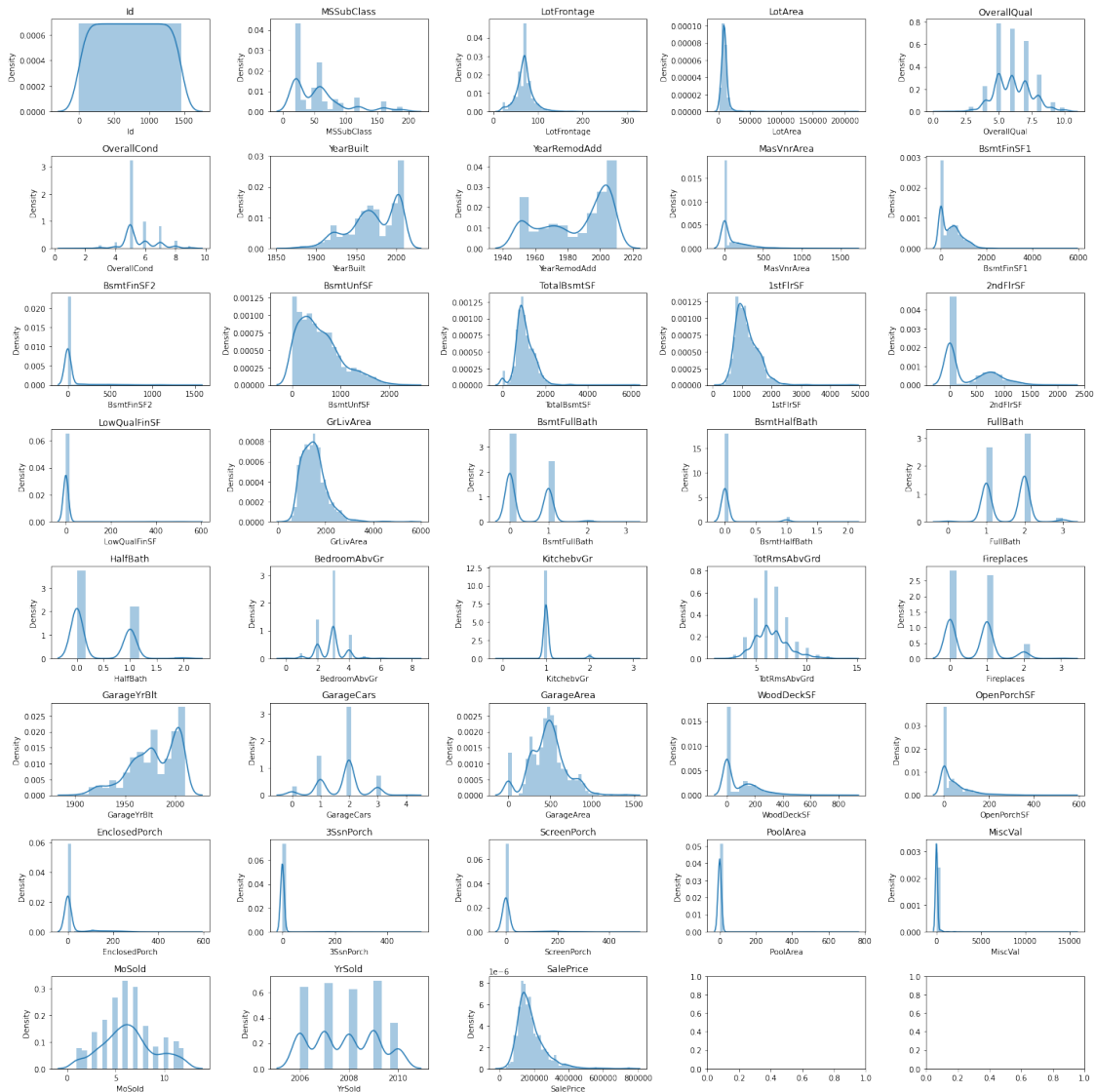
After implementation of Simple Imputer on numeric data : False

```
[12]: # b. Identify the skewness and distribution
sns.distplot(impute_numeric_data['SalePrice'],color='g')
plt.show()
```



```
[13]: fig, axes = plt.subplots(nrows=8, ncols=5)
      axes = axes.flatten()
      fig.set_size_inches(20, 20)

      for ax, col in zip(axes, impute_numeric_data.columns):
          sns.distplot(impute_numeric_data[col], ax=ax)
          ax.set_title(col)
      plt.tight_layout()
```



```
[14]: # c. Identify significant variables using a correlation matrix
data.corr()
```

```
[14]:
```

	Id	MSSubClass	LotFrontage	LotArea	OverallQual	\
Id	1.000000	0.011156	-0.010601	-0.033226	-0.028365	
MSSubClass	0.011156	1.000000	-0.386347	-0.139781	0.032628	
LotFrontage	-0.010601	-0.386347	1.000000	0.426095	0.251646	
LotArea	-0.033226	-0.139781	0.426095	1.000000	0.105806	
OverallQual	-0.028365	0.032628	0.251646	0.105806	1.000000	
OverallCond	0.012609	-0.059316	-0.059213	-0.005636	-0.091932	
YearBuilt	-0.012713	0.027850	0.123349	0.014228	0.572323	
YearRemodAdd	-0.021998	0.040581	0.088866	0.013788	0.550684	
MasVnrArea	-0.050298	0.022936	0.193458	0.104160	0.411876	

BsmtFinSF1	-0.005024	-0.069836	0.233633	0.214103	0.239666
BsmtFinSF2	-0.005968	-0.065649	0.049900	0.111170	-0.059119
BsmtUnfSF	-0.007940	-0.140759	0.132644	-0.002618	0.308159
TotalBsmtSF	-0.015415	-0.238518	0.392075	0.260833	0.537808
1stFlrSF	0.010496	-0.251758	0.457181	0.299475	0.476224
2ndFlrSF	0.005590	0.307886	0.080177	0.050986	0.295493
LowQualFinSF	-0.044230	0.046474	0.038469	0.004779	-0.030429
GrLivArea	0.008273	0.074853	0.402797	0.263116	0.593007
BsmtFullBath	0.002289	0.003491	0.100949	0.158155	0.111098
BsmtHalfBath	-0.020155	-0.002333	-0.007234	0.048046	-0.040150
FullBath	0.005587	0.131608	0.198769	0.126031	0.550600
HalfBath	0.006784	0.177354	0.053532	0.014259	0.273458
BedroomAbvGr	0.037719	-0.023438	0.263170	0.119690	0.101676
KitchenAbvGr	0.002951	0.281721	-0.006069	-0.017784	-0.183882
TotRmsAbvGrd	0.027239	0.040380	0.352096	0.190015	0.427452
Fireplaces	-0.019772	-0.045569	0.266639	0.271364	0.396765
GarageYrBlt	0.000072	0.085072	0.070250	-0.024947	0.547766
GarageCars	0.016570	-0.040110	0.285691	0.154871	0.600671
GarageArea	0.017634	-0.098672	0.344997	0.180403	0.562022
WoodDeckSF	-0.029643	-0.012579	0.088521	0.171698	0.238923
OpenPorchSF	-0.000477	-0.006100	0.151972	0.084774	0.308819
EnclosedPorch	0.002889	-0.012037	0.010700	-0.018340	-0.113937
3SsnPorch	-0.046635	-0.043825	0.070029	0.020423	0.030371
ScreenPorch	0.001330	-0.026030	0.041383	0.043160	0.064886
PoolArea	0.057044	0.008283	0.206167	0.077672	0.065166
MiscVal	-0.006242	-0.007683	0.003368	0.038068	-0.031406
MoSold	0.021172	-0.013585	0.011200	0.001205	0.070815
YrSold	0.000712	-0.021407	0.007450	-0.014261	-0.027347
SalePrice	-0.021917	-0.084284	0.351799	0.263843	0.790982

	OverallCond	YearBuilt	YearRemodAdd	MasVnrArea	BsmtFinSF1 \
Id	0.012609	-0.012713	-0.021998	-0.050298	-0.005024
MSSubClass	-0.059316	0.027850	0.040581	0.022936	-0.069836
LotFrontage	-0.059213	0.123349	0.088866	0.193458	0.233633
LotArea	-0.005636	0.014228	0.013788	0.104160	0.214103
OverallQual	-0.091932	0.572323	0.550684	0.411876	0.239666
OverallCond	1.000000	-0.375983	0.073741	-0.128101	-0.046231
YearBuilt	-0.375983	1.000000	0.592855	0.315707	0.249503
YearRemodAdd	0.073741	0.592855	1.000000	0.179618	0.128451
MasVnrArea	-0.128101	0.315707	0.179618	1.000000	0.264736
BsmtFinSF1	-0.046231	0.249503	0.128451	0.264736	1.000000
BsmtFinSF2	0.040229	-0.049107	-0.067759	-0.072319	-0.050117
BsmtUnfSF	-0.136841	0.149040	0.181133	0.114442	-0.495251
TotalBsmtSF	-0.171098	0.391452	0.291066	0.363936	0.522396
1stFlrSF	-0.144203	0.281986	0.240379	0.344501	0.445863
2ndFlrSF	0.028942	0.010308	0.140024	0.174561	-0.137079
LowQualFinSF	0.025494	-0.183784	-0.062419	-0.069071	-0.064503

GrLivArea	-0.079686	0.199010	0.287389	0.390857	0.208171
BsmtFullBath	-0.054942	0.187599	0.119470	0.085310	0.649212
BsmtHalfBath	0.117821	-0.038162	-0.012337	0.026673	0.067418
FullBath	-0.194149	0.468271	0.439046	0.276833	0.058543
HalfBath	-0.060769	0.242656	0.183331	0.201444	0.004262
BedroomAbvGr	0.012980	-0.070651	-0.040581	0.102821	-0.107355
KitchenAbvGr	-0.087001	-0.174800	-0.149598	-0.037610	-0.081007
TotRmsAbvGrd	-0.057583	0.095589	0.191740	0.280682	0.044316
Fireplaces	-0.023820	0.147716	0.112581	0.249070	0.260011
GarageYrBlt	-0.324297	0.825667	0.642277	0.252691	0.153484
GarageCars	-0.185758	0.537850	0.420622	0.364204	0.224054
GarageArea	-0.151521	0.478954	0.371600	0.373066	0.296970
WoodDeckSF	-0.003334	0.224880	0.205726	0.159718	0.204306
OpenPorchSF	-0.032589	0.188686	0.226298	0.125703	0.111761
EnclosedPorch	0.070356	-0.387268	-0.193919	-0.110204	-0.102303
3SsnPorch	0.025504	0.031355	0.045286	0.018796	0.026451
ScreenPorch	0.054811	-0.050364	-0.038740	0.061466	0.062021
PoolArea	-0.001985	0.004950	0.005829	0.011723	0.140491
MiscVal	0.068777	-0.034383	-0.010286	-0.029815	0.003571
MoSold	-0.003511	0.012398	0.021490	-0.005965	-0.015727
YrSold	0.043950	-0.013618	0.035743	-0.008201	0.014359
SalePrice	-0.077856	0.522897	0.507101	0.477493	0.386420

	...	WoodDeckSF	OpenPorchSF	EnclosedPorch	3SsnPorch	\
Id	...	-0.029643	-0.000477	0.002889	-0.046635	
MSSubClass	...	-0.012579	-0.006100	-0.012037	-0.043825	
LotFrontage	...	0.088521	0.151972	0.010700	0.070029	
LotArea	...	0.171698	0.084774	-0.018340	0.020423	
OverallQual	...	0.238923	0.308819	-0.113937	0.030371	
OverallCond	...	-0.003334	-0.032589	0.070356	0.025504	
YearBuilt	...	0.224880	0.188686	-0.387268	0.031355	
YearRemodAdd	...	0.205726	0.226298	-0.193919	0.045286	
MasVnrArea	...	0.159718	0.125703	-0.110204	0.018796	
BsmtFinSF1	...	0.204306	0.111761	-0.102303	0.026451	
BsmtFinSF2	...	0.067898	0.003093	0.036543	-0.029993	
BsmtUnfSF	...	-0.005316	0.129005	-0.002538	0.020764	
TotalBsmtSF	...	0.232019	0.247264	-0.095478	0.037384	
1stFlrSF	...	0.235459	0.211671	-0.065292	0.056104	
2ndFlrSF	...	0.092165	0.208026	0.061989	-0.024358	
LowQualFinSF	...	-0.025444	0.018251	0.061081	-0.004296	
GrLivArea	...	0.247433	0.330224	0.009113	0.020643	
BsmtFullBath	...	0.175315	0.067341	-0.049911	-0.000106	
BsmtHalfBath	...	0.040161	-0.025324	-0.008555	0.035114	
FullBath	...	0.187703	0.259977	-0.115093	0.035353	
HalfBath	...	0.108080	0.199740	-0.095317	-0.004972	
BedroomAbvGr	...	0.046854	0.093810	0.041570	-0.024478	
KitchenAbvGr	...	-0.090130	-0.070091	0.037312	-0.024600	

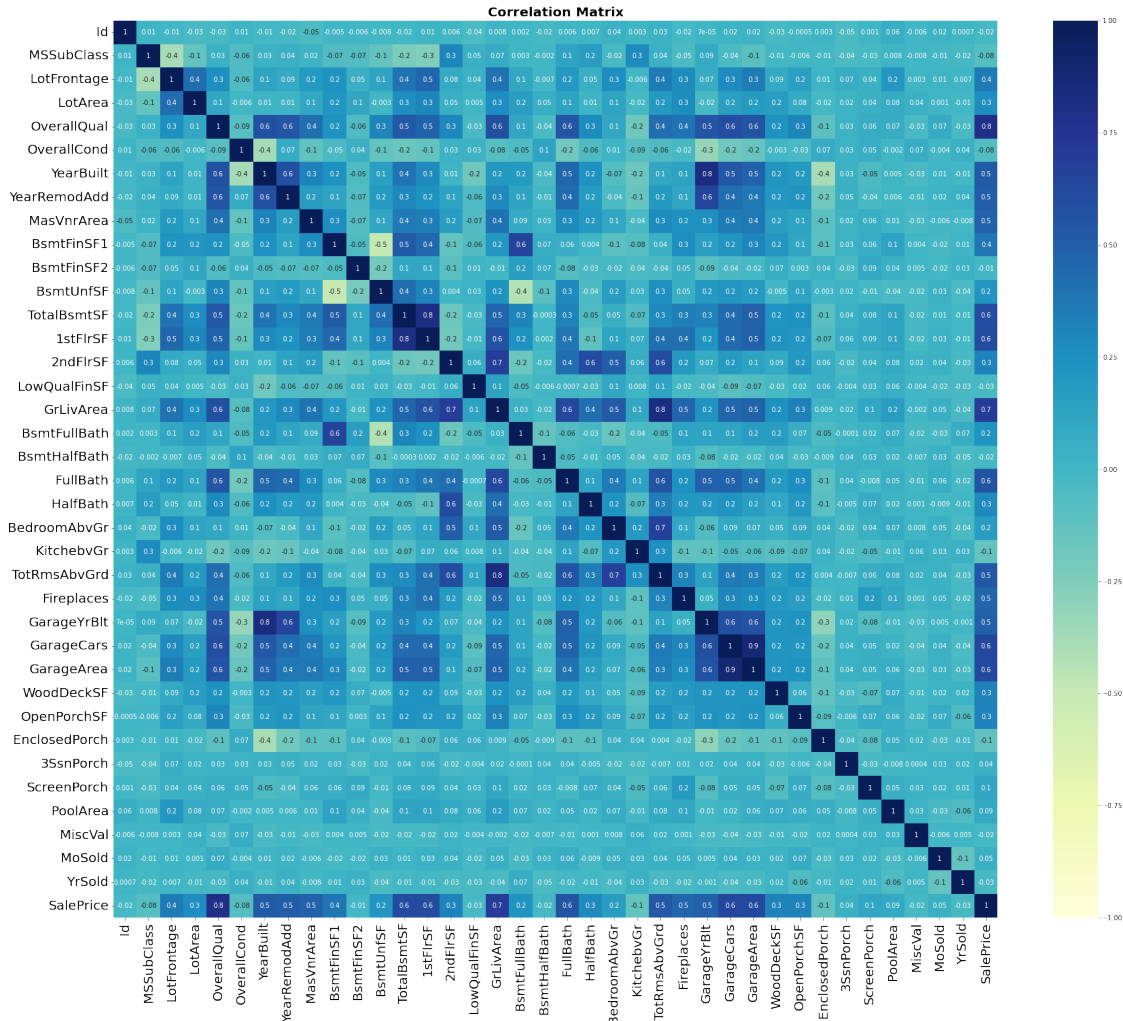
TotRmsAbvGrd	...	0.165984	0.234192	0.004151	-0.006683
Fireplaces	...	0.200019	0.169405	-0.024822	0.011257
GarageYrBlt	...	0.224577	0.228425	-0.297003	0.023544
GarageCars	...	0.226342	0.213569	-0.151434	0.035765
GarageArea	...	0.224666	0.241435	-0.121777	0.035087
WoodDeckSF	...	1.000000	0.058661	-0.125989	-0.032771
OpenPorchSF	...	0.058661	1.000000	-0.093079	-0.005842
EnclosedPorch	...	-0.125989	-0.093079	1.000000	-0.037305
3SsnPorch	...	-0.032771	-0.005842	-0.037305	1.000000
ScreenPorch	...	-0.074181	0.074304	-0.082864	-0.031436
PoolArea	...	0.073378	0.060762	0.054203	-0.007992
MiscVal	...	-0.009551	-0.018584	0.018361	0.000354
MoSold	...	0.021011	0.071255	-0.028887	0.029474
YrSold	...	0.022270	-0.057619	-0.009916	0.018645
SalePrice	...	0.324413	0.315856	-0.128578	0.044584

	ScreenPorch	PoolArea	MiscVal	MoSold	YrSold	SalePrice
Id	0.001330	0.057044	-0.006242	0.021172	0.000712	-0.021917
MSSubClass	-0.026030	0.008283	-0.007683	-0.013585	-0.021407	-0.084284
LotFrontage	0.041383	0.206167	0.003368	0.011200	0.007450	0.351799
LotArea	0.043160	0.077672	0.038068	0.001205	-0.014261	0.263843
OverallQual	0.064886	0.065166	-0.031406	0.070815	-0.027347	0.790982
OverallCond	0.054811	-0.001985	0.068777	-0.003511	0.043950	-0.077856
YearBuilt	-0.050364	0.004950	-0.034383	0.012398	-0.013618	0.522897
YearRemodAdd	-0.038740	0.005829	-0.010286	0.021490	0.035743	0.507101
MasVnrArea	0.061466	0.011723	-0.029815	-0.005965	-0.008201	0.477493
BsmtFinSF1	0.062021	0.140491	0.003571	-0.015727	0.014359	0.386420
BsmtFinSF2	0.088871	0.041709	0.004940	-0.015211	0.031706	-0.011378
BsmtUnfSF	-0.012579	-0.035092	-0.023837	0.034888	-0.041258	0.214479
TotalBsmtSF	0.084489	0.126053	-0.018479	0.013196	-0.014969	0.613581
1stFlrSF	0.088758	0.131525	-0.021096	0.031372	-0.013604	0.605852
2ndFlrSF	0.040606	0.081487	0.016197	0.035164	-0.028700	0.319334
LowQualFinSF	0.026799	0.062157	-0.003793	-0.022174	-0.028921	-0.025606
GrLivArea	0.101510	0.170205	-0.002416	0.050240	-0.036526	0.708624
BsmtFullBath	0.023148	0.067616	-0.023047	-0.025361	0.067049	0.227122
BsmtHalfBath	0.032121	0.020025	-0.007367	0.032873	-0.046524	-0.016844
FullBath	-0.008106	0.049604	-0.014290	0.055872	-0.019669	0.560664
HalfBath	0.072426	0.022381	0.001290	-0.009050	-0.010269	0.284108
BedroomAbvGr	0.044300	0.070703	0.007767	0.046544	-0.036014	0.168213
KitchenAbvGr	-0.051613	-0.014525	0.062341	0.026589	0.031687	-0.135907
TotRmsAbvGrd	0.059383	0.083757	0.024763	0.036907	-0.034516	0.533723
Fireplaces	0.184530	0.095074	0.001409	0.046357	-0.024096	0.466929
GarageYrBlt	-0.075418	-0.014501	-0.032417	0.005337	-0.001014	0.486362
GarageCars	0.050494	0.020934	-0.043080	0.040522	-0.039117	0.640409
GarageArea	0.051412	0.061047	-0.027400	0.027974	-0.027378	0.623431
WoodDeckSF	-0.074181	0.073378	-0.009551	0.021011	0.022270	0.324413
OpenPorchSF	0.074304	0.060762	-0.018584	0.071255	-0.057619	0.315856

EnclosedPorch	-0.082864	0.054203	0.018361	-0.028887	-0.009916	-0.128578
3SsnPorch	-0.031436	-0.007992	0.000354	0.029474	0.018645	0.044584
ScreenPorch	1.000000	0.051307	0.031946	0.023217	0.010694	0.111447
PoolArea	0.051307	1.000000	0.029669	-0.033737	-0.059689	0.092404
MiscVal	0.031946	0.029669	1.000000	-0.006495	0.004906	-0.021190
MoSold	0.023217	-0.033737	-0.006495	1.000000	-0.145721	0.046432
YrSold	0.010694	-0.059689	0.004906	-0.145721	1.000000	-0.028923
SalePrice	0.111447	0.092404	-0.021190	0.046432	-0.028923	1.000000

[38 rows x 38 columns]

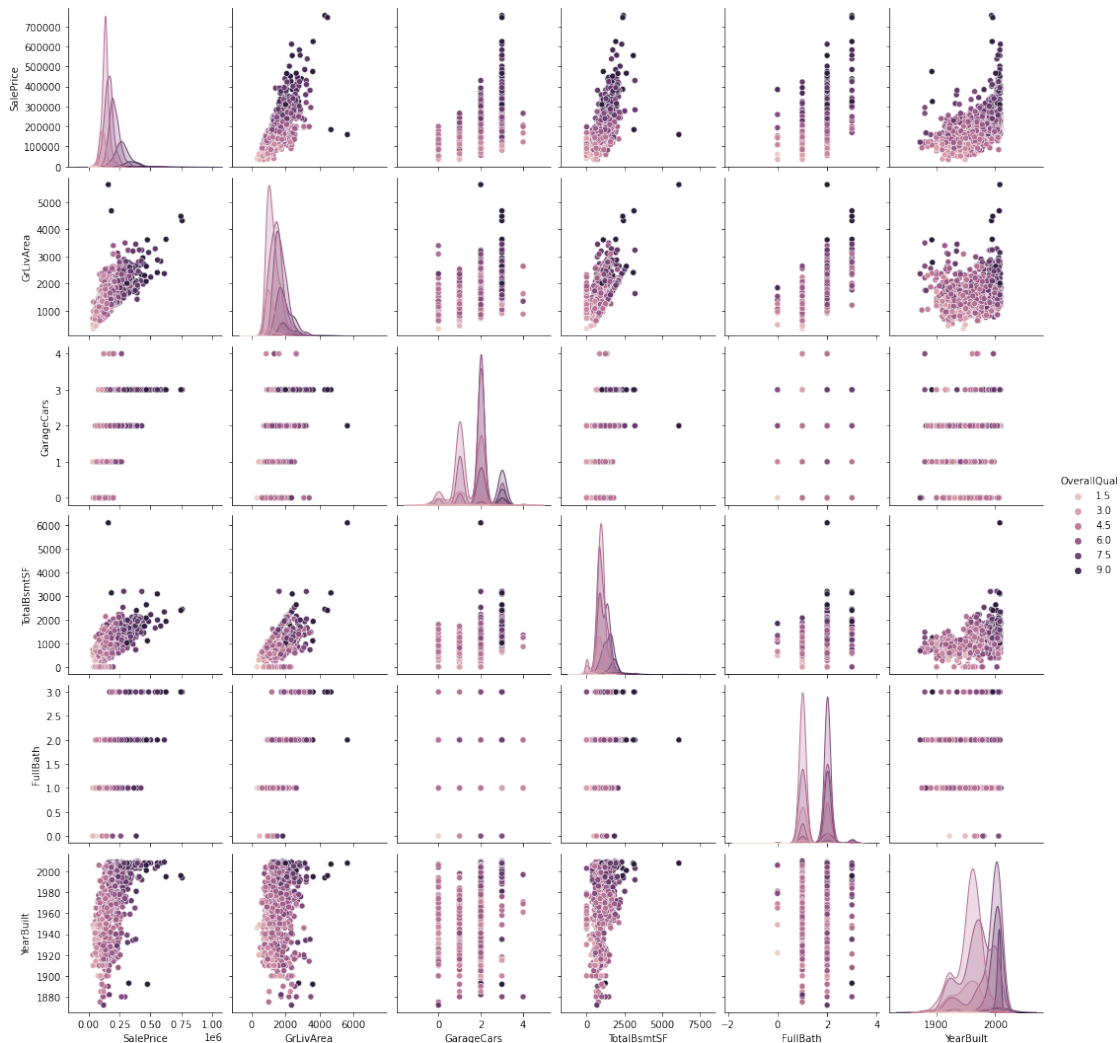
```
[15]: plt.figure(figsize=(30,25))
sns.heatmap(data.corr(), cmap='YlGnBu', annot=True, fmt='.1g', vmax=1, vmin=-1)
plt.tick_params(labelsize=20)
plt.title('Correlation Matrix', fontsize=20, weight='bold')
plt.show()
```



```
[16]: # d. Pair plot for distribution and density
```

```
cols = ['SalePrice', 'OverallQual', 'GrLivArea', 'GarageCars', 'TotalBsmtSF',  
        ↪ 'FullBath', 'YearBuilt']  
sns.pairplot(impute_numeric_data[cols], hue='OverallQual')
```

```
[16]: <seaborn.axisgrid.PairGrid at 0x2e5abff69a0>
```



4 4. EDA of categorical variables

- Missing value treatment
- Count plot and box plot for bivariate analysis

c. Identify significant variables using p-values and Chi-Square values

```
[17]: # a. Missing value treatment.

class DataFrameImputer(TransformerMixin):

    def __init__(self):
        """Impute missing values.

        Columns of dtype object are imputed with the most frequent value
        in column.

        Columns of other types are imputed with mean of column.

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

        self.fill = pd.Series([X[c].value_counts().index[0]
                               if X[c].dtype == np.dtype('O') else X[c].mode() for c in X],
                               index=X.columns)

        return self

    def transform(self, X, y=None):
        return X.fillna(self.fill)

impute_categorical_data = pd.DataFrame(data_categorical)
impute_categorical_data = DataFrameImputer().fit_transform(impute_categorical_data)
```

```
[18]: print('Before implementation of Simple Imputer on categorical data :
      ↪', data_categorical.isna().sum().any())
print('After implementation of Simple Imputer on categorical data :
      ↪', impute_categorical_data.isna().sum().any())
```

Before implementation of Simple Imputer on categorical data : True

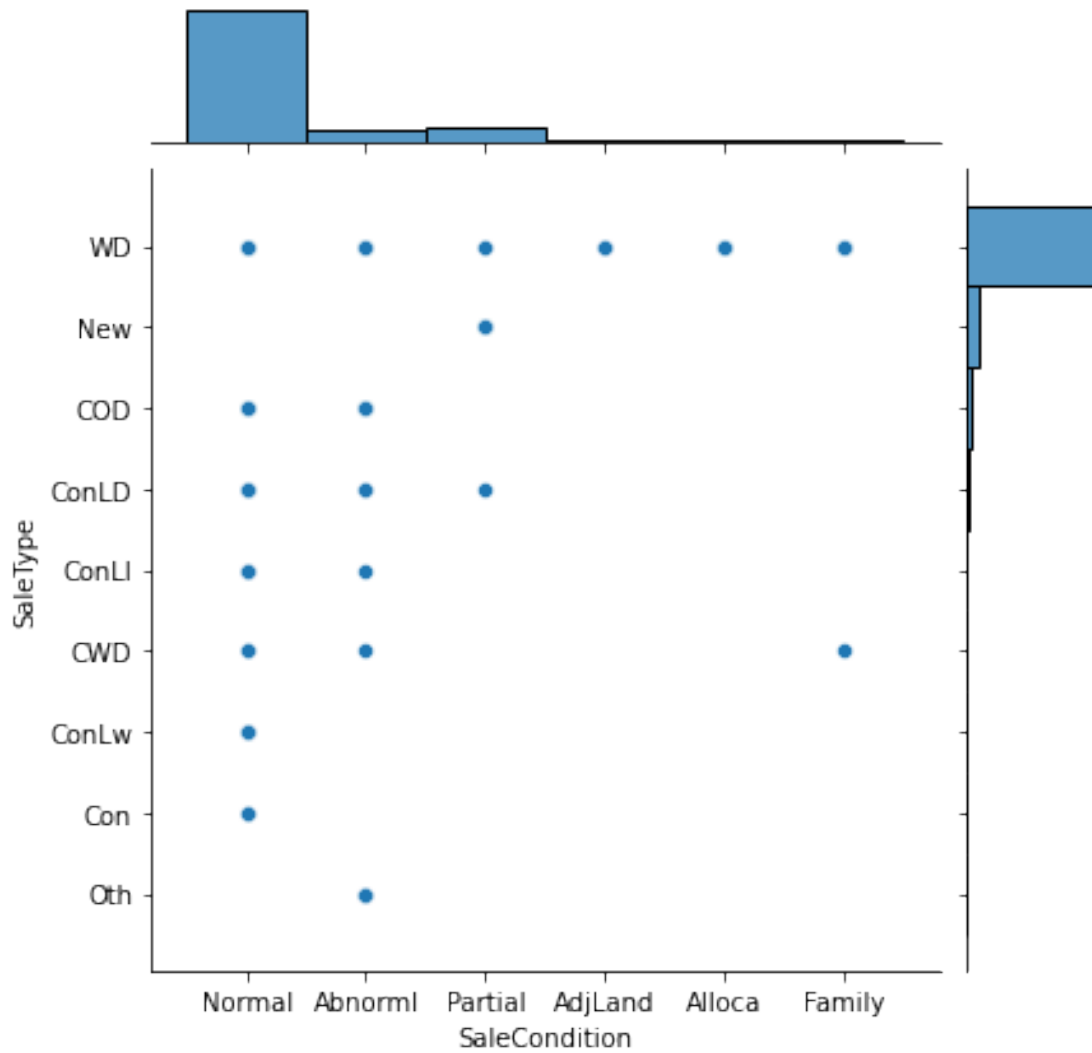
After implementation of Simple Imputer on categorical data : False

```
[19]: # b. Count plot and box plot for bivariate analysis

plt.figure(figsize=(10,5))
sns.
    ↪ jointplot('SaleCondition', 'SaleType', data=impute_categorical_data, kind='scatter')
```

```
[19]: <seaborn.axisgrid.JointGrid at 0x2e5ab6b6340>
```

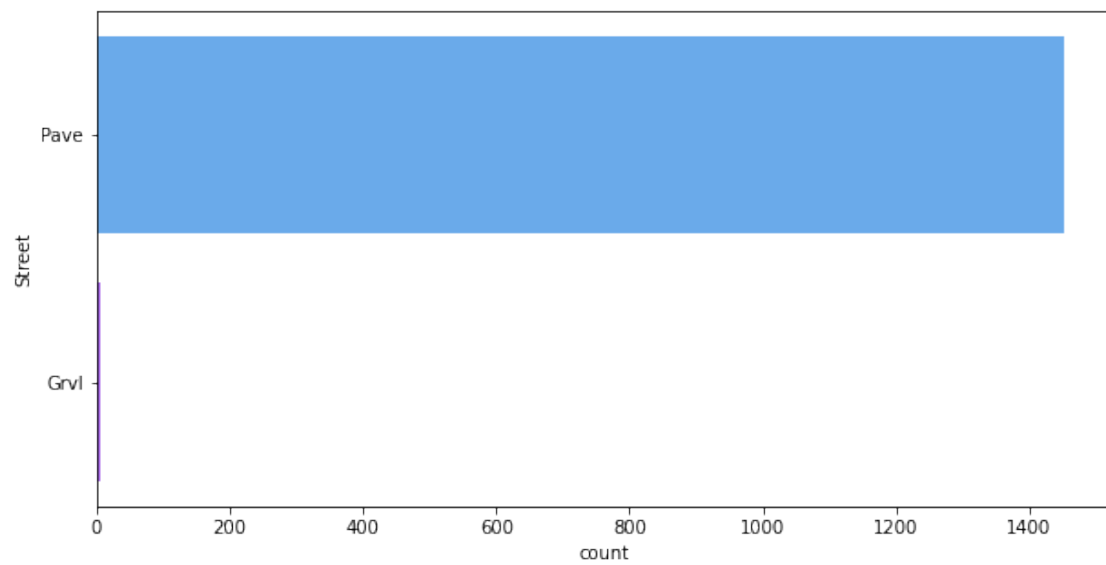
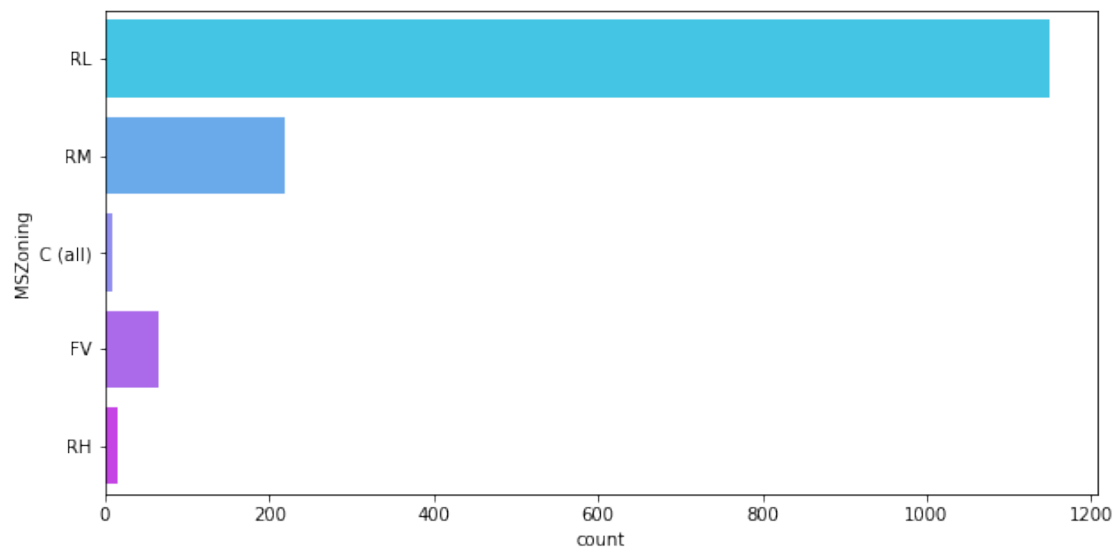
<Figure size 720x360 with 0 Axes>

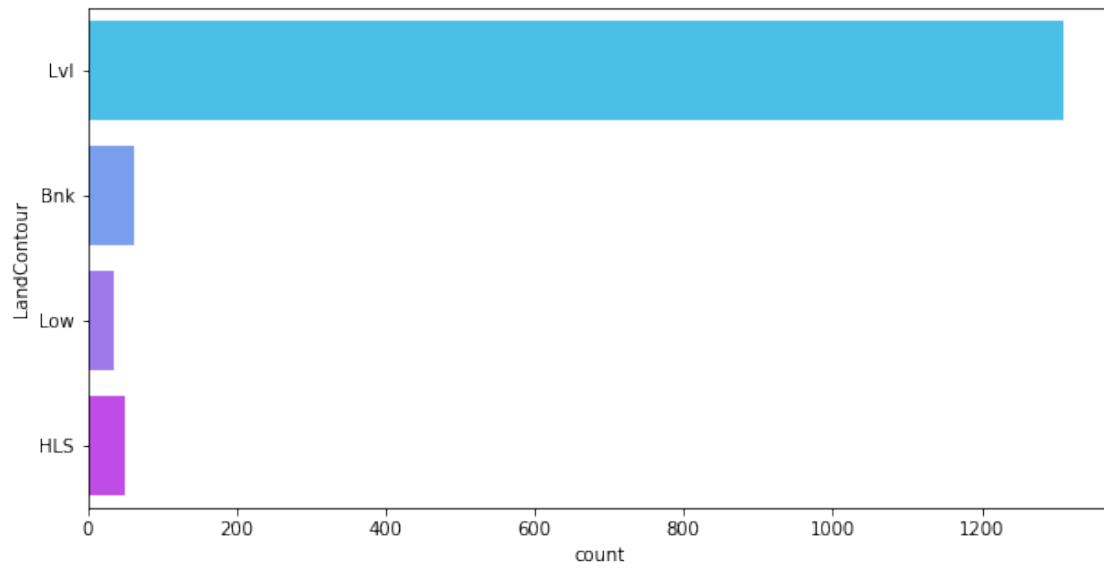
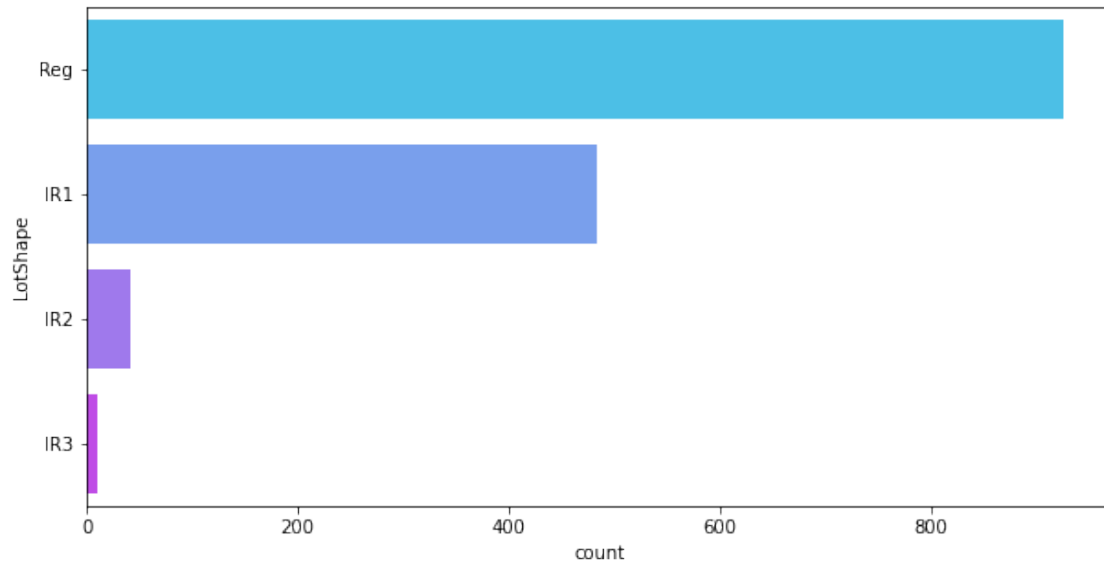


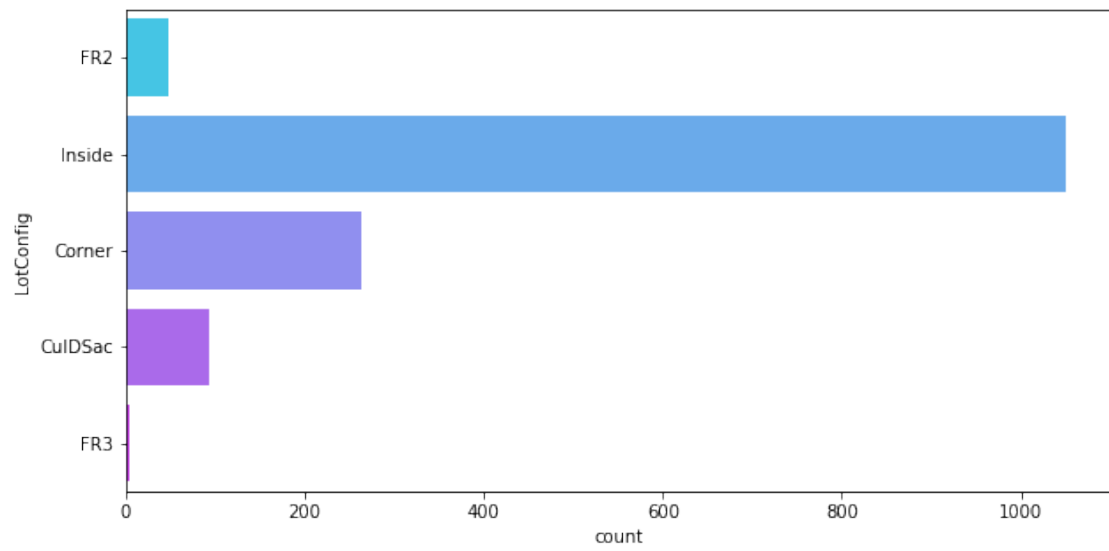
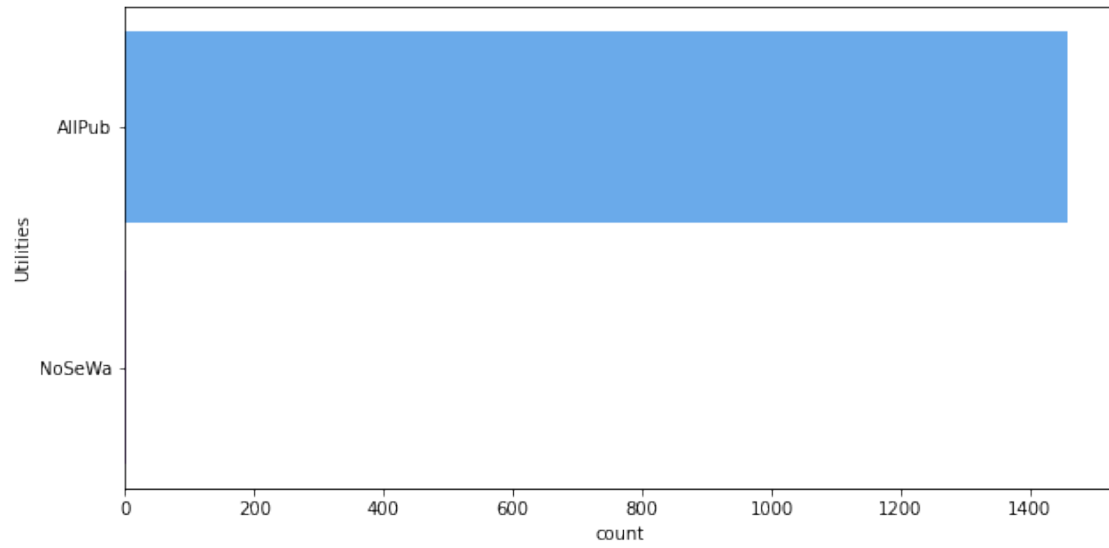
```
[20]: cols = [ 'MSZoning', 'Street', 'LotShape', 'LandContour', '
↳ 'Utilities', 'LotConfig', 'LandSlope', 'Neighborhood',
      'Condition1', 'Condition2', 'BldgType', 'HouseStyle', 'RoofStyle', '
↳ 'RoofMatl', 'Exterior1st', 'Exterior2nd',
      'MasVnrType', 'ExterQual', 'ExterCond', 'Foundation', 'BsmtQual', '
↳ 'BsmtCond', 'BsmtExposure', 'BsmtFinType1',
      'BsmtFinType2', 'Heating', 'HeatingQC', 'CentralAir', 'Electrical', '
↳ 'KitchenQual', 'Function1', 'FireplaceQu',
      'GarageType', 'GarageFinish', 'GarageQual', 'GarageCond', 'PavedDrive', '
↳ 'SaleType', 'SaleCondition']

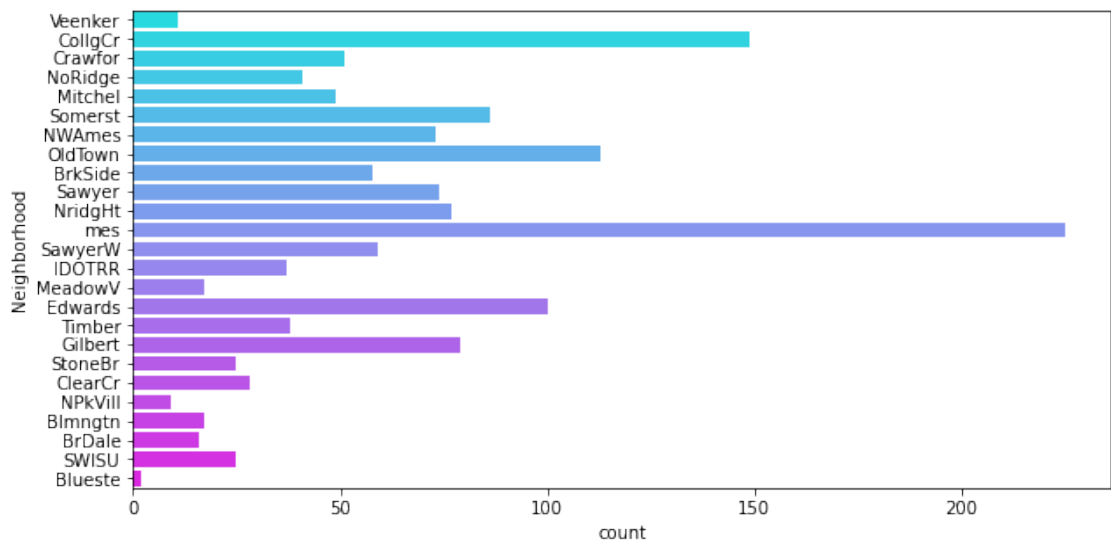
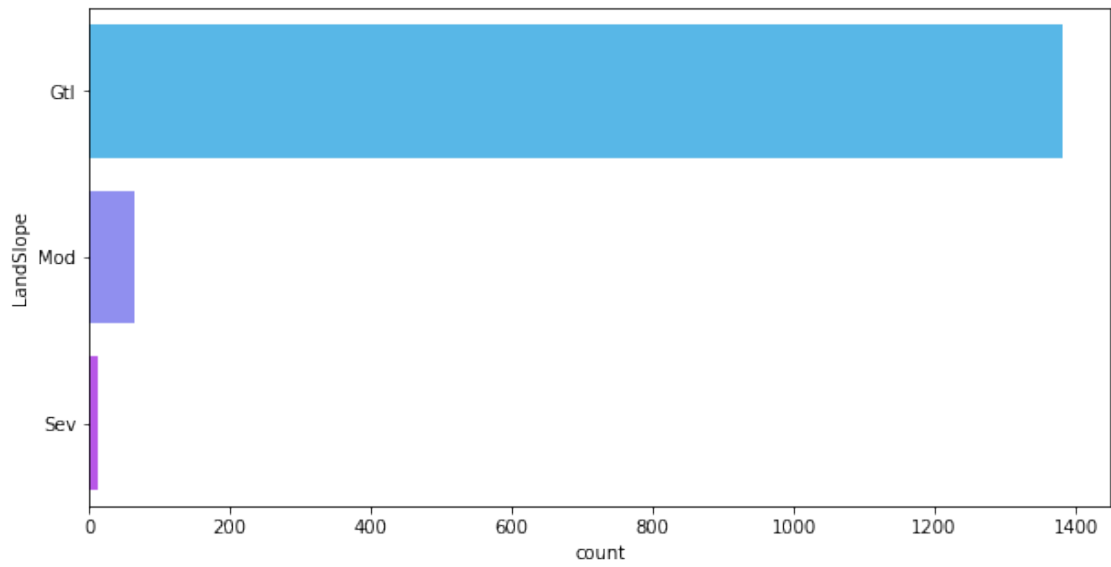
for i in cols:
    fig, ax = plt.subplots(1,1, figsize=(10,5))
```

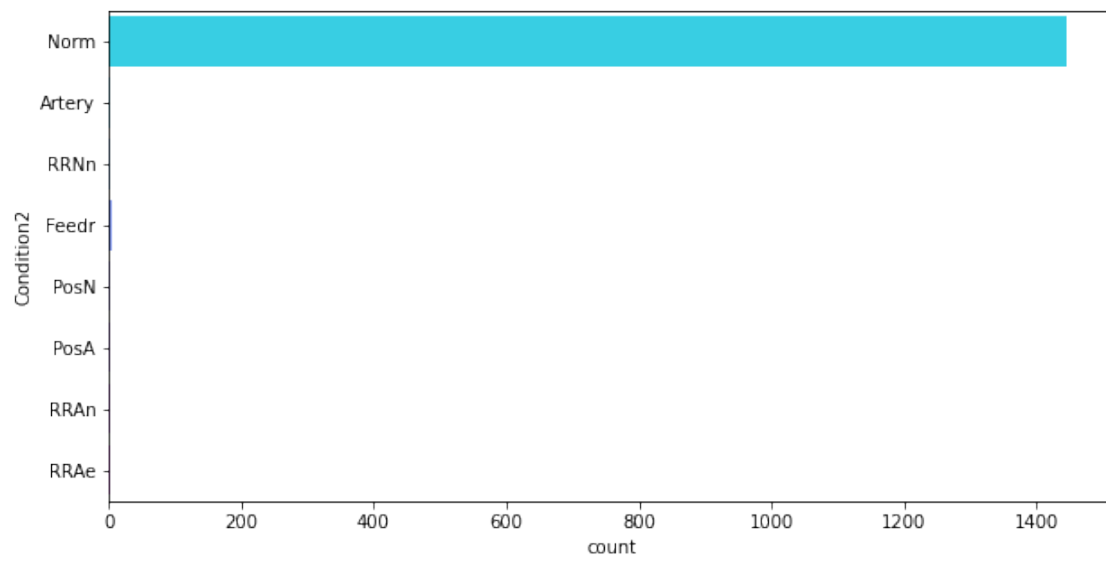
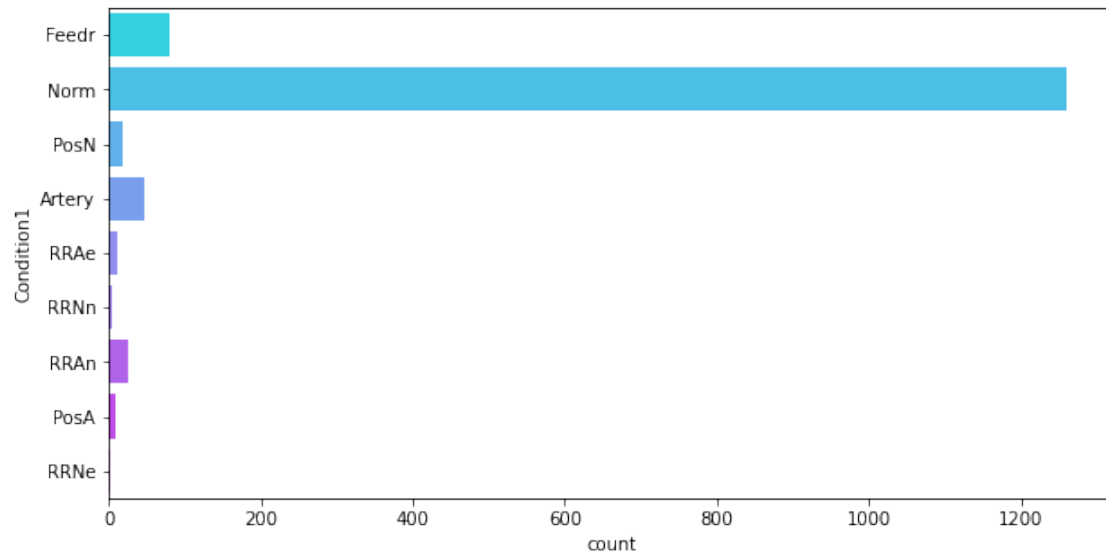
```
sns.countplot(data=impute_categoric_data, y=impute_categoric_data[i][1:],  
palette='cool')
```

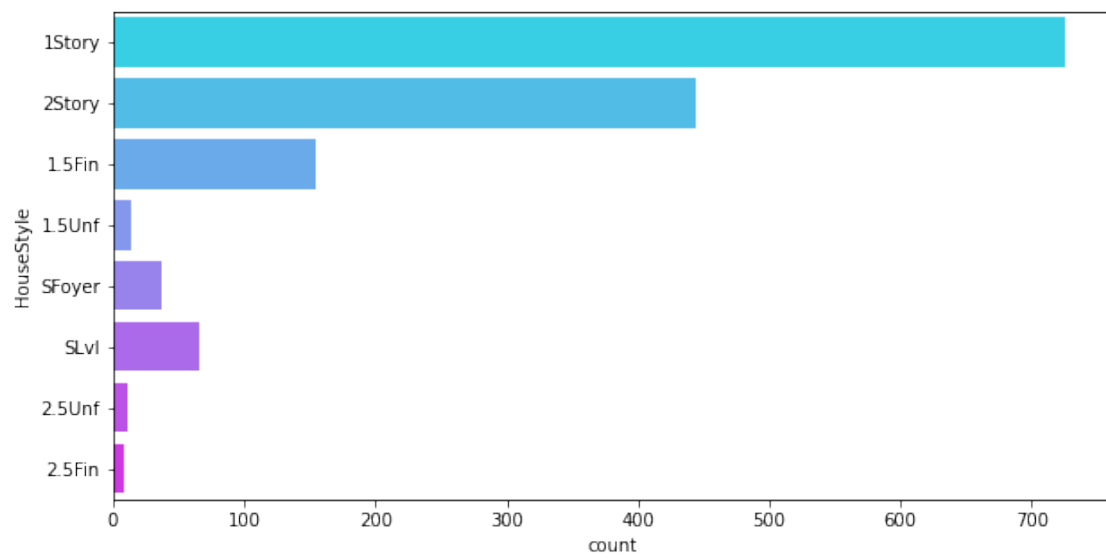
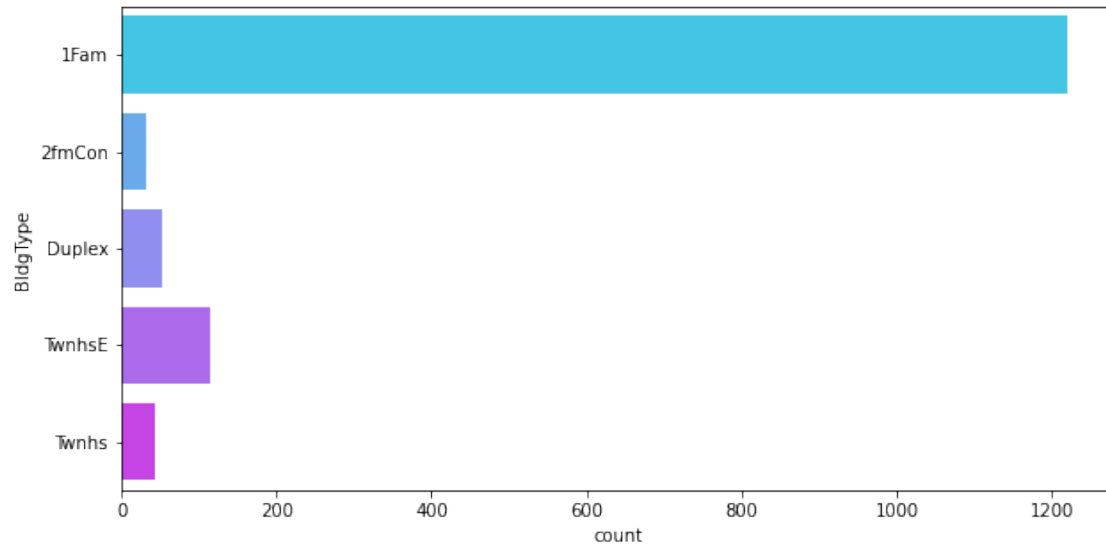


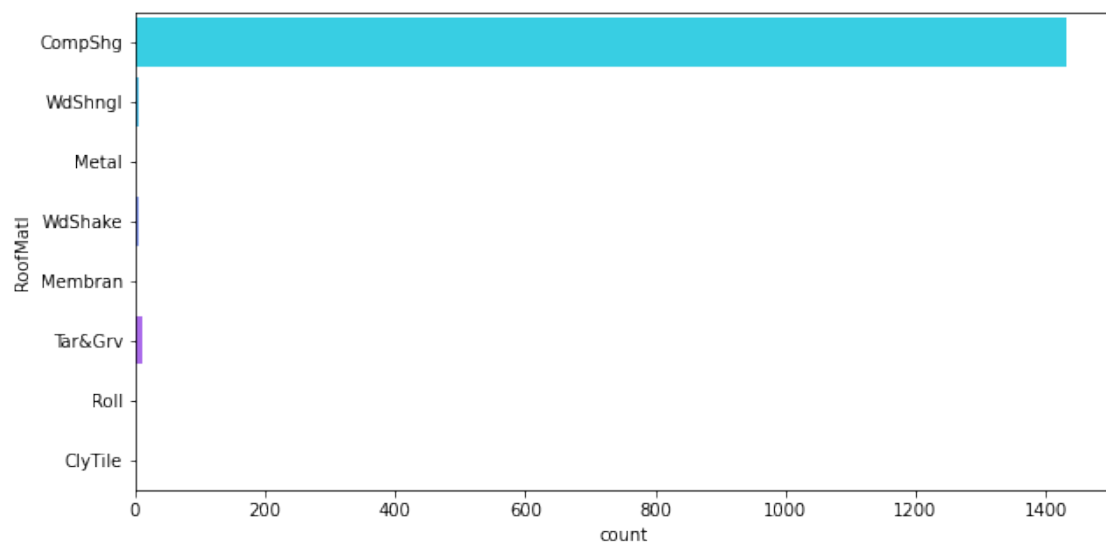
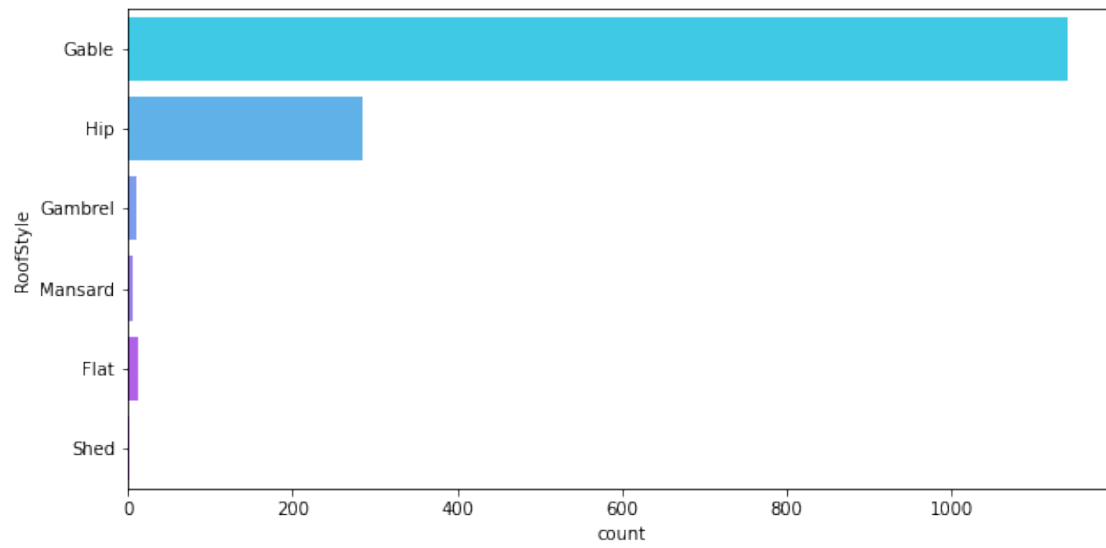


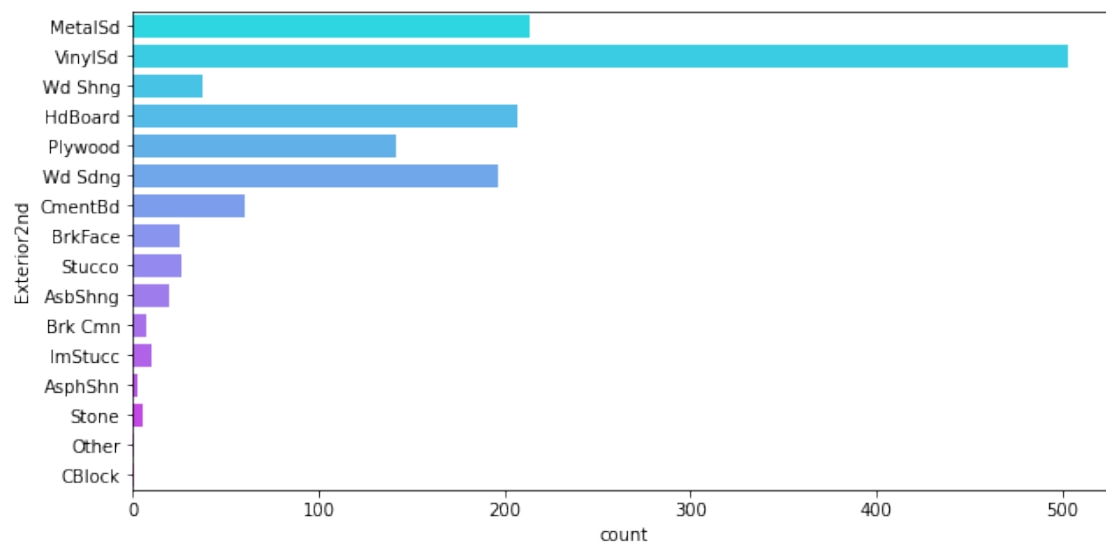
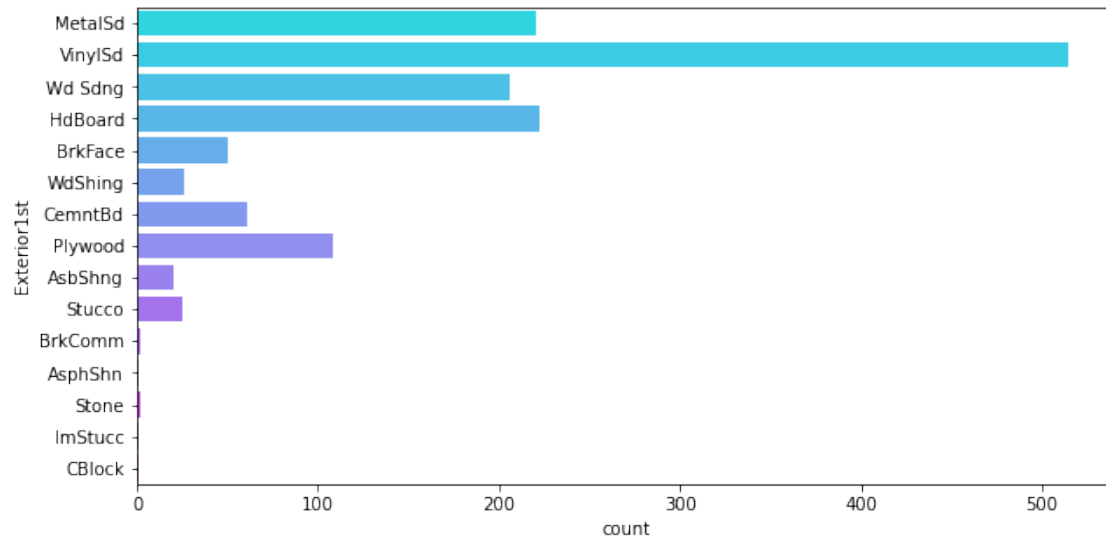


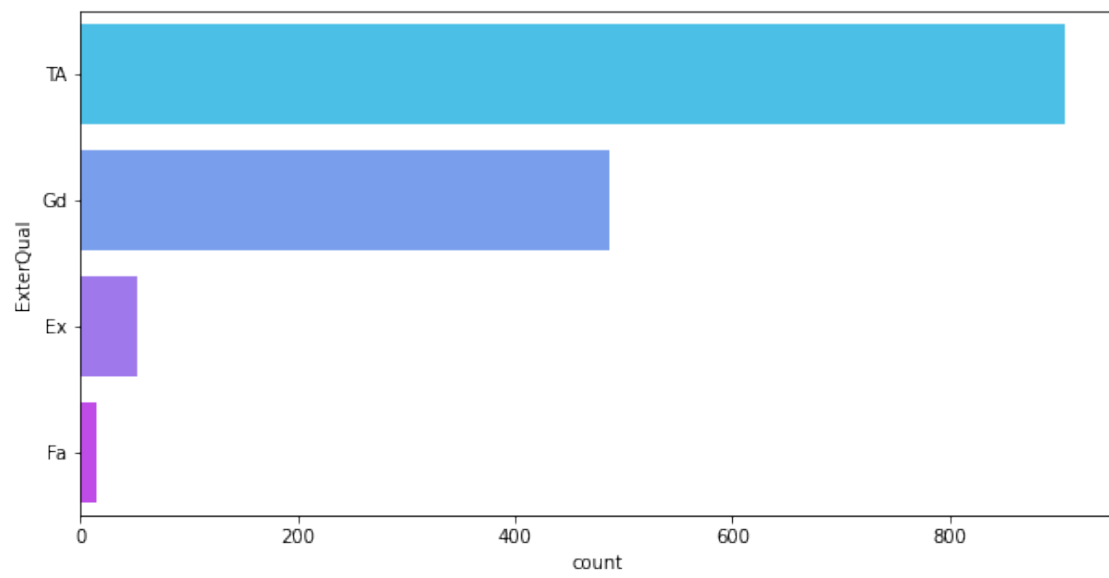
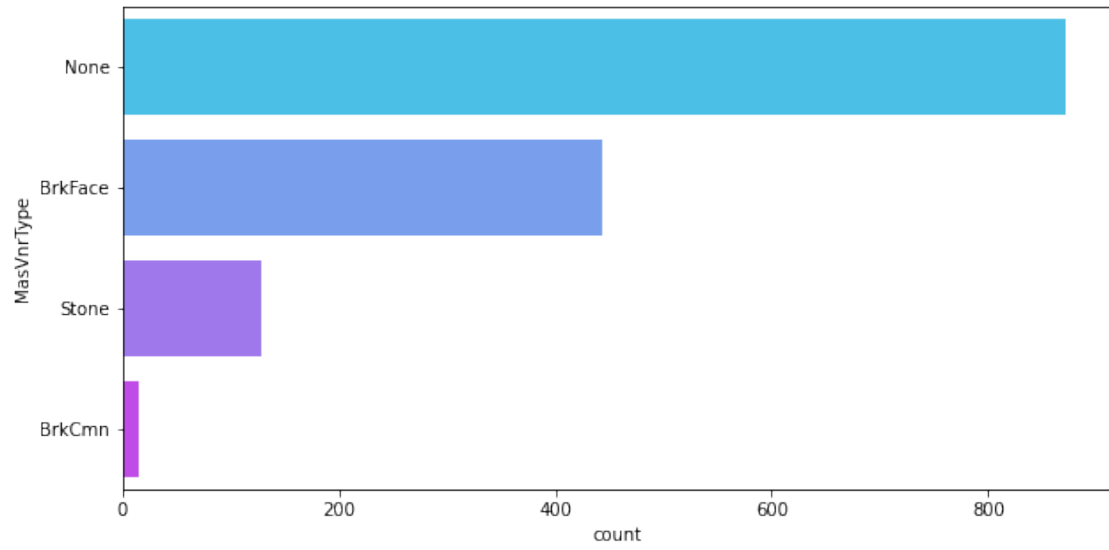


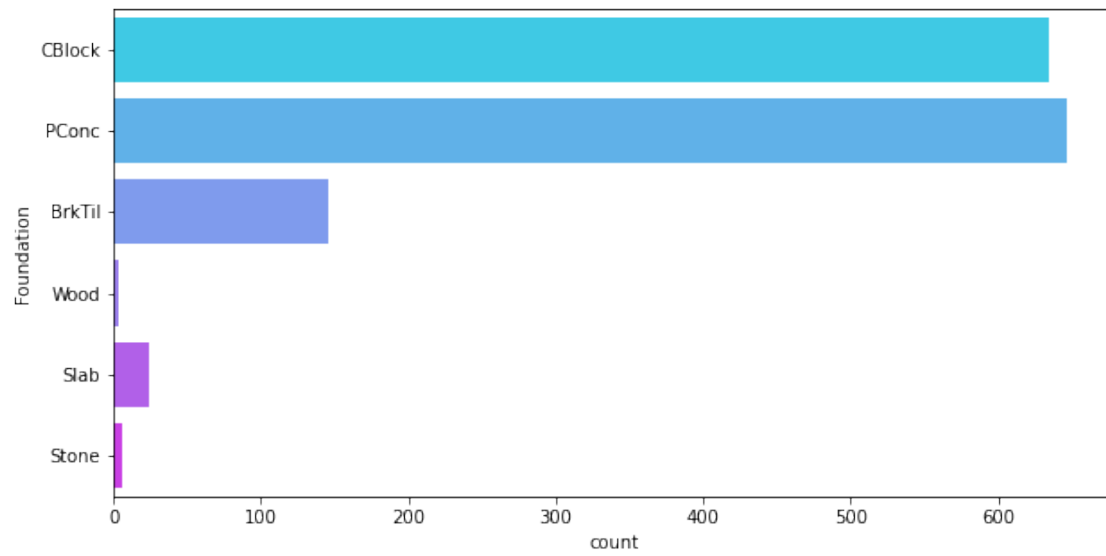
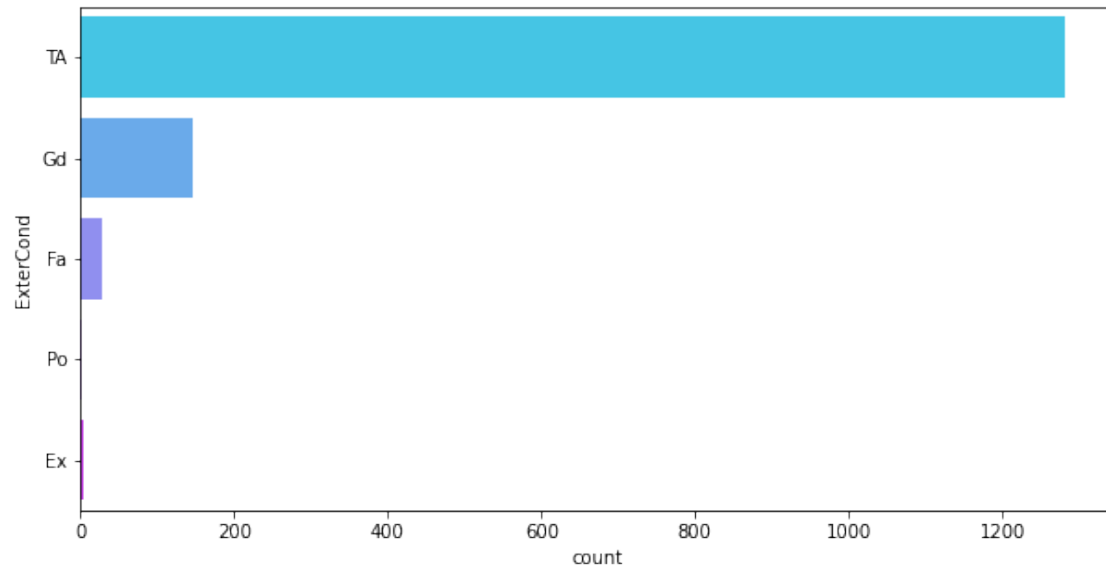


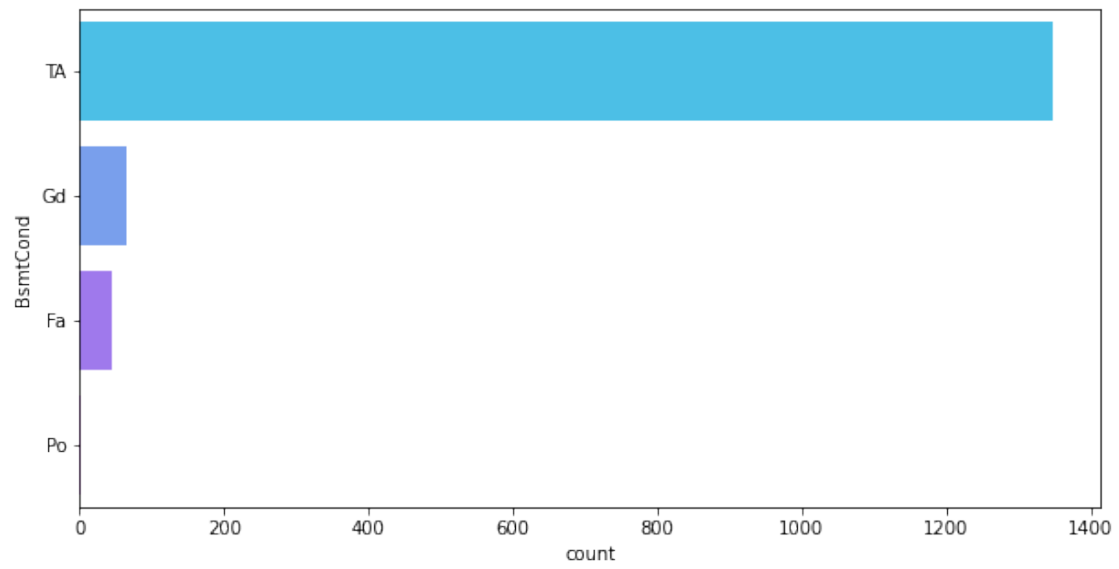
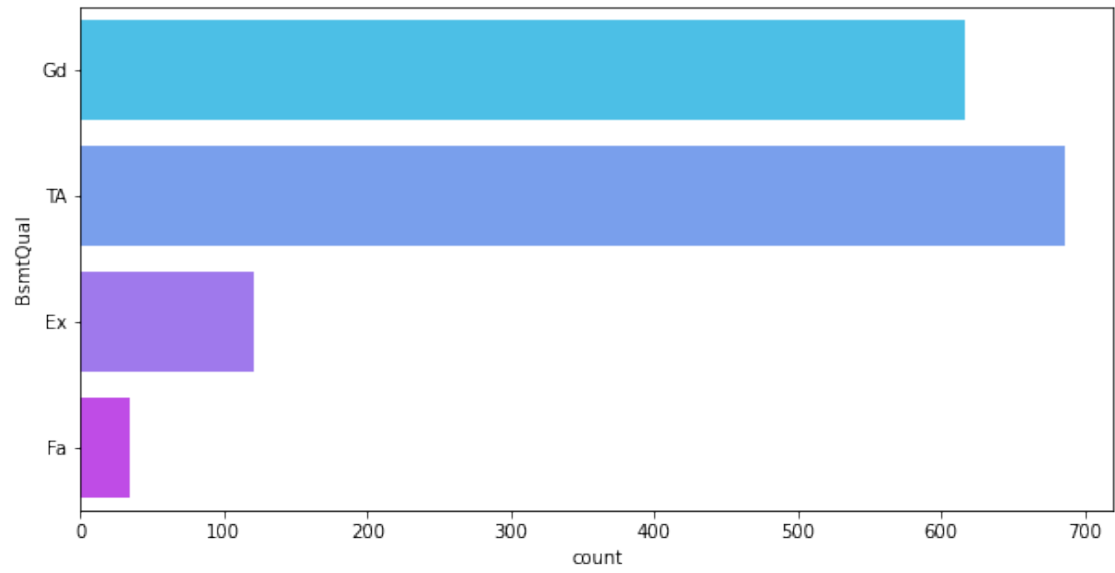


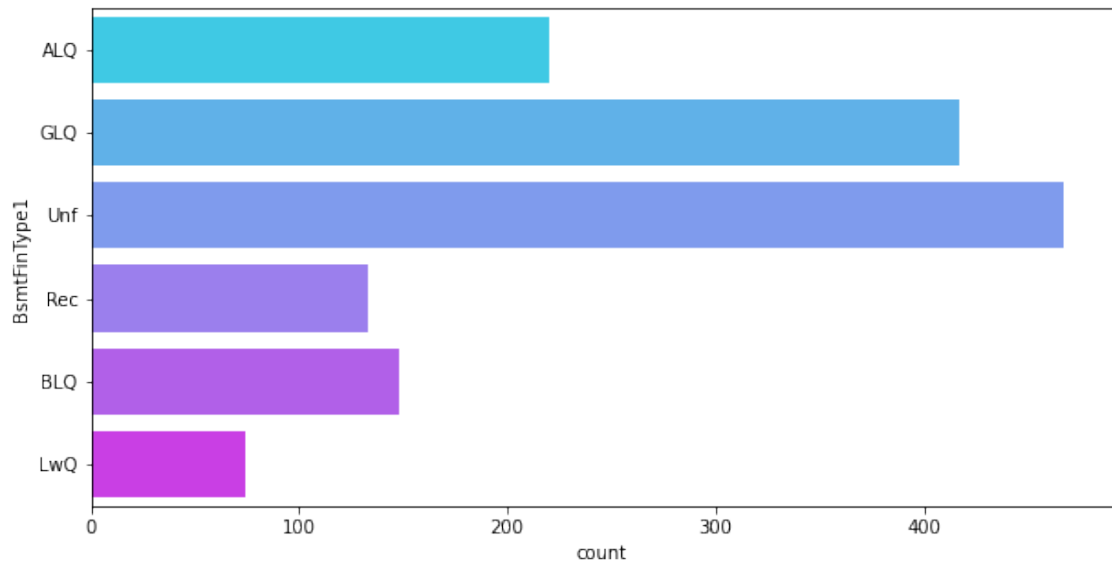
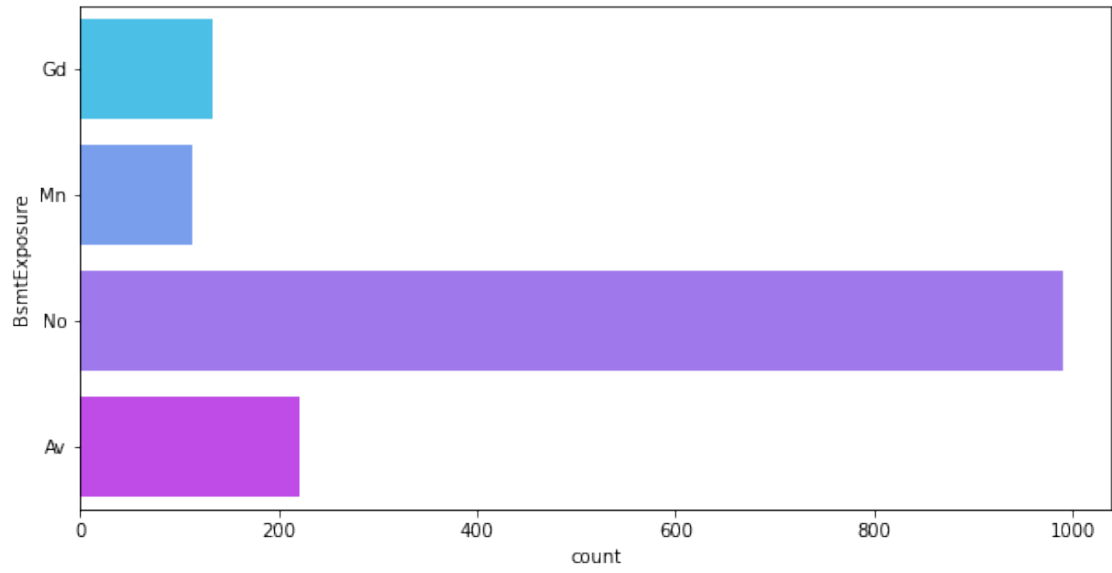


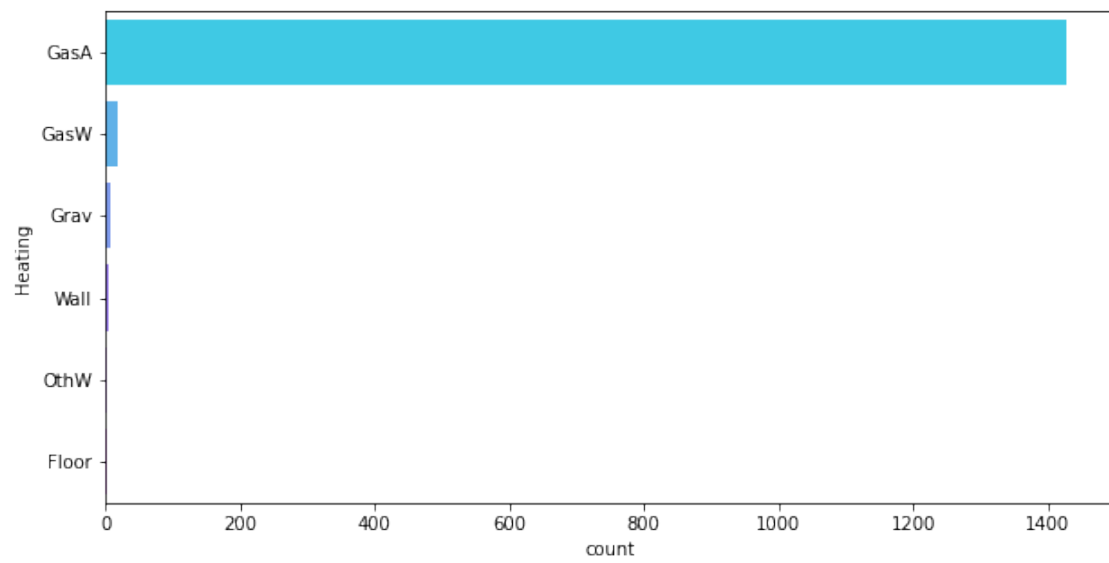
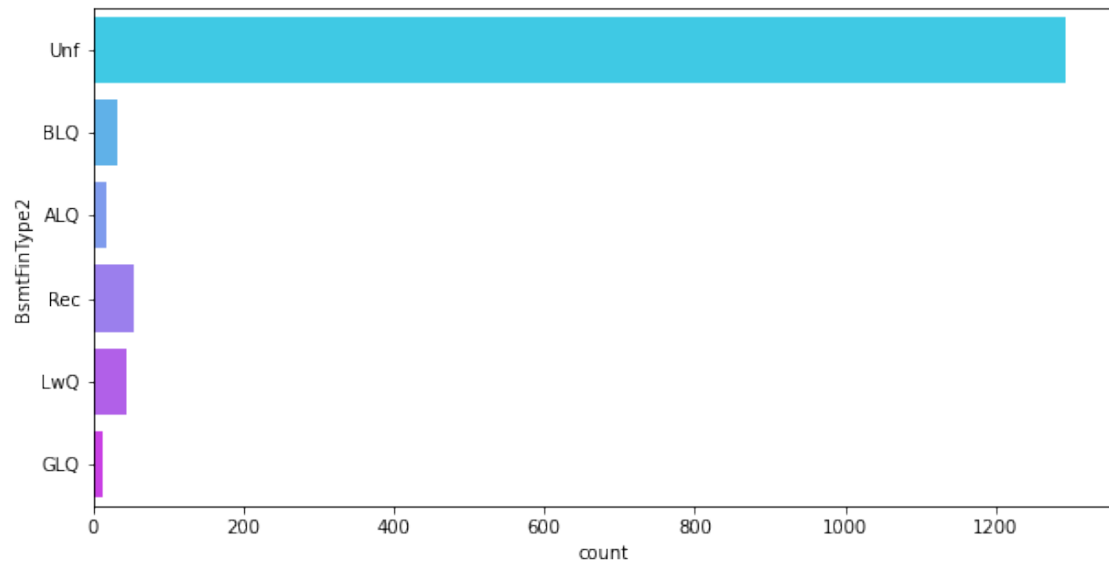


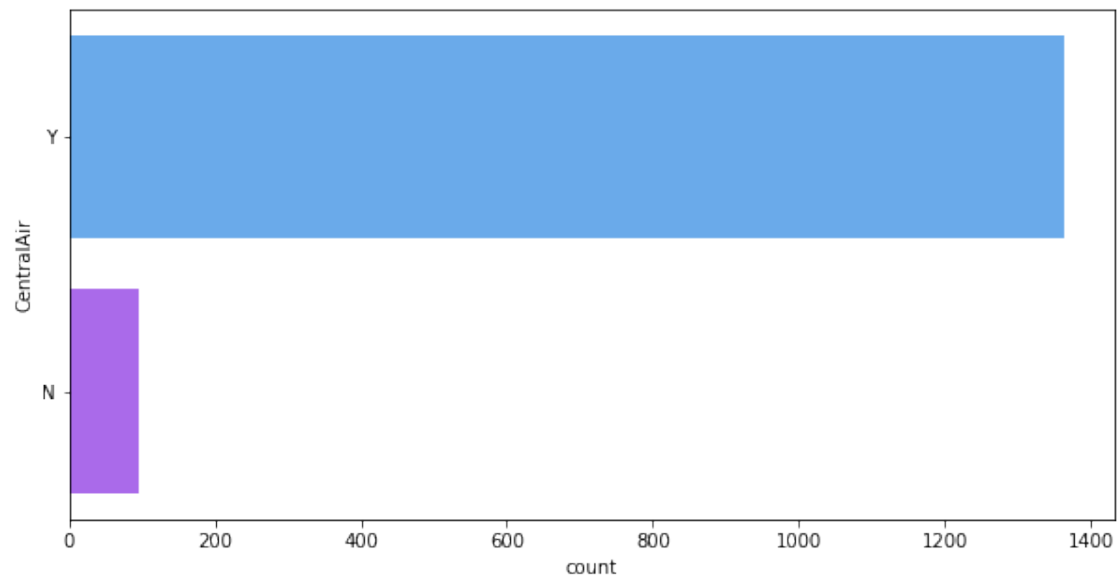
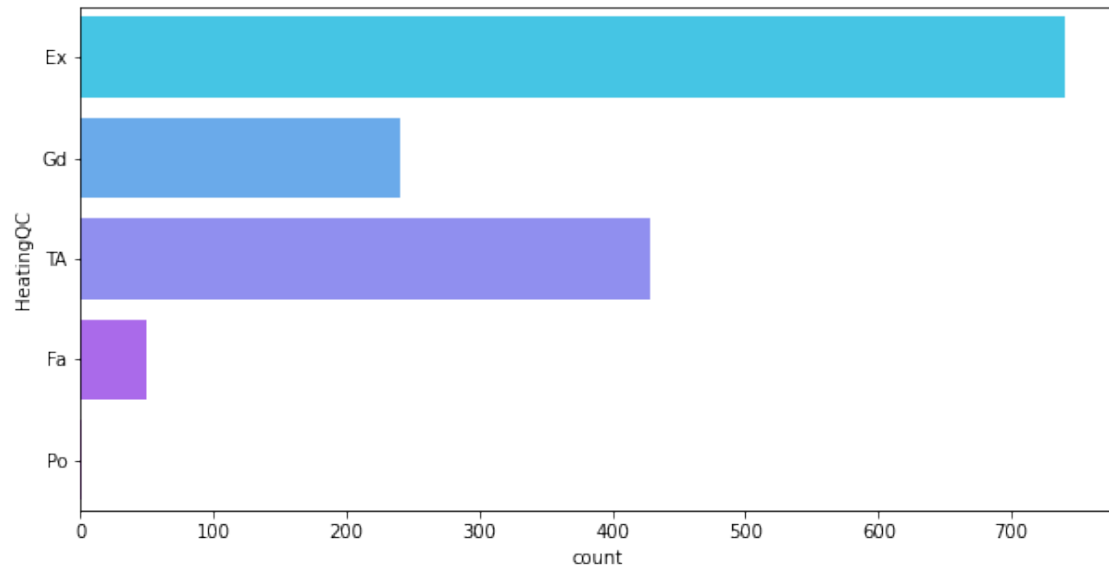


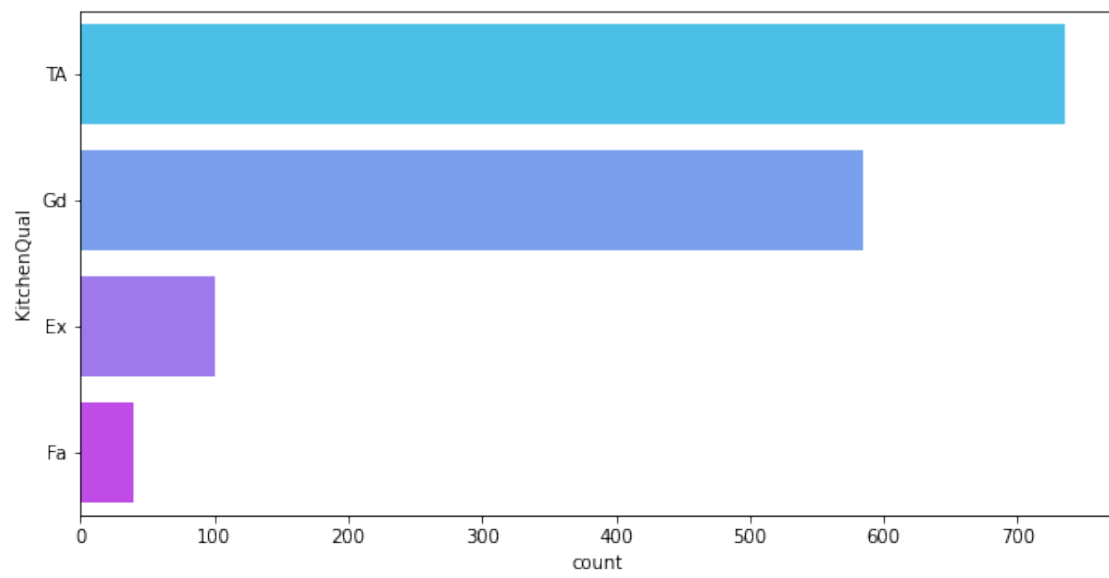
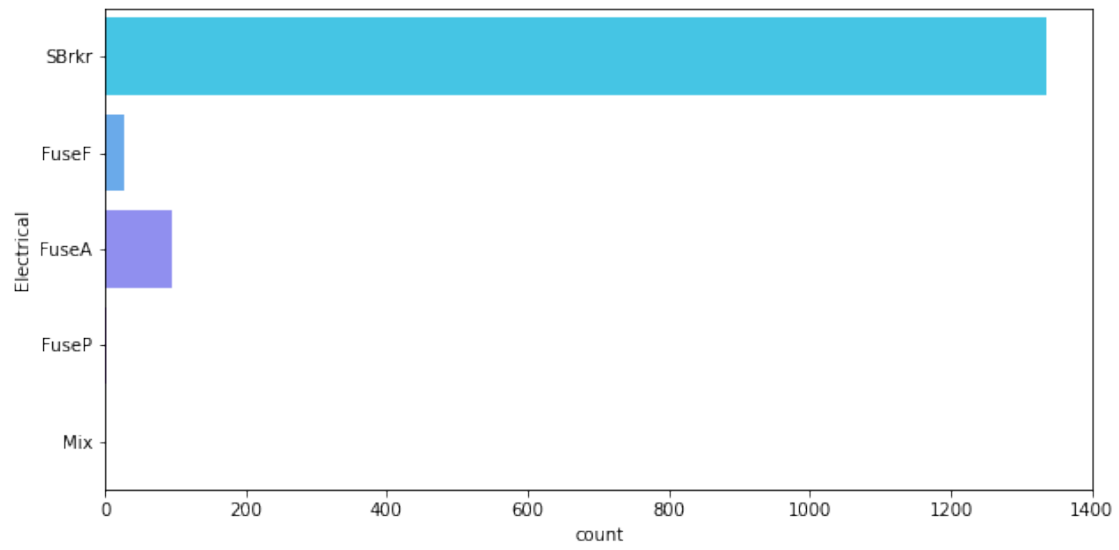


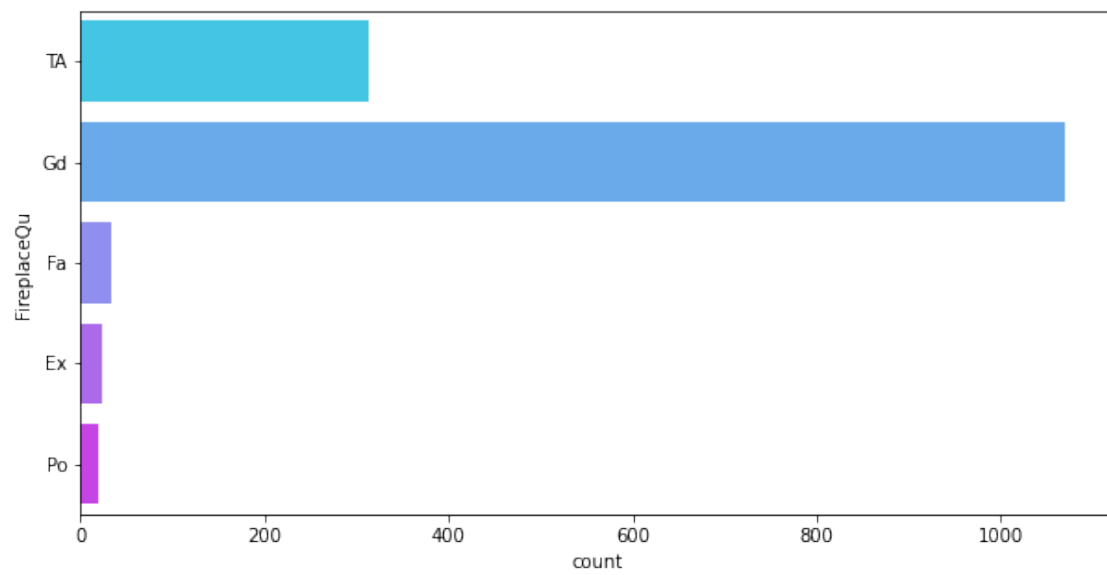
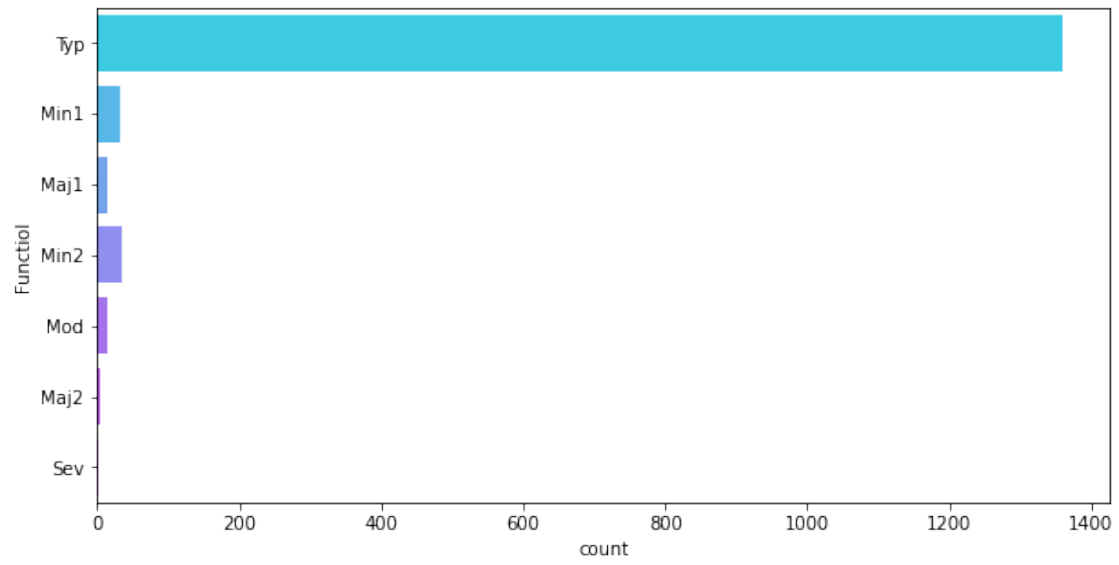


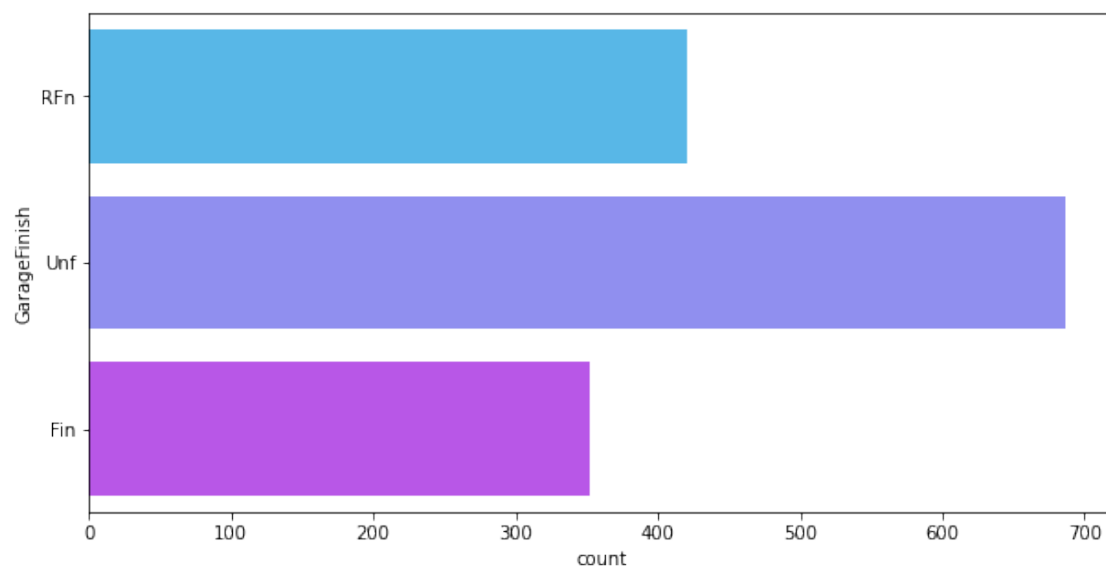
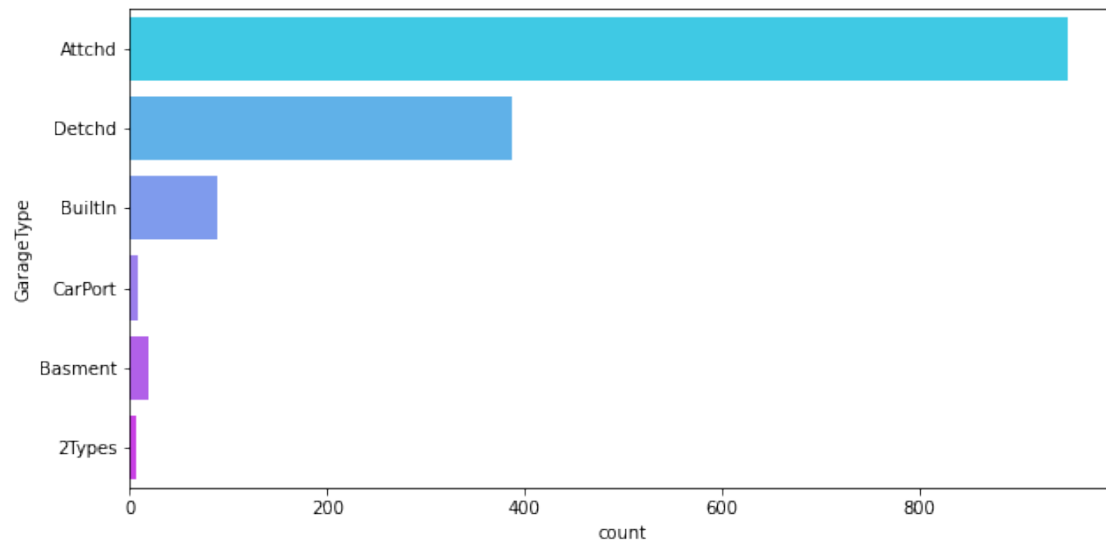


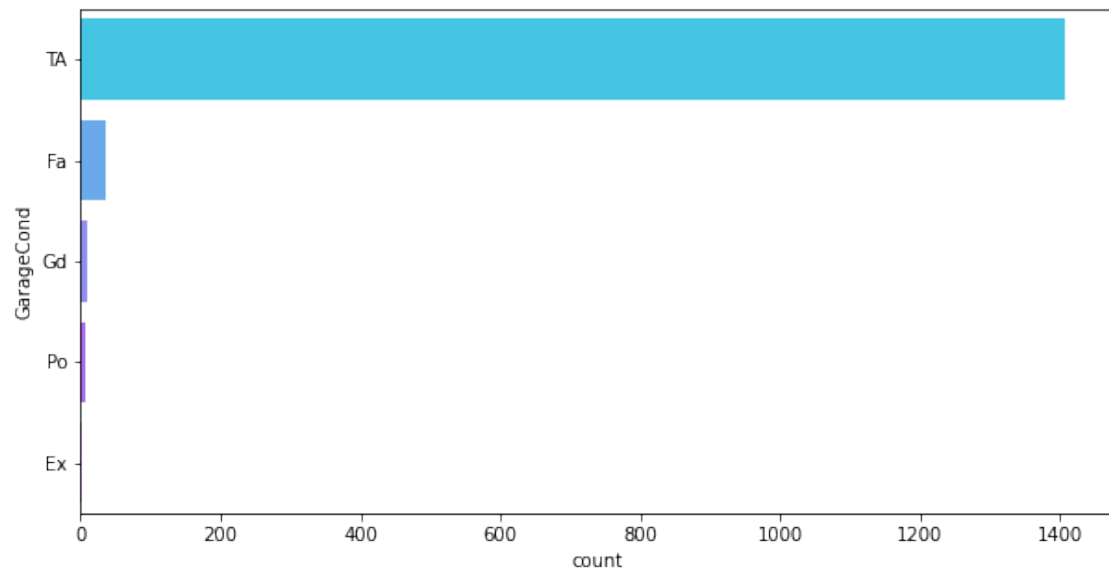
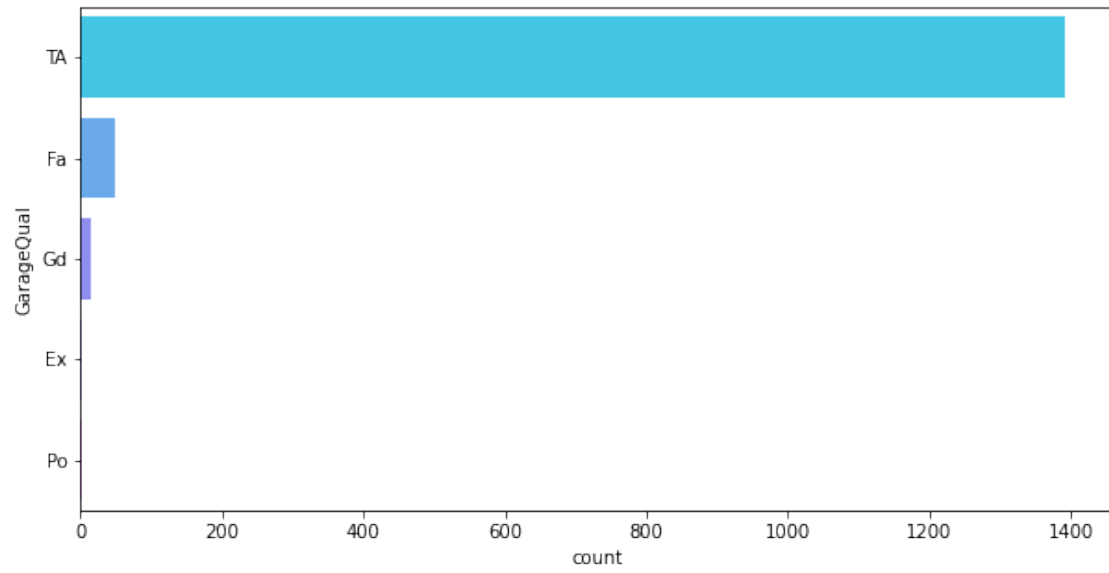


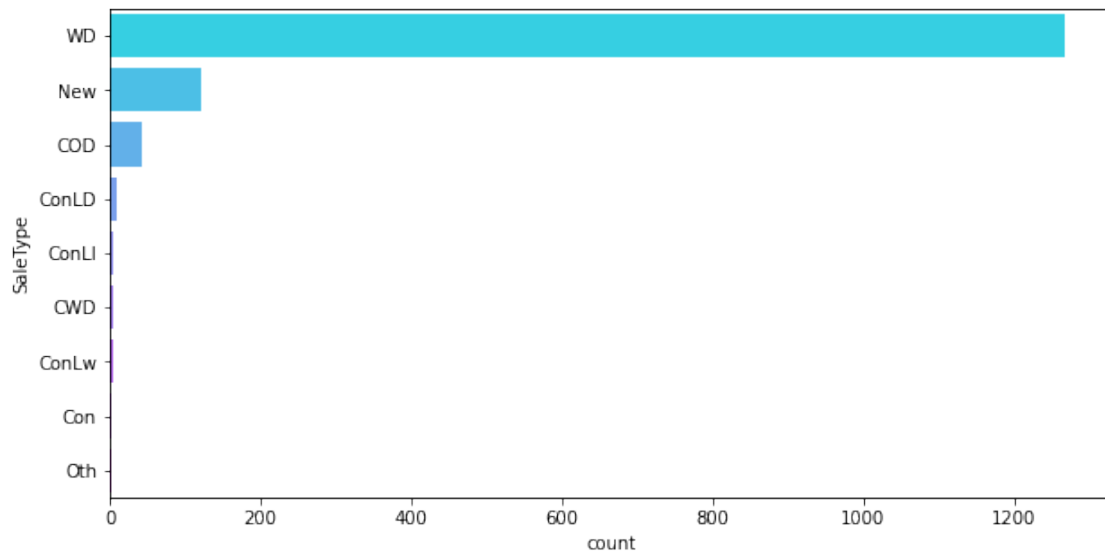
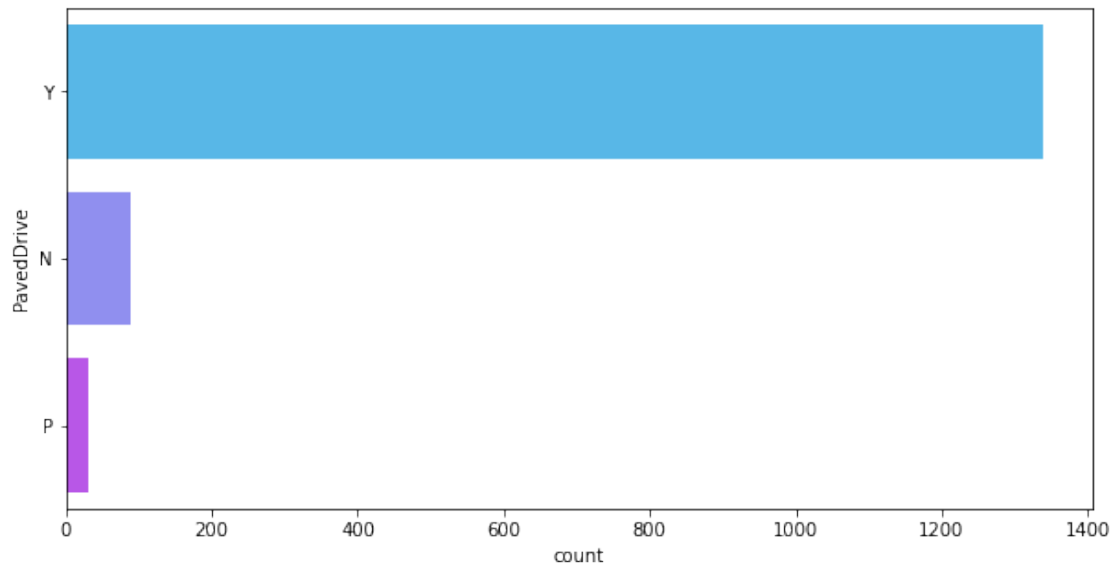


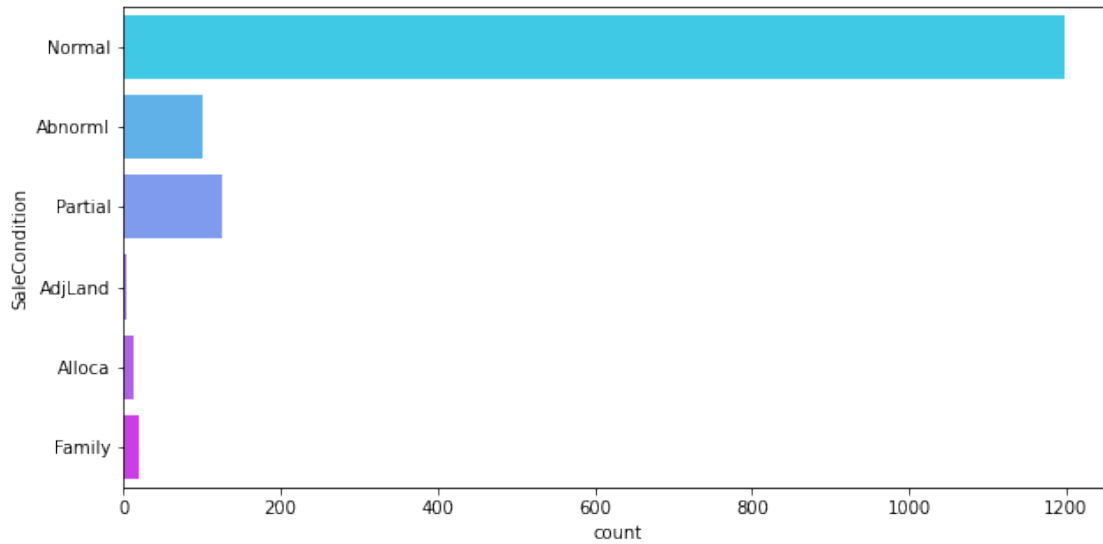






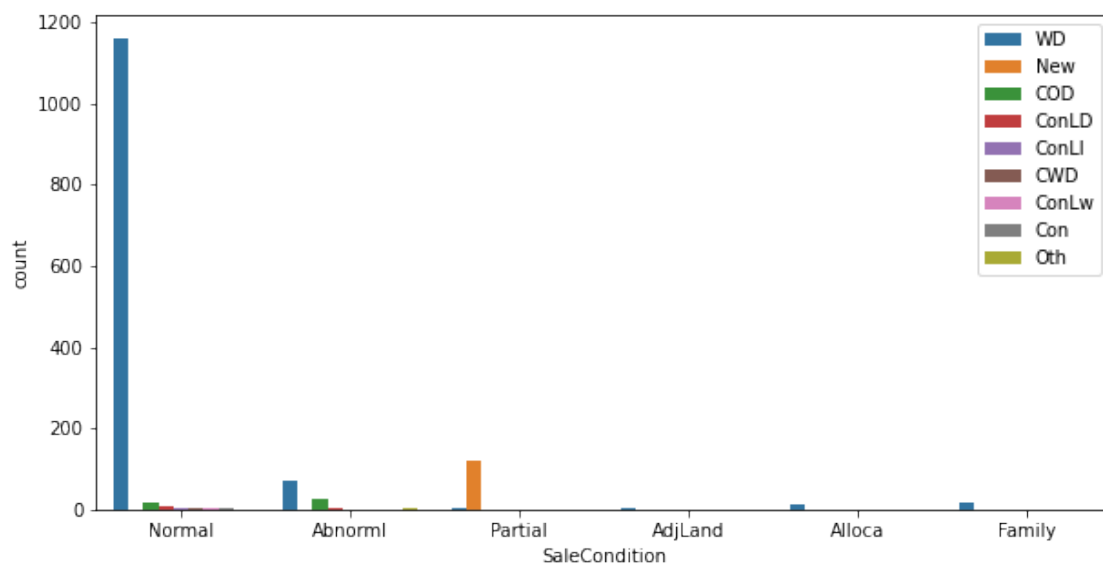






```
[21]: plt.figure(figsize=(10,5))
sns.countplot(impute_categoric_data.SaleCondition, hue = 'SaleType',
↳data=impute_categoric_data)
plt.legend(loc='upper right')
```

[21]: <matplotlib.legend.Legend at 0x2e5ab90c460>



```
[23]: # c. Identify significant variables using p-values and Chi-Square values
```

```

from scipy import stats

data_cross = pd.
    ↳ crosstab(impute_categoric_data['SaleType'], impute_categoric_data['SaleCondition'])
coeff, pval, dof, expec = stats.chi2_contingency(data_cross)
print("chisquare", coeff)
print("Pvalue", pval)
print("DOF", dof)
print("Expected", expec)

```

```

chisquare 1652.6750772643866
Pvalue 0.0
DOF 40
Expected [[2.97465753e+00 1.17808219e-01 3.53424658e-01 5.89041096e-01
 3.52835616e+01 3.68150685e+00]
 [2.76712329e-01 1.09589041e-02 3.28767123e-02 5.47945205e-02
 3.28219178e+00 3.42465753e-01]
 [1.38356164e-01 5.47945205e-03 1.64383562e-02 2.73972603e-02
 1.64109589e+00 1.71232877e-01]
 [6.22602740e-01 2.46575342e-02 7.39726027e-02 1.23287671e-01
 7.38493151e+00 7.70547945e-01]
 [3.45890411e-01 1.36986301e-02 4.10958904e-02 6.84931507e-02
 4.10273973e+00 4.28082192e-01]
 [3.45890411e-01 1.36986301e-02 4.10958904e-02 6.84931507e-02
 4.10273973e+00 4.28082192e-01]
 [8.43972603e+00 3.34246575e-01 1.00273973e+00 1.67123288e+00
 1.00106849e+02 1.04452055e+01]
 [2.07534247e-01 8.21917808e-03 2.46575342e-02 4.10958904e-02
 2.46164384e+00 2.56849315e-01]
 [8.76486301e+01 3.47123288e+00 1.04136986e+01 1.73561644e+01
 1.03963425e+03 1.08476027e+02]]

```

```

[24]: if pval < 0.05:
        print("Alter Hypo----->relation exist")
    else:
        print("Null Hypo----->No relation")

```

Alter Hypo----->relation exist

5. Combine all the significant categorical and numerical variables

```

[25]: df_combine = pd.concat([impute_numeric_data, impute_categoric_data], axis=1)
df_combine.
    ↳ drop(['Id', 'LotFrontage', 'LotArea', 'TotRmsAbvGrd', 'YearBuilt', 'MasVnrArea', 'BsmtFinSF1', 'BsmtFinSF2', 'BsmtUnfSF', 'TotalBsmtSF', 'SaleType', 'SaleCondition'])

```

```

        \
        ↳ 'TotalBsmtSF', 'EnclosedPorch', 'YearRemodAdd', 'GarageYrBlt', 'BsmtFullBath', 'BsmtUnfSF', 'MSSu
        \
        ↳ '1stFlrSF', '2ndFlrSF', 'PoolQC', 'MiscFeature', 'Fence', 'Alley', 'GrLivArea', 'GarageArea',
        'WoodDeckSF', 'OpenPorchSF', 'ScreenPorch'], axis=1, \
        ↳ inplace=True)
df_combine.head()

```

```

[25]: OverallQual OverallCond LowQualFinSF BsmtHalfBath FullBath HalfBath \
0      7.0      5.0      0.0      0.0      2.0      1.0
1      6.0      8.0      0.0      1.0      2.0      0.0
2      7.0      5.0      0.0      0.0      2.0      1.0
3      7.0      5.0      0.0      0.0      1.0      0.0
4      8.0      5.0      0.0      0.0      2.0      1.0

```

```

    BedroomAbvGr KitchenQual \
0      3.0      1.0      0.0      2.0 ...      Gd
1      3.0      1.0      1.0      2.0 ...      TA
2      3.0      1.0      1.0      2.0 ...      Gd
3      3.0      1.0      1.0      3.0 ...      Gd
4      4.0      1.0      1.0      3.0 ...      Gd

```

```

    Functional FireplaceQual GarageType GarageFinish GarageQual GarageCond \
0      Typ      Gd      Attchd      RFn      TA      TA
1      Typ      TA      Attchd      RFn      TA      TA
2      Typ      TA      Attchd      RFn      TA      TA
3      Typ      Gd      Detchd      Unf      TA      TA
4      Typ      TA      Attchd      RFn      TA      TA

```

```

    PavedDrive SaleType SaleCondition
0      Y      WD      Normal
1      Y      WD      Normal
2      Y      WD      Normal
3      Y      WD      Abnorml
4      Y      WD      Normal

```

[5 rows x 55 columns]

6. Plot box plot for the new dataset to find the variables with outliers

```

[26]: cols = ['OverallQual', 'OverallCond', 'LowQualFinSF', 'BsmtHalfBath', \
        ↳ 'FullBath', 'HalfBath', 'BedroomAbvGr', 'KitchenQual',
        'Fireplaces', 'GarageCars', '3SsnPorch', 'PoolArea', 'MiscVal', 'MoSold', \
        ↳ 'YrSold', 'MSZoning', 'Street', 'LotShape',

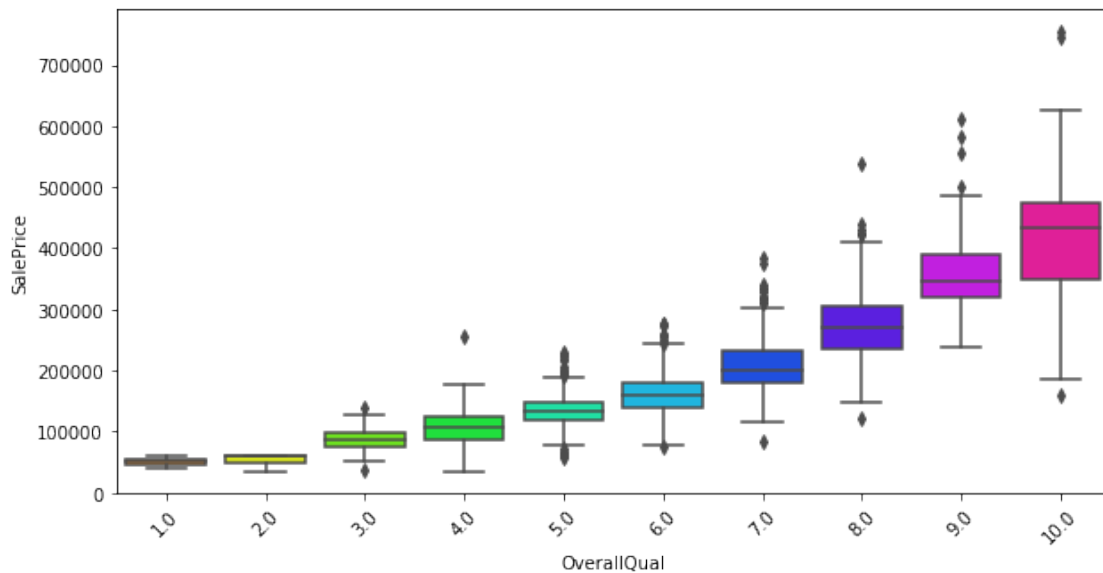
```

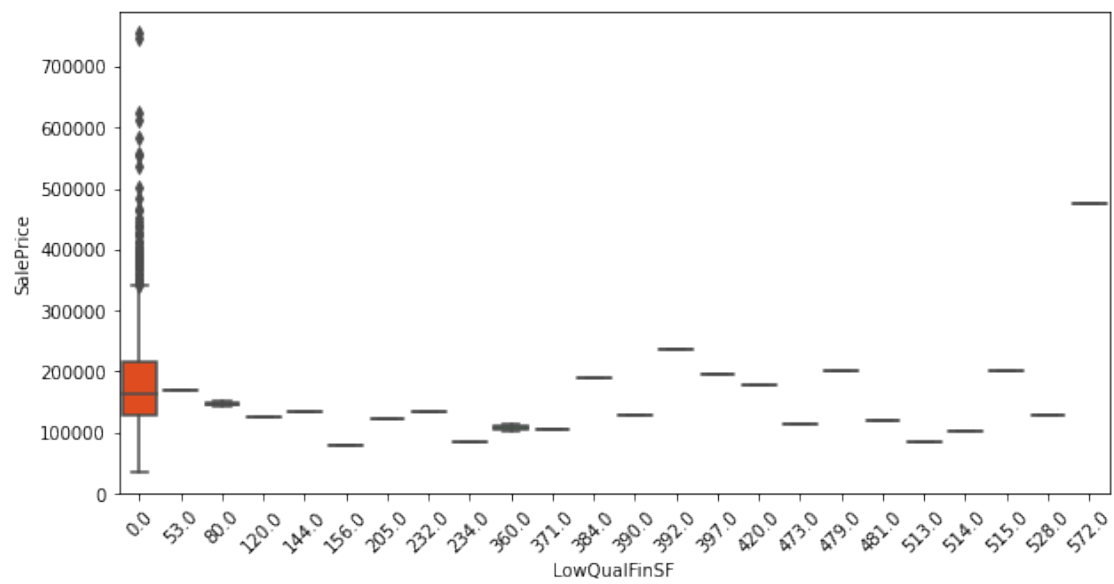
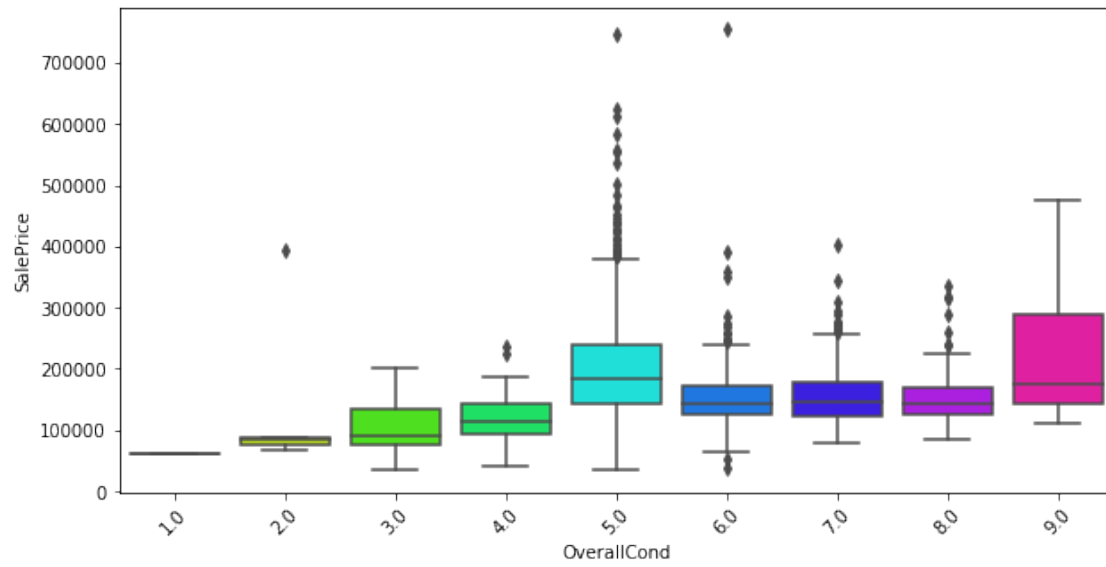
```

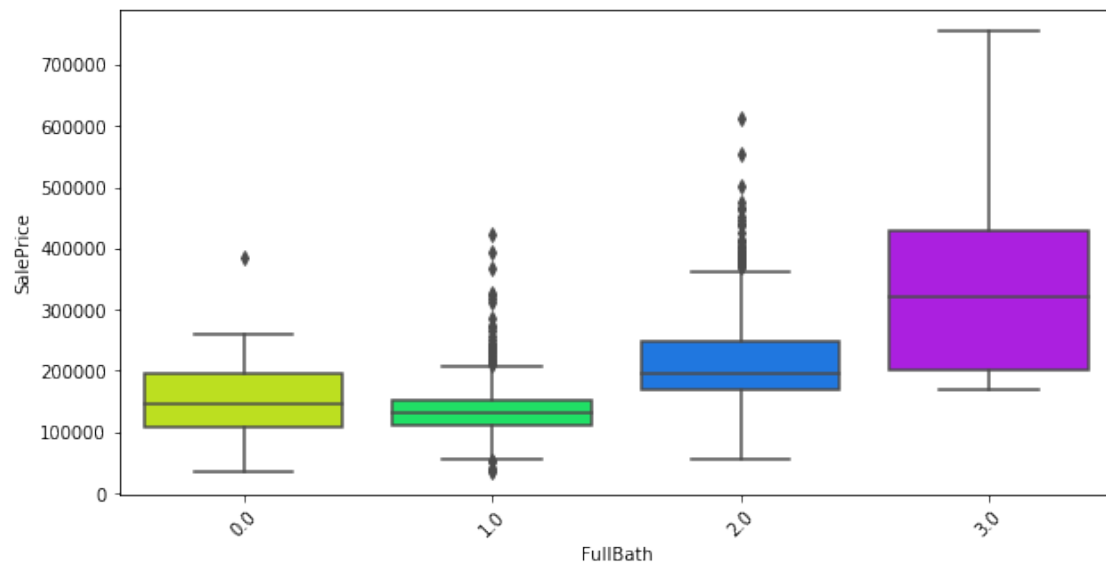
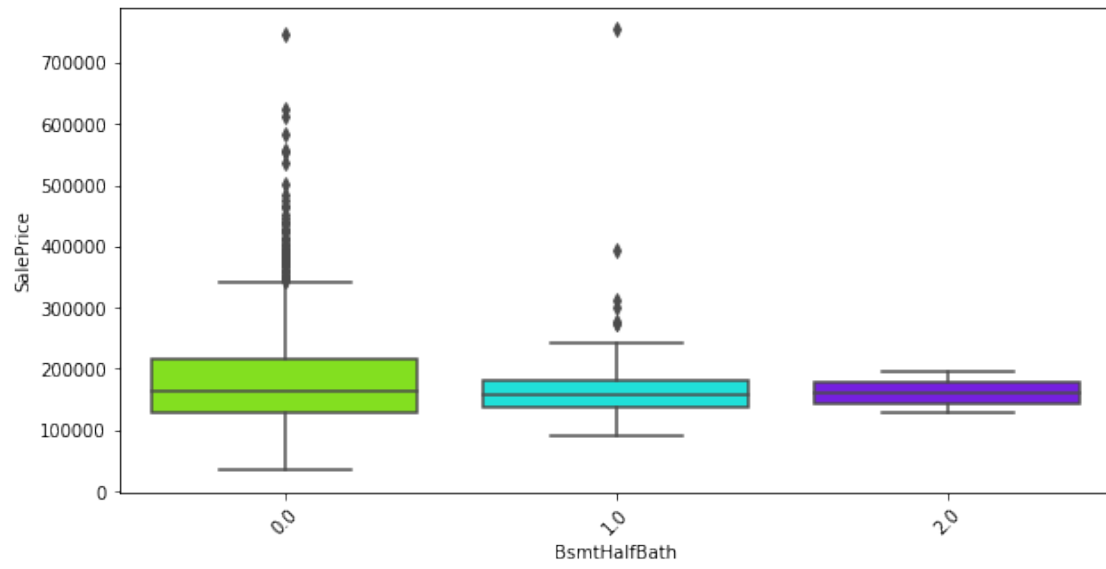
    'LandContour','Utilities','LotConfig','LandSlope','Neighborhood',
    ↪ 'Condition1', 'Condition2', 'BldgType',
    'HouseStyle', 'RoofStyle', 'RoofMatl', 'Exterior1st',
    ↪ 'Exterior2nd','MasVnrType', 'ExterQual', 'ExterCond',
    'Foundation', 'BsmtQual','BsmtCond', 'BsmtExposure', 'BsmtFinType1',
    ↪ 'BsmtFinType2', 'Heating','HeatingQC',
    'CentralAir', 'Electrical', 'KitchenQual', 'Function1','FireplaceQu',
    ↪ 'GarageType', 'GarageFinish', 'GarageQual',
    'GarageCond','PavedDrive', 'SaleType', 'SaleCondition']

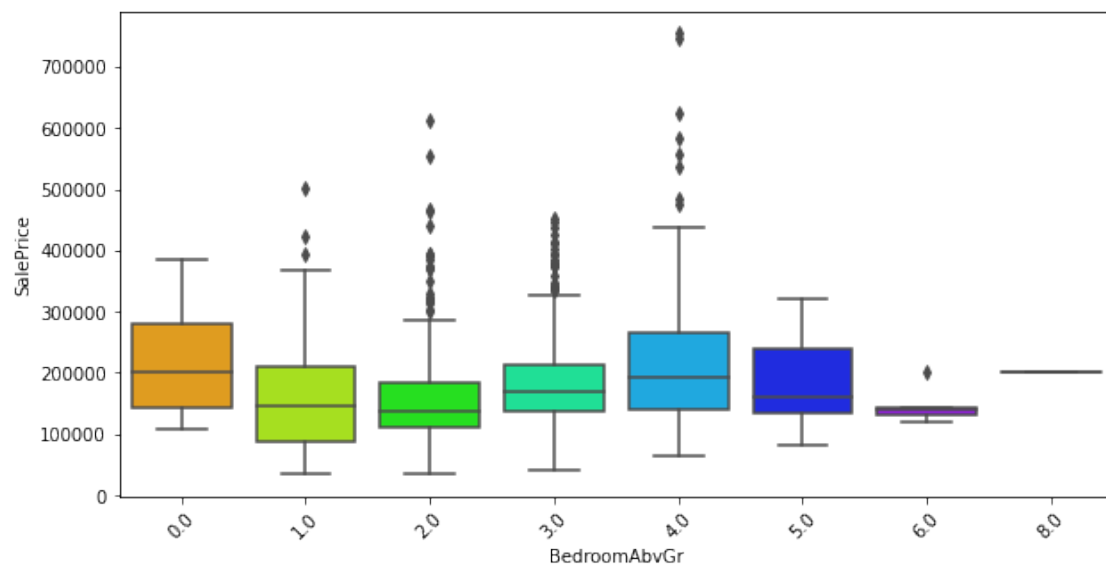
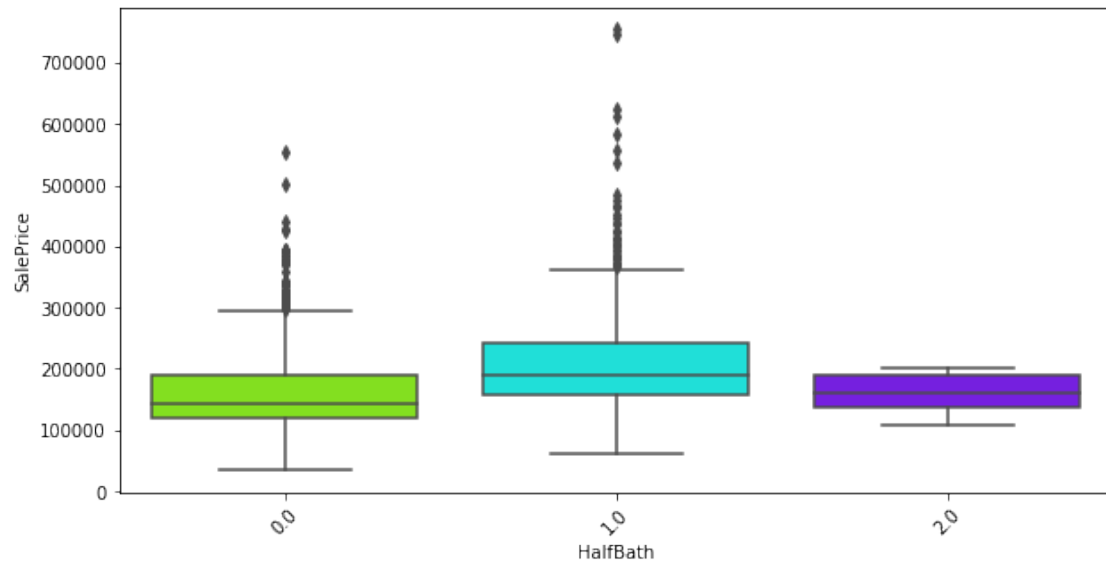
for i in cols:
    fig, ax = plt.subplots(1,1, figsize=(10,5))
    plt.xticks(rotation = 45)
    sns.boxplot(df_combine[i][1:], 'SalePrice', data=df_combine,palette='hsv')

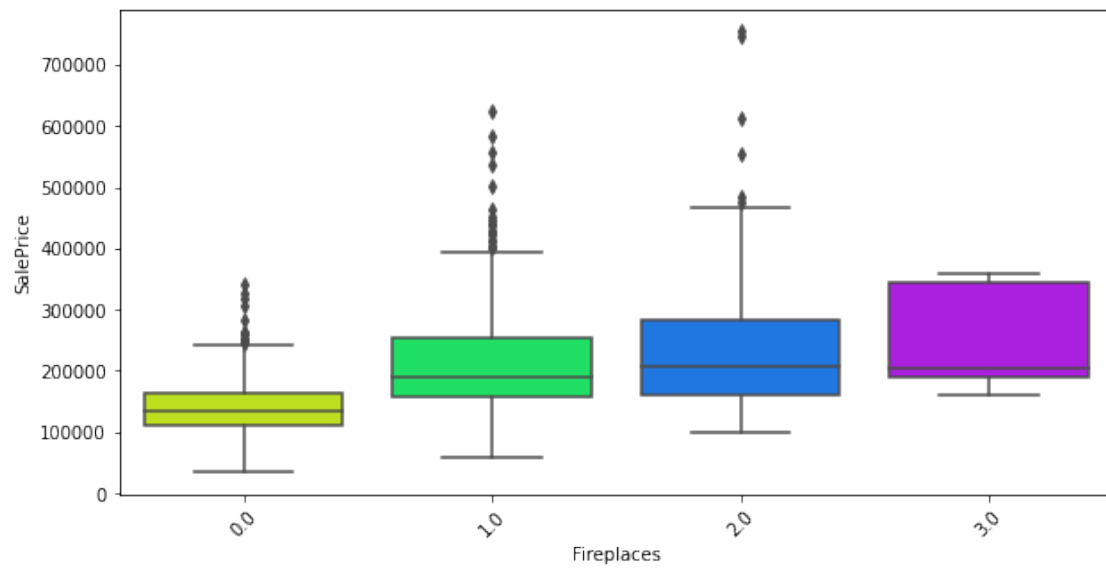
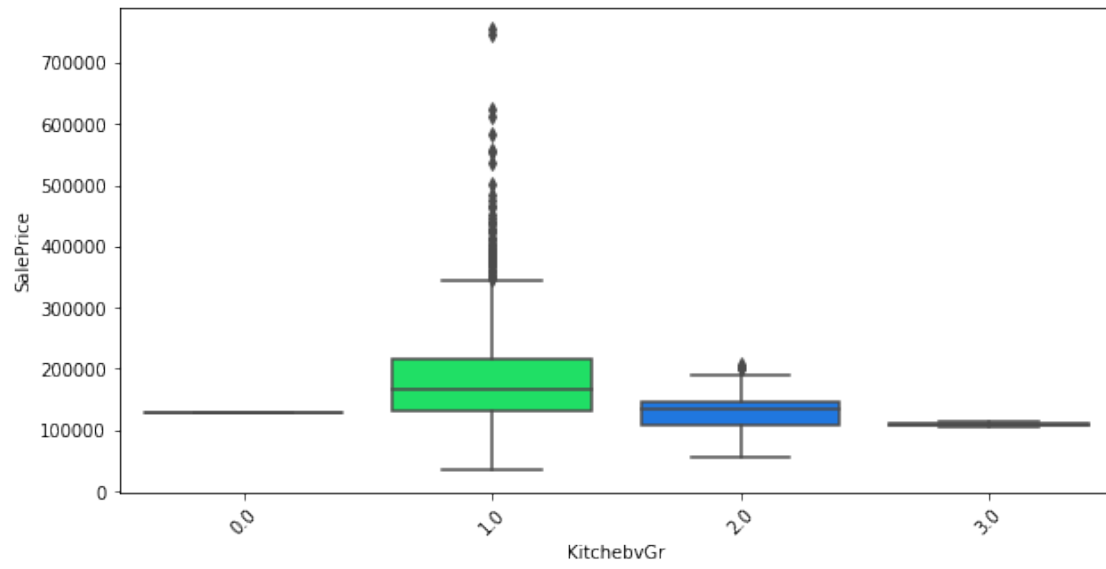
```

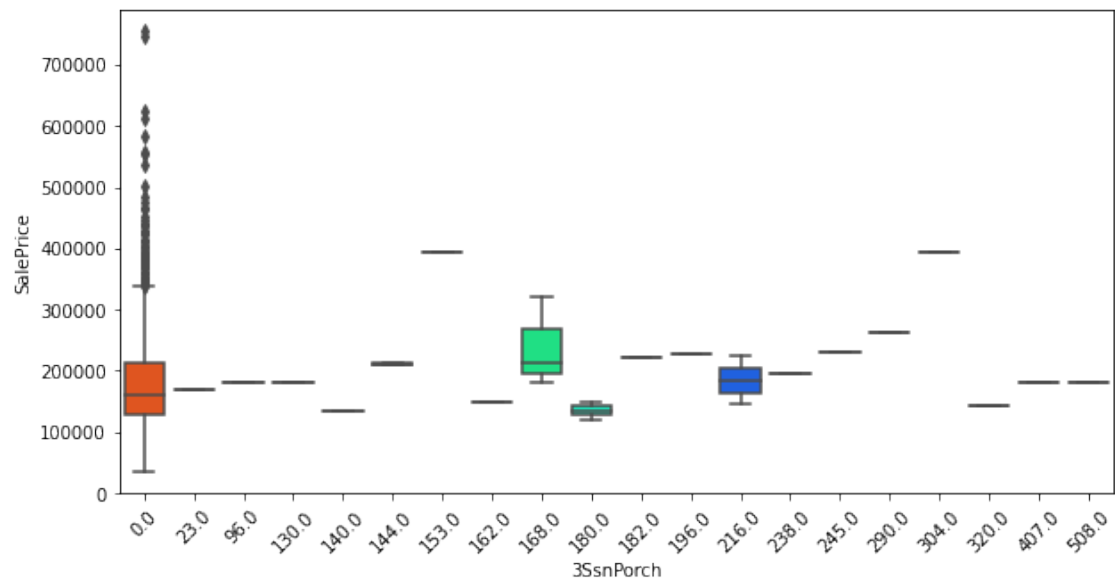
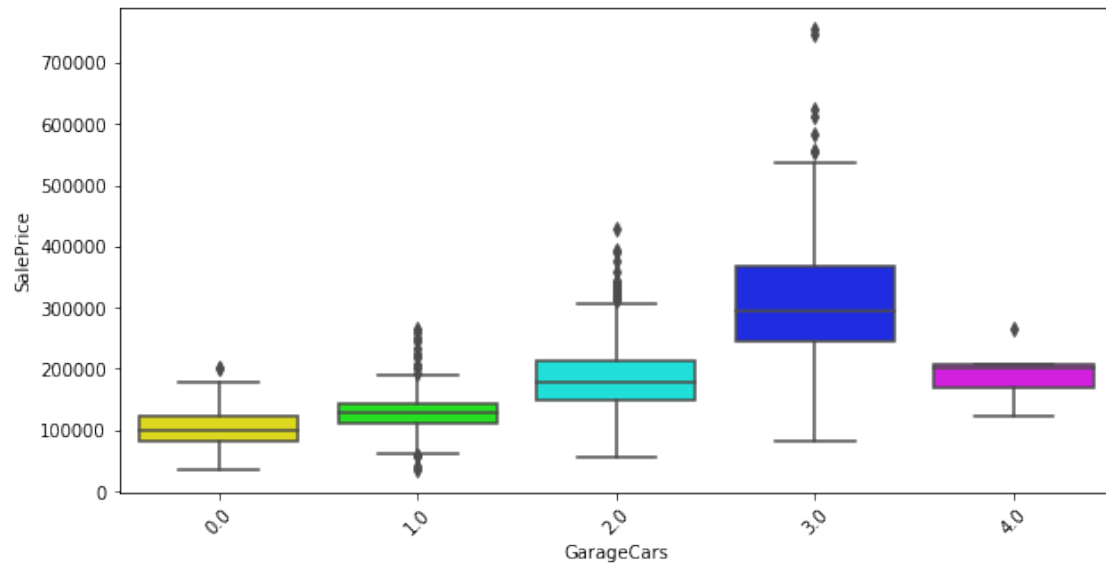


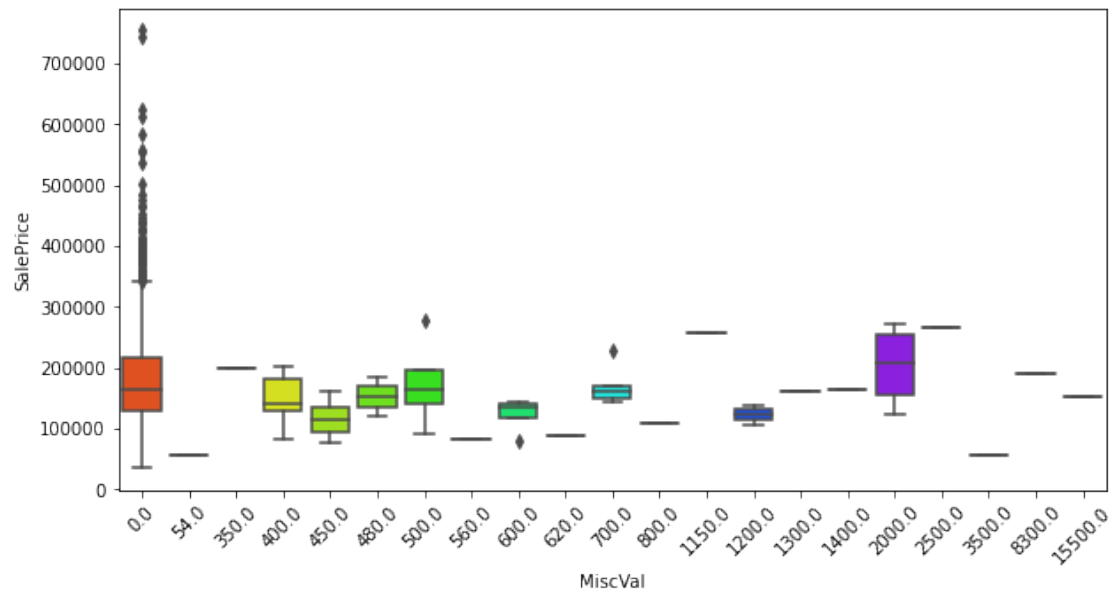
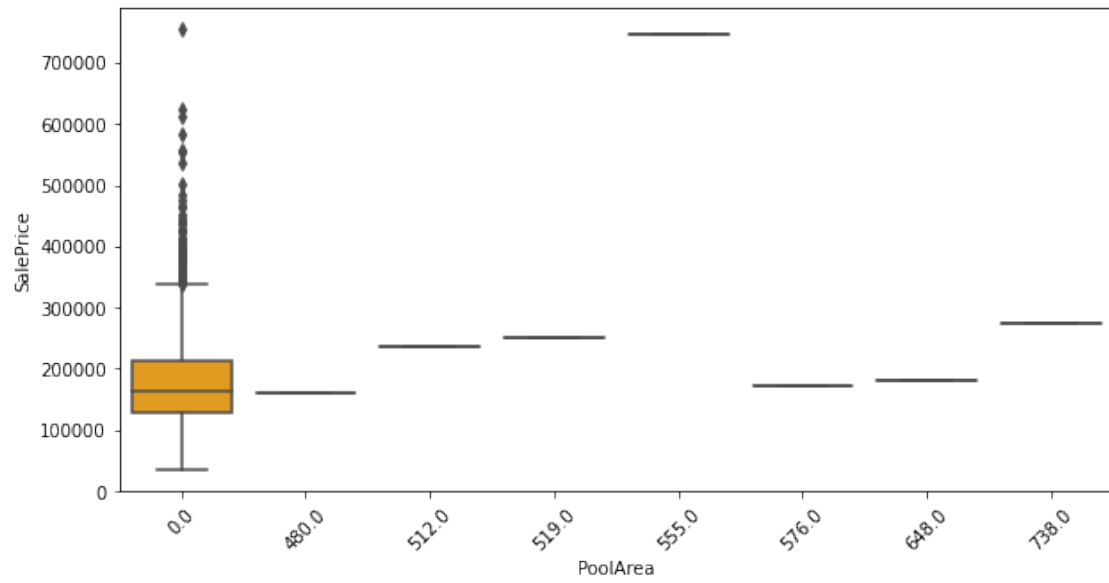


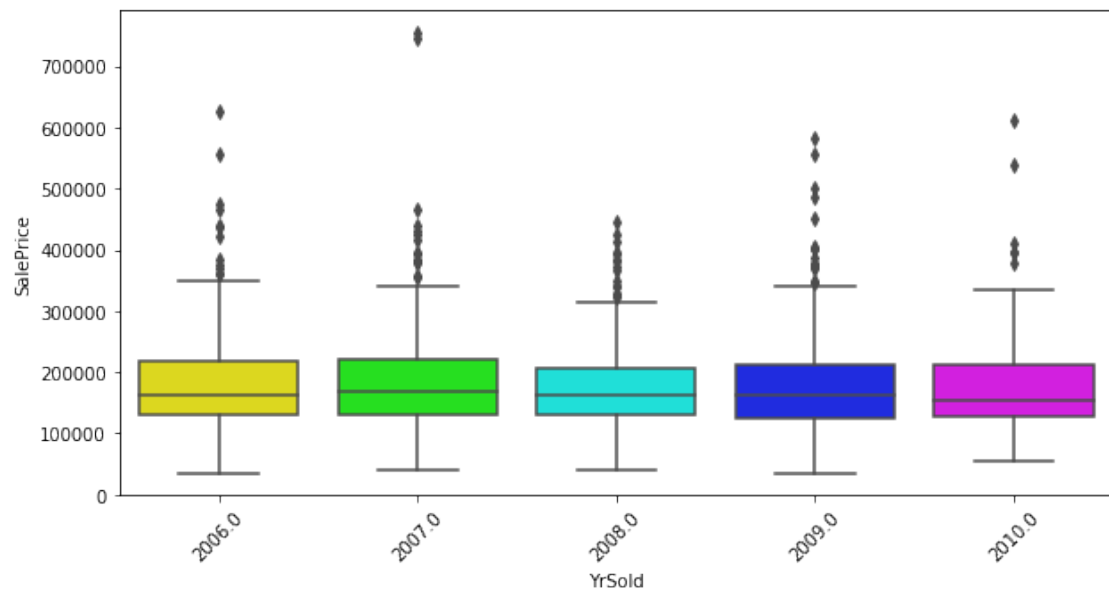
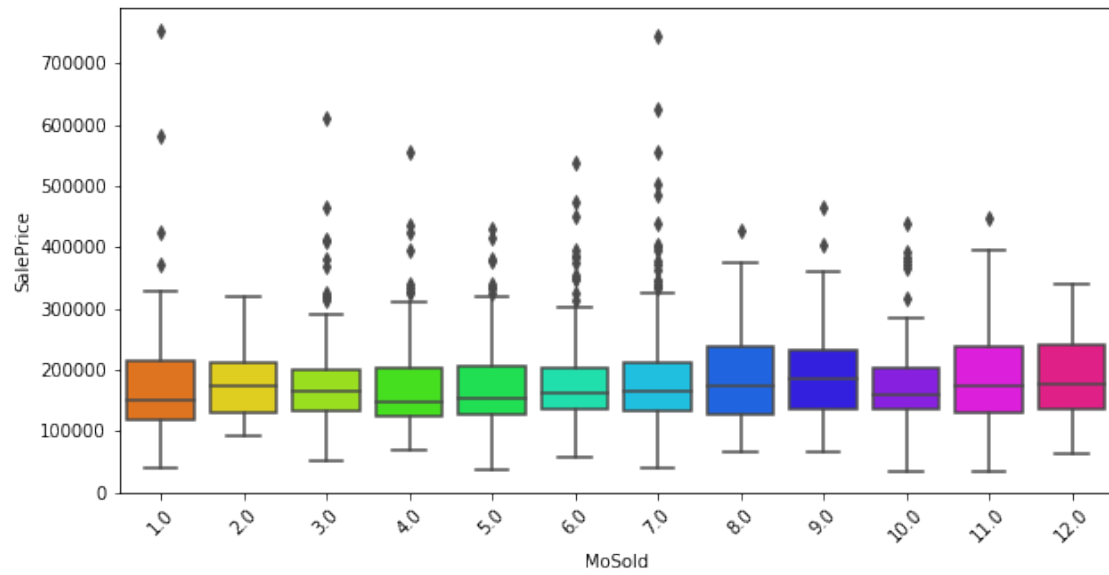


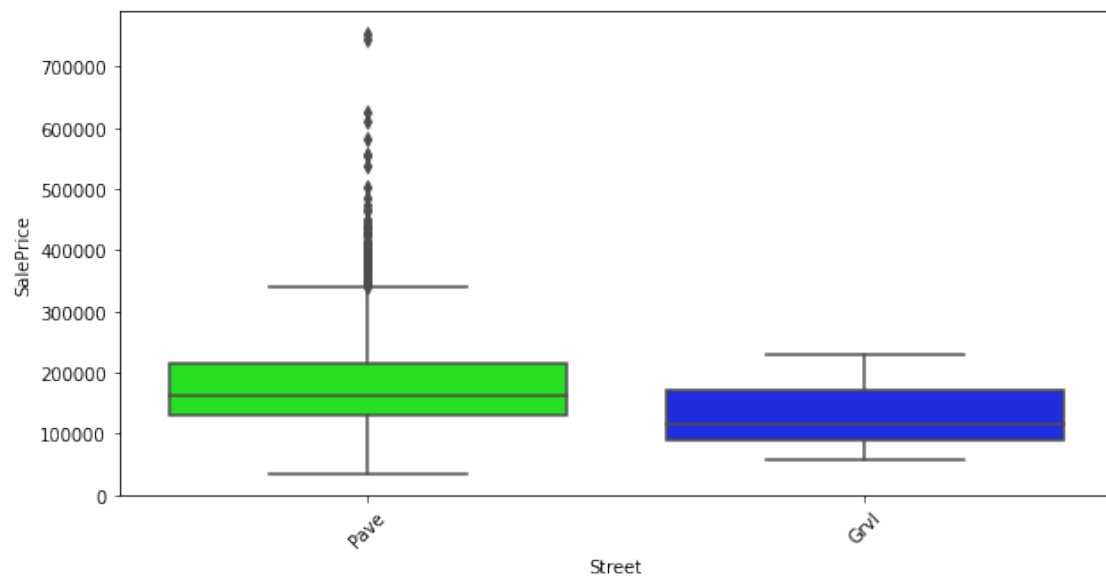
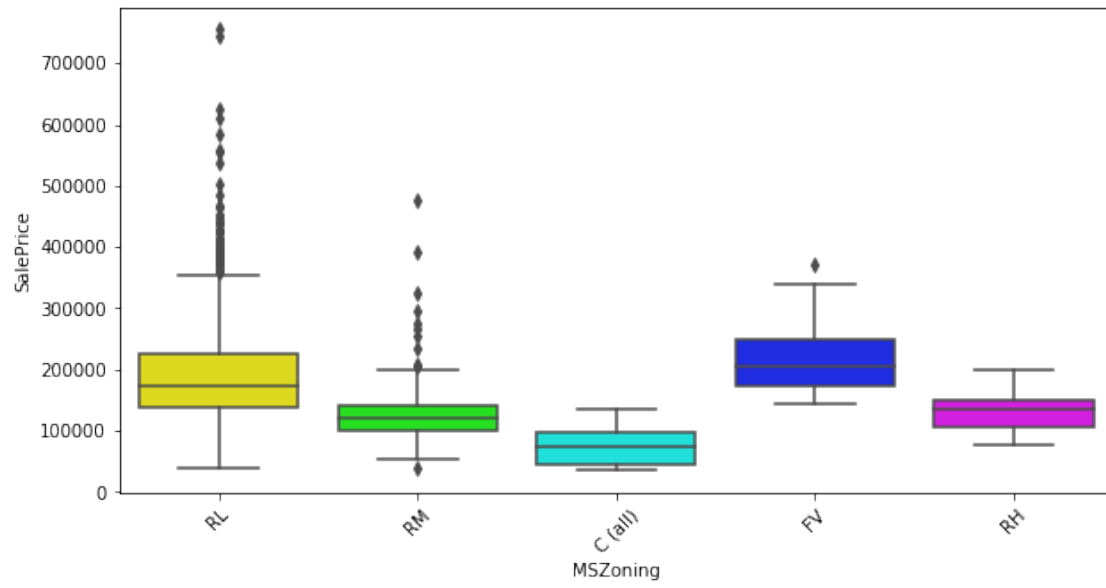


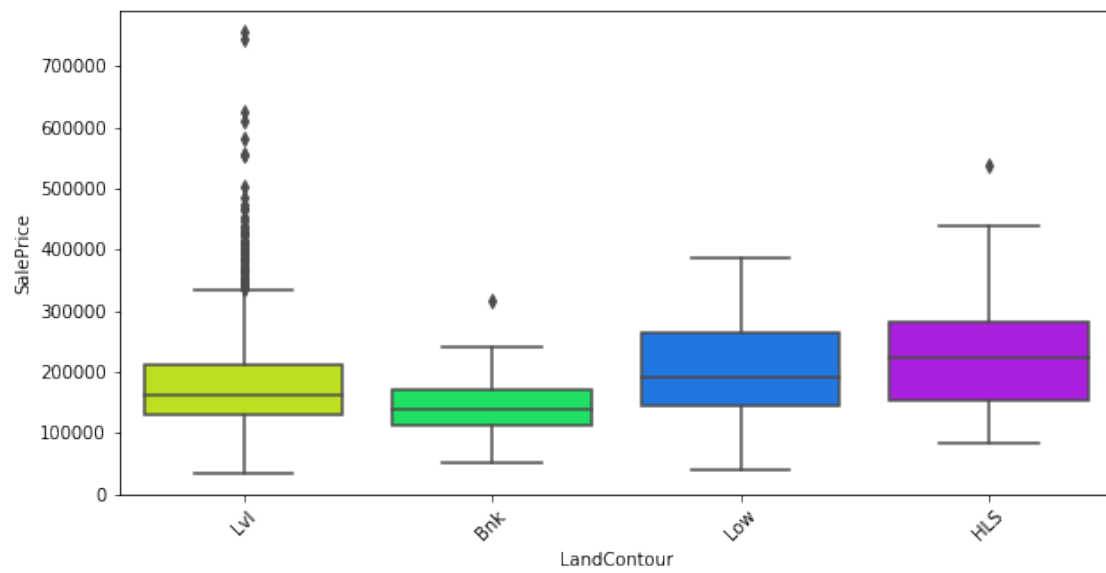
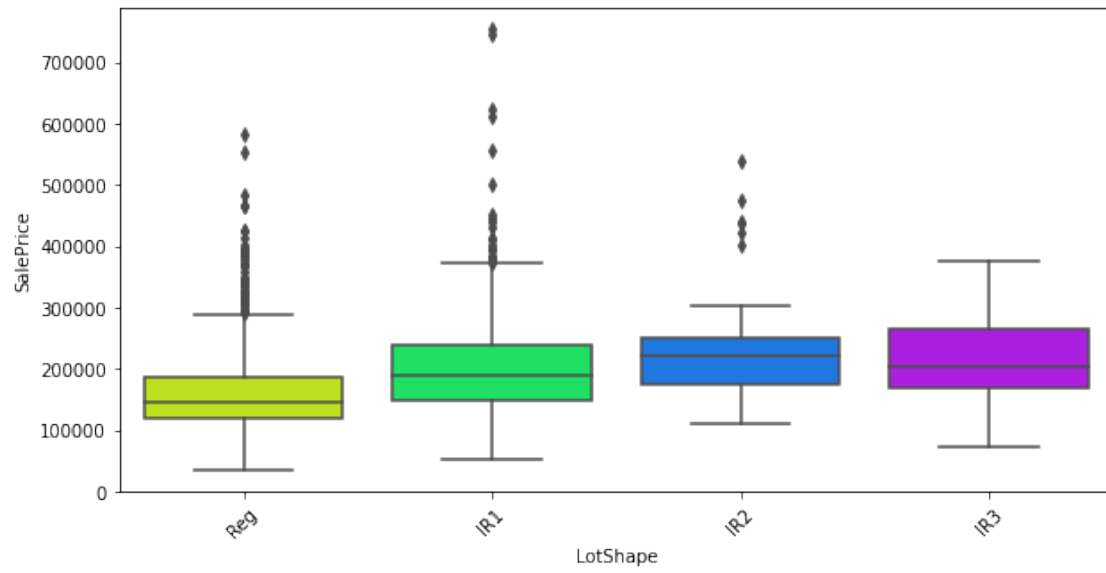


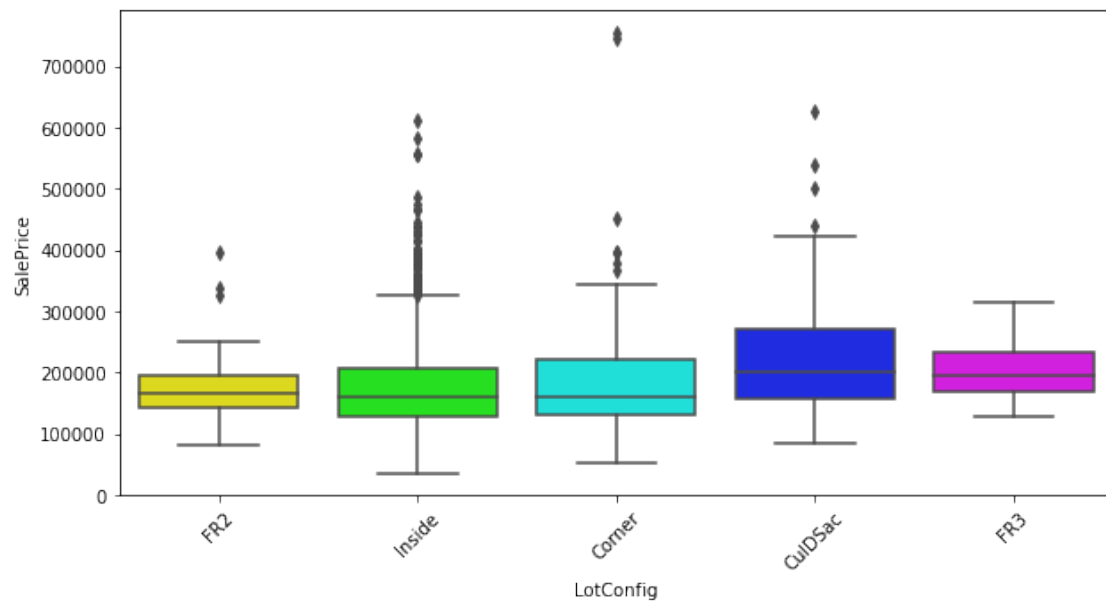
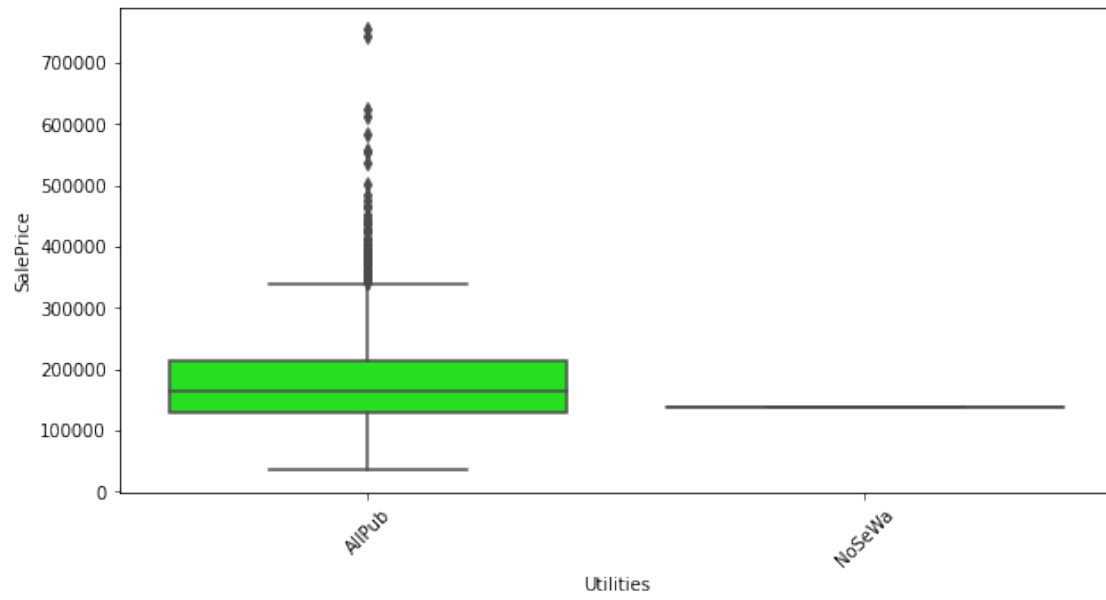


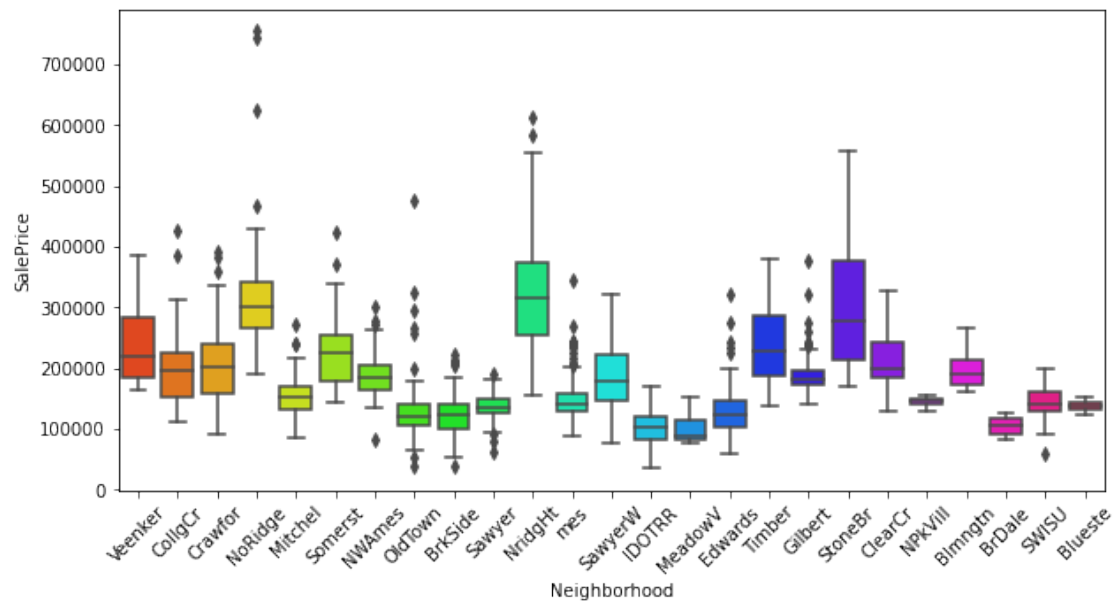
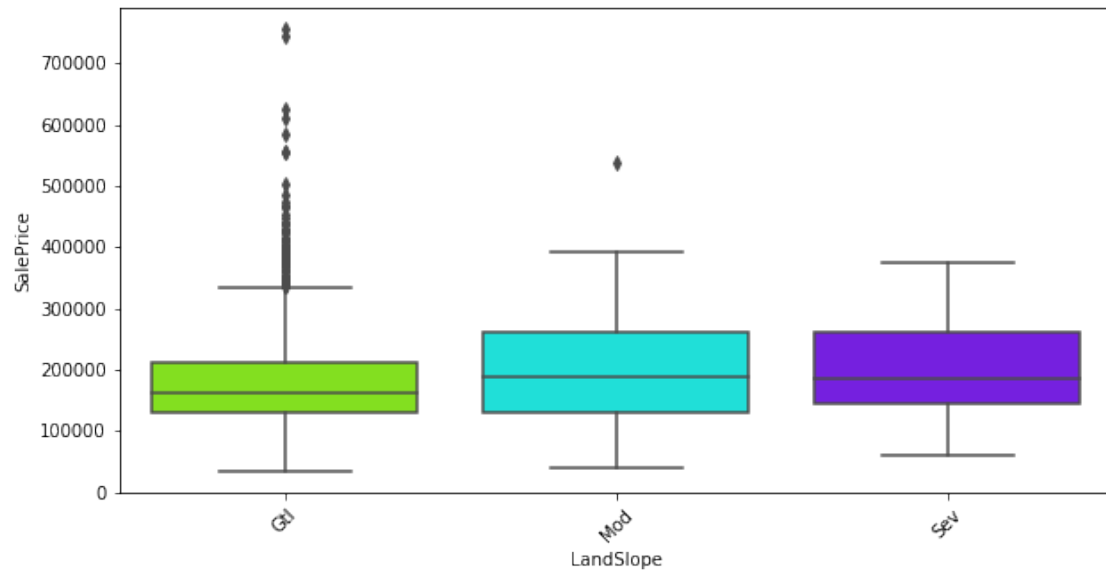


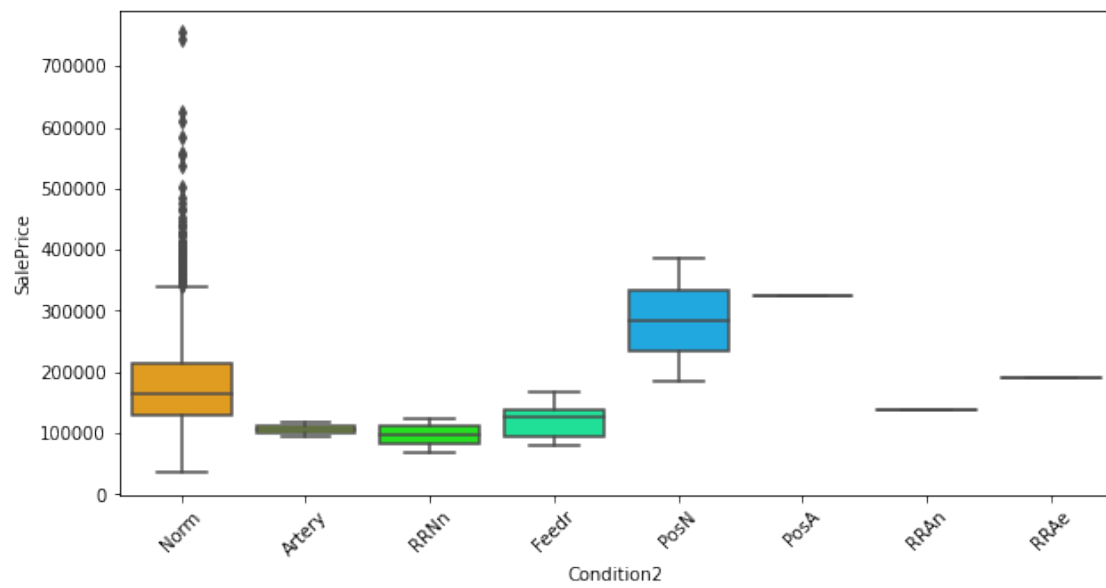
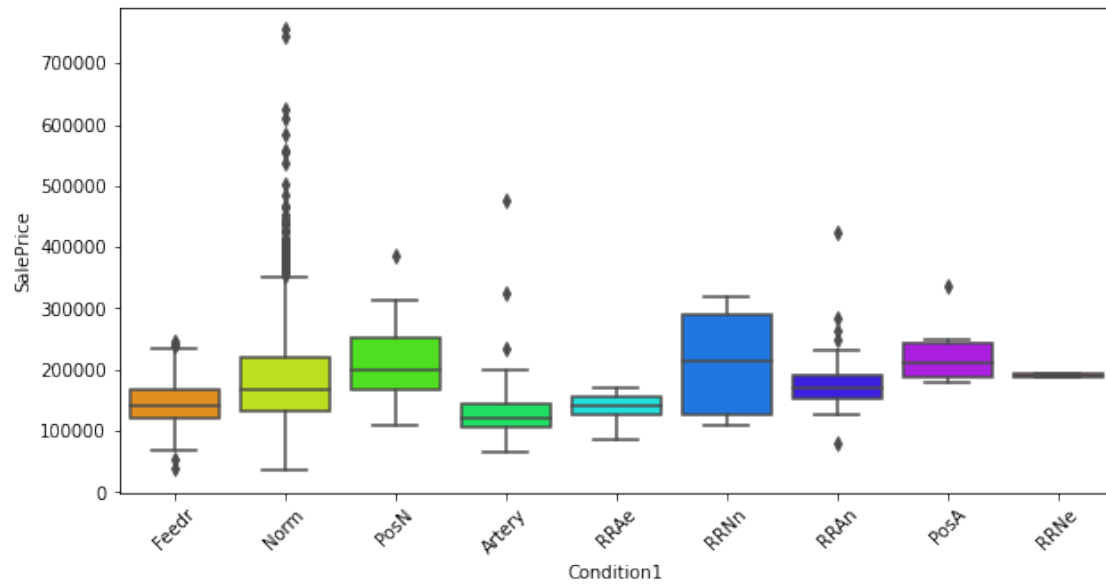


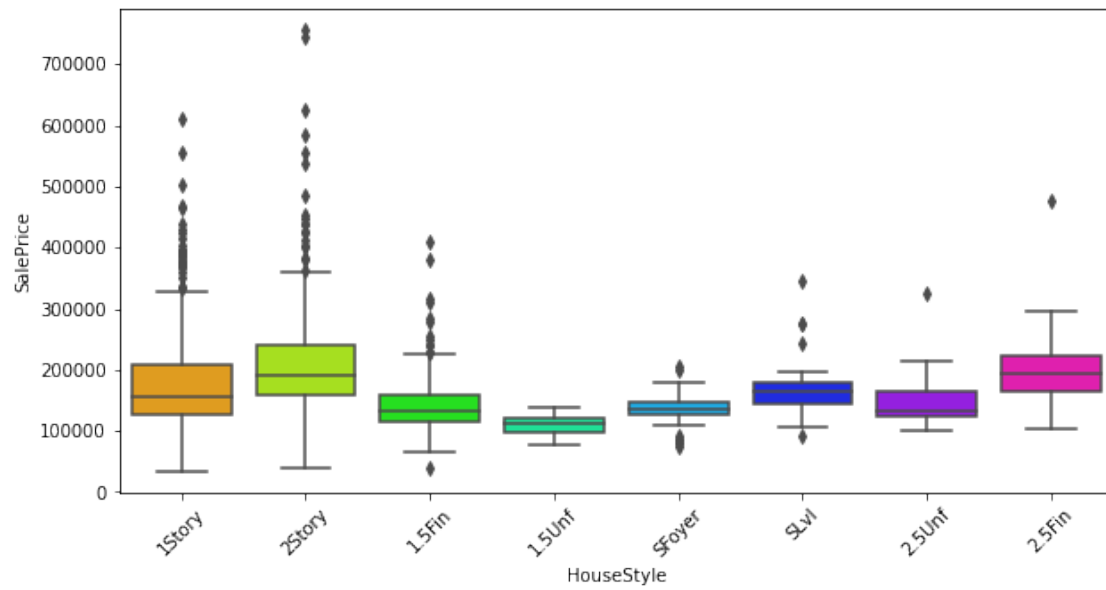
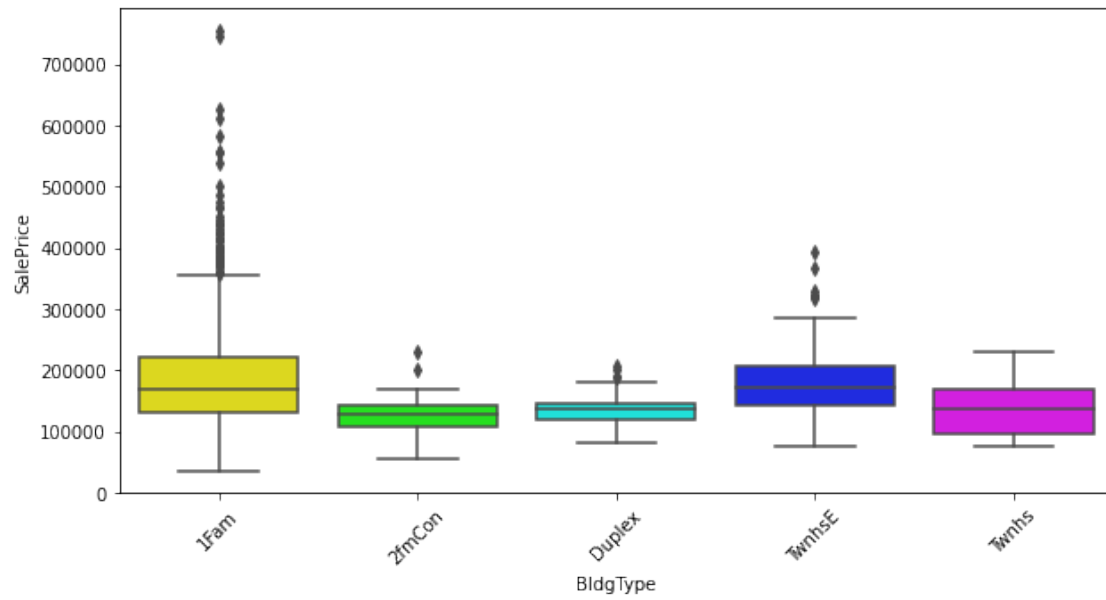


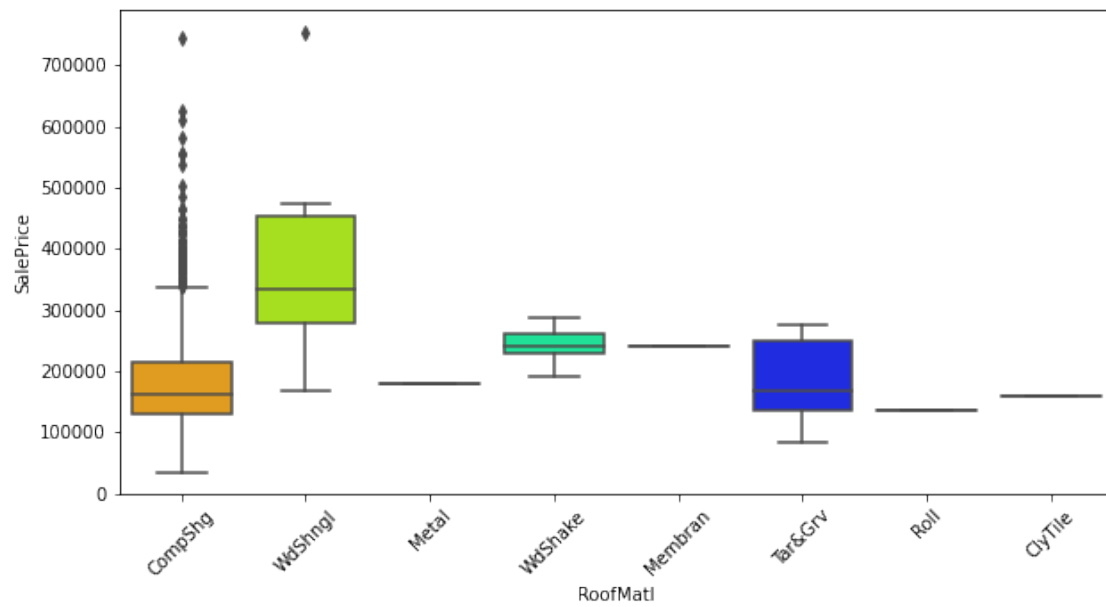
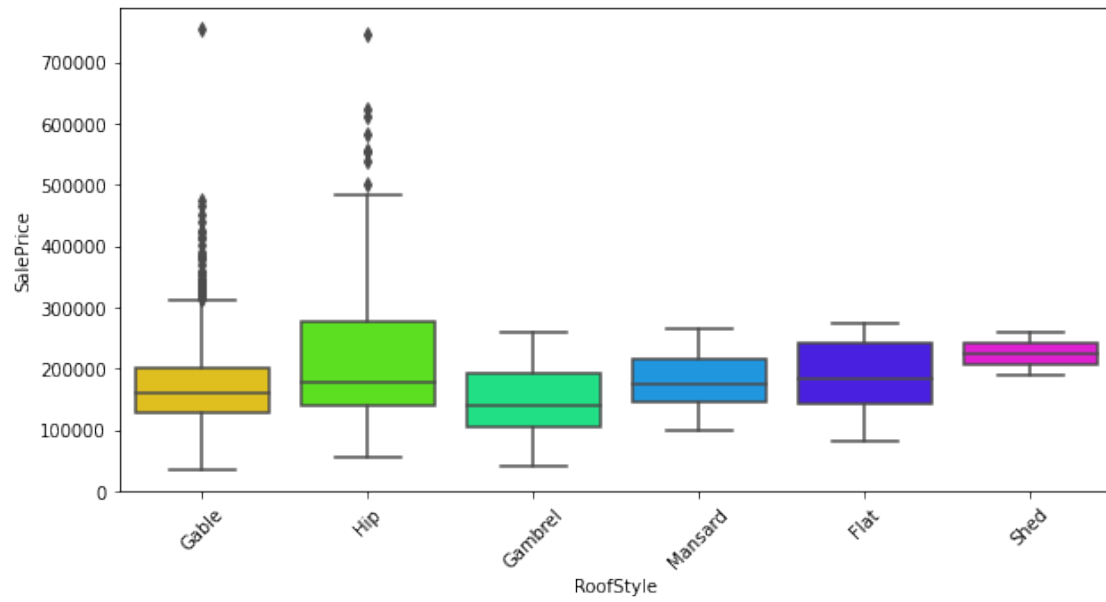


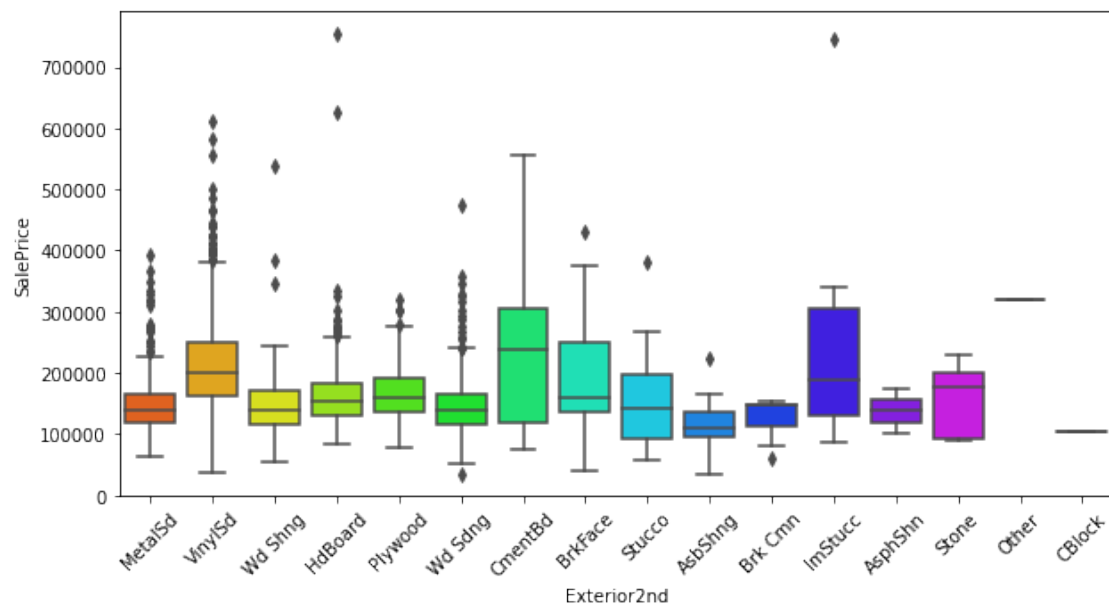
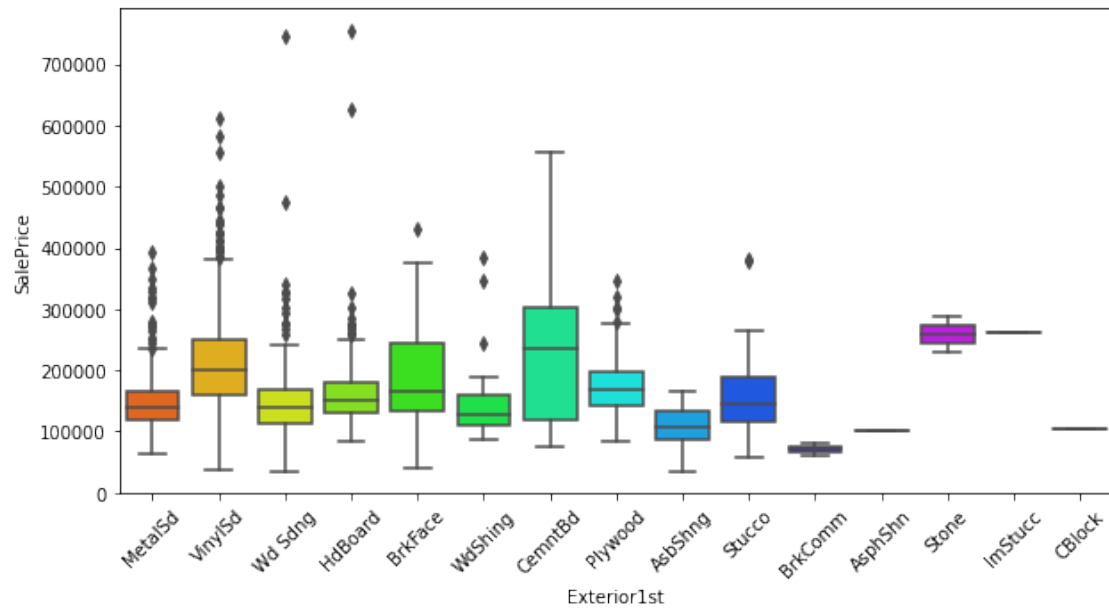


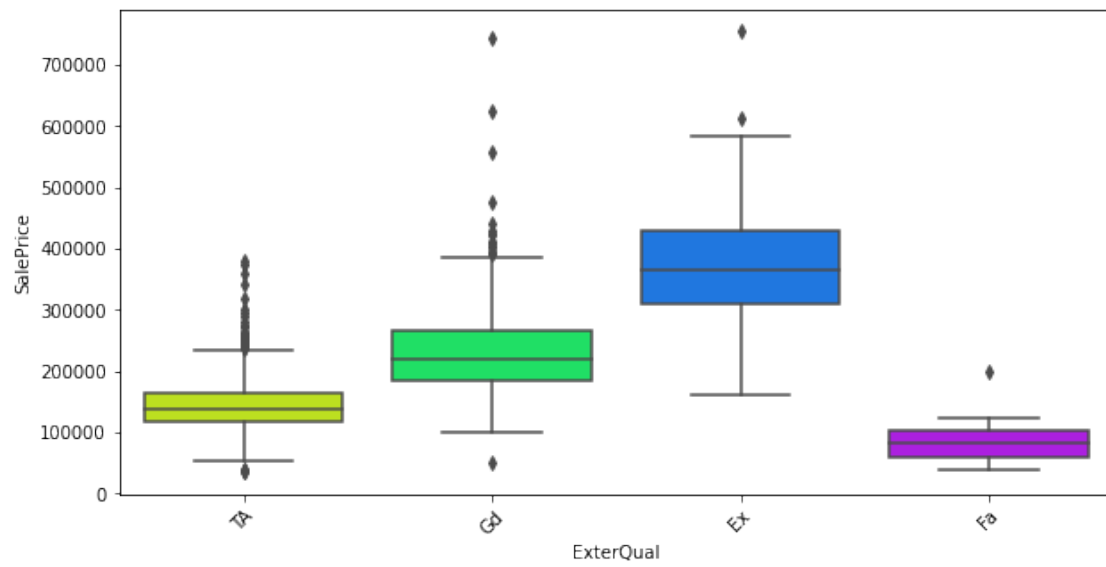
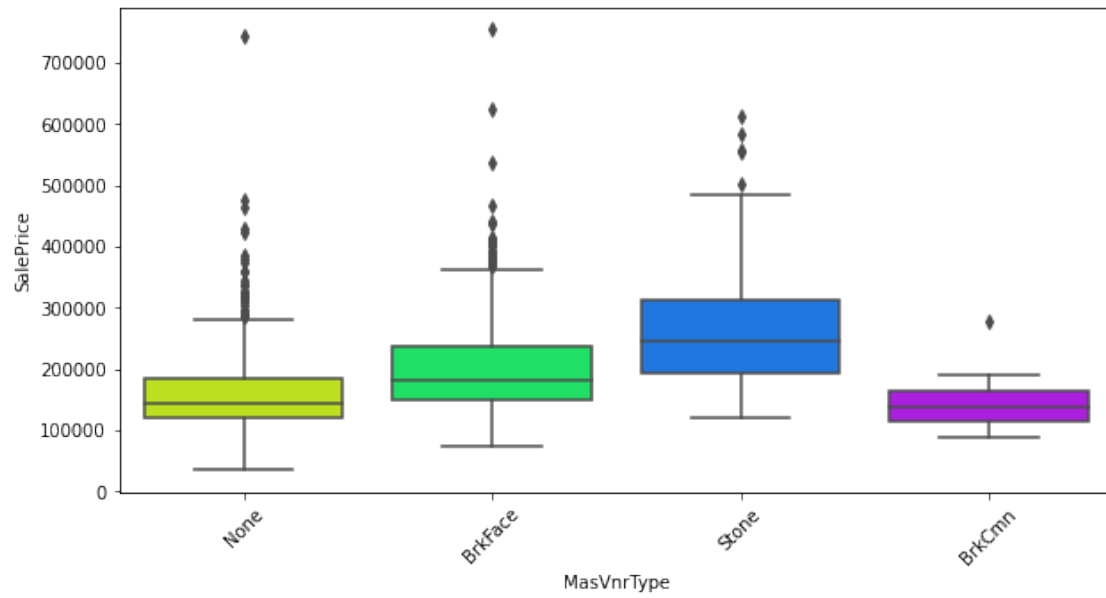


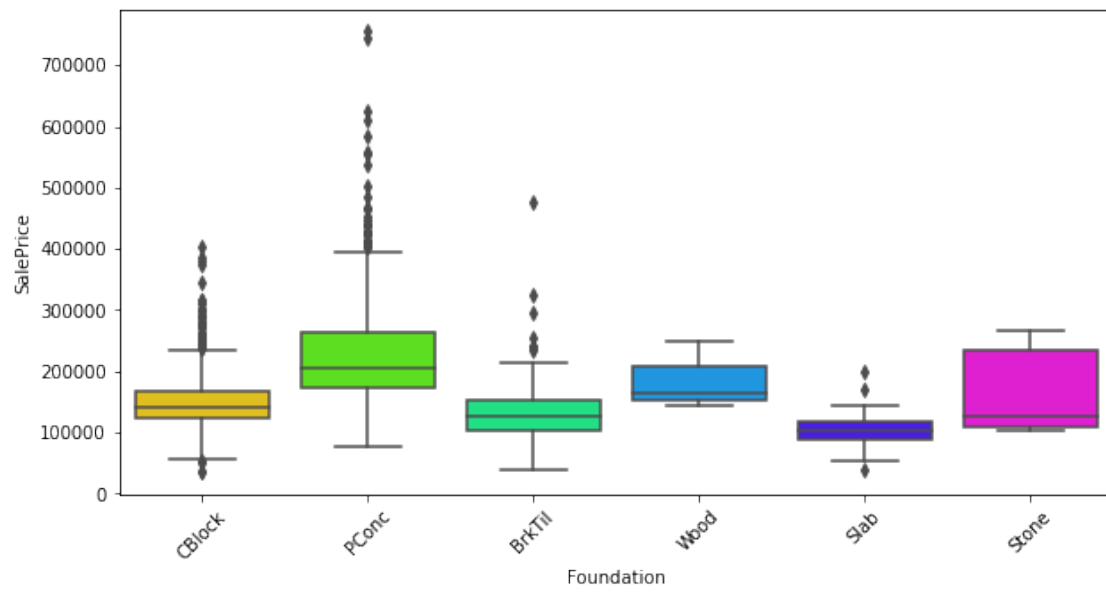
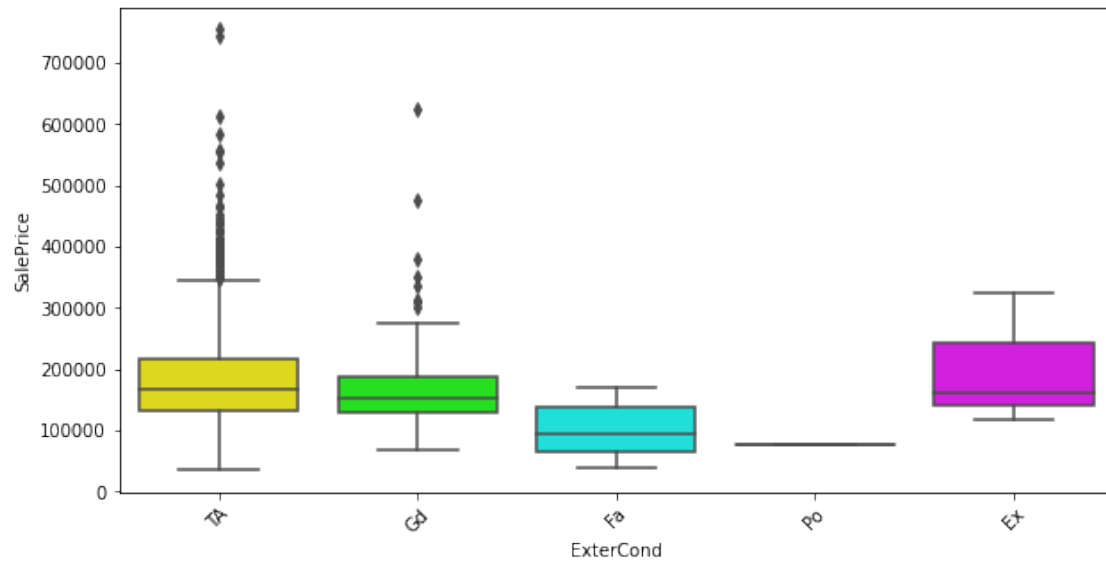


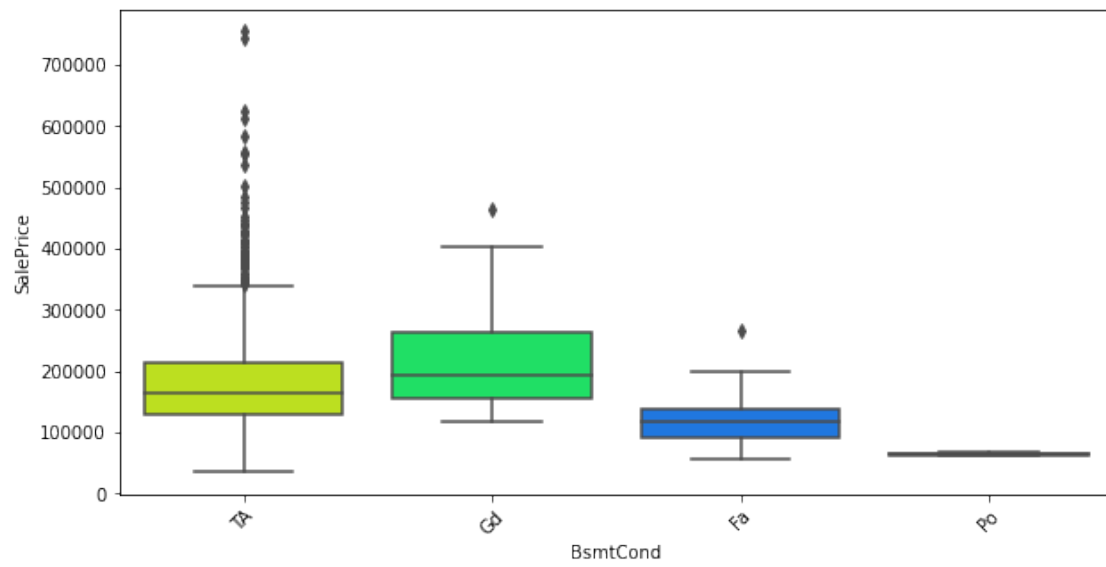
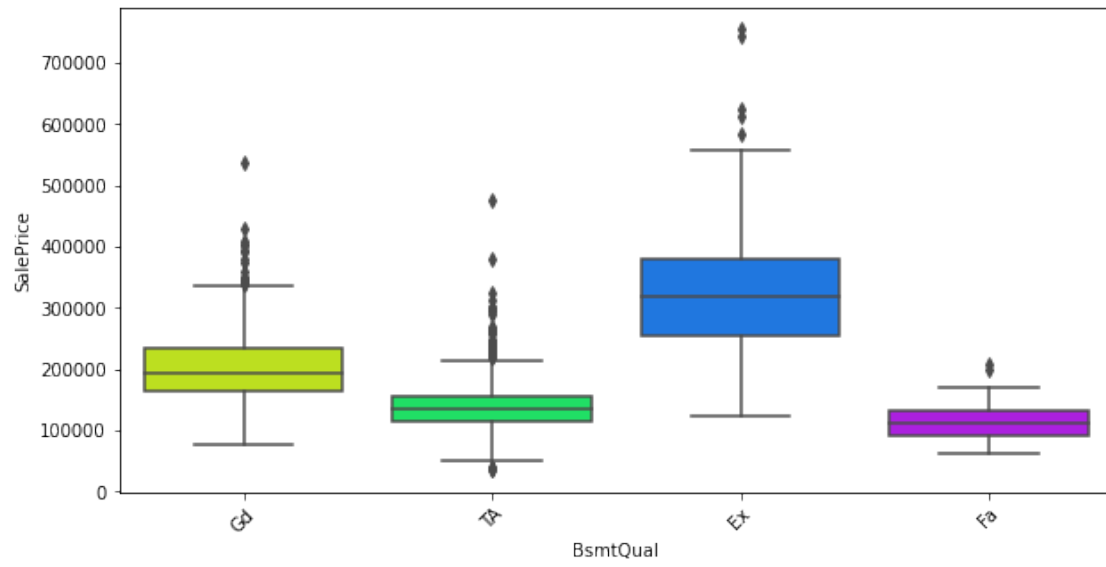


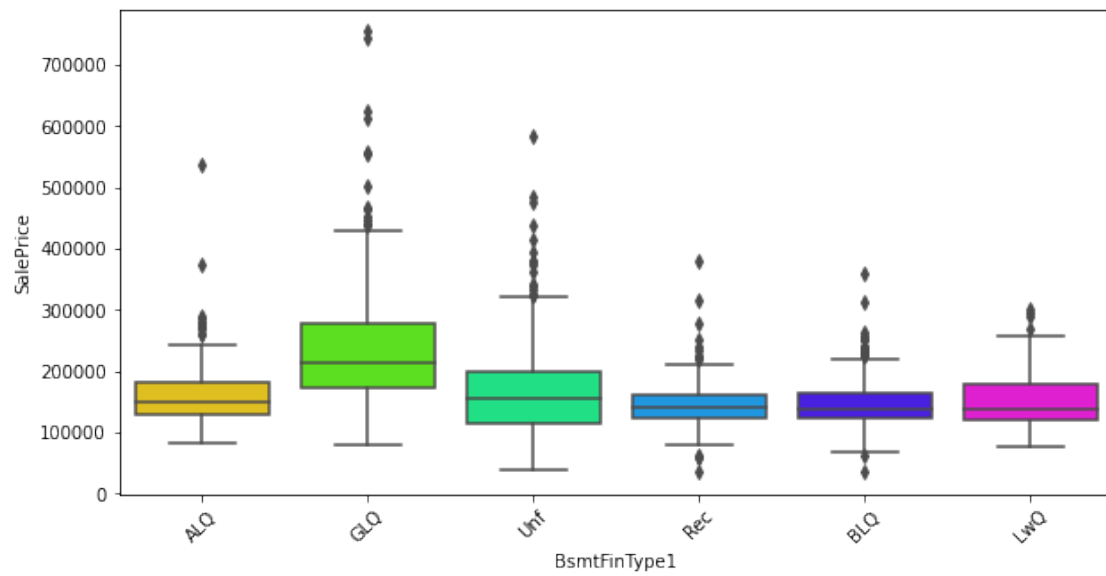
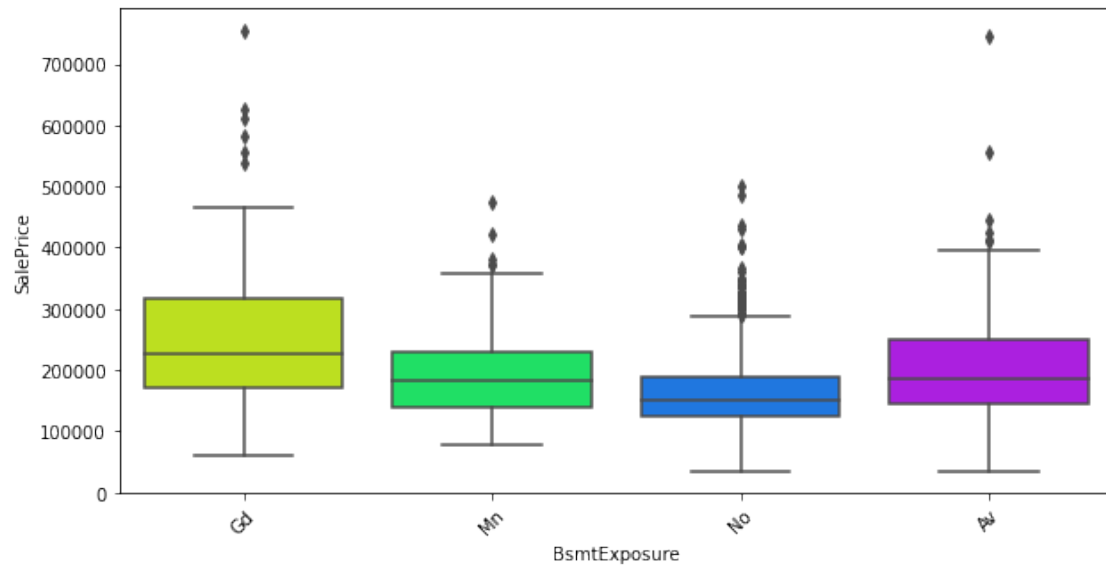


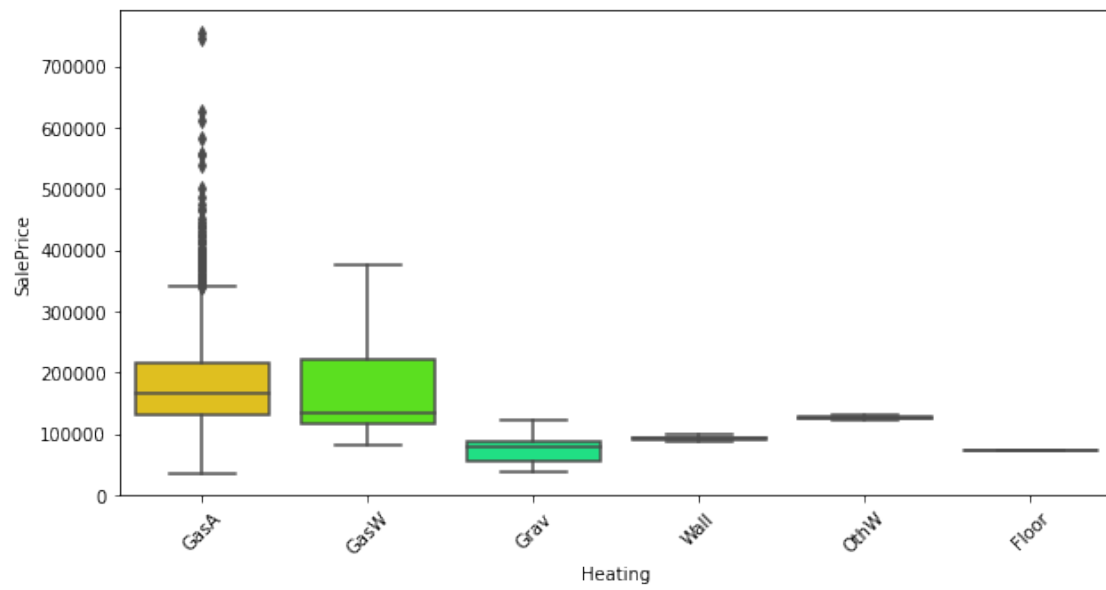
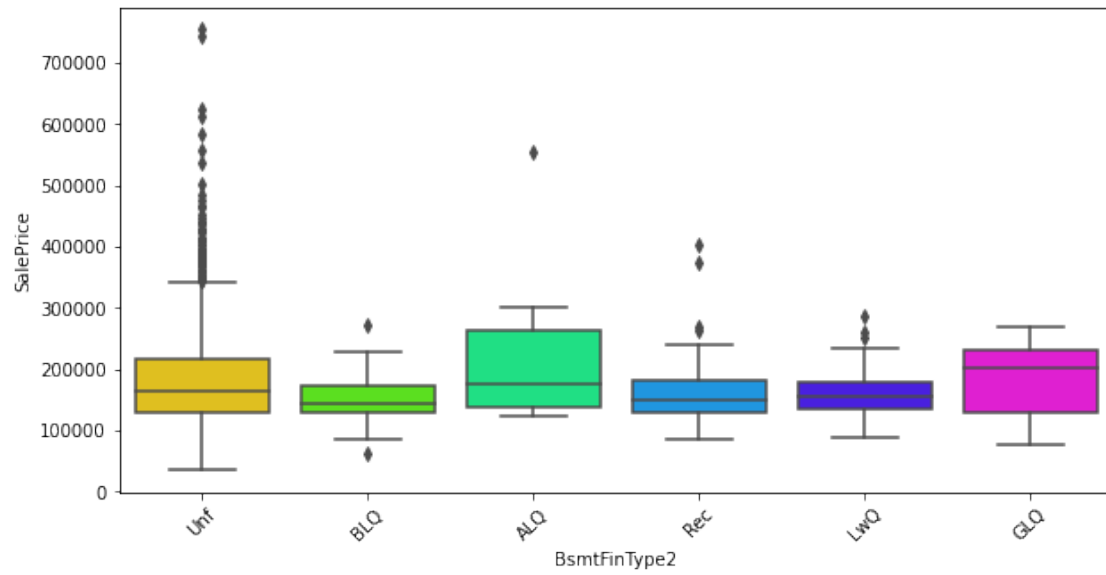


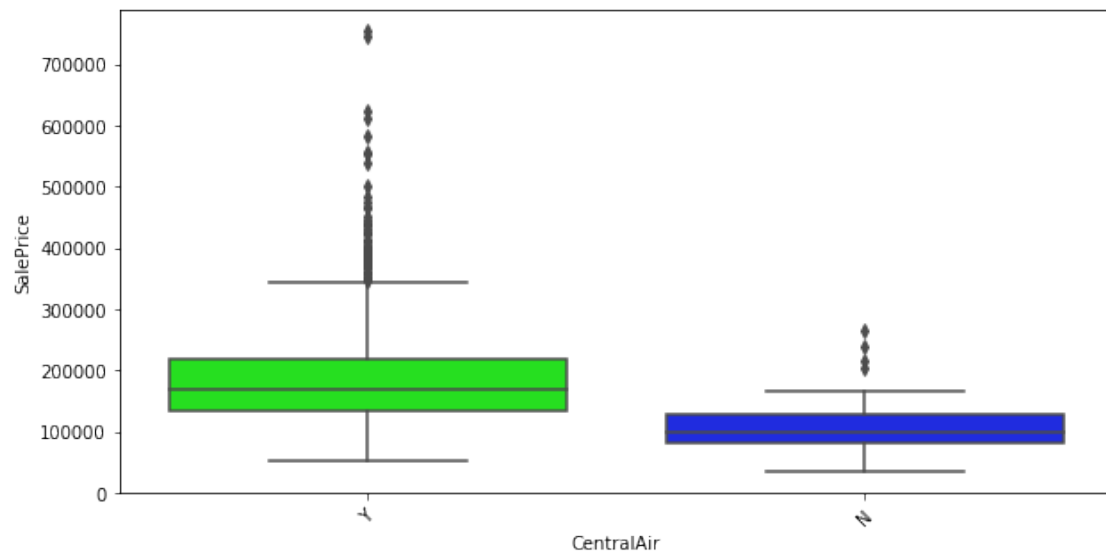
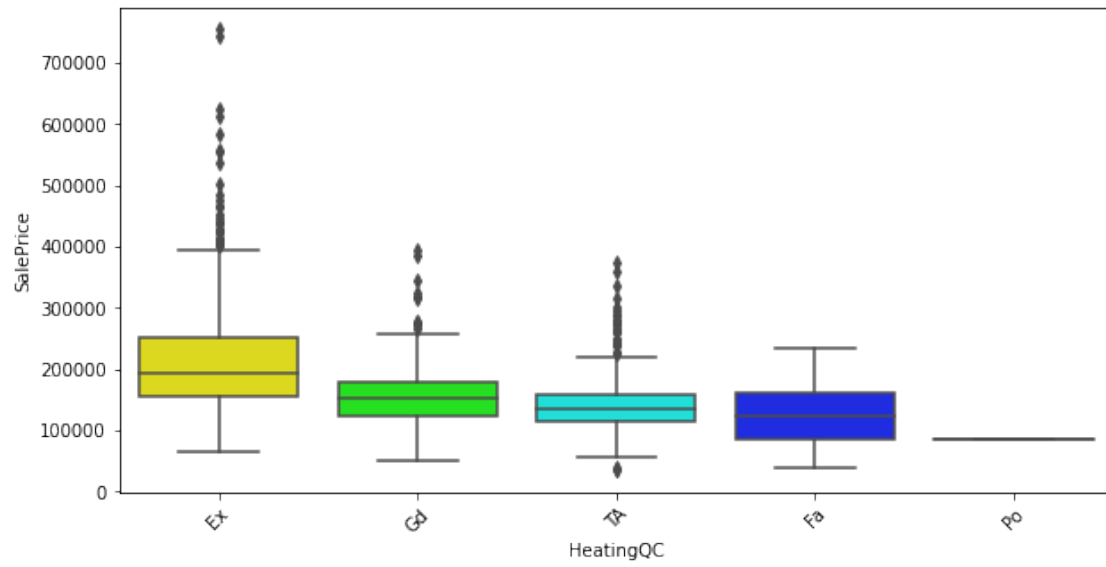


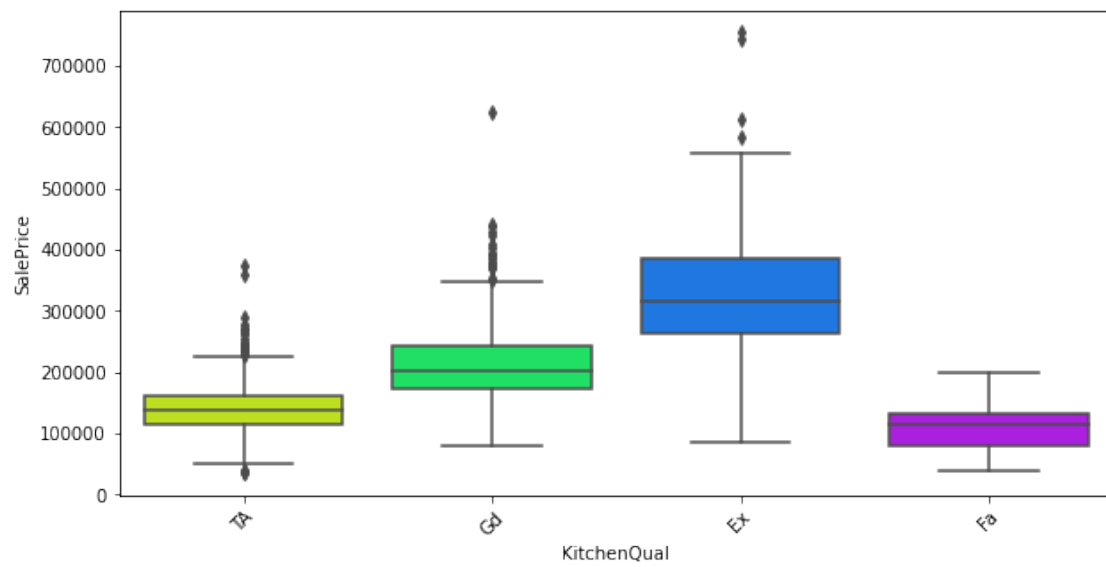
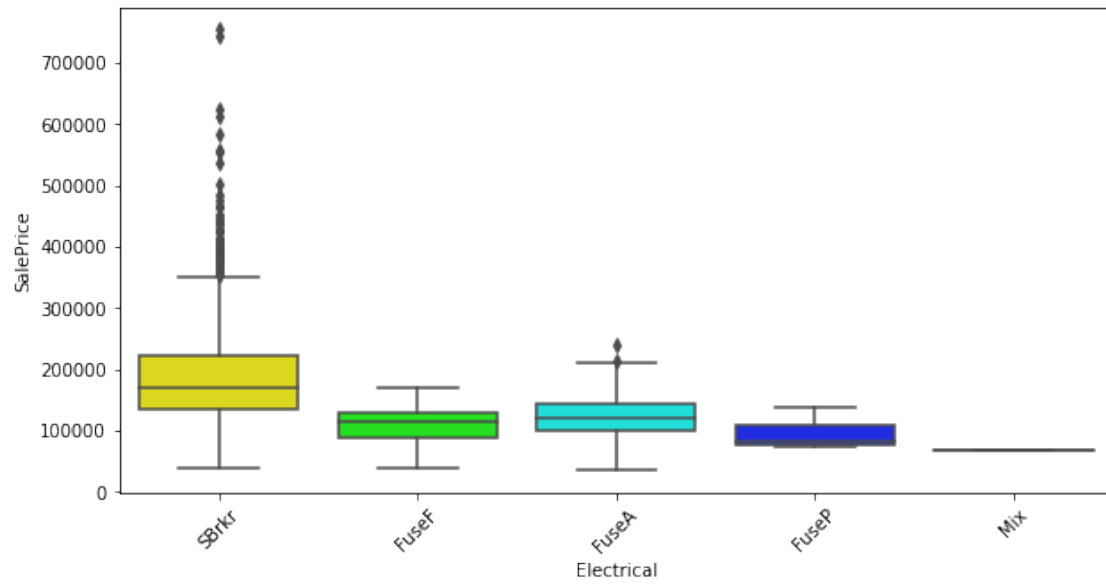


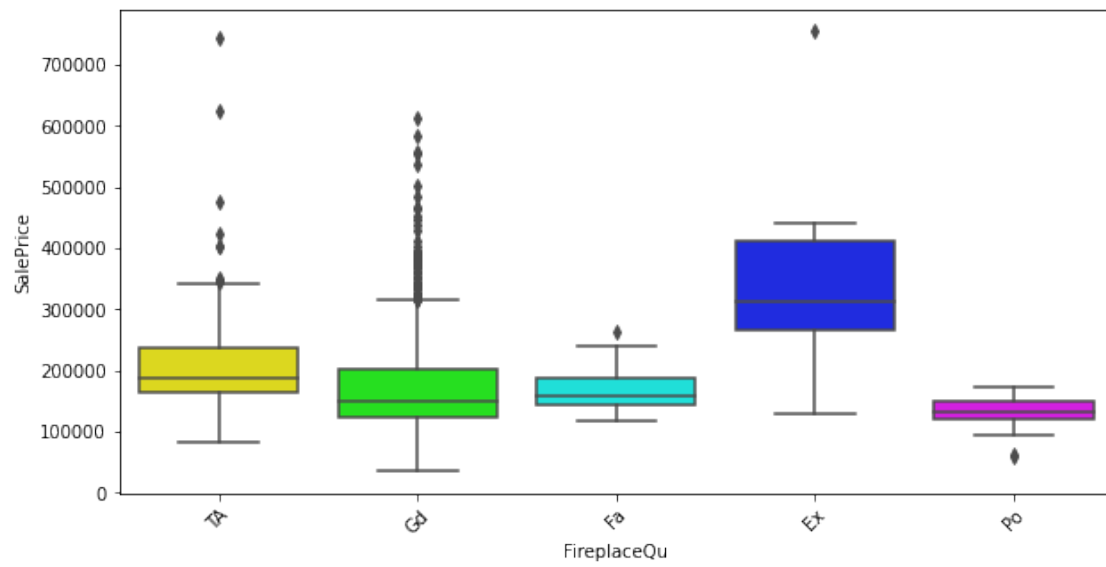
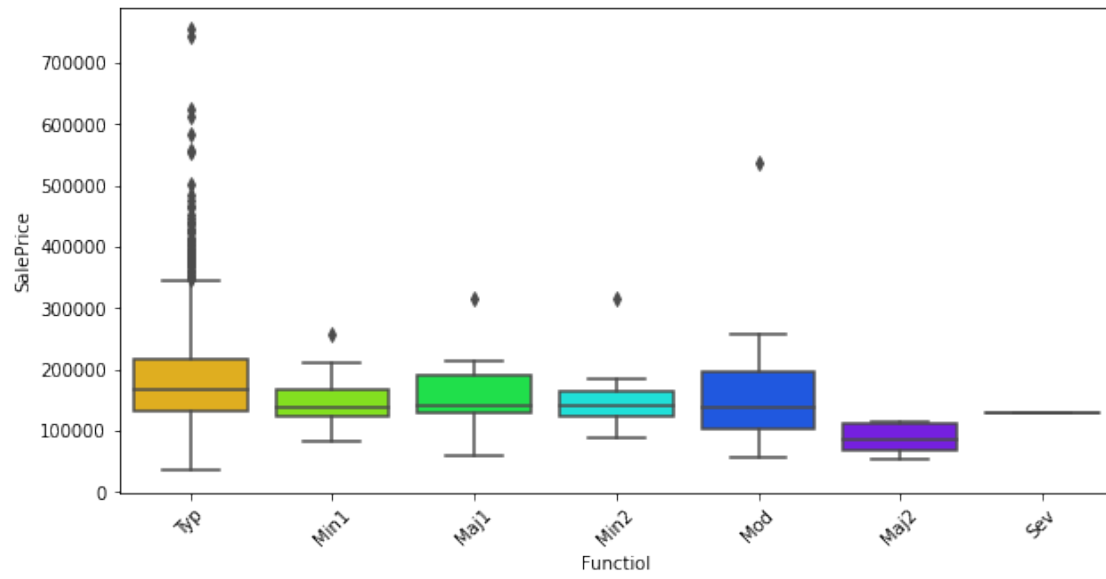


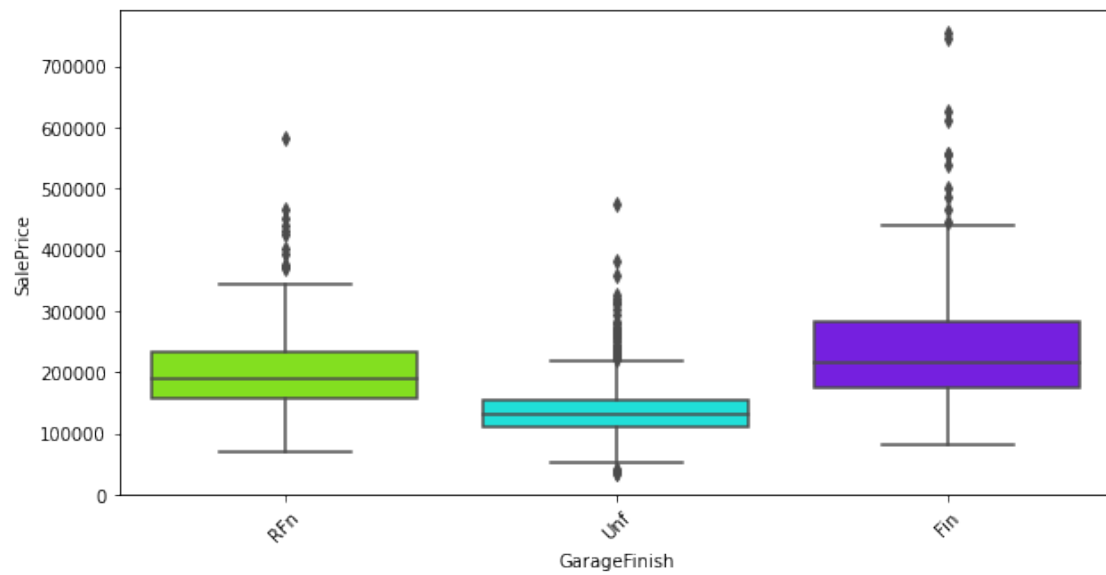
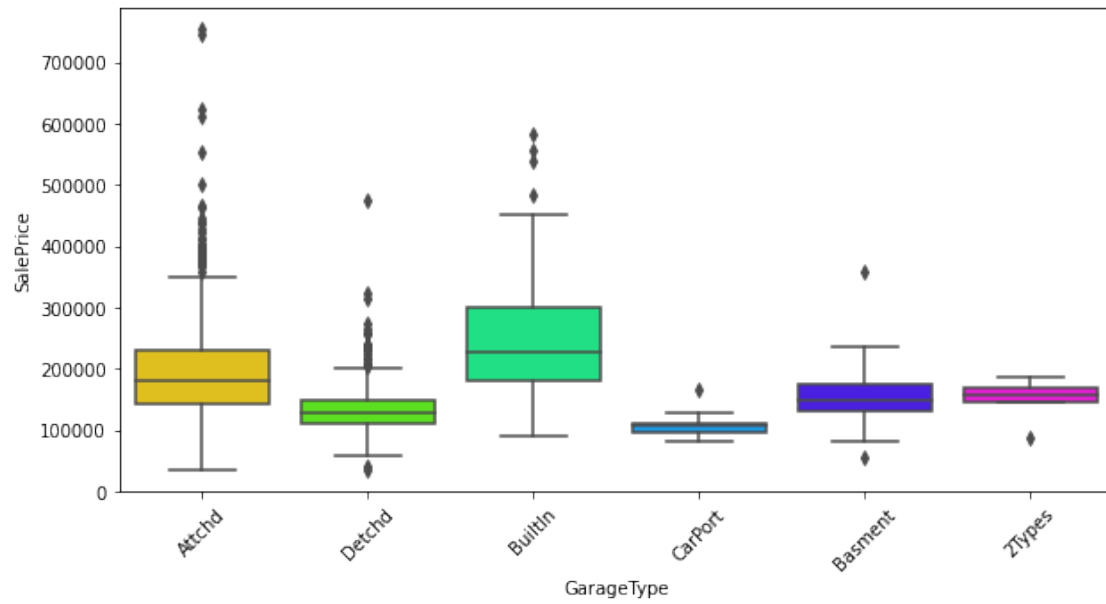


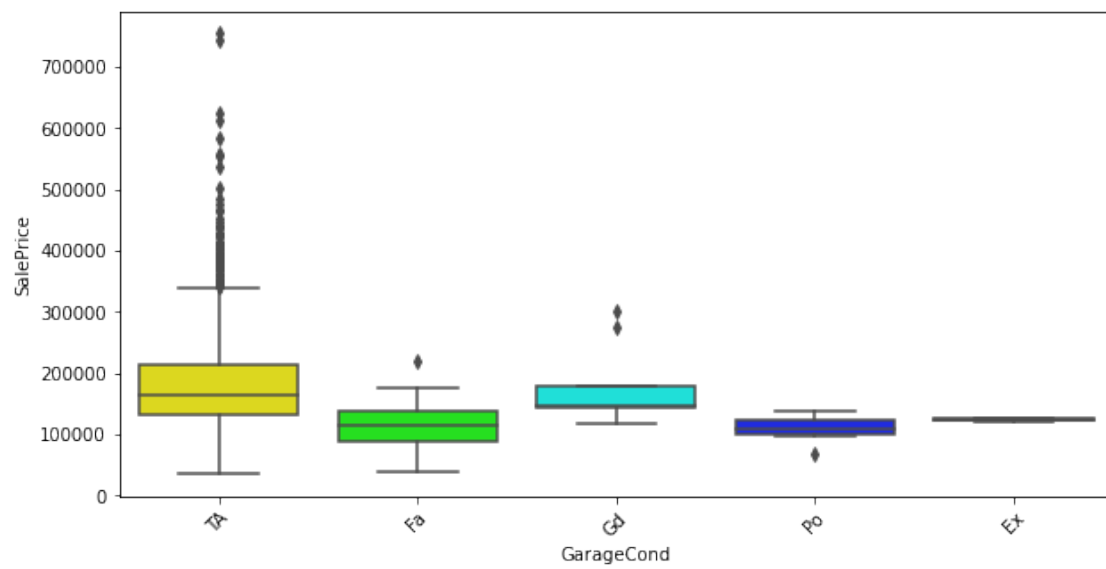
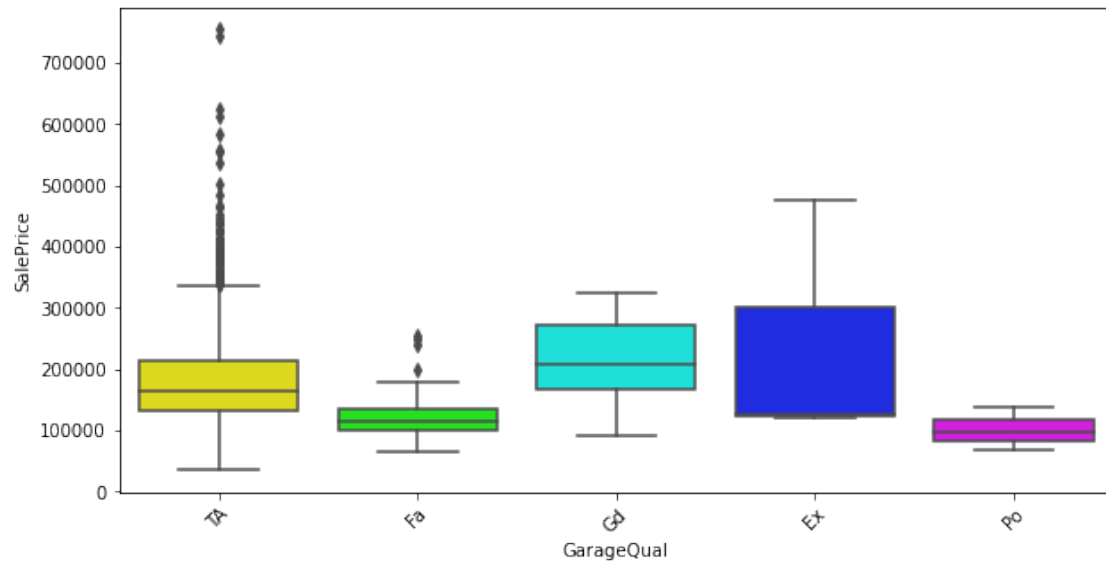


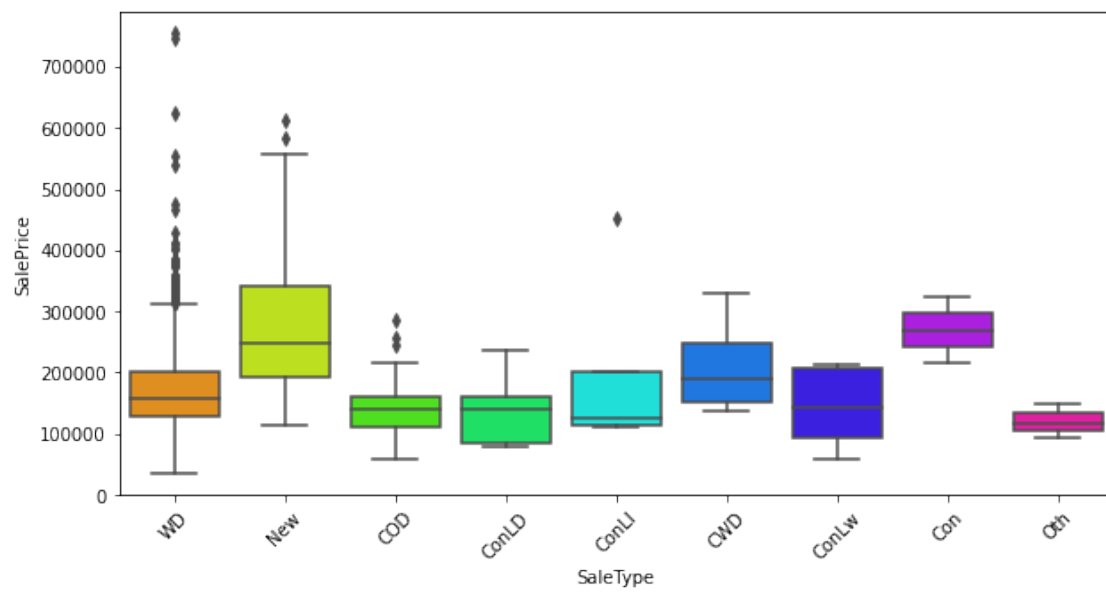
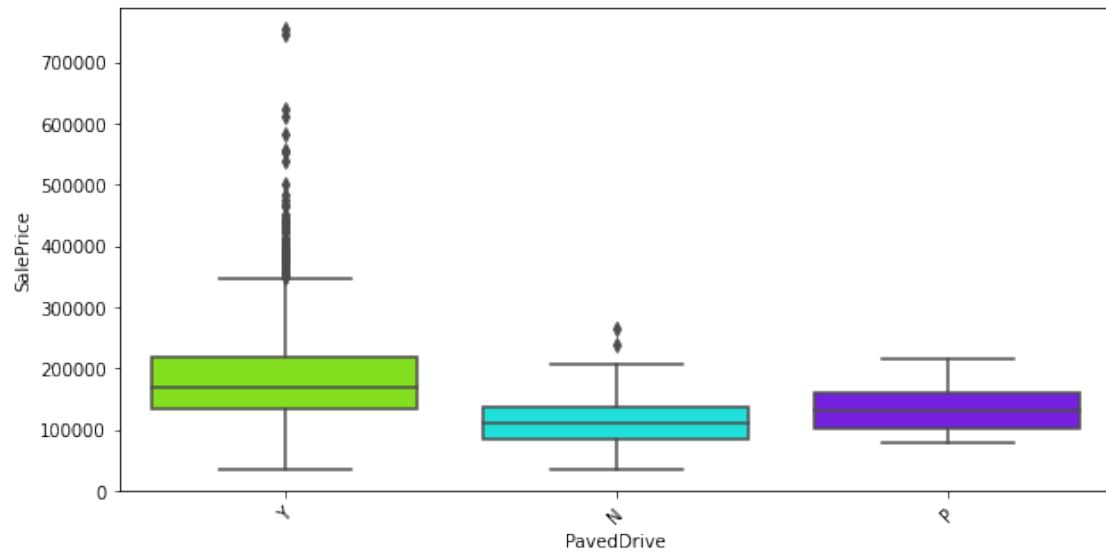


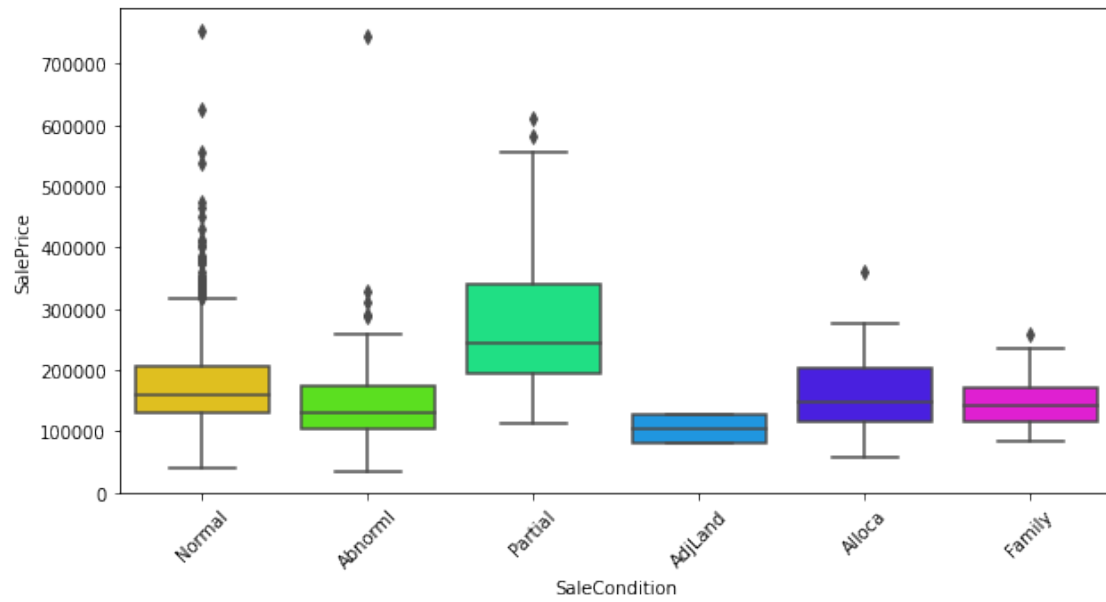












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