Missind values imputation using sklean

for Numeric and Categorical data

```
In [1]: import numpy as np
   import pandas as pd
   import matplotlib.pyplot as plt
   import seaborn as sns
   from sklearn.impute import SimpleImputer

In [2]: train=pd.read_csv("train.csv")
   test=pd.read_csv("test.csv")

   print('train dataset shape :-',train.shape)
   print('test dataset shape :-',test.shape)

   train dataset shape :- (1460, 81)
   test dataset shape :- (1459, 80)
```

```
train.head()
In [3]:
Out[3]:
            Id MSSubClass MSZoning LotFrontage LotArea Street Alley LotShape LandContour Utilities ... PoolArea PoolQC Fence Mis
          0 1
                       60
                                 RL
                                                                                           AllPub ...
                                           65.0
                                                   8450
                                                         Pave
                                                               NaN
                                                                         Reg
                                                                                      Lvl
                                                                                                           0
                                                                                                                NaN
                                                                                                                       NaN
          1 2
                                                         Pave
                        20
                                 RL
                                           80.0
                                                   9600
                                                               NaN
                                                                         Reg
                                                                                           AllPub ...
                                                                                                                NaN
                                                                                                                       NaN
                                                                                           AllPub ...
          2 3
                                 RL
                                                                         IR1
                       60
                                           68.0
                                                  11250
                                                         Pave
                                                               NaN
                                                                                                                NaN
                                                                                                                       NaN
          3 4
                                                                                           AllPub ...
                       70
                                 RL
                                           60.0
                                                   9550
                                                         Pave
                                                               NaN
                                                                         IR1
                                                                                      Lvl
                                                                                                           0
                                                                                                                NaN
                                                                                                                       NaN
                                                                                           AllPub ...
          4 5
                       60
                                 RL
                                                                         IR1
                                           84.0
                                                  14260
                                                         Pave
                                                               NaN
                                                                                      Lvl
                                                                                                           0
                                                                                                                NaN
                                                                                                                       NaN
         5 rows × 81 columns
In [4]: x train=train.drop(columns='SalePrice')
         y_train=train['SalePrice']
         print('train dataset shape :-',x train.shape)
         print('test dataset shape :-',y train.shape)
         train dataset shape :- (1460, 80)
         test dataset shape :- (1460,)
```

numerical values inputation

In [17]:	<pre>x_train[num_vars].isnull().sum()</pre>			
Out[17]:	Id	0		
	MSSubClass	0		
	LotFrontage	259		
	LotArea	0		
	OverallQual	0		
	OverallCond	0		
	YearBuilt	0		
	YearRemodAdd	0		
	MasVnrArea	8		
	BsmtFinSF1	0		
	BsmtFinSF2	0		
	BsmtUnfSF	0		
	TotalBsmtSF	0		
	1stFlrSF	0		
	2ndFlrSF	0		
	LowQualFinSF	0		
	GrLivArea	0		
	BsmtFullBath	0		
	BsmtHalfBath	0		
	FullBath	0		
	HalfBath	0		
	BedroomAbvGr	0		
	KitchenAbvGr	0		
	TotRmsAbvGrd	0		
	Fireplaces	0		
	GarageYrBlt	81		
	GarageCars	0		
	GarageArea	0		
	WoodDeckSF	0		
	OpenPorchSF	0		
	EnclosedPorch	0		
	3SsnPorch	0		
	ScreenPorch	0		
	PoolArea	0		
	MiscVal	0		
	MoSold	0		

```
dtype: int64
In [20]: imputer mean=SimpleImputer(strategy="mean")
         # imputer mean=SimpleImputer(strategy="constant", fill value=999....jo valuse tuze impute karni heoo values
        imputer mean.fit(x train[num vars])
In [22]:
Out[22]: SimpleImputer()
In [23]: imputer_mean.statistics_
Out[23]: array([7.30500000e+02, 5.68972603e+01, 7.00499584e+01, 1.05168281e+04,
                6.09931507e+00, 5.57534247e+00, 1.97126781e+03, 1.98486575e+03,
                1.03685262e+02, 4.43639726e+02, 4.65493151e+01, 5.67240411e+02,
                1.05742945e+03, 1.16262671e+03, 3.46992466e+02, 5.84452055e+00,
                1.51546370e+03, 4.25342466e-01, 5.75342466e-02, 1.56506849e+00,
                3.82876712e-01, 2.86643836e+00, 1.04657534e+00, 6.51780822e+00,
                6.13013699e-01, 1.97850616e+03, 1.76712329e+00, 4.72980137e+02,
                9.42445205e+01, 4.66602740e+01, 2.19541096e+01, 3.40958904e+00,
                1.50609589e+01, 2.75890411e+00, 4.34890411e+01, 6.32191781e+00,
```

YrSold

0

2.00781575e+031)

In [32]: |x_train[num_vars].isnull().sum() Out[32]: Id 0 MSSubClass 0 LotFrontage 0 0 LotArea OverallQual 0 OverallCond 0 YearBuilt 0 YearRemodAdd 0 0 MasVnrArea BsmtFinSF1 0 BsmtFinSF2 BsmtUnfSF 0 TotalBsmtSF 0 1stFlrSF 0 2ndFlrSF 0 0 LowQualFinSF GrLivArea 0 BsmtFullBath 0 BsmtHalfBath 0 FullBath 0 HalfBath 0 0 BedroomAbvGr 0 KitchenAbvGr TotRmsAbvGrd 0 Fireplaces 0 GarageYrBlt 0 GarageCars 0 GarageArea 0 WoodDeckSF 0 OpenPorchSF 0 EnclosedPorch 0 3SsnPorch 0 ScreenPorch 0 PoolArea 0 MiscVal 0 MoSold 0

YrSold 0

dtype: int64

In [34]: test[num_vars].isnull().sum() Out[34]: Id 0 MSSubClass 0 LotFrontage 0 0 LotArea OverallQual 0 OverallCond 0 YearBuilt 0 YearRemodAdd 0 0 MasVnrArea BsmtFinSF1 0 BsmtFinSF2 BsmtUnfSF 0 TotalBsmtSF 0 1stFlrSF 0 2ndFlrSF 0 0 LowQualFinSF GrLivArea 0 BsmtFullBath 0 BsmtHalfBath 0 FullBath 0 HalfBath 0 0 BedroomAbvGr 0 KitchenAbvGr TotRmsAbvGrd 0 Fireplaces 0 GarageYrBlt 0 0 GarageCars GarageArea 0 WoodDeckSF 0 OpenPorchSF 0 EnclosedPorch 0 3SsnPorch 0 ScreenPorch 0 PoolArea 0 MiscVal 0 MoSold 0

YrSold 0 dtype: int64

categorical missing values imputation valuee

In [38]:	<pre>x_train[cat_vars].isnull().sum()</pre>			
Out[38]:	MSZoning	0		
	Street	0		
	Alley	1369		
	LotShape	0		
	LandContour	0		
	Utilities	0		
	LotConfig	0		
	LandSlope	0		
	Neighborhood	0		
	Condition1	0		
	Condition2	0		
	BldgType	0		
	HouseStyle	0		
	RoofStyle	0		
	RoofMatl	0		
	Exterior1st	0		
	Exterior2nd	0		
	MasVnrType	8		
	ExterQual	0		
	ExterCond	0		
	Foundation	0		
	BsmtQual	37		
	BsmtCond	37		
	BsmtExposure	38		
	BsmtFinType1	37		
	BsmtFinType2	38		
	Heating	0		
	HeatingQC	0		
	CentralAir	0		
	Electrical	1		
	KitchenQual	0		
	Functional	0		
	FireplaceQu	690		
	GarageType	81		
	GarageFinish	81		
	GarageQual	81		

```
PavedDrive
         PoolQC
                          1453
         Fence
                          1179
         MiscFeature
                          1406
         SaleType
         SaleCondition
         dtype: int64
In [62]: imputer mode=SimpleImputer(strategy="most frequent") #... most frequent =mode
In [64]: imputer mode.fit(x train[cat vars])
Out[64]: SimpleImputer(strategy='most frequent')
In [69]: imputer mode.statistics
Out[69]: array(['RL', 'Pave', 'Grvl', 'Reg', 'Lvl', 'AllPub', 'Inside', 'Gtl',
                 'NAmes', 'Norm', 'Norm', '1Fam', '1Story', 'Gable', 'CompShg',
                 'VinylSd', 'VinylSd', 'None', 'TA', 'TA', 'PConc', 'TA', 'TA',
                 'No', 'Unf', 'Unf', 'GasA', 'Ex', 'Y', 'SBrkr', 'TA', 'Typ', 'Gd',
                 'Attchd', 'Unf', 'TA', 'TA', 'Y', 'Gd', 'MnPrv', 'Shed', 'WD',
                 'Normal'], dtype=object)
In [70]: imputer mode.transform(x train[cat vars])
Out[70]: array([['RL', 'Pave', 'Grvl', ..., 'Shed', 'WD', 'Normal'],
                ['RL', 'Pave', 'Grvl', ..., 'Shed', 'WD', 'Normal']], dtype=object)
```

GarageCond

81

In [50]: x_train[cat_vars]=imputer_mode.transform(x_train[cat_vars])
 test[cat_vars]=imputer_mode.transform(test[cat_vars])

In [51]:	<pre>x_train[cat_vars].isnull().sum</pre>		
Out[51]:	MSZoning	0	
	Street	0	
	Alley	0	
	LotShape	0	
	LandContour	0	
	Utilities	0	
	LotConfig	0	
	LandSlope	0	
	Neighborhood	0	
	Condition1	0	
	Condition2	0	
	BldgType	0	
	HouseStyle	0	
	RoofStyle	0	
	RoofMatl	0	
	Exterior1st	0	
	Exterior2nd	0	
	MasVnrType	0	
	ExterQual	0	
	ExterCond	0	
	Foundation	0	
	BsmtQual	0	
	BsmtCond	0	
	BsmtExposure	0	
	BsmtFinType1	0	
	BsmtFinType2	0	
	Heating	0	
	HeatingQC	0	
	CentralAir	0	
	Electrical	0	
	KitchenQual	0	
	Functional	0	
	FireplaceQu	0	
	GarageType	0	
	GarageFinish	0	
	GarageQual	0	

GarageCond	0
PavedDrive	0
PoolQC	0
Fence	0
MiscFeature	0
SaleType	0
SaleCondition	0
dtype: int64	

In [52]: test[cat_vars].isnull().sum() Out[52]: MSZoning 0 Street 0 Alley 0 LotShape LandContour Utilities 0 LotConfig 0 LandSlope 0 Neighborhood 0 Condition1 0 Condition2 BldgType 0 0 HouseStyle RoofStyle 0 RoofMat1 0 Exterior1st 0 Exterior2nd 0 MasVnrType 0 ExterQual 0 ExterCond Foundation 0 **BsmtQual** 0 **BsmtCond** 0 0 BsmtExposure BsmtFinType1 0 BsmtFinType2 0 0 Heating HeatingQC 0 CentralAir 0 Electrical 0 KitchenQual 0 Functional 0 FireplaceQu 0 GarageType 0 GarageFinish 0 GarageQual 0

```
GarageCond 0
PavedDrive 0
PoolQC 0
Fence 0
MiscFeature SaleType 0
SaleCondition 0
dtype: int64
```

```
In [53]: x_train[cat_vars].isnull().sum().sum()
Out[53]: 0
In [ ]:
```

localhost:8888/notebooks/mL imp/data cleaning/Missind values imputation using sklean .ipynb