## **Data Cleaning**

## Missing value imputation by Mean, Median

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



data\_df=r"https://drive.google.com/open?id=1BiGZSedP4BIIuTbVTBodOhVgFImaz08"

In [2]: data=pd.read\_csv("train.csv")
 data

Out[2]:

	ld	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	LandContour	Utilities	 PoolArea	PoolQC	Fenc
0	1	60	RL	65.0	8450	Pave	NaN	Reg	LvI	AllPub	 0	NaN	Nal
1	2	20	RL	80.0	9600	Pave	NaN	Reg	LvI	AllPub	 0	NaN	Nal
2	3	60	RL	68.0	11250	Pave	NaN	IR1	LvI	AllPub	 0	NaN	Nal
3	4	70	RL	60.0	9550	Pave	NaN	IR1	LvI	AllPub	 0	NaN	Nal
4	5	60	RL	84.0	14260	Pave	NaN	IR1	LvI	AllPub	 0	NaN	Nal
1455	1456	60	RL	62.0	7917	Pave	NaN	Reg	LvI	AllPub	 0	NaN	Nal
1456	1457	20	RL	85.0	13175	Pave	NaN	Reg	LvI	AllPub	 0	NaN	MnPr
1457	1458	70	RL	66.0	9042	Pave	NaN	Reg	LvI	AllPub	 0	NaN	GdPr
1458	1459	20	RL	68.0	9717	Pave	NaN	Reg	LvI	AllPub	 0	NaN	Nal
1459	1460	20	RL	75.0	9937	Pave	NaN	Reg	LvI	AllPub	 0	NaN	Nal

1460 rows × 81 columns

In [3]: data.shape

Out[3]: (1460, 81)

In [4]: pd.set\_option("display.max\_columns",None)
pd.set\_option("display.max\_rows",None)

## In [5]: data.info()

RangeIndex: 1460 entries, 0 to 1459 Data columns (total 81 columns): Column Non-Null Count # Dtype 0 Id1460 non-null int64 1 MSSubClass 1460 non-null int64 1460 non-null 2 MSZoning object 1201 non-null LotFrontage 3 float64 4 LotArea 1460 non-null int64 object 5 Street 1460 non-null Alley 91 non-null 6 object LotShape object 7 1460 non-null LandContour 1460 non-null object Utilities 9 1460 non-null object LotConfig 1460 non-null object 10 LandSlope 1460 non-null object 11 Neighborhood 1460 non-null object 12 13 Condition1 1460 non-null object 1460 non-null Condition2 object 14 BldgType 15 1460 non-null object 16 HouseStyle 1460 non-null object OverallQual 1460 non-null 17 int64 OverallCond 1460 non-null 18 int64 1460 non-null 19 YearBuilt int64 YearRemodAdd 1460 non-null int64 20 RoofStyle 1460 non-null object 21 RoofMat1 object 22 1460 non-null Exterior1st 1460 non-null object Exterior2nd 1460 non-null object 24 1452 non-null 25 MasVnrType object MasVnrArea 1452 non-null float64 ExterQual 1460 non-null object 27 1460 non-null object 28 ExterCond 29 Foundation 1460 non-null object **BsmtQual** 1423 non-null object

<class 'pandas.core.frame.DataFrame'>

	· ·	. ,
BsmtCond	1423 non-null	object
BsmtExposure	1422 non-null	object
BsmtFinType1	1423 non-null	object
BsmtFinSF1	1460 non-null	int64
BsmtFinType2	1422 non-null	object
BsmtFinSF2	1460 non-null	int64
BsmtUnfSF	1460 non-null	int64
TotalBsmtSF	1460 non-null	int64
Heating	1460 non-null	object
HeatingQC	1460 non-null	object
CentralAir	1460 non-null	object
Electrical	1459 non-null	object
1stFlrSF	1460 non-null	int64
2ndFlrSF	1460 non-null	int64
LowQualFinSF	1460 non-null	int64
GrLivArea	1460 non-null	int64
BsmtFullBath	1460 non-null	int64
BsmtHalfBath	1460 non-null	int64
FullBath	1460 non-null	int64
	1460 non-null	int64
BedroomAbvGr	1460 non-null	int64
KitchenAbvGr		int64
KitchenQual		object
TotRmsAbvGrd	1460 non-null	int64
Functional	1460 non-null	object
Fireplaces	1460 non-null	int64
FireplaceQu	770 non-null	object
GarageType	1379 non-null	object
•	1379 non-null	float64
GarageFinish	1379 non-null	object
GarageCars	1460 non-null	int64
GarageArea	1460 non-null	int64
GarageQual	1379 non-null	object
GarageCond	1379 non-null	object
	1460 non-null	object
WoodDeckSF	1460 non-null	int64
OpenPorchSF	1460 non-null	int64
EnclosedPorch	1460 non-null	int64
3SsnPorch	1460 non-null	int64
	BsmtExposure BsmtFinType1 BsmtFinSF1 BsmtFinSF2 BsmtUnfSF TotalBsmtSF Heating HeatingQC CentralAir Electrical 1stFlrSF 2ndFlrSF LowQualFinSF GrLivArea BsmtFullBath BsmtHalfBath FullBath HalfBath HalfBath BedroomAbvGr KitchenQual TotRmsAbvGrd Fireplaces Fireplaces FireplaceQu GarageType GarageYrBlt GarageFinish GarageCars GarageArea GarageQual GarageCond PavedDrive WoodDeckSF OpenPorchSF	BsmtFinType1 1422 non-null BsmtFinSF1 1460 non-null BsmtFinSF2 1460 non-null BsmtFinSF2 1460 non-null BsmtUnfSF 1460 non-null TotalBsmtSF 1460 non-null Heating 1460 non-null Heating 1460 non-null Electrical 1459 non-null StFlrSF 1460 non-null LowQualFinSF 1460 non-null ContralAir 1460 non-null Electrical 1459 non-null BsmtFulBath 1460 non-null BsmtFulBath 1460 non-null BsmtHalfBath 1460 non-null BsmtHalfBath 1460 non-null FulBath 1460 non-null KitchenAbvGr 1460 non-null KitchenQual 1460 non-null TotRmsAbvGrd 1460 non-null Fireplaces 1460 non-null Fireplaces 1460 non-null FireplaceQu 770 non-null GarageType 1379 non-null GarageType 1379 non-null GarageFinish 1379 non-null GarageCond 1379 non-null GarageCond 1379 non-null GarageCond 1379 non-null PavedDrive 1460 non-null DodDeckSF 1460 non-null NoodDeckSF 1460 non-null HoodDeckSF 1460 non-null

70	ScreenPorch	1460 non-null	int64
71	PoolArea	1460 non-null	int64
72	PoolQC	7 non-null	object
73	Fence	281 non-null	object
74	MiscFeature	54 non-null	object
75	MiscVal	1460 non-null	int64
76	MoSold	1460 non-null	int64
77	YrSold	1460 non-null	int64
78	SaleType	1460 non-null	object
79	SaleCondition	1460 non-null	object
80	SalePrice	1460 non-null	int64
dtype	es: float64(3),	int64(35), object	ct(43)

memory usage: 678.7+ KB

In [6]:	data.isnull().	sum()
Out[6]:	Id	0
	MSSubClass	0
	MSZoning	0
	LotFrontage	259
	LotArea	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
	OverallQual	0
	OverallCond	0
	YearBuilt	0
	YearRemodAdd	0
	RoofStyle	0
	RoofMatl	0
	Exterior1st	0
	Exterior2nd	0
	MasVnrType	8
	MasVnrArea	8 0
	ExterQual ExterCond	0
	Foundation	0
		37
	BsmtQual BsmtCond	37 37
	BsmtExposure	38
	BsmtFinType1	36 37
	BsmtFinSF1	0
	BsmtFinType2	38
	Pamer Till Aher	טכ

BsmtFinSF2	0
BsmtUnfSF	0
TotalBsmtSF	0
Heating	0
HeatingQC	0
CentralAir	0
Electrical	1
1stFlrSF	0
2ndFlrSF	0
LowQualFinSF	0
GrLivArea	0
BsmtFullBath	0
BsmtHalfBath	0
FullBath	0
HalfBath	0
BedroomAbvGr	0
KitchenAbvGr	0
KitchenQual	0
TotRmsAbvGrd	0
Functional	0
Fireplaces	0
FireplaceQu	690
GarageType	81
GarageYrBlt	81
GarageFinish	81
GarageCars	0
GarageArea	0
GarageQual	81
GarageCond	81
PavedDrive	0
WoodDeckSF	0
OpenPorchSF	0
EnclosedPorch	0
3SsnPorch	0
ScreenPorch	0
PoolArea	0
Poo1QC	1453
Fence	1179
MiscFeature	1406

MiscVal 0
MoSold 0
YrSold 0
SaleType 0
SaleCondition 0
SalePrice 0
dtype: int64

In [7]: data.isnull().sum().sum()

Out[7]: 6965

In [8]: data.head()

Out[8]:

	ld	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	LandContour	Utilities	LotConfig	LandSlope	Neighborho
0	1	60	RL	65.0	8450	Pave	NaN	Reg	LvI	AllPub	Inside	Gtl	Coll
1	2	20	RL	80.0	9600	Pave	NaN	Reg	LvI	AllPub	FR2	Gtl	Veer
2	3	60	RL	68.0	11250	Pave	NaN	IR1	Lvl	AllPub	Inside	Gtl	Coll
3	4	70	RL	60.0	9550	Pave	NaN	IR1	Lvl	AllPub	Corner	Gtl	Crav
4	5	60	RL	84.0	14260	Pave	NaN	IR1	LvI	AllPub	FR2	Gtl	NoRi

```
In [9]: | null_var=data.isnull().sum()/data.shape[0]*100
null_var
```

Out[9]:	Id	0.000000
	MSSubClass	0.000000
	MSZoning	0.000000
	LotFrontage	17.739726
	LotArea	0.000000
	Street	0.000000
	Alley	93.767123
	LotShape	0.000000
	LandContour	0.000000
	Utilities	0.000000
	LotConfig	0.000000
	LandSlope	0.000000
	Neighborhood	0.000000
	Condition1	0.000000
	Condition2	0.000000
	BldgType	0.000000
	HouseStyle	0.000000
	OverallQual	0.000000
	OverallCond	0.000000
	YearBuilt	0.000000
	YearRemodAdd	0.000000
	RoofStyle	0.000000
	RoofMatl	0.000000
	Exterior1st	0.000000
	Exterior2nd	0.000000
	MasVnrType	0.547945
	MasVnrArea	0.547945
	ExterQual	0.000000
	ExterCond	0.000000
	Foundation	0.000000
	BsmtQual	2.534247
	BsmtCond	2.534247
	BsmtExposure	2.602740
	BsmtFinType1	2.534247
	BsmtFinSF1	0.000000

BsmtFinType2	2.602740
BsmtFinSF2	0.000000
BsmtUnfSF	0.000000
TotalBsmtSF	0.000000
Heating	0.000000
HeatingQC	0.000000
CentralAir	0.000000
Electrical	0.068493
1stFlrSF	0.000000
2ndF1rSF	0.000000
LowQualFinSF	0.000000
GrLivArea	0.000000
BsmtFullBath	0.000000
BsmtHalfBath	0.000000
FullBath	0.000000
HalfBath	0.000000
BedroomAbvGr	0.000000
KitchenAbvGr	0.000000
KitchenQual	0.000000
TotRmsAbvGrd	0.000000
Functional	0.000000
Fireplaces	0.000000
FireplaceQu	47.260274
GarageType	5.547945
GarageYrBlt	5.547945
GarageFinish	5.547945
GarageCars	0.000000
GarageArea	0.000000
GarageQual	5.547945
GarageCond	5.547945
PavedDrive	0.000000
WoodDeckSF	0.000000
OpenPorchSF	0.000000
EnclosedPorch	0.000000
3SsnPorch	0.000000
ScreenPorch	0.000000
PoolArea	0.000000
PoolQC	99.520548
Fence	80.753425

```
MiscFeature
                          96.301370
         MiscVal
                           0.000000
         MoSold
                           0.000000
         YrSold
                           0.000000
         SaleType
                           0.000000
         SaleCondition
                           0.000000
         SalePrice
                           0.000000
         dtype: float64
In [10]: drop cloumn = null var[null var >20].keys()
         drop cloumn
Out[10]: Index(['Alley', 'FireplaceQu', 'PoolQC', 'Fence', 'MiscFeature'], dtype='object')
In [11]:
         data2=data.drop(columns=drop cloumn)
In [12]:
        data2.shape
Out[12]: (1460, 76)
In [13]: data3_num=data2.select_dtypes(include=["int64","float64"])
```

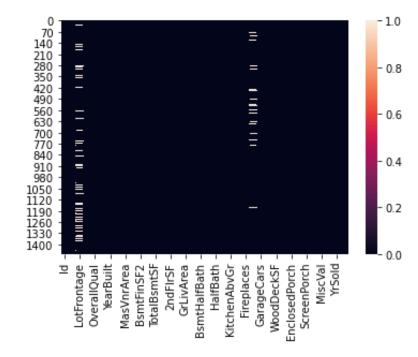
In [14]: data3\_num.head()

Out[14]:

	ld	MSSubClass	LotFrontage	LotArea	OverallQual	OverallCond	YearBuilt	YearRemodAdd	MasVnrArea	BsmtFinSF1	BsmtFinSF2
0	1	60	65.0	8450	7	5	2003	2003	196.0	706	0
1	2	20	80.0	9600	6	8	1976	1976	0.0	978	0
2	3	60	68.0	11250	7	5	2001	2002	162.0	486	0
3	4	70	60.0	9550	7	5	1915	1970	0.0	216	0
4	5	60	84.0	14260	8	5	2000	2000	350.0	655	0

In [15]: sns.heatmap(data3\_num.isnull())

## Out[15]: <AxesSubplot:>



In [17]: data3\_num.isnull().sum()/data.shape[0]\*100

Out[17]: Id 0.000000 MSSubClass 0.000000 LotFrontage 17.739726 LotArea 0.000000 OverallQual 0.000000 OverallCond 0.000000 0.000000 YearBuilt YearRemodAdd 0.000000 0.547945 MasVnrArea BsmtFinSF1 0.000000 0.000000 BsmtFinSF2 **BsmtUnfSF** 0.000000 TotalBsmtSF 0.000000 1stFlrSF 0.000000 0.000000 2ndFlrSF LowQualFinSF 0.000000 0.000000 GrLivArea 0.000000 BsmtFullBath BsmtHalfBath 0.000000 0.000000 FullBath 0.000000 HalfBath 0.000000 BedroomAbvGr KitchenAbvGr 0.000000 TotRmsAbvGrd 0.000000 Fireplaces 0.000000 GarageYrBlt 5.547945 GarageCars 0.000000 GarageArea 0.000000 WoodDeckSF 0.000000 OpenPorchSF 0.000000 EnclosedPorch 0.000000 0.000000 3SsnPorch ScreenPorch 0.000000 PoolArea 0.000000 0.000000 MiscVal

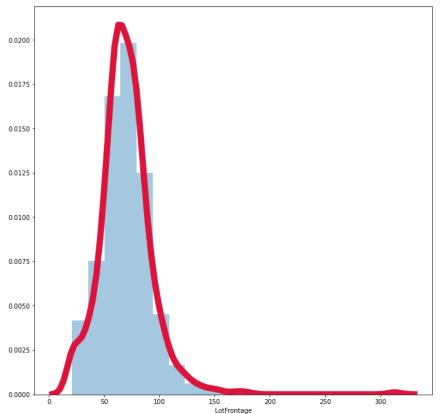
MoSold 0.000000 YrSold 0.000000 SalePrice 0.000000

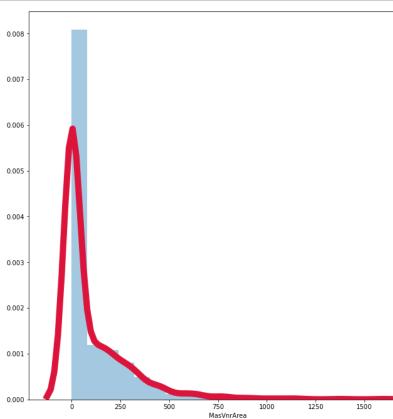
dtype: float64

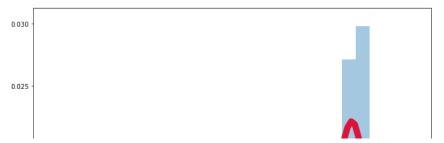
```
In [19]: missin_num_var=[var for var in data3_num.columns if data3_num[var].isnull().sum()>0]
missin_num_var
```

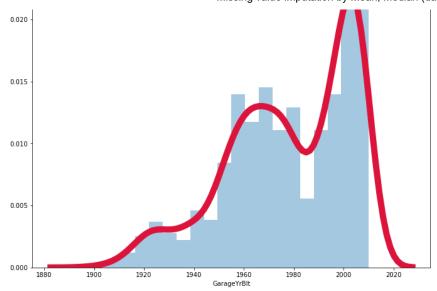
Out[19]: ['LotFrontage', 'MasVnrArea', 'GarageYrBlt']

```
In [22]: plt.figure(figsize=(25,25))
    for i, var in enumerate(missin_num_var):
        plt.subplot(2,2,i+1)
        sns.distplot(data3_num[var],bins=20,kde_kws={"linewidth":10,"color":"#DC143C"})
    # sns.distplot(data3[var],bins=20)
```







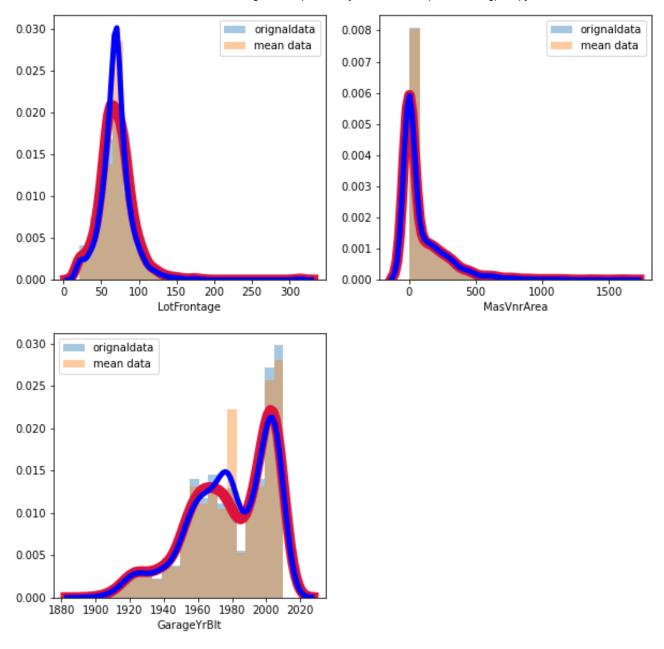


In [23]: data4\_num\_mean=data3\_num.fillna(data3\_num.mean())

In [25]: | data4\_num\_mean.isnull().sum().sum()

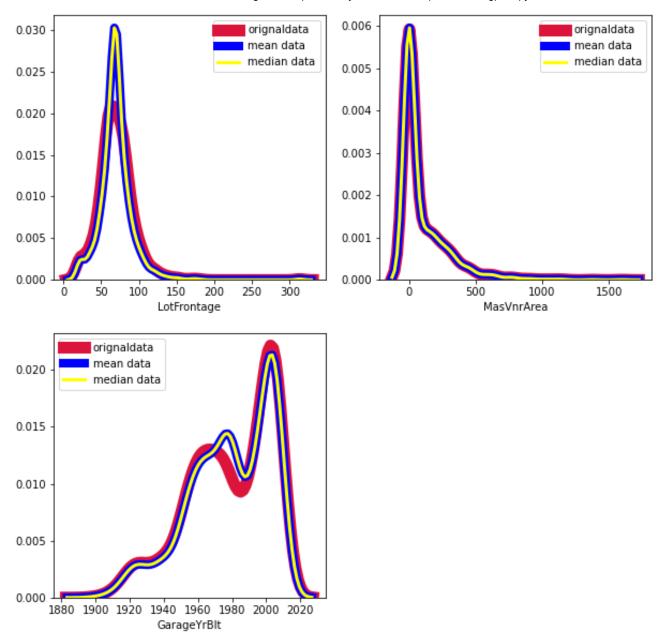
Out[25]: 0

```
In [33]: plt.figure(figsize=(10,10))
for i, var in enumerate(missin_num_var):
    plt.subplot(2,2,i+1)
    sns.distplot(data3_num[var],bins=20,kde_kws={"linewidth":10,"color":"#DC143C"},label="orignaldata")
    sns.distplot(data4_num_mean[var],bins=20,kde_kws={"linewidth":5,"color":"blue"},label="mean data")
    plt.legend()
```

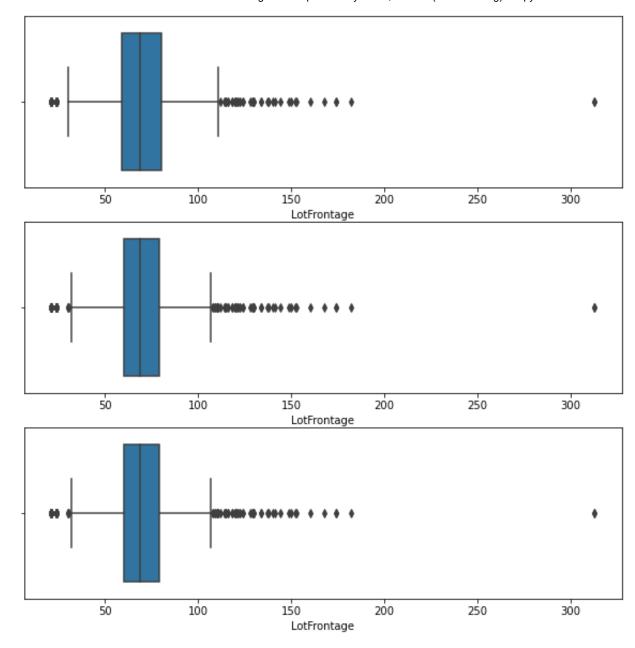


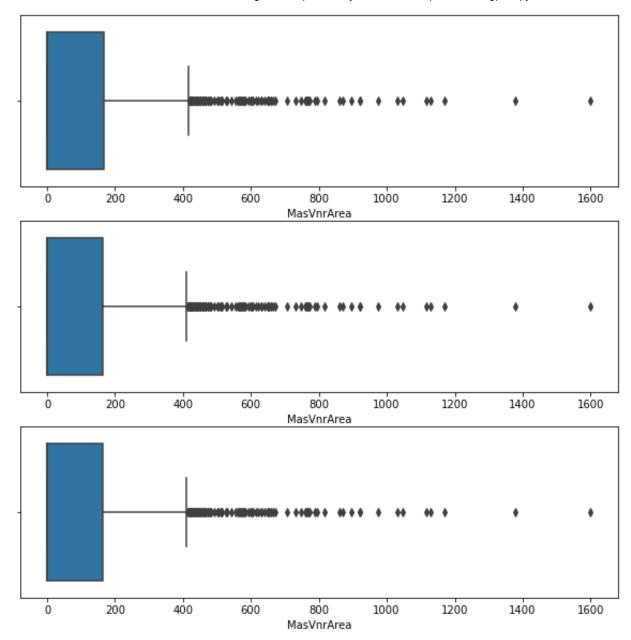
In [40]: data5\_num\_median=data3\_num.fillna(data3\_num.median())

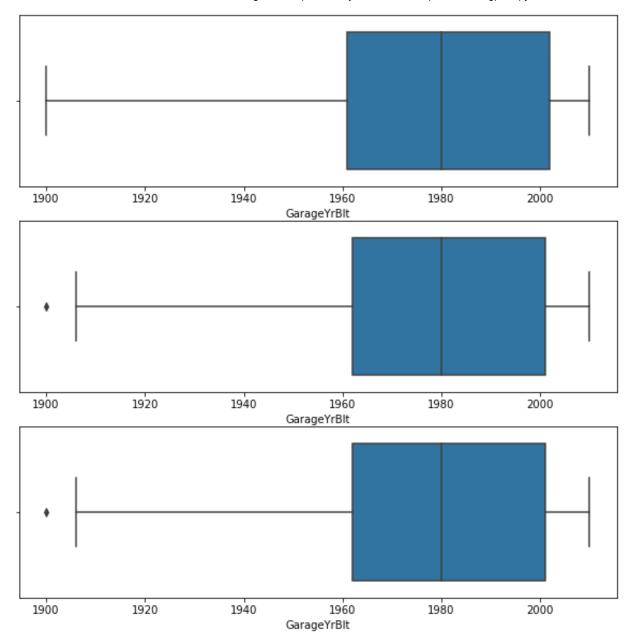
```
In [47]: plt.figure(figsize=(10,10))
for i, var in enumerate(missin_num_var):
    plt.subplot(2,2,i+1)
    sns.distplot(data3_num[var],bins=20,hist=False,kde_kws={"linewidth":11,"color":"#DC143C"},label="origna sns.distplot(data4_num_mean[var],bins=20,hist=False,kde_kws={"linewidth":8,"color":"blue"},label="mean sns.distplot(data5_num_median[var],bins=20,hist=False,kde_kws={"linewidth":3,"color":"yellow"},label="mplt.legend()
```



```
In [52]: for i, var in enumerate(missin_num_var):
    plt.figure(figsize=(10,10))
    plt.subplot(3,1,1)
    sns.boxplot(data[var])
    plt.subplot(3,1,2)
    sns.boxplot(data4_num_mean[var])
    plt.subplot(3,1,3)
    sns.boxplot(data5_num_median[var])
```







In [55]: data\_concat=pd.concat([data[missin\_num\_var],data4\_num\_mean[missin\_num\_var],data5\_num\_median[missin\_num\_var]

In [59]: con\_data=data\_concat[data\_concat.isnull().any(axis=1)]

In [61]: con\_data.head(25)

Out[61]:

	LotFrontage	MasVnrArea	GarageYrBlt	LotFrontage	MasVnrArea	GarageYrBlt	LotFrontage	MasVnrArea	GarageYrBlt
7	NaN	240.0	1973.0	69.0	240.0	1973.0	69.0	240.0	1973.0
12	NaN	0.0	1962.0	69.0	0.0	1962.0	69.0	0.0	1962.0
14	NaN	212.0	1960.0	69.0	212.0	1960.0	69.0	212.0	1960.0
16	NaN	180.0	1970.0	69.0	180.0	1970.0	69.0	180.0	1970.0
24	NaN	0.0	1968.0	69.0	0.0	1968.0	69.0	0.0	1968.0
31	NaN	0.0	1966.0	69.0	0.0	1966.0	69.0	0.0	1966.0
39	65.0	0.0	NaN	65.0	0.0	1980.0	65.0	0.0	1980.0
42	NaN	0.0	1983.0	69.0	0.0	1983.0	69.0	0.0	1983.0
43	NaN	0.0	1977.0	69.0	0.0	1977.0	69.0	0.0	1977.0
48	33.0	0.0	NaN	33.0	0.0	1980.0	33.0	0.0	1980.0
50	NaN	0.0	1997.0	69.0	0.0	1997.0	69.0	0.0	1997.0
64	NaN	573.0	1998.0	69.0	573.0	1998.0	69.0	573.0	1998.0
66	NaN	287.0	1970.0	69.0	287.0	1970.0	69.0	287.0	1970.0
76	NaN	0.0	1956.0	69.0	0.0	1956.0	69.0	0.0	1956.0
78	72.0	0.0	NaN	72.0	0.0	1980.0	72.0	0.0	1980.0
84	NaN	22.0	1995.0	69.0	22.0	1995.0	69.0	22.0	1995.0
88	105.0	0.0	NaN	105.0	0.0	1980.0	105.0	0.0	1980.0
89	60.0	0.0	NaN	60.0	0.0	1980.0	60.0	0.0	1980.0
95	NaN	68.0	1993.0	69.0	68.0	1993.0	69.0	68.0	1993.0
99	77.0	0.0	NaN	77.0	0.0	1980.0	77.0	0.0	1980.0
100	NaN	28.0	1977.0	69.0	28.0	1977.0	69.0	28.0	1977.0

	LotFrontage	MasVnrArea	GarageYrBlt	LotFrontage	MasVnrArea	GarageYrBlt	LotFrontage	MasVnrArea	GarageYrBlt
104	NaN	600.0	1951.0	69.0	600.0	1951.0	69.0	600.0	1951.0
108	85.0	0.0	NaN	85.0	0.0	1980.0	85.0	0.0	1980.0
111	NaN	0.0	2000.0	69.0	0.0	2000.0	69.0	0.0	2000.0
113	NaN	184.0	1953.0	69.0	184.0	1953.0	69.0	184.0	1953.0

localhost:8888/notebooks/mL imp/data cleaning/Missing value imputation by Mean%2C Median (data cleaning).ipynb