```
In [1]:
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
```

In [2]:

import io
%cd "C:\Users\gouth\OneDrive\Desktop\Big Mart\HousePrice"

C:\Users\gouth\OneDrive\Desktop\Big Mart\HousePrice

In [3]:

housetrain=pd.read_csv("train.csv")

In [4]:

housetest=pd.read_csv("test.csv")

In [5]:

print(housetrain.shape)
print(housetest.shape)

(1460, 81) (1459, 80) In [6]:

housetrain.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1460 entries, 0 to 1459
Data columns (total 81 columns):

_		ries, 0 to 1459	
Data #	columns (total Column	81 columns): Non-Null Count	Dtype
0	Id	1460 non-null	int64
1	MSSubClass	1460 non-null	int64
2	MSZoning	1460 non-null	object
3	LotFrontage	1201 non-null	float64
4	LotArea	1460 non-null	int64
5	Street	1460 non-null	object
6 7	Alley LotShape	91 non-null 1460 non-null	object object
8	LandContour	1460 non-null	object
9	Utilities	1460 non-null	object
10	LotConfig	1460 non-null	object
11	LandSlope	1460 non-null	object
12	Neighborhood	1460 non-null	object
13 14	Condition1	1460 non-null	object
15	Condition2 BldgType	1460 non-null 1460 non-null	object object
16	HouseStyle	1460 non-null	object
17	OverallQual	1460 non-null	int64
18	OverallCond	1460 non-null	int64
19	YearBuilt	1460 non-null	int64
20	YearRemodAdd	1460 non-null	int64
21 22	RoofStyle RoofMatl	1460 non-null 1460 non-null	object
23	Exterior1st	1460 non-null 1460 non-null	object object
24	Exterior2nd	1460 non-null	object
25	MasVnrType	1452 non-null	object
26	MasVnrArea	1452 non-null	float64
27	ExterQual	1460 non-null	object
28	ExterCond	1460 non-null	object
29	Foundation	1460 non-null	object
30 31	BsmtQual BsmtCond	1423 non-null 1423 non-null	object object
32	BsmtExposure	1422 non-null	object
33	BsmtFinType1	1423 non-null	object
34	BsmtFinSF1	1460 non-null	int64
35	BsmtFinType2	1422 non-null	object
36	BsmtFinSF2	1460 non-null	int64
37	BsmtUnfSF	1460 non-null	int64
38	TotalBsmtSF	1460 non-null	int64
39 40	Heating HeatingQC	1460 non-null 1460 non-null	object object
41	CentralAir	1460 non-null	object
42	Electrical	1459 non-null	object
43	1stFlrSF	1460 non-null	int64
44	2ndFlrSF	1460 non-null	int64
45	LowQualFinSF	1460 non-null	int64
46	GrLivArea	1460 non-null	int64
47 48	BsmtFullBath BsmtHalfBath	1460 non-null 1460 non-null	int64 int64
49	FullBath	1460 non-null	int64
50	HalfBath	1460 non-null	int64
51	BedroomAbvGr	1460 non-null	int64
52	KitchenAbvGr	1460 non-null	int64
53	KitchenQual	1460 non-null	object
54	TotRmsAbvGrd	1460 non-null	int64
55 56	Functional Fireplaces	1460 non-null 1460 non-null	object int64
57	FireplaceQu	770 non-null	object
58	GarageType	1379 non-null	object
59	GarageYrBlt	1379 non-null	float64
60	GarageFinish	1379 non-null	object
61	GarageCars	1460 non-null	int64
62	GarageArea	1460 non-null	int64
63 64	GarageQual GarageCond	1379 non-null 1379 non-null	object object
65	PavedDrive	1460 non-null	object
66	WoodDeckSF	1460 non-null	int64
67	OpenPorchSF	1460 non-null	int64
68	EnclosedPorch	1460 non-null	int64
69	3SsnPorch	1460 non-null	int64
70 71	ScreenPorch	1460 non-null	int64
71 72	PoolArea PoolQC	1460 non-null 7 non-null	int64
72 73	Fence	281 non-null	object 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 80	SaleCondition SalePrice	1460 non-null	object
80 dtvne	salerrice es: float64(3),	1460 non-null int64(35), object	int64 +(43)
	ry usage: 924.0+		- (-)
	. 3		

housetest.info()

In [7]:

<class 'pandas.core.frame.DataFrame'> RangeIndex: 1459 entries, 0 to 1458 Data columns (total 80 columns): Non-Null Count Dtvpe # Column --a Τd 1459 non-null int64 MSSubClass 1 1459 non-null int64 MSZoning 2 1455 non-null object 3 LotFrontage 1232 non-null float64 4 LotArea 1459 non-null int64 5 Street 1459 non-null object 6 Alley 107 non-null object LotShape 1459 non-null object 8 LandContour 1459 non-null object 9 Utilities 1457 non-null object LotConfig 1459 non-null object LandSlope 1459 non-null 11 object Neighborhood 1459 non-null 12 object 13 Condition1 1459 non-null object 1459 non-null 14 Condition2 object 1459 non-null 15 BldgType object 16 HouseStyle 1459 non-null object OverallQual 1459 non-null 17 int64 18 OverallCond 1459 non-null int64 YearBuilt 1459 non-null 19 int64 YearRemodAdd 1459 non-null int64 20 RoofStyle 1459 non-null 21 object RoofMatl 1459 non-null 22 object 23 Exterior1st 1458 non-null object 24 Exterior2nd 1458 non-null object 1443 non-null 25 MasVnrTvne object 26 MasVnrArea 1444 non-null float64 1459 non-null 27 ExterOual object 28 1459 non-null ExterCond object 29 Foundation 1459 non-null object 30 BsmtOual 1415 non-null object 31 **BsmtCond** 1414 non-null object 32 BsmtExposure 1415 non-null object 33 BsmtFinType1 1417 non-null object 34 BsmtFinSF1 1458 non-null float64 35 BsmtFinType2 1417 non-null object 36 BsmtFinSF2 1458 non-null float64 37 **BsmtUnfSF** 1458 non-null float64 38 TotalBsmtSF 1458 non-null float64 39 Heating 1459 non-null object HeatingQC 40 1459 non-null object 41 CentralAir 1459 non-null object 42 Electrical 1459 non-null object 43 1stFlrSF 1459 non-null int64 44 2ndFlrSF 1459 non-null int64 1459 non-null 45 LowQualFinSF int64 46 GrLivArea 1459 non-null int64 47 **BsmtFullBath** 1457 non-null float64 48 BsmtHalfBath 1457 non-null float64 49 FullBath 1459 non-null int64 50 HalfBath 1459 non-null int64 51 BedroomAbvGr 1459 non-null int64 KitchenAbvGr 1459 non-null 52 int64 53 KitchenQual 1458 non-null object 54 TotRmsAbvGrd 1459 non-null int64 1457 non-null 55 Functional object 56 Fireplaces 1459 non-null int64 57 FireplaceQu 729 non-null object 58 1383 non-null GarageType object float64 59 GarageYrBlt 1381 non-null 60 GarageFinish 1381 non-null object 1458 non-null 61 GarageCars float64 62 GarageArea 1458 non-null float64 63 ${\tt GarageQual}$ 1381 non-null object 64 GarageCond 1381 non-null object 65 PavedDrive 1459 non-null object 66 WoodDeckSF 1459 non-null int64 67 OpenPorchSF 1459 non-null int64 68 ${\tt EnclosedPorch}$ 1459 non-null int64 69 3SsnPorch 1459 non-null int64 70 ScreenPorch 1459 non-null int64 71 PoolArea 1459 non-null int64 72 Poo1QC 3 non-null object 73 290 non-null Fence object 74 MiscFeature 51 non-null object 75 MiscVal 1459 non-null int64 76 MoSold 1459 non-null int64 77 YrSold 1459 non-null int64 SaleType 1458 non-null object SaleCondition 1459 non-null object dtypes: float64(11), int64(26), object(43) memory usage: 912.0+ KB

```
In [8]:
housetest["SalePrice"]="test"

In [9]:
combinedf=pd.concat([housetrain,housetest],axis=0)

In [10]:
pd.set_option("display.max_rows",82)
```

In [11]:

combinedf.isnull().sum().sort_values(ascending=False)

Out[11]:

PoolQC	2909
MiscFeature	2814
Alley	2721
Fence	2348
FireplaceQu	1420
LotFrontage	486
GarageFinish	159
GarageQual	159
GarageCond	159
	159
GarageYrBlt	
GarageType	157
BsmtExposure	82
BsmtCond	82
BsmtQual	81
BsmtFinType2	80
BsmtFinType1	79
MasVnrType	24
	23
MasVnrArea	
MSZoning	4
Functional	2
Utilities	2
BsmtHalfBath	2
BsmtFullBath	2
GarageArea	1
BsmtFinSF1	1
SaleType	1
GarageCars	1
BsmtUnfSF	1
	1
Electrical	
Exterior2nd	1
Exterior1st	1
KitchenQual	1
•	
TotalBsmtSF	1
BsmtFinSF2	1
TotRmsAbvGrd	0
Fireplaces	0
Id	0
BedroomAbvGr	0
PavedDrive	0
WoodDeckSF	0
OpenPorchSF	0
•	
EnclosedPorch	0
3SsnPorch	0
ScreenPorch	0
PoolArea	0
MiscVal	0
MoSold	0
YrSold	0
SaleCondition	0
KitchenAbvGr	0
HeatingQC	0
	0
HalfBath	0
FullBath	0
LotArea	0
	0
Street	
LotShape	0
LandContour	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
ExterQual	0
ExterCond	0
Foundation	0
Heating	0
MSSubClass	0
CentralAir	0
1stFlrSF	0
2ndFlrSF	0
LowQualFinSF	0
	0
GrLivArea	
SalePrice	0
dtype: int64	

```
In [12]:
```

```
objectcols=combinedf.select_dtypes(include=["object"])
numericcols=combinedf.select_dtypes(include=np.number)
```

In [13]:

```
print(objectcols.shape)
print(numericcols.shape)
```

(2919, 44) (2919, 37)

In [14]:

objectcols.isnull().sum().sort_values(ascending=False)

Out[14]:

```
PoolQC
                 2909
MiscFeature
                 2814
Alley
                 2721
Fence
                 2348
FireplaceQu
                 1420
GarageCond
                  159
GarageQual
                  159
GarageFinish
GarageType
                  157
BsmtCond
                   82
BsmtExposure
BsmtQual
                   81
BsmtFinType2
                   80
BsmtFinType1
MasVnrType
                   24
MSZoning
Utilities
Functional
Exterior2nd
                    1
Electrical
                    1
SaleType
                    1
Exterior1st
                    1
KitchenQual
RoofStyle
                    0
Neighborhood
                    0
SaleCondition
                    0
ExterQual
                    0
LotShape
                    0
LandContour
                    0
ExterCond
                    0
PavedDrive
                    0
LotConfig
                    0
LandSlope
Condition1
                    0
HouseStyle
Condition2
                    0
Foundation
RoofMat1
                    0
Street
CentralAir
HeatingQC
Heating
                    0
BldgType
                    0
SalePrice
```

In [15]:

dtype: int64

combinedf.columns

Out[15]:

```
In [16]:
notavailablecols=['PoolQC','MiscFeature','Alley','Fence','FireplaceQu']
In [17]:
for col in notavailablecols:objectcols[col]=objectcols[col].fillna("notavailable")
garage_list=[col for col in objectcols if col.startswith("Gar")]
garage_list
Out[18]:
['GarageType', 'GarageFinish', 'GarageQual', 'GarageCond']
In [19]:
for col in garage_list:
    freq=objectcols[col].value_counts(dropna=False)
    print(freq)
Attchd
           1723
Detchd
            779
BuiltIn
            186
            157
NaN
Basment
             36
             23
2Types
CarPort
             15
Name: GarageType, dtype: int64
Unf
       1230
RFn
        811
Fin
        719
NaN
        159
Name: GarageFinish, dtype: int64
       2604
TΑ
       159
NaN
Fa
        124
Gd
         24
Ро
          5
Ex
          3
Name: GarageQual, dtype: int64
TΑ
       2654
NaN
        159
Fa
         74
Gd
         15
Ро
         14
Name: GarageCond, dtype: int64
In [20]:
Basement_list=[col for col in objectcols if col.startswith("Bsm")]
Basement_list
Out[20]:
```

['BsmtQual', 'BsmtCond', 'BsmtExposure', 'BsmtFinType1', 'BsmtFinType2']

In [21]:

```
for col in Basement_list:
    freq=objectcols[col].value_counts(dropna=False)
    print(freq)
TΑ
       1283
Gd
       1209
        258
Ex
Fa
         88
NaN
         81
Name: BsmtQual, dtype: int64
TΑ
       2606
Gd
        122
Fa
        104
NaN
         82
Ро
          5
Name: BsmtCond, dtype: int64
No
       1904
Gd
        276
Mn
        239
NaN
Name:
      BsmtExposure, dtype: int64
Unf
GLQ
       849
ALQ
       429
Rec
       288
BLQ
       269
Lw0
       154
NaN
        79
Name: BsmtFinType1, dtype: int64
Unf
       2493
        105
Rec
LwQ
         87
NaN
         80
BLQ
         68
ALQ
         52
GLQ
         34
Name: BsmtFinType2, dtype: int64
In [22]:
```

```
for col in objectcols.columns:
    objectcols[col]=objectcols[col].fillna(
        objectcols[col].value_counts().idxmax())
```

```
In [23]:
```

```
numericcols.isnull().sum().sort_values(ascending=False)
Out[23]:
LotFrontage
                         486
GarageYrBlt
                         159
MasVnrArea
                           23
BsmtHalfBath
                            2
BsmtFullBath
                            2
BsmtFinSF2
                            1
{\tt GarageCars}
                            1
GarageArea
                            1
{\tt TotalBsmtSF}
                            1
BsmtUnfSF
                            1
BsmtFinSF1
                            1
KitchenAbvGr
                            0
3SsnPorch
                            0
EnclosedPorch
                            0
OpenPorchSF
                            0
WoodDeckSF
                            0
ScreenPorch
PoolArea
                            0
MiscVal
                            0
MoSold
                            0
Fireplaces
                            0
TotRmsAbvGrd
                            0
                            0
Ιd
BedroomAbvGr
                            0
HalfBath
                            0
FullBath
                            0
MSSubClass
                            0
GrLivArea
                            0
LowQualFinSF
                            0
2ndFlrSF
                            0
1stFlrSF
                            0
                            0
YearRemodAdd
                            0
YearBuilt
                            0
OverallCond
OverallQual
                            0
                            a
LotArea
YrSold
                            0
dtype: int64
In [24]:
numericcols.columns
Out[24]:
Index(['Id', 'MSSubClass', 'LotFrontage', 'LotArea', 'OverallQual',
    'OverallCond', 'YearBuilt', 'YearRemodAdd', 'MasVnrArea', 'BsmtFinSF1',
    'BsmtFinSF2', 'BsmtUnfSF', 'TotalBsmtSF', '1stFlrSF', '2ndFlrSF',
    'LowQualFinSF', 'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath',
    'HalfBath', 'BedroomAbvGr', 'KitchenAbvGr', 'TotRmsAbvGrd',
    'Fireplaces', 'GarageYrBlt', 'GarageCars', 'GarageArea', 'WoodDeckSF',
    'OpenPorchSF', 'EnclosedPorch', '3SsnPorch', 'ScreenPorch', 'PoolArea',
    'MiscVal' 'Mosold' 'YrSold']
           'MiscVal', 'MoSold', 'YrSold'],
         dtype='object')
In [25]:
categorycols=numericcols[['OverallQual','OverallCond','YearBuilt','YearRemodAdd','GarageYrBlt','MoSold','YrSold']]
In [26]:
numericcols=numericcols.drop(['OverallQual','OverallCond','YearBuilt','YearRemodAdd','GarageYrBlt','MoSold','YrSold'],axis=1)
In [27]:
print(categorycols.shape)
print(numericcols.shape)
(2919, 7)
(2919, 30)
```

```
In [28]:
```

```
categorycols.isnull().sum().sort_values(ascending=False)
```

Out[28]:

GarageYrBlt 159
OverallQual 0
OverallCond 0
YearBuilt 0
YearRemodAdd 0
MoSold 0
YrSold 0
dtype: int64

In [29]:

categorycols.GarageYrBlt=categorycols.GarageYrBlt.fillna(categorycols.GarageYrBlt.value_counts().idxmax())

In [30]:

```
# Median Imputation for Numeric Cols
for col in numericcols.columns:
    numericcols[col]=numericcols[col].fillna(numericcols[col].median())
```

In [31]:

from sklearn.preprocessing import LabelEncoder

In [32]:

numericcols["SalePrice"]=objectcols.SalePrice

In [33]:

objectcols=objectcols.drop("SalePrice",axis=1)

In [34]:

le=LabelEncoder()

In [35]:

objectcols_encode=objectcols.apply(le.fit_transform)

In [36]:

categorycols_encode=categorycols.apply(le.fit_transform)

In [37]:

numericcols.head()

Out[37]:

	ld	MSSubClass	LotFrontage	LotArea	MasVnrArea	BsmtFinSF1	BsmtFinSF2	BsmtUnfSF	TotalBsmtSF	1stFIrSF	 GarageCars	GarageArea	Wood
0	1	60	65.0	8450	196.0	706.0	0.0	150.0	856.0	856	 2.0	548.0	
1	2	20	80.0	9600	0.0	978.0	0.0	284.0	1262.0	1262	 2.0	460.0	
2	3	60	68.0	11250	162.0	486.0	0.0	434.0	920.0	920	 2.0	608.0	
3	4	70	60.0	9550	0.0	216.0	0.0	540.0	756.0	961	 3.0	642.0	
4	5	60	84.0	14260	350.0	655.0	0.0	490.0	1145.0	1145	 3.0	836.0	

5 rows × 31 columns

```
In [38]:
# Check for Multicollinearity
numericcols.drop('SalePrice',axis=1).corr()
Out[38]:
```

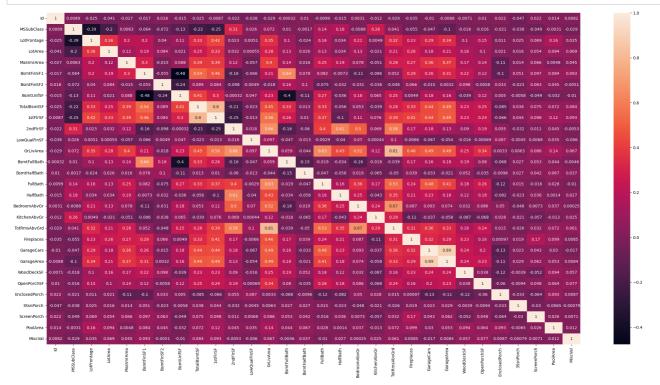
	ld	MSSubClass	LotFrontage	LotArea	MasVnrArea	BsmtFinSF1	BsmtFinSF2	BsmtUnfSF	TotalBsmtSF	1stFlrSF	 Fireplaces	GarageCar
ld	1.000000	0.008931	-0.024710	-0.040746	-0.026737	-0.016988	0.018170	-0.014509	-0.024960	-0.008678	 -0.035236	-0.01006
MSSubClass	0.008931	1.000000	-0.389469	-0.201730	0.006309	-0.064254	-0.072431	-0.125913	-0.219893	-0.248641	 -0.055151	-0.04656
LotFrontage	-0.024710	-0.389469	1.000000	0.361426	0.200474	0.201697	0.040330	0.106921	0.330765	0.423217	 0.231731	0.29063
LotArea	-0.040746	-0.201730	0.361426	1.000000	0.124728	0.194050	0.084107	0.021400	0.254150	0.332460	 0.261185	0.18041
MasVnrArea	-0.026737	0.006309	0.200474	0.124728	1.000000	0.301427	-0.014580	0.087991	0.393662	0.392367	 0.273129	0.35765
BsmtFinSF1	-0.016988	-0.064254	0.201697	0.194050	0.301427	1.000000	-0.055028	-0.477387	0.536471	0.458091	 0.293095	0.25551
BsmtFinSF2	0.018170	-0.072431	0.040330	0.084107	-0.014580	-0.055028	1.000000	-0.238215	0.089423	0.084389	 0.065707	-0.01475
BsmtUnfSF	-0.014509	-0.125913	0.106921	0.021400	0.087991	-0.477387	-0.238215	1.000000	0.412291	0.296623	 0.004881	0.18006 ⁻
TotalBsmtSF	-0.024960	-0.219893	0.330765	0.254150	0.393662	0.536471	0.089423	0.412291	1.000000	0.801638	 0.332948	0.43790
4												.

In [39]:

import seaborn as sns

In [40]:

```
plt.figure(figsize=(30,15))
sns.heatmap(numericcols.drop('SalePrice',axis=1).corr(),annot=True)
plt.show()
```



In [41]:

from sklearn.preprocessing import StandardScaler

In [42]:

scaler=StandardScaler()

In [43]:

```
numericcols_scaled=scaler.fit_transform(numericcols.drop(
'SalePrice',axis=1))
```

In [44]:

numericcols_scaled=pd.DataFrame(numericcols_scaled, columns=numericcols.columns[0:30])

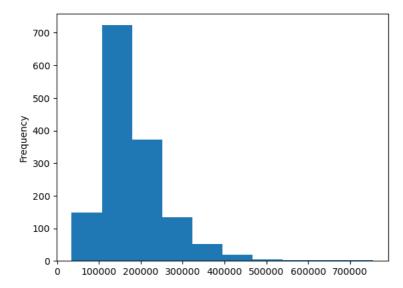
```
In [45]:
numericcols_scaled=numericcols_scaled.reset_index()
In [46]:
objectcols_encode=objectcols_encode.reset_index()
In [47]:
categorycols encode=categorycols encode.reset index()
In [48]:
combined f\_clean = pd.concat([numeric cols\_scaled, object cols\_encode, object cols\_encode]) \\
categorycols_encode],axis=1)
In [49]:
numericcols=numericcols.reset_index()
In [50]:
combinedf_clean['SalePrice']=numericcols.SalePrice
In [51]:
housetrain_df=combinedf_clean[combinedf_clean.SalePrice!='test']
housetest_df=combinedf_clean[combinedf_clean.SalePrice=='test']
In [52]:
housetest_df=housetest_df.drop('SalePrice',axis=1)
In [53]:
\# Split data into X and Y
y=housetrain_df.SalePrice
x=housetrain_df.drop(['Id','SalePrice','index'],axis=1)
In [54]:
housetest_df=housetest_df.drop(['Id','index'],axis=1)
In [55]:
housetest_df.head()
Out[55]:
      MSSubClass LotFrontage
                              LotArea
                                      MasVnrArea BsmtFinSF1 BsmtFinSF2 BsmtUnfSF TotalBsmtSF
                                                                                                 1stFIrSF
                                                                                                         2ndFlrSF
                                                                                                                     MiscFeature
                                                                                                                                 SaleTy
                                                                                                                              4
 1460
         -0.873616
                     0.511940
                             0.184371
                                         -0.567016
                                                     0.058407
                                                                0.558290
                                                                           -0.661680
                                                                                       -0.385270 -0.671897
                                                                                                         -0.785025
 1461
         -0.873616
                    0.558857
                             0.519791
                                         0.036930
                                                     1.057404
                                                                -0.293025
                                                                           -0.352165
                                                                                       0
 1462
         0.067331
                     0.230438
                             0.464374
                                         -0.567016
                                                     0.767585
                                                                -0.293025
                                                                           -0.964368
                                                                                       -0.280871 -0.590326
                                                                                                          0.850426
                                                                                                                              4
         0.067331
                                                                -0.293025
 1463
                    0.418106 -0.024109
                                         -0.455174
                                                     0.352617
                                                                           -0.538784
                                                                                       -0.285410 -0.595424 0.796766
                                                                                                                              4
         1.478753
                    -1.223988 -0.654748
                                         -0.567016
                                                                -0.293025
 1464
                                                    -0.391691
                                                                           1.038379
                                                                                       5 rows × 79 columns
4
In [56]:
y.dtype
Out[56]:
dtype('0')
In [57]:
y=y.astype('int64')
```

In [58]:

y.plot(kind="hist")

Out[58]:

<AxesSubplot:ylabel='Frequency'>

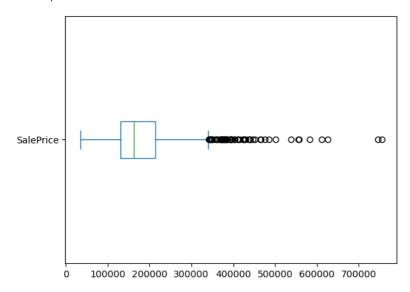


In [59]:

y.plot(kind="box",vert=False)

Out[59]:

<AxesSubplot:>



```
In [60]:
y.plot(kind="density")
Out[60]:
<AxesSubplot:ylabel='Density'>
       1e-6
    7
    6
    5
    2
    1
    0
              -0.2
                       0.0
                                0.2
                                          0.4
                                                   0.6
                                                            0.8
                                                                     1.0
                                                                           1e6
In [61]:
from sklearn.linear_model import LinearRegression
In [62]:
reg=LinearRegression()
In [63]:
regmodel=reg.fit(x,y)
In [64]:
regmodel.score(x,y)
Out[64]:
0.8545634224507179
In [65]:
regtestpredict=regmodel.predict(housetest_df)
In [66]:
regtestpredict
Out[66]:
array([106189.76326881, 156987.78817192, 167242.08767411, ..., 143662.33304422, 115297.31625419, 242815.16584994])
In [67]:
pd.DataFrame(regtestpredict).to_csv("reg.csv")
In [ ]:
```