

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
df=pd.read_csv('train.csv')
```

```
df.head()
```

	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
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]
```

```
df.isnull().sum()
```

Id	0
MSSubClass	0
MSZoning	0
LotFrontage	259
LotArea	0

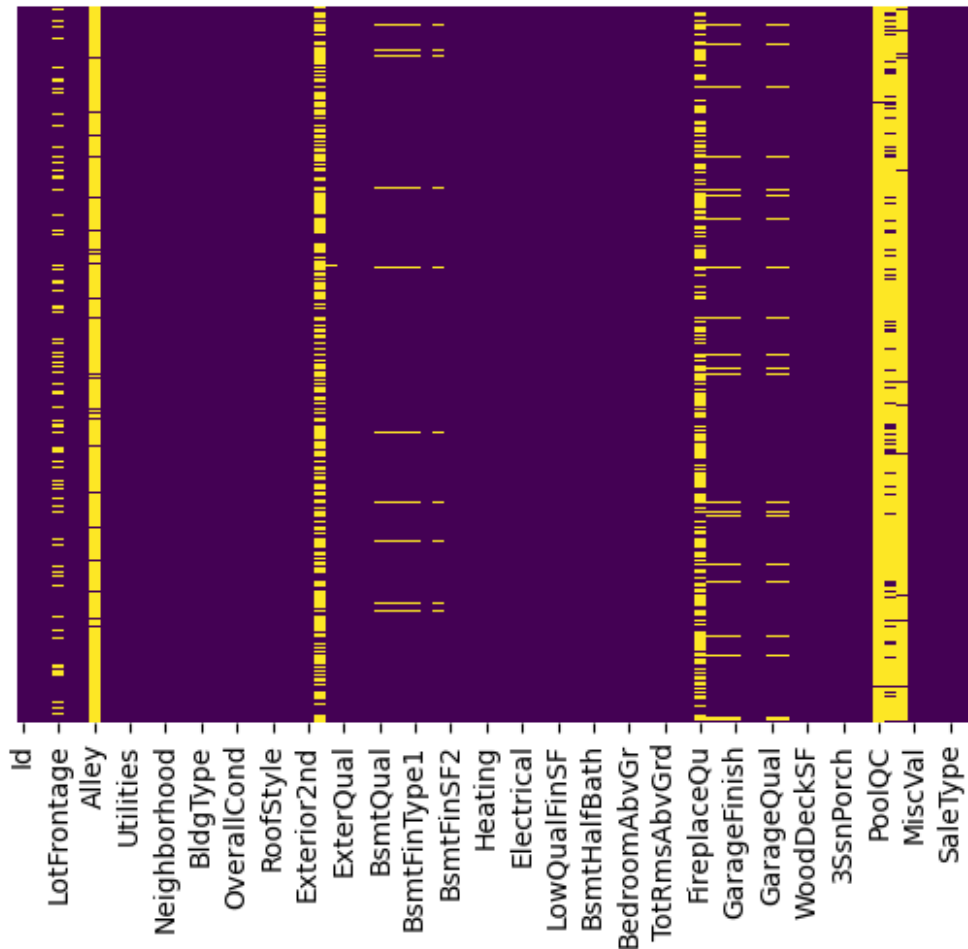
```

    ...
MoSold      0
YrSold      0
SaleType    0
SaleCondition 0
SalePrice   0
Length: 81, dtype: int64

sns.heatmap(df.isnull(),yticklabels=False,cbar=False,cmap='viridis')

<Axes: >

```



```

df.shape
(1460, 81)

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1460 entries, 0 to 1459

```

```

Data columns (total 81 columns):
#   Column              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   Alley                91 non-null     object
7   LotShape             1460 non-null   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  Condition1           1460 non-null   object
14  Condition2           1460 non-null   object
15  BldgType             1460 non-null   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  RoofStyle            1460 non-null   object
22  RoofMatl             1460 non-null   object
23  Exterior1st          1460 non-null   object
24  Exterior2nd          1460 non-null   object
25  MasVnrType           588 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  BsmtQual             1423 non-null   object
31  BsmtCond             1423 non-null   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  Heating              1460 non-null   object
40  HeatingQC            1460 non-null   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	BsmtFullBath	1460	non-null	int64
48	BsmtHalfBath	1460	non-null	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	Functional	1460	non-null	object
56	Fireplaces	1460	non-null	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	GarageQual	1379	non-null	object
64	GarageCond	1379	non-null	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	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

dtypes: float64(3), int64(35), object(43)

memory usage: 924.0+ KB

```
df['LotFrontage']=df['LotFrontage'].fillna(df['LotFrontage'].mean())
```

```
df.drop(['Alley'],axis=1,inplace=True)
```

```
df['BsmtCond']=df['BsmtCond'].fillna(df['BsmtCond'].mode()[0])
```

```
df['BsmtQual']=df['BsmtQual'].fillna(df['BsmtQual'].mode()[0])
```

```
df['FireplaceQu']=df['FireplaceQu'].fillna(df['FireplaceQu'].mode()[0])
```

```
df['GarageType']=df['GarageType'].fillna(df['GarageType'].mode()[0])
```

```
df.drop(['GarageYrBlt'],axis=1,inplace=True)
```

```

df['GarageFinish']=df['GarageFinish'].fillna(df['GarageFinish'].mode()[0])
df['GarageQual']=df['GarageQual'].fillna(df['GarageQual'].mode()[0])
df['GarageCond']=df['GarageCond'].fillna(df['GarageCond'].mode()[0])

df.drop(['PoolQC','Fence','MiscFeature'],axis=1,inplace=True)

df.shape

(1460, 76)

df.drop(['Id'],axis=1,inplace=True)

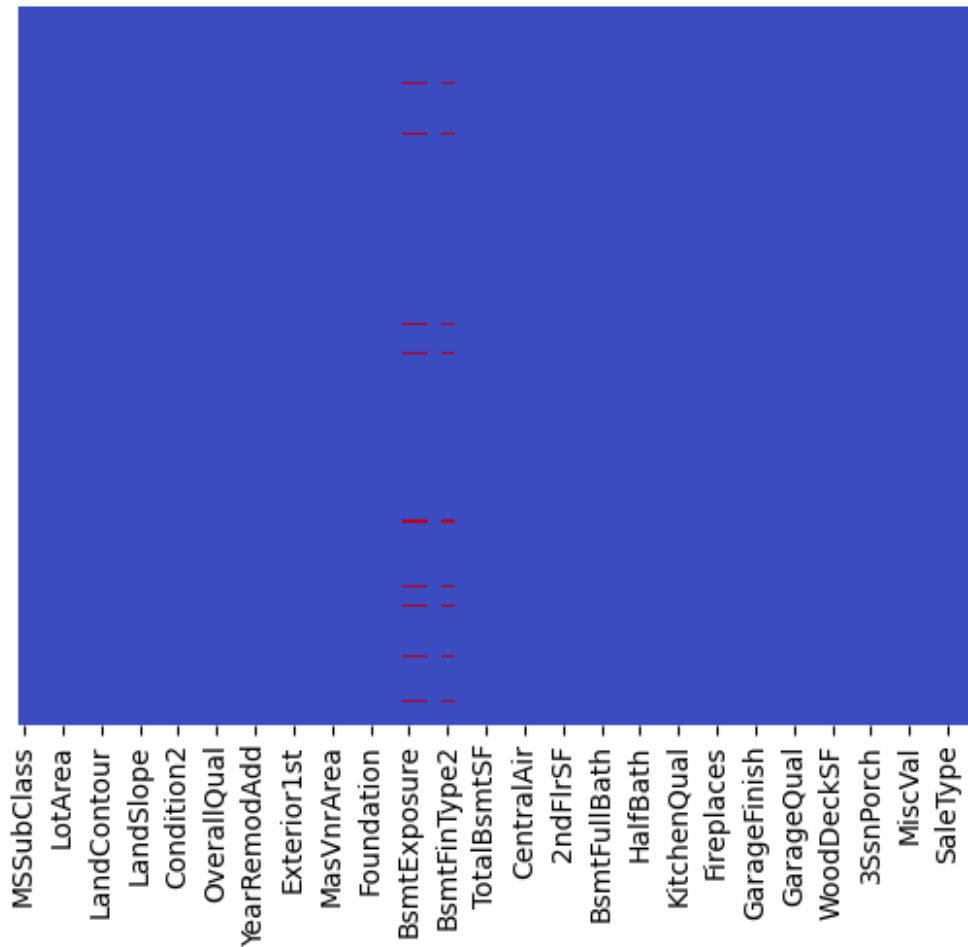
df.isnull().sum()
MSSubClass      0
MSZoning         0
LotFrontage     0
LotArea         0
Street          0
..
MoSold          0
YrSold          0
SaleType        0
SaleCondition   0
SalePrice       0
Length: 75, dtype: int64

df['MasVnrType']=df['MasVnrType'].fillna(df['MasVnrType'].mode()[0])
df['MasVnrArea']=df['MasVnrArea'].fillna(df['MasVnrArea'].mode()[0])

sns.heatmap(df.isnull(),yticklabels=False,cbar=False,cmap='coolwarm')

<Axes: >

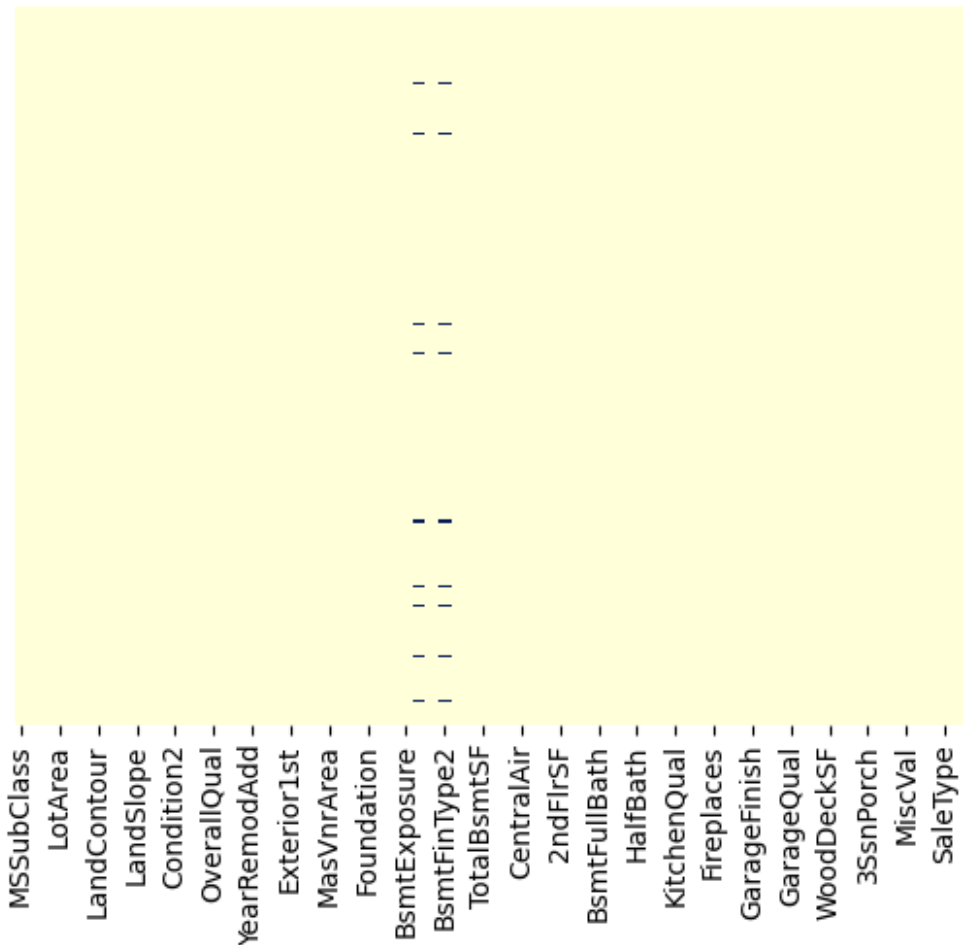
```



```
df['BsmtExposure']=df['BsmtExposure'].fillna(df['BsmtExposure'].mode()[0])

sns.heatmap(df.isnull(),yticklabels=False,cbar=False,cmap='YlGnBu')

<Axes: >
```



```
df['BsmtFinType2']=df['BsmtFinType2'].fillna(df['BsmtFinType2'].mode()[0])
```

```
df.dropna(inplace=True)
```

```
df.shape
```

```
(1421, 75)
```

```
df.head()
```

	MSSubClass	MSZoning	LotFrontage	LotArea	Street	LotShape
LandContour \						
0	60	RL	65.0	8450	Pave	Reg
Lvl						
1	20	RL	80.0	9600	Pave	Reg
Lvl						
2	60	RL	68.0	11250	Pave	IR1
Lvl						
3	70	RL	60.0	9550	Pave	IR1
Lvl						

4	60	RL	84.0	14260	Pave	IR1
---	----	----	------	-------	------	-----

Lvl

	Utilities	LotConfig	LandSlope	...	EnclosedPorch	3SsnPorch
ScreenPorch	\					
0	AllPub	Inside	Gtl	...	0	0
0						
1	AllPub	FR2	Gtl	...	0	0
0						
2	AllPub	Inside	Gtl	...	0	0
0						
3	AllPub	Corner	Gtl	...	272	0
0						
4	AllPub	FR2	Gtl	...	0	0
0						

	PoolArea	MiscVal	MoSold	YrSold	SaleType	SaleCondition	SalePrice
0	0	0	2	2008	WD	Normal	208500
1	0	0	5	2007	WD	Normal	181500
2	0	0	9	2008	WD	Normal	223500
3	0	0	2	2006	WD	Abnorml	140000
4	0	0	12	2008	WD	Normal	250000

[5 rows x 75 columns]

```
cat_cols = df.select_dtypes(include='object').columns
print(cat_cols)
```

```
Index(['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',
      'Functional', 'FireplaceQu', 'GarageType', 'GarageFinish',
      'GarageQual',
      'GarageCond', 'PavedDrive', 'SaleType', 'SaleCondition'],
      dtype='object')
```



```

from sklearn.preprocessing import LabelEncoder

le = LabelEncoder()
for col in cat_cols:
    df[col] = le.fit_transform(df[col])

X = df.drop('SalePrice', axis=1)
y = df['SalePrice']

from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.2, random_state=42)

from sklearn.preprocessing import StandardScaler

scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)

from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score
import numpy as np

lr = LinearRegression()
lr.fit(X_train, y_train)

# Predictions
y_pred = lr.predict(X_test)

# Evaluation
rmse = np.sqrt(mean_squared_error(y_test, y_pred))
r2 = r2_score(y_test, y_pred)

print(f"RMSE: {rmse:.2f}")
print(f"R2 Score: {r2:.2f}")

RMSE: 30606.95
R2 Score: 0.84

# Example: Use the first row from test data
sample = X_test[0].reshape(1, -1)

# Predict price
predicted_price = lr.predict(sample)
print(f"Predicted House Price: ${predicted_price[0]:,.2f}")

Predicted House Price: $371,907.06

```