

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

```
In [2]: 1 df = pd.read_csv('train.csv')
        2 df.head()
```

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

5 rows × 81 columns

```
In [3]: 1 df.columns
```

```
Index(['Id', 'MSSubClass', 'MSZoning', 'LotFrontage', 'LotArea', 'Street',
       'Alley', 'LotShape', 'LandContour', 'Utilities', 'LotConfig',
       'LandSlope', 'Neighborhood', 'Condition1', 'Condition2', 'BldgType',
       'HouseStyle', 'OverallQual', 'OverallCond', 'YearBuilt', 'YearRemodAdd',
       'RoofStyle', 'RoofMatl', 'Exterior1st', 'Exterior2nd', 'MasVnrType',
       'MasVnrArea', 'ExterQual', 'ExterCond', 'Foundation', 'BsmtQual',
       'BsmtCond', 'BsmtExposure', 'BsmtFinType1', 'BsmtFinSF1',
       'BsmtFinType2', 'BsmtFinSF2', 'BsmtUnfSF', 'TotalBsmtSF', 'Heating',
       'HeatingQC', 'CentralAir', 'Electrical', '1stFlrSF', '2ndFlrSF',
       'LowQualFinSF', 'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath',
       'HalfBath', 'BedroomAbvGr', 'KitchenAbvGr', 'KitchenQual',
       'TotRmsAbvGrd', 'Functional', 'Fireplaces', 'FireplaceQu', 'GarageType',
       'GarageYrBlt', 'GarageFinish', 'GarageCars', 'GarageArea', 'GarageQual',
       'GarageCond', 'PavedDrive', 'WoodDeckSF', 'OpenPorchSF',
       'EnclosedPorch', '3SsnPorch', 'ScreenPorch', 'PoolArea', 'PoolQC',
       'Fence', 'MiscFeature', 'MiscVal', 'MoSold', 'YrSold', 'SaleType',
       'SaleCondition', 'SalePrice'],
      dtype='object')
```

```
In [4]: 1 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	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	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

```

```
In [5]: 1 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

```

```
In [6]: 1 df.shape
```

```
(1460, 81)
```

```
In [7]: 1 X = df.drop(columns='SalePrice')
        2 y = df['SalePrice']
```

```
In [8]: 1 numeric_columns = []
2 object_columns = []
3
4 for column in X.columns:
5     if pd.api.types.is_numeric_dtype(df[column]):
6         numeric_columns.append(column)
7     elif pd.api.types.is_object_dtype(df[column]):
8         object_columns.append(column)
9
10 print(len(numeric_columns), len(object_columns))
```

37 43

```
In [9]: 1 from sklearn.impute import SimpleImputer
2 from sklearn.compose import ColumnTransformer
3 from sklearn.preprocessing import OrdinalEncoder
4 from sklearn.pipeline import Pipeline,make_pipeline
5 from sklearn.preprocessing import StandardScaler
6 from sklearn.linear_model import LinearRegression
```

```
In [10]: 1 handle_numerical = Pipeline(steps=[
2     ('impute_numerical',SimpleImputer(strategy='mean')),
3     ('scaling_numerical',StandardScaler())
4 ])
```

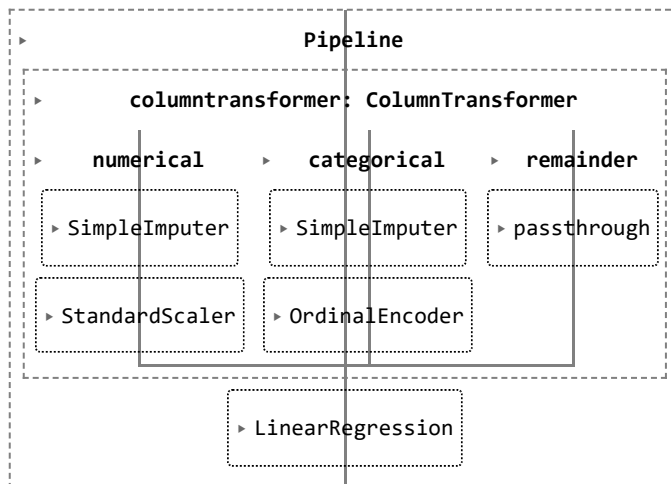
```
In [11]: 1 handle_categorical = Pipeline(steps=[
2     ('handle_categorical',SimpleImputer(strategy='most_frequent')),
3     ('encode_categorical',OrdinalEncoder())
4 ])
```

```
In [12]: 1 preprocessing = ColumnTransformer(transformers=[
2     ('numerical',handle_numerical,numeric_columns),
3     ('categorical',handle_categorical,object_columns)
4 ],remainder='passthrough')
```

```
In [13]: 1 model = LinearRegression()
```

```
In [14]: 1 pipe = make_pipeline(preprocessing,model)
```

```
In [15]: 1 pipe.fit(X,y)
```



```
In [16]: 1 X_test = pd.read_csv('tesst.csv')
```

```
In [17]: 1 X_test.shape
```

(1459, 80)

```
In [18]: 1 sub = pd.read_csv('saample_submissions.csv')
2 sub.head()
```

	Id	SalePrice
0	1461	169277.052498
1	1462	187758.393989
2	1463	183583.683570
3	1464	179317.477511
4	1465	150730.079977

```
In [19]: 1 y_pred = pipe.predict(X_test)
```

```
In [20]: 1 id_df = X[['Id']]
2
3 y_pred_df = pd.DataFrame({'SalePrice': y_pred})
4
5 y_pred_final = pd.concat([id_df, y_pred_df], axis=1)
```

```
In [21]: 1 pd.DataFrame(y_pred_final)
```

	Id	SalePrice
0	1	104990.090203
1	2	157182.090203
2	3	164342.090203
3	4	183484.090203
4	5	191310.090203
...
1455	1456	57438.090203
1456	1457	132985.436864
1457	1458	116621.009512
1458	1459	244657.009512
1459	1460	NaN

1460 rows × 2 columns

```
In [22]: 1 y_pred_final.to_csv('Output.csv', index=False)
```

```
In [ ]: 1
```