

```
from google.colab import drive

drive.mount('/content/drive')

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
import plotly.graph_objects as go

import warnings
warnings.filterwarnings('ignore')

df = pd.read_csv('/content/drive/MyDrive/MELBOURNE_HOUSE_PRICES_LESS (1).csv')
```

```
df.head()
```

| | Suburb | Address | Rooms | Type | Price | Method | SellerG | Date | Postcode | Reg |
|---|------------|---------------|-------|------|-----------|--------|----------|-----------|----------|-----|
| 0 | Abbotsford | 49 Lithgow St | 3 | h | 1490000.0 | S | Jellis | 1/04/2017 | 3067 | Met |
| 1 | Abbotsford | 59A Turner St | 3 | h | 1220000.0 | S | Marshall | 1/04/2017 | 3067 | Met |
| 2 | Abbotsford | 119B Yarra St | 3 | h | 1420000.0 | S | Nelson | 1/04/2017 | 3067 | Met |

```
df.tail()
```

| | Suburb | Address | Rooms | Type | Price | Method | SellerG | Date | Postcode |
|-------|---------------|-----------------|-------|------|----------|--------|---------|------------|----------|
| 63018 | Roxburgh Park | 3 Carr Pl | 3 | h | 566000.0 | S | Raine | 31/03/2018 | 3064 |
| 63019 | Roxburgh Park | 9 Parker Ct | 3 | h | 500000.0 | S | Raine | 31/03/2018 | 3064 |
| 63020 | Roxburgh Park | 5 Parkinson Ave | 3 | h | 545000.0 | S | Raine | 31/03/2018 | 3064 |

```
df.shape
```

(63023, 13)

```
df.columns
```

Index(['Suburb', 'Address', 'Rooms', 'Type', 'Price', 'Method', 'SellerG', 'Date', 'Postcode', 'Regionname', 'Propertycount', 'Distance', 'CouncilArea'], dtype='object')

```
df.duplicated().sum()
```

2

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

Suburb 0
Address 0
Rooms 0
Type 0
Price 14590
Method 0
SellerG 0
Date 0
Postcode 0
Regionname 0

```
Propertycount      0
Distance            0
CouncilArea        0
dtype: int64
```

df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 63023 entries, 0 to 63022
Data columns (total 13 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Suburb          63023 non-null object
1   Address         63023 non-null object
2   Rooms           63023 non-null int64
3   Type            63023 non-null object
4   Price           48433 non-null float64
5   Method          63023 non-null object
6   SellerG         63023 non-null object
7   Date            63023 non-null object
8   Postcode        63023 non-null int64
9   Regionname      63023 non-null object
10  Propertycount   63023 non-null int64
11  Distance        63023 non-null float64
12  CouncilArea     63023 non-null object
dtypes: float64(2), int64(3), object(8)
memory usage: 6.3+ MB
```

df.describe()

| | Rooms | Price | Postcode | Propertycount | Distance |
|-------|--------------|--------------|--------------|---------------|--------------|
| count | 63023.000000 | 4.843300e+04 | 63023.000000 | 63023.000000 | 63023.000000 |
| mean | 3.110595 | 9.978982e+05 | 3125.673897 | 7617.728131 | 12.684829 |
| std | 0.957551 | 5.934989e+05 | 125.626877 | 4424.423167 | 7.592015 |
| min | 1.000000 | 8.500000e+04 | 3000.000000 | 39.000000 | 0.000000 |
| 25% | 3.000000 | 6.200000e+05 | 3056.000000 | 4380.000000 | 7.000000 |
| 50% | 3.000000 | 8.300000e+05 | 3107.000000 | 6795.000000 | 11.400000 |
| 75% | 4.000000 | 1.220000e+06 | 3163.000000 | 10412.000000 | 16.700000 |
| max | 31.000000 | 1.120000e+07 | 3980.000000 | 21650.000000 | 64.100000 |

df.nunique()

```
Suburb      380
Address    57754
Rooms       14
Type        3
Price      3417
Method      9
SellerG     476
Date       112
Postcode   225
Regionname  8
Propertycount 368
Distance   180
CouncilArea 34
dtype: int64
```

```
#cols_to_fill_zero = ['Price']
#df[cols_to_fill_zero] = df[cols_to_fill_zero].fillna(0)
#df.isna().sum()
```

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

df.isna().sum()

```
Suburb      0
Address     0
Rooms       0
Type        0
Price       0
Method      0
```

```
SellerG      0
Date         0
Postcode     0
Regionname   0
Propertycount 0
Distance     0
CouncilArea  0
dtype: int64
```

```
df = pd.get_dummies(df, drop_first = True)
```

```
df.head()
```

| | Rooms | Price | Postcode | Propertycount | Distance | Suburb_Aberfeldie | Suburb_Airport_West |
|---|-------|-----------|----------|---------------|----------|-------------------|---------------------|
| 0 | 3 | 1490000.0 | 3067 | 4019 | 3.0 | 0 | 0 |
| 1 | 3 | 1220000.0 | 3067 | 4019 | 3.0 | 0 | 0 |
| 2 | 3 | 1420000.0 | 3067 | 4019 | 3.0 | 0 | 0 |
| 3 | 3 | 1515000.0 | 3040 | 1543 | 7.5 | 1 | 0 |
| 4 | 2 | 670000.0 | 3042 | 3464 | 10.4 | 0 | 0 |

5 rows × 8 columns

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

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 = 2)

from sklearn.linear_model import LinearRegression

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

reg.score(X_test, y_test)

reg.score(X_train, y_train)

from sklearn import linear_model
lasso_reg = linear_model.Lasso(alpha = 50, max_iter=100, tol=0.1)
lasso_reg.fit(X_train, y_train)

lasso_reg.score(X_test, y_test)

lasso_reg.score(X_train, y_train)

from sklearn.linear_model import Ridge
ridge_reg = Ridge(alpha = 50, max_iter=100, tol=0.1)
ridge_reg.fit(X_train, y_train)

ridge_reg.score(X_test, y_test)

ridge_reg.score(X_train, y_train)
```

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