20250608 01

June 8, 2025

```
[1]: # Load dataset
     import pandas as pd
     data = pd.read_csv('melb_data.csv')
     data.head()
[1]:
            Suburb
                              Address
                                       Rooms Type
                                                        Price Method SellerG \
        Abbotsford
                         85 Turner St
                                            2
                                                    1480000.0
                                                                    S
                                                                       Biggin
                                                 h
        Abbotsford
                                            2
     1
                      25 Bloomburg St
                                                    1035000.0
                                                                       Biggin
        Abbotsford
                         5 Charles St
                                            3
                                                                       Biggin
                                                    1465000.0
        Abbotsford
                    40 Federation La
     3
                                            3
                                                     850000.0
                                                                   ΡI
                                                                       Biggin
        Abbotsford
                          55a Park St
                                                    1600000.0
                                                                       Nelson
                                                           Landsize
                                                                      BuildingArea
             Date
                  Distance
                             Postcode
                                           Bathroom
                                                      Car
        3/12/2016
                                                               202.0
     0
                         2.5
                                3067.0
                                                 1.0
                                                      1.0
                                                                               NaN
       4/02/2016
                         2.5
                                3067.0
                                                 1.0
                                                     0.0
                                                               156.0
                                                                              79.0
     1
     2 4/03/2017
                         2.5
                                3067.0
                                                 2.0
                                                     0.0
                                                               134.0
                                                                             150.0
     3 4/03/2017
                                                 2.0
                                                      1.0
                         2.5
                                3067.0
                                                               94.0
                                                                               NaN
     4 4/06/2016
                         2.5
                                3067.0
                                                 1.0
                                                      2.0
                                                               120.0
                                                                             142.0
        YearBuilt
                   CouncilArea Lattitude
                                           Longtitude
                                                                    Regionname
     0
                                 -37.7996
                                              144.9984
              NaN
                          Yarra
                                                        Northern Metropolitan
     1
           1900.0
                                -37.8079
                                              144.9934
                                                        Northern Metropolitan
                          Yarra
     2
           1900.0
                                -37.8093
                                                        Northern Metropolitan
                          Yarra
                                              144.9944
     3
              NaN
                          Yarra
                                -37.7969
                                              144.9969
                                                        Northern Metropolitan
     4
                                -37.8072
                                                        Northern Metropolitan
           2014.0
                          Yarra
                                              144.9941
       Propertycount
     0
              4019.0
     1
              4019.0
     2
              4019.0
     3
              4019.0
     4
              4019.0
     [5 rows x 21 columns]
```

```
[2]: # Set features and target
     X = data.drop('Price', axis = 1)
     y = data['Price']
[3]: # Split features into numeric and categorical
     # Add .columns to only return the col names, otherwise it will return the whole,
      \rightarrow dataset
     num_cols = X.select_dtypes(include = ['int64', 'float64']).columns
     cat_cols = X.select_dtypes(include = ['object']).columns
[4]: # Handle missing values
     from sklearn.impute import SimpleImputer
     # Initialize the imputer with mean
     imputer_num = SimpleImputer(strategy = 'mean')
     # Fit and transform the numeric data
     X_num = pd.DataFrame(imputer_num.fit_transform(X[num_cols]), columns = num_cols)
     # Create dummy variables for categorical columns, including NaNs
     X_cat = pd.get_dummies(X[cat_cols], dummy_na = True)
[5]: # Put them back together
     X_processed = pd.concat([X_num, X_cat], axis = 1)
[6]: # Now we try the clean dataset
     from sklearn.model_selection import train_test_split
     from sklearn.linear_model import LinearRegression
     # Split data
     X_train, X_test, y_train, y_test = train_test_split(X_processed, y,_
      ⇒random state = 42)
     # Train model
     model = LinearRegression()
     model.fit(X_train, y_train)
     # Evaluate, which is default to using R2
     print("Model R<sup>2</sup> Score:", model.score(X_test, y_test))
    Model R<sup>2</sup> Score: -2.0371647495946337
[8]: # Oof, now we drop some cols
     high_card_cols = [col for col in cat_cols if X[col].nunique() > 50]
     X = X.drop(columns = high_card_cols)
     cat_cols = [col for col in cat_cols if col not in high_card_cols]
     X_cat = pd.get_dummies(X[cat_cols], dummy_na = True)
```

```
X_processed = pd.concat([X_num, X_cat], axis = 1)

X_train, X_test, y_train, y_test = train_test_split(X_processed, y,u_arandom_state = 42)

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

# Evaluate, which is default to using R2
print("Model R2 Score:", model_retry.score(X_test, y_test))
```

Model R² Score: 0.6508737168099696