

# real-estate

May 22, 2024

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[2]: import pandas as pd

# Load the dataset
data_path = './marketing.csv'
df = pd.read_csv(data_path)

# Display the first five rows of the dataframe
df.head()
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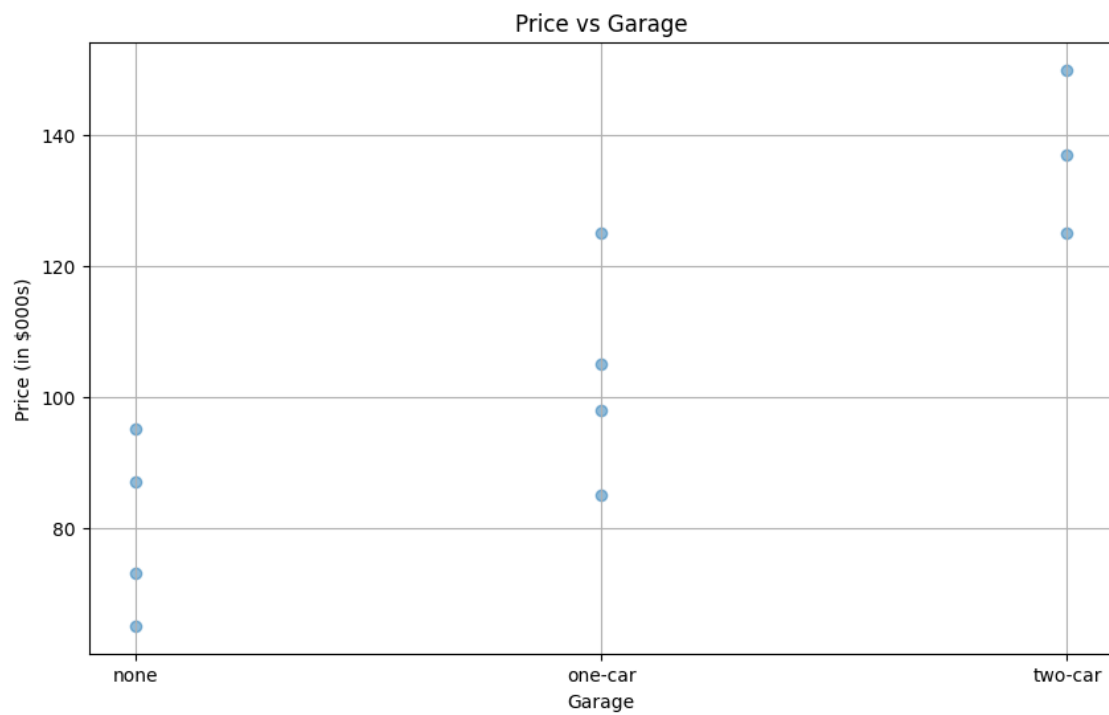
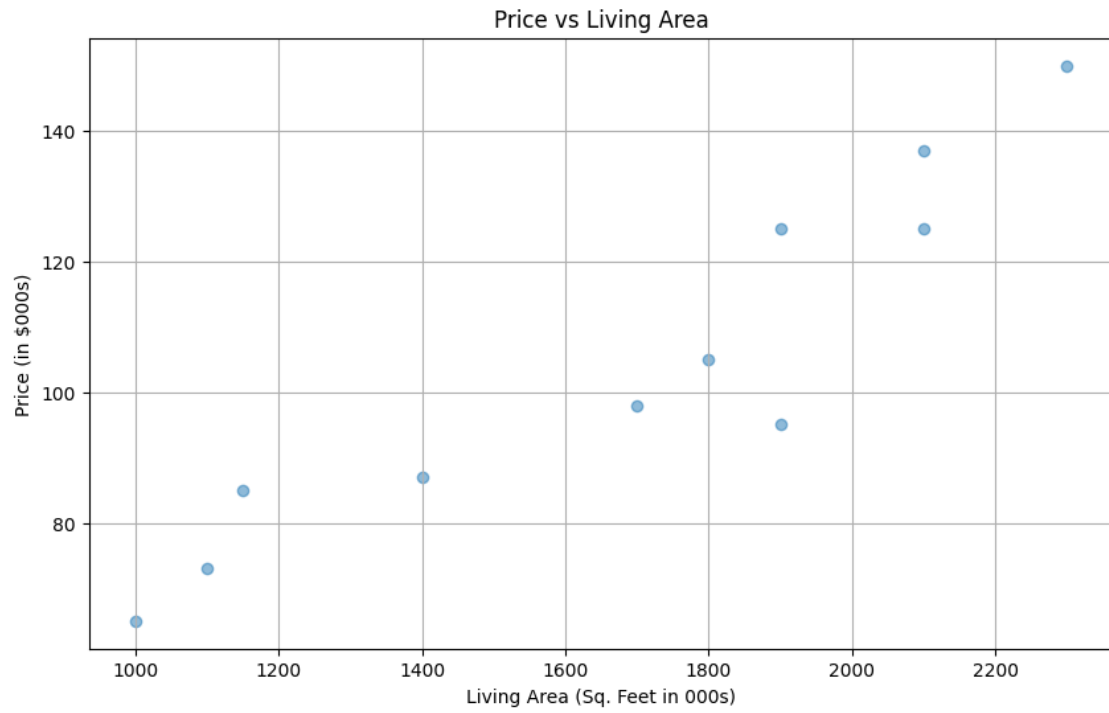
```
[2]:
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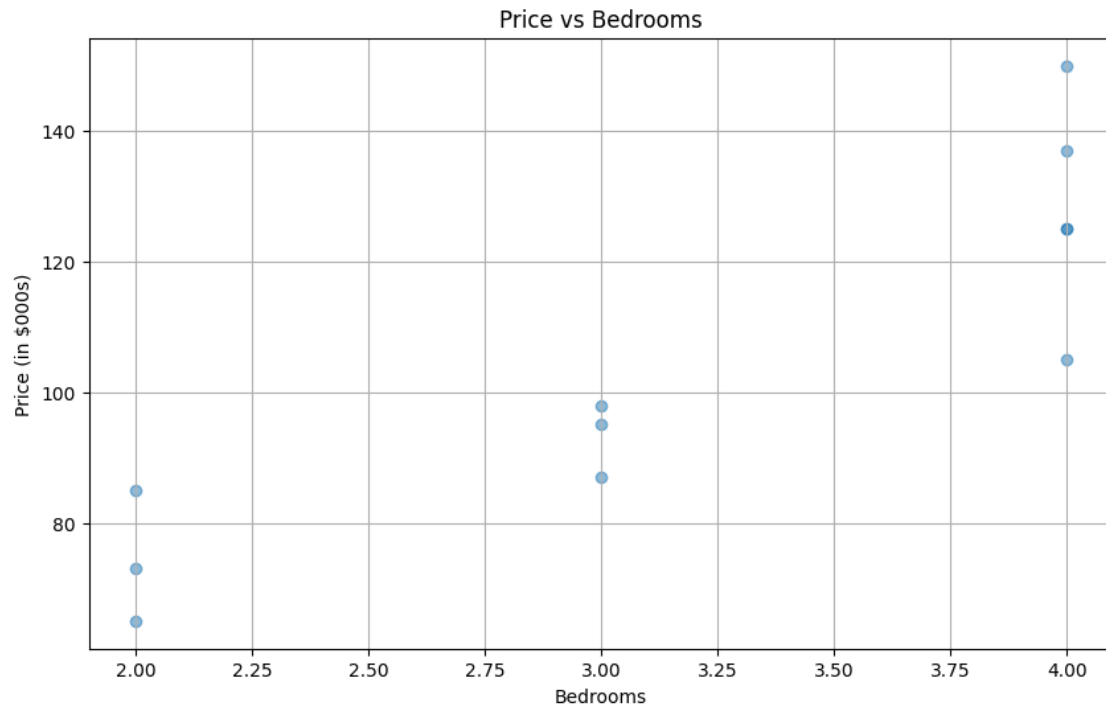
|   | Sq. Feet (in 000s) | Garage  | Bedrooms | Price (in \$000s) |
|---|--------------------|---------|----------|-------------------|
| 0 | 1000               | none    | 2        | 65                |
| 1 | 1100               | none    | 2        | 73                |
| 2 | 1150               | one-car | 2        | 85                |
| 3 | 1400               | none    | 3        | 87                |
| 4 | 1700               | one-car | 3        | 98                |

```
[3]: import matplotlib.pyplot as plt

# Scatterplot for Price vs Living Area
plt.figure(figsize=(10, 6))
plt.scatter(df['Sq. Feet (in 000s)'], df['Price (in $000s)'], alpha=0.5)
plt.title('Price vs Living Area')
plt.xlabel('Living Area (Sq. Feet in 000s)')
plt.ylabel('Price (in $000s)')
plt.grid(True)
plt.show()

# Scatterplots for Price vs other attributes
for column in df.columns:
    if column != 'Price (in $000s)' and column != 'Sq. Feet (in 000s)':
        plt.figure(figsize=(10, 6))
        plt.scatter(df[column], df['Price (in $000s)'], alpha=0.5)
        plt.title(f'Price vs {column}')
        plt.xlabel(column)
        plt.ylabel('Price (in $000s)')
        plt.grid(True)
        plt.show()
```





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[4]: from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression

# Define the feature (Living Area) and target (Price)
X = df[['Sq. Feet (in 000s)']]
y = df['Price (in $000s)']

# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random_state=42)

# Create and fit the linear regression model
model = LinearRegression()
model.fit(X_train, y_train)

# Display the model coefficients
print(f'Intercept: {model.intercept_}')
print(f'Coefficient: {model.coef_[0]}')
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Intercept: 13.826900855095985
Coefficient: 0.05368153455049684
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[5]: # Calculate the R2 score
r_squared = model.score(X_test, y_test)
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print(f'Coefficient of Determination (R^2): {r_squared}')
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Coefficient of Determination (R^2): 0.9442619843108755

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[8]: # Example prediction
living_area_example = pd.DataFrame({'Sq. Feet (in 000s)': [2.0]}) # Create a
    ↳ DataFrame with the same column name
predicted_price = model.predict(living_area_example)
print(f'Predicted Price for {living_area_example["Sq. Feet (in 000s)"][0]}000
    ↳ sq ft: ${predicted_price[0]:.2f}000')
```

Predicted Price for 2.0000 sq ft: \$13.93000

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[7]: import numpy as np
from sklearn.metrics import mean_squared_error

# Calculate predictions on the test set
y_pred = model.predict(X_test)

# Calculate the Mean Squared Error
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
print(f'Mean Squared Error (MSE): {mse}')
print(f'Root Mean Squared Error (RMSE): {rmse}')

# Summary of the model quality
print(f'The R^2 value of {r_squared} indicates that the model explains
    ↳ {r_squared*100:.2f}% of the variance in the selling price based on the
    ↳ living area.')
print(f'The RMSE value of {rmse} provides an estimate of the average prediction
    ↳ error in thousands of dollars.')
```

Mean Squared Error (MSE): 48.35582516674265

Root Mean Squared Error (RMSE): 6.953835284700282

The R^2 value of 0.9442619843108755 indicates that the model explains 94.43% of the variance in the selling price based on the living area.

The RMSE value of 6.953835284700282 provides an estimate of the average prediction error in thousands of dollars.