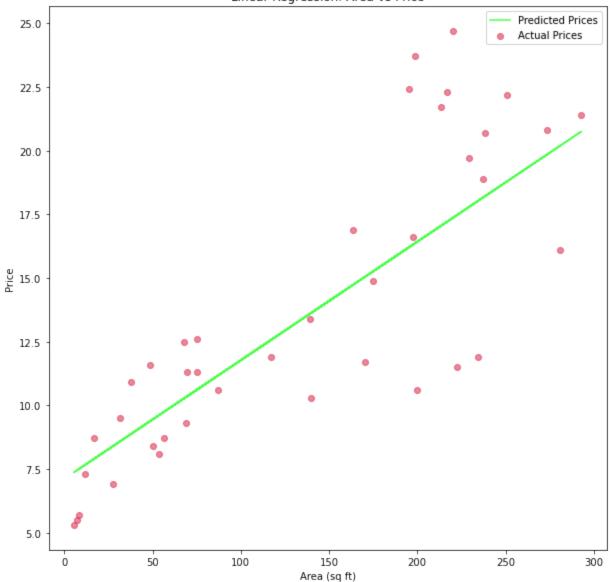
```
import pandas as pd
import numpy as np
import mglearn
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score
```

```
Linear Regression
In [2]:
         #Load the dataset
         df = pd.read_csv('Advertising Budget and Sales.csv')
         df = df.drop(df.columns[0], axis=1)
In [3]:
         df.head()
           TV Ad Budget ($) Radio Ad Budget ($) Newspaper Ad Budget ($) Sales ($)
Out[3]:
         0
                      230.1
                                         37.8
                                                                69.2
                                                                         22.1
         1
                      44.5
                                         39.3
                                                                45.1
                                                                         10.4
         2
                                         45.9
                      17.2
                                                                69.3
                                                                          9.3
         3
                      151.5
                                                                58.5
                                                                         18.5
                                         41.3
                      180.8
                                         10.8
                                                                58.4
                                                                         12.9
In [4]:
         df = df.rename(columns={
              'TV Ad Budget ($)' : 'TV_ads_budget',
              'Radio Ad Budget ($)' : 'Ra_ads_budget',
              'Newspaper Ad Budget ($)' : 'Ne_ads_budget',
              'Sales ($)' : 'Sales'
         })
In [5]:
         #Data preprocessing
         X = df[['TV_ads_budget']] # Independent variable(s)
         y = df['Sales'] # Dependent variable (target)
In [6]:
         #Splitting 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=4
In [7]:
         #Fitting the Linear Regression model
```

```
model = LinearRegression()
          model.fit(X_train, y_train)
Out[7]: LinearRegression()
 In [8]:
          #Making predictions and evaluating the model
          y_pred = model.predict(X_test)
          mse = mean_squared_error(y_test, y_pred)
          r2 = r2_score(y_test, y_pred)
 In [9]:
          print(f"Mean Squared Error: {mse:.3f}")
          print(f"R2 Score: {r2:.3f}")
         Mean Squared Error: 10.205
         R<sup>2</sup> Score: 0.677
In [10]:
          #Visualizing the Results
          plt.figure(figsize=(10,10))
          plt.scatter(X_test, y_test, color='#DC143C', label='Actual Prices', alpha = 0.5)
          plt.plot(X_test, y_pred, color='#00FF00', label='Predicted Prices', alpha = 0.7)
          plt.xlabel('Area (sq ft)')
          plt.ylabel('Price')
          plt.title('Linear Regression: Area vs Price')
          plt.legend()
          plt.show()
```





Multiple Linear Regression

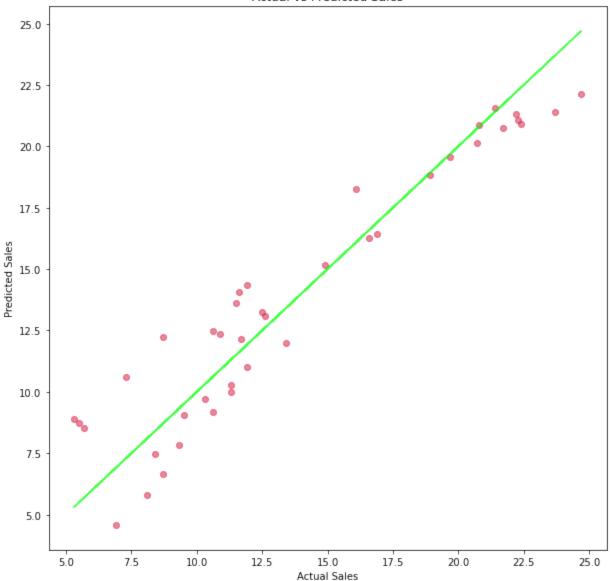
```
In [11]: #Define features and target variable
    X = df[['TV_ads_budget', 'Ra_ads_budget', 'Ne_ads_budget']] # Independent variable(s)
    y = df['Sales'] # Dependent variable (target)

In [12]: #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=4)

In [13]: #Fit a multiple linear regression model
    model = LinearRegression()
    model.fit(X_train, y_train)
```

```
Out[13]: LinearRegression()
In [14]:
          #Make predictions
          y_pred = model.predict(X_test)
In [15]:
          #Evaluate the model
          mse = mean_squared_error(y_test, y_pred)
          r2 = r2_score(y_test, y_pred)
In [16]:
          print(f"Mean Squared Error: {mse:.3f}")
          print(f"R2 Score: {r2:.3f}")
         Mean Squared Error: 3.174
         R<sup>2</sup> Score: 0.899
In [17]:
          # Visualizing Actual vs Predicted
          plt.figure(figsize=(10, 10))
          plt.scatter(y_test, y_pred, color='#DC143C', alpha = 0.5)
          plt.plot(y_test, y_test, color='#00FF00', alpha = 0.7)
          plt.xlabel('Actual Sales')
          plt.ylabel('Predicted Sales')
          plt.title('Actual vs Predicted Sales')
          plt.show()
```

Actual vs Predicted Sales



Multiple Linear Regression (L2 & L1)

```
In [18]: from sklearn.linear_model import Ridge, Lasso
    from sklearn.model_selection import train_test_split

In [19]: #Initialize Ridge and Lasso regression models
    ridge_model = Ridge(alpha=100)
    lasso_model = Lasso(alpha=1)

In [20]: #Fit both models to the training data
    ridge_model.fit(X_train, y_train)
    lasso_model.fit(X_train, y_train)
Out[20]: Lasso(alpha=1)
```

```
In [21]:
          #Make predictions with both models
          ridge pred = ridge model.predict(X test)
          lasso_pred = lasso_model.predict(X_test)
In [22]:
          #Evaluate both models using Mean Squared Error and R<sup>2</sup> Score
          ridge_mse = mean_squared_error(y_test, ridge_pred)
          lasso_mse = mean_squared_error(y_test, lasso_pred)
          ridge_r2 = r2_score(y_test, ridge_pred)
          lasso_r2 = r2_score(y_test, lasso_pred)
In [23]:
          print(f"Ridge Regression - Mean Squared Error: {ridge_mse:.3f}, R2 Score: {ridge_r2:.3f
          print(f"Lasso Regression - Mean Squared Error: {lasso_mse:.3f}, R2 Score: {lasso_r2:.3f
         Ridge Regression - Mean Squared Error: 3.174, R<sup>2</sup> Score: 0.899
         Lasso Regression - Mean Squared Error: 3.144, R<sup>2</sup> Score: 0.900
         KNN Regression
In [24]:
          from sklearn.neighbors import KNeighborsRegressor
In [25]:
          #Define features and target variable
          X = df[['TV_ads_budget', 'Ra_ads_budget', 'Ne_ads_budget']] # Independent variable(s)
                           # Dependent variable (target)
In [26]:
          #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=4
In [27]:
          #Fit a KNN regression model
          knn_model = KNeighborsRegressor(n_neighbors=k)
          knn_model.fit(X_train, y_train)
Out[27]: KNeighborsRegressor()
In [28]:
          #Make predictions
          y_pred = knn_model.predict(X_test)
In [29]:
          #Evaluate the model
          mse = mean_squared_error(y_test, y_pred)
          r2 = r2_score(y_test, y_pred)
```

```
In [30]: print(f"Mean Squared Error: {mse:.3f}")
    print(f"R<sup>2</sup> Score: {r2:.3f}")

Mean Squared Error: 2.821
    R<sup>2</sup> Score: 0.911
```

Decision Tree Regression

```
In [31]:
          from sklearn.tree import DecisionTreeRegressor
In [32]:
          X = df[['TV_ads_budget', 'Ra_ads_budget', 'Ne_ads_budget']] # Independent variable(s)
          y = df['Sales']
                           # Dependent variable (target)
In [33]:
          X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=4
In [34]:
          model = DecisionTreeRegressor(random_state=42)
          model.fit(X_train, y_train)
Out[34]: DecisionTreeRegressor(random_state=42)
In [35]:
          y_pred = model.predict(X_test)
In [36]:
          mse = mean_squared_error(y_test, y_pred)
          r2 = r2_score(y_test, y_pred)
In [37]:
          print(f"Mean Squared Error: {mse:.3f}")
          print(f"R2 Score: {r2:.3f}")
         Mean Squared Error: 2.175
         R<sup>2</sup> Score: 0.931
```

Random Forest Regression

```
In [38]: from sklearn.ensemble import RandomForestRegressor

In [39]: X = df[['TV_ads_budget', 'Ra_ads_budget', 'Ne_ads_budget']] # Independent variable(s)
y = df['Sales'] # Dependent variable (target)

In [40]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=4)
In [41]: rf_model = RandomForestRegressor(n_estimators=100, random_state=42)
rf_model.fit(X_train, y_train)
```

```
RandomForestRegressor(random_state=42)
Out[41]:
In [42]:
           y_pred = rf_model.predict(X test)
In [43]:
           mse = mean_squared_error(y_test, y_pred)
           r2 = r2_score(y_test, y_pred)
In [44]:
           print(f"Mean Squared Error: {mse:.3f}")
           print(f"R2 Score: {r2:.3f}")
          Mean Squared Error: 0.591
          R<sup>2</sup> Score: 0.981
         Summary
In [45]:
           summary = {
               'Linear Regression' : [10.205, 0.677],
               'Multiple Linear Regression' : [3.174, 0.899],
               'Ridge Multiple Linear Regression' : [3.174, 0.899],
               'Lasso Multiple Linear Regression' : [3.144, 0.900],
               'KNN Multiple Regression' : [2.821, 0.911],
               'Decision Tree Regression': [2.175, 0.931],
               'Randomo Forest Regression' : [0.591, 0.981]
           }
In [46]:
           df_summary = pd.DataFrame.from_dict(summary, orient='index', columns=['MSE','R2 Score']
In [47]:
           df summary.reset_index(inplace=True)
           df_summary.rename(columns={'index': 'Model'})
                                  Model
                                           MSE R<sup>2</sup> Score
Out[47]:
          0
                         Linear Regression 10.205
                                                   0.677
          1
                  Multiple Linear Regression
                                          3.174
                                                   0.899
             Ridge Multiple Linear Regression
                                          3.174
                                                   0.899
             Lasso Multiple Linear Regression
                                          3.144
                                                   0.900
          4
                   KNN Multiple Regression
                                          2.821
                                                   0.911
          5
                    Decision Tree Regression
                                          2.175
                                                   0.931
```

Randomo Forest Regression

0.591

0.981

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