

```

#decision tree
import matplotlib.pyplot as plt
from sklearn.tree import DecisionTreeRegressor
from sklearn.model_selection import train_test_split
import pandas as pd
import numpy as np
from sklearn import tree

# Load the data
data = pd.read_csv('C:/Users/HP/Downloads/Real estate - Real estate.csv')

# Drop unnecessary columns and remove rows with missing values
df = data.drop(["No", "X1 transaction date", "X5 latitude", "X6 longitude"],
axis=1).dropna()

# Remove duplicate rows
df = df.drop_duplicates()

# Split the data into input features (X) and target variable (Y)
X = df[['X3 distance to the nearest MRT station']]
Y = df['Y house price of unit area']

# 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=0)

# Create and fit the DecisionTreeRegressor
regressor = DecisionTreeRegressor(random_state=0)
regressor.fit(X_train, Y_train)

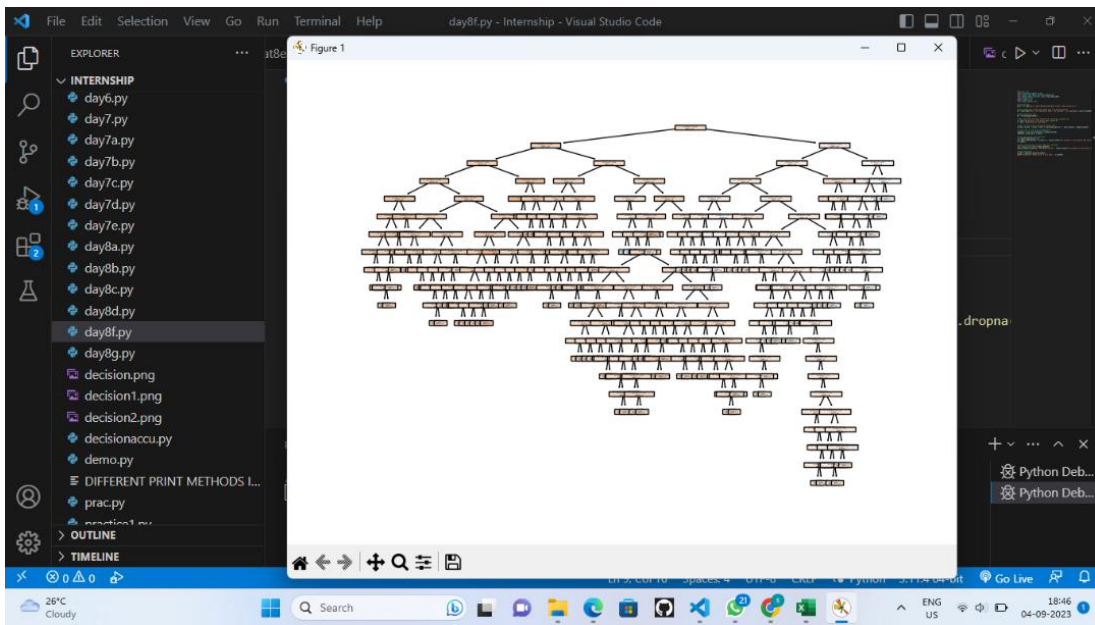
# Plot the decision tree (optional)
plt.figure(figsize=(10, 6))
tree.plot_tree(regressor, filled=True, feature_names=['X3 distance to the nearest MRT
station'])
plt.show()

# Export the decision tree to a DOT file (optional)
from sklearn.tree import export_graphviz
export_graphviz(regressor, out_file='tree.dot', feature_names=['X3 distance to the nearest
MRT station'])

# Make predictions
y_pred = regressor.predict([[390]])
print("Predicted Y house price of unit area:", y_pred[0])

```

screenshot:



```
File Edit Selection View Go Run Terminal Help day8f.py - Internship - Visual Studio Code

EXPLORER
INTERNSHIP
  day6.py
  day7.py
  day7a.py
  day7b.py
  day7c.py
  day7d.py
  day7e.py
  day8a.py
  day8b.py
  day8c.py
  day8d.py
  day8f.py
  day8g.py
  decision.png
  decision1.png
  decision2.png
  decisionaccu.py
  demo.py
  DIFFERENT PRINT METHODS L...
  prac.py
  practical1.py
OUTLINE
TIMELINE

day8f.py > ...
1 #decision tree
2 import matplotlib.pyplot as plt
3 from sklearn.tree import DecisionTreeRegressor
4 from sklearn.model_selection import train_test_split
5 import pandas as pd
6 import numpy as np
7 from sklearn import tree
8
9 # Load the data
10 data = pd.read_csv('C:/Users/HP/Downloads/Real estate - Real estate.csv')
11
12 # Drop unnecessary columns and remove rows with missing values
13 df = data.drop(["No", "X1 transaction date", "X5 latitude", "X6 longitude"], axis=1).dropna()
14
15 # Remove duplicate rows
16 df = df.drop_duplicates()
17
18 # Split the data into input features (X) and target variable (Y)
```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL

```
./...debugpy\launcher' '59092' '--' 'C:\Users\HP\Desktop\Internship\day8f.py'
C:\Users\HP\AppData\Roaming\Python\Python311\site-packages\sklearn\base
does not have valid feature names, but DecisionTreeRegressor was fitte
warnings.warn(
Predicted Y house price of unit area: 40.6
PS C:\Users\HP\Desktop\Internship>

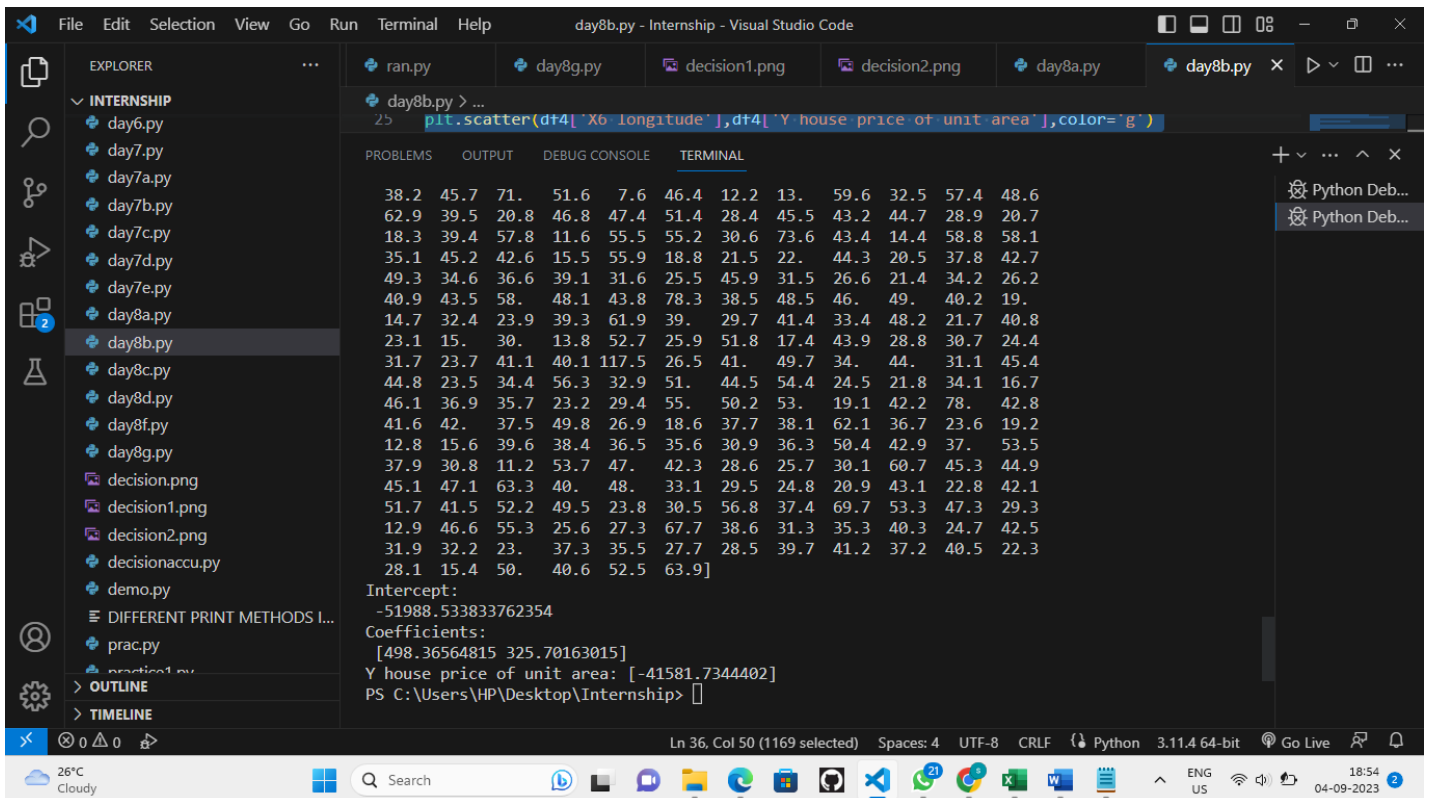
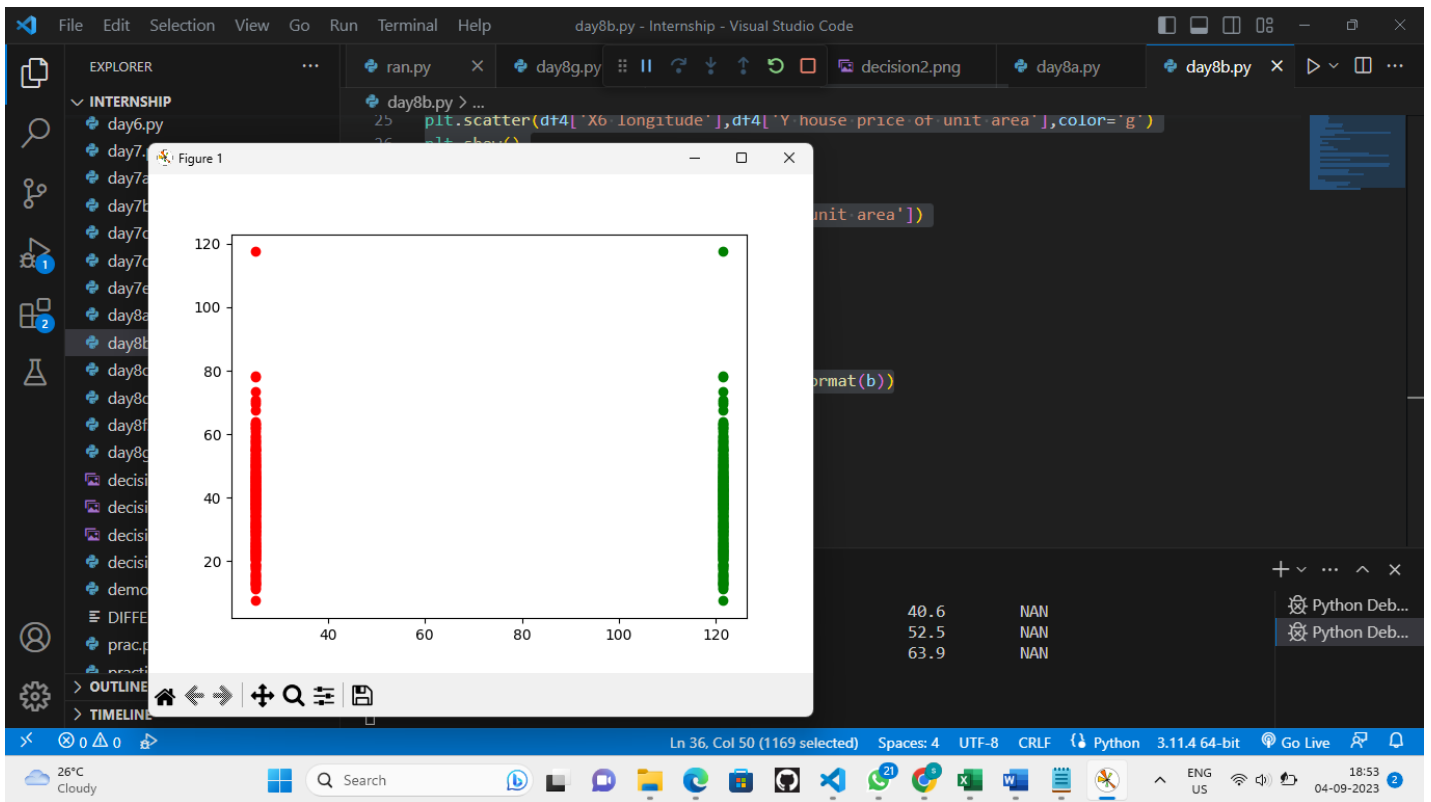
Battery saver
Battery saver is on
Consider plugging in your device.
```

Visual Studio Code interface showing the code for day8f.py. The Explorer panel on the left lists files under 'INTERNSHIP'. The main editor displays the Python code for a decision tree. The bottom status bar shows '26°C Cloudy' and the date '04-09-2023'. A 'Battery saver' notification is visible in the bottom right corner.

```

#multilinear model
import pandas as pd
import numpy as np
import numpy as np
from sklearn import linear_model
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings("ignore")
data = pd.read_csv(r"C:/Users/HP/Downloads/Real estate - Real estate (1).csv")
df1 = pd.DataFrame(data)
print(df1)
df2 = df1.drop(["Unnamed: 8"],axis = 1)
df2.isnull()
df3 = df2.dropna()
print(df3)
print(df3.duplicated())
df3 = df3.drop_duplicates()
print(df3)
df4=df3.duplicated(subset=['Y house price of unit area'])
df4 = df3.drop_duplicates(subset=['Y house price of unit area'],keep='last')
print(df4)
df4.reset_index(inplace=True,drop=True)
print(df4)
plt.scatter(df4['X5 latitude'],df4['Y house price of unit area'],color='red')
plt.scatter(df4['X6 longitude'],df4['Y house price of unit area'],color='g')
plt.show()
regr = linear_model.LinearRegression()
x=df4[['X5 latitude','X6 longitude']]
y = np.asanyarray(df4['Y house price of unit area'])
print(x)
print(y)
Y=regr.fit(x,y)
print('Intercept: \n', regr.intercept_)
print('Coefficients: \n', regr.coef_)
b = regr.predict([[15,9]])
print("Y house price of unit area: {}".format(b))

```



```
#linear model
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_absolute_error, mean_squared_error
from math import sqrt

# Load the dataset
data = pd.read_csv(r"C:\Users\HP\Downloads\Real estate - Real estate (1).csv")
```

```
df1 = pd.DataFrame(data)

# Drop the "Unnamed: 8" column
df1 = df1.drop(["Unnamed: 8"], axis=1)

# Remove rows with missing values
df2 = df1.dropna()

# Remove duplicate rows
df3 = df2.drop_duplicates()

# Reset the index
df3.reset_index(inplace=True, drop=True)

# Plot the data
plt.scatter(df3['X2 house age'], df3['Y house price of unit area'], color='red')
plt.xlabel('X2 house age')
plt.ylabel('Y house price of unit area')
plt.show()

# Define the features (X) and target (y)
X = df3[['X2 house age']]
y = df3['Y house price of unit area']

# 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 a linear regression model
regr = LinearRegression()

# Train the model
regr.fit(x_train, y_train)

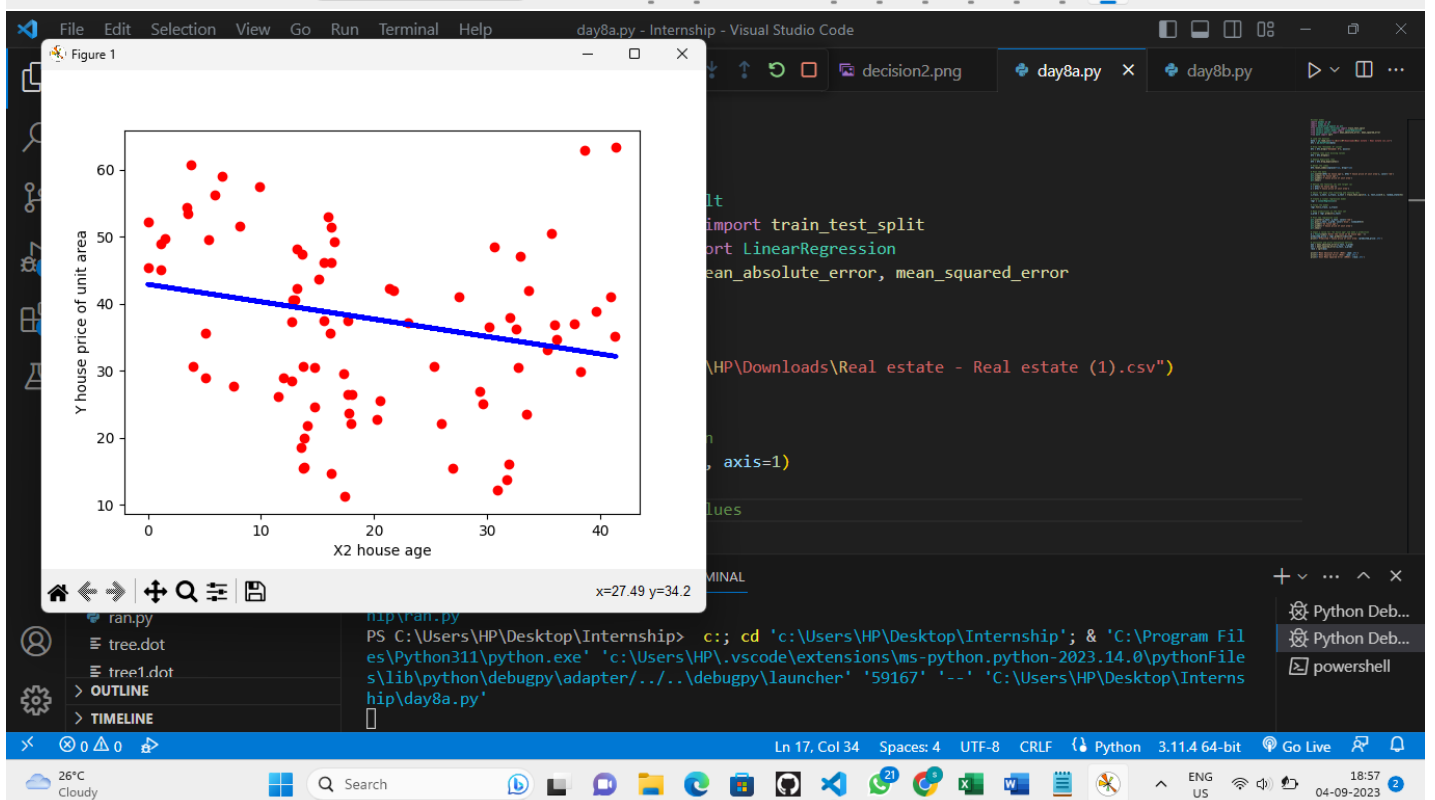
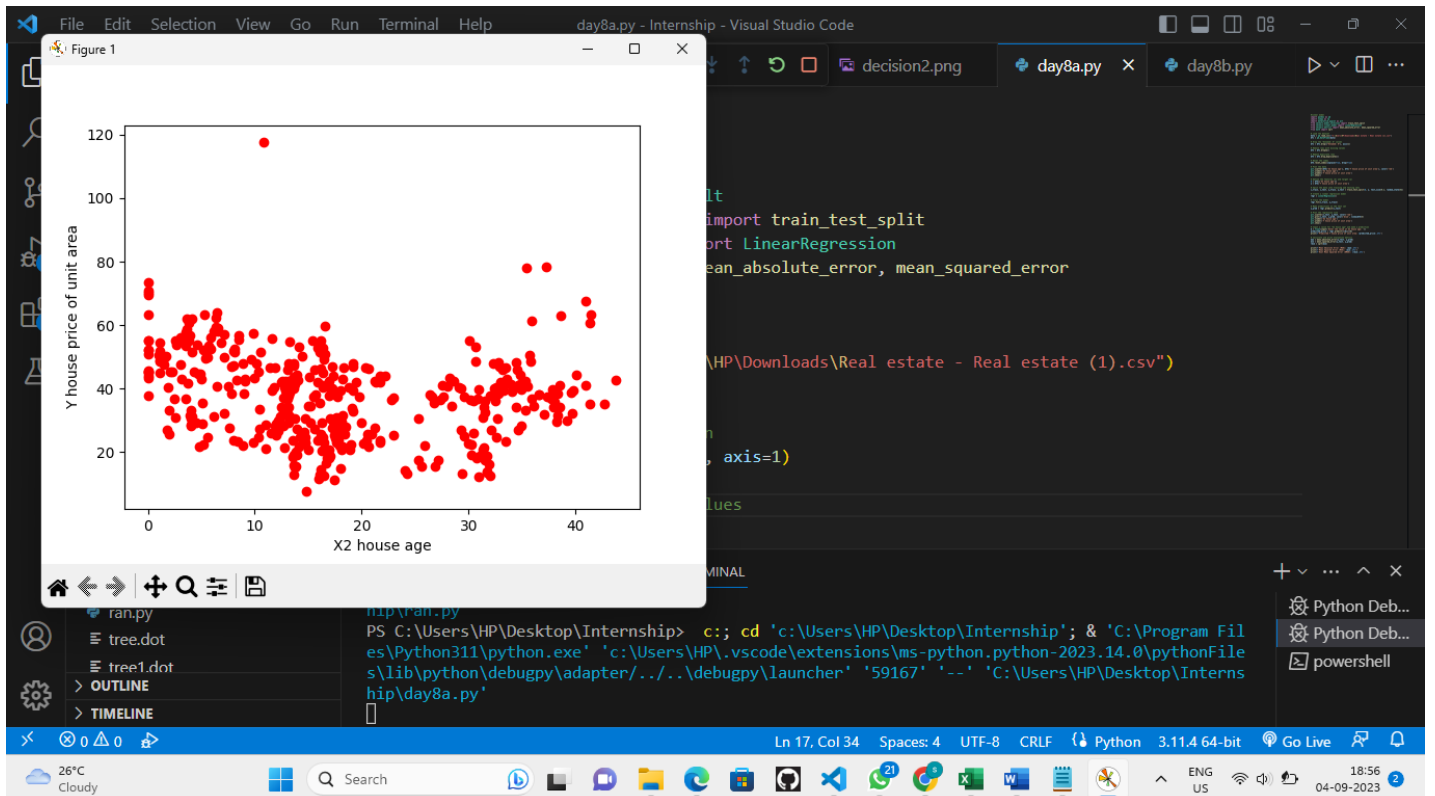
# Make predictions on the test set
y_pred = regr.predict(x_test)

# Plot the regression line
plt.scatter(x_test, y_test, color='red')
plt.plot(x_test, y_pred, color='blue', linewidth=3)
plt.xlabel('X2 house age')
plt.ylabel('Y house price of unit area')
plt.show()

# Input a value for "X2 house age" and make a prediction
d = float(input('Enter the value of X2 house age: '))
predicted_price = regr.predict([[d]])[0]
print(f'Predicted Y house price of unit area: {predicted_price:.2f}')

# Calculate and print evaluation metrics
mae = mean_absolute_error(y_test, y_pred)
mse = mean_squared_error(y_test, y_pred)
rmse = sqrt(mse)

print(f'Mean Absolute Error (MAE): {mae:.2f}')
print(f'Mean Squared Error (MSE): {mse:.2f}')
print(f'Root Mean Squared Error (RMSE): {rmse:.2f}')
```



The screenshot shows the Visual Studio Code interface with a file explorer on the left containing files like `day8a.py`, `day8b.py`, and `decision.png`. The main editor displays `day8a.py` with the following code:

```
56 d = float(input('Enter the value of X2 house age: '))
57 predicted_price = regr.predict([[d]])[0]
58 print(f'Predicted Y house price of unit area: {predicted_price:.2f}')
59
60 # Calculate and print evaluation metrics
61 mae = mean_absolute_error(y_test, y_pred)
62 mse = mean_squared_error(y_test, y_pred)
63 rmse = sqrt(mse)
64
65 print(f'Mean Absolute Error (MAE): {mae:.2f}')
66 print(f'Mean Squared Error (MSE): {mse:.2f}')
67 print(f'Root Mean Squared Error (RMSE): {rmse:.2f}')
68
```

The bottom panel shows the terminal output:

```
warnings.warn(
Predicted Y house price of unit area: 33.53
Mean Absolute Error (MAE): 10.62
Mean Squared Error (MSE): 165.21
Root Mean Squared Error (RMSE): 12.85
PS C:\Users\HP\Desktop\Internship>
```

```
import pandas as pd
import numpy as np
from sklearn import linear_model
from sklearn.tree import DecisionTreeRegressor
from sklearn.model_selection import train_test_split
from sklearn.metrics import r2_score, mean_squared_error

# Load the data
data = pd.read_csv('C:/Users/HP/Downloads/Real estate - Real estate.csv')

# Drop unnecessary columns and remove rows with missing values
df = data.drop(["No", "X1 transaction date", "X5 latitude", "X6 longitude"],
axis=1).dropna()

# Remove duplicate rows
df = df.drop_duplicates()

# Split the data into input features (X) and target variable (Y)
X_linear = df[['X3 distance to the nearest MRT station']]
X_multilinear = df[['X3 distance to the nearest MRT station', 'X4 number of convenience
stores']]
X_tree = df[['X3 distance to the nearest MRT station']]
Y = df['Y house price of unit area']

# Split the data into training and testing sets
X_linear_train, X_linear_test, Y_train, Y_test = train_test_split(X_linear, Y,
test_size=0.2, random_state=0)
X_multilinear_train, X_multilinear_test, _, _ = train_test_split(X_multilinear, Y,
test_size=0.2, random_state=0)
X_tree_train, X_tree_test, _, _ = train_test_split(X_tree, Y, test_size=0.2, random_state=0)
```

```
# Create and fit the Linear Regression model
regr_linear = linear_model.LinearRegression()
regr_linear.fit(X_linear_train, Y_train)

# Make predictions using Linear Regression
y_pred_linear = regr_linear.predict(X_linear_test)

# Calculate R-squared and MSE for Linear Regression
r2_linear = r2_score(Y_test, y_pred_linear)
mse_linear = mean_squared_error(Y_test, y_pred_linear)

# Create and fit the Multilinear Regression model
regr_multilinear = linear_model.LinearRegression()
regr_multilinear.fit(X_multilinear_train, Y_train)

# Make predictions using Multilinear Regression
y_pred_multilinear = regr_multilinear.predict(X_multilinear_test)

# Calculate R-squared and MSE for Multilinear Regression
r2_multilinear = r2_score(Y_test, y_pred_multilinear)
mse_multilinear = mean_squared_error(Y_test, y_pred_multilinear)

# Create and fit the Decision Tree Regressor model
regressor_tree = DecisionTreeRegressor(random_state=0)
regressor_tree.fit(X_tree_train, Y_train)

# Make predictions using Decision Tree Regressor
y_pred_tree = regressor_tree.predict(X_tree_test)

# Calculate R-squared and MSE for Decision Tree Regressor
r2_tree = r2_score(Y_test, y_pred_tree)
mse_tree = mean_squared_error(Y_test, y_pred_tree)

# Create an accuracy table
accuracy_table = pd.DataFrame({
    'Model': ['Simple Linear Regression', 'Multilinear Regression', 'Decision Tree
Regressor'],
    'R-squared': [r2_linear, r2_multilinear, r2_tree],
    'MSE': [mse_linear, mse_multilinear, mse_tree]
})

print(accuracy_table)
```



```
57 # Calculate R-squared and MSE for Decision Tree Regressor
58 r2_tree = r2_score(Y_test, y_pred_tree)
59 mse_tree = mean_squared_error(Y_test, y_pred_tree)
60
61 # Create an accuracy table
62 accuracy_table = pd.DataFrame({
63     'Model': ['Simple Linear Regression', 'Multilinear Regression', 'Decision Tree Regression'],
64     'R-squared': [r2_linear, r2_multilinear, r2_tree],
65     'MSE': [mse_linear, mse_multilinear, mse_tree]
66 })
67
68 print(accuracy_table)
```

	Model	R-squared	MSE
0	Simple Linear Regression	0.515951	84.078815
1	Multilinear Regression	0.571179	74.485765
2	Decision Tree Regressor	0.333552	115.761375

Based on R-squared values, the Multilinear Regression model (0.571179) has the highest R-squared, which suggests it explains more variance in the data compared to the other models. Therefore, it seems to be the most accurate among the three models in terms of explaining the variation in the dependent variable.

Based on MSE values, the Multilinear Regression model also has the lowest MSE (74.485765), which indicates it has the smallest error in predicting the target variable compared to the other models. So, it is the most accurate in terms of minimizing prediction error.

In conclusion, based on both R-squared and MSE, the Multilinear Regression model appears to be the most accurate among the three models provided.