- Linear Regression and Boston Data Set
 - · Load Boston Data Set
 - · Assign data to X and target to y

import pandas as pd
df= pd.read_csv("/content/boston_house_prices.csv")

df.head()

	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	PTRATIO	В	LSTAT	MEDV	
0	0.00632	18.0	2.31	0	0.538	6.575	65.2	4.0900	1	296	15.3	396.90	4.98	24.0	th
1	0.02731	0.0	7.07	0	0.469	6.421	78.9	4.9671	2	242	17.8	396.90	9.14	21.6	
2	0.02729	0.0	7.07	0	0.469	7.185	61.1	4.9671	2	242	17.8	392.83	4.03	34.7	
3	0.03237	0.0	2.18	0	0.458	6.998	45.8	6.0622	3	222	18.7	394.63	2.94	33.4	
4	0.06905	0.0	2.18	0	0.458	7.147	54.2	6.0622	3	222	18.7	396.90	5.33	36.2	
	1 2 3	0 0.006321 0.02731	 0 0.00632 18.0 1 0.02731 0.0 2 0.02729 0.0 3 0.03237 0.0 	0 0.00632 18.0 2.31 1 0.02731 0.0 7.07 2 0.02729 0.0 7.07 3 0.03237 0.0 2.18	0 0.00632 18.0 2.31 0 1 0.02731 0.0 7.07 0 2 0.02729 0.0 7.07 0 3 0.03237 0.0 2.18 0	0 0.00632 18.0 2.31 0 0.538 1 0.02731 0.0 7.07 0 0.469 2 0.02729 0.0 7.07 0 0.469 3 0.03237 0.0 2.18 0 0.458	0 0.00632 18.0 2.31 0 0.538 6.575 1 0.02731 0.0 7.07 0 0.469 6.421 2 0.02729 0.0 7.07 0 0.469 7.185 3 0.03237 0.0 2.18 0 0.458 6.998	0 0.00632 18.0 2.31 0 0.538 6.575 65.2 1 0.02731 0.0 7.07 0 0.469 6.421 78.9 2 0.02729 0.0 7.07 0 0.469 7.185 61.1 3 0.03237 0.0 2.18 0 0.458 6.998 45.8	0 0.00632 18.0 2.31 0 0.538 6.575 65.2 4.0900 1 0.02731 0.0 7.07 0 0.469 6.421 78.9 4.9671 2 0.02729 0.0 7.07 0 0.469 7.185 61.1 4.9671 3 0.03237 0.0 2.18 0 0.458 6.998 45.8 6.0622	0 0.00632 18.0 2.31 0 0.538 6.575 65.2 4.0900 1 1 0.02731 0.0 7.07 0 0.469 6.421 78.9 4.9671 2 2 0.02729 0.0 7.07 0 0.469 7.185 61.1 4.9671 2 3 0.03237 0.0 2.18 0 0.458 6.998 45.8 6.0622 3	0 0.00632 18.0 2.31 0 0.538 6.575 65.2 4.0900 1 296 1 0.02731 0.0 7.07 0 0.469 6.421 78.9 4.9671 2 242 2 0.02729 0.0 7.07 0 0.469 7.185 61.1 4.9671 2 242 3 0.03237 0.0 2.18 0 0.458 6.998 45.8 6.0622 3 222	0 0.00632 18.0 2.31 0 0.538 6.575 65.2 4.0900 1 296 15.3 1 0.02731 0.0 7.07 0 0.469 6.421 78.9 4.9671 2 242 17.8 2 0.02729 0.0 7.07 0 0.469 7.185 61.1 4.9671 2 242 17.8 3 0.03237 0.0 2.18 0 0.458 6.998 45.8 6.0622 3 222 18.7	0 0.00632 18.0 2.31 0 0.538 6.575 65.2 4.0900 1 296 15.3 396.90 1 0.02731 0.0 7.07 0 0.469 6.421 78.9 4.9671 2 242 17.8 396.90 2 0.02729 0.0 7.07 0 0.469 7.185 61.1 4.9671 2 242 17.8 392.83 3 0.03237 0.0 2.18 0 0.458 6.998 45.8 6.0622 3 222 18.7 394.63	0 0.00632 18.0 2.31 0 0.538 6.575 65.2 4.0900 1 296 15.3 396.90 4.98 1 0.02731 0.0 7.07 0 0.469 6.421 78.9 4.9671 2 242 17.8 396.90 9.14 2 0.02729 0.0 7.07 0 0.469 7.185 61.1 4.9671 2 242 17.8 392.83 4.03 3 0.03237 0.0 2.18 0 0.458 6.998 45.8 6.0622 3 222 18.7 394.63 2.94	0 0.00632 18.0 2.31 0 0.538 6.575 65.2 4.0900 1 296 15.3 396.90 4.98 24.0 1 0.02731 0.0 7.07 0 0.469 6.421 78.9 4.9671 2 242 17.8 396.90 9.14 21.6 2 0.02729 0.0 7.07 0 0.469 7.185 61.1 4.9671 2 242 17.8 392.83 4.03 34.7 3 0.03237 0.0 2.18 0 0.458 6.998 45.8 6.0622 3 222 18.7 394.63 2.94 33.4

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

· Check the shape of the dataset

df.shape

(506, 14)

• Split the dataset into test and train datasets, you can set the random seed to 42 by random_state=42

```
from sklearn.model_selection import train_test_split
```

```
# Assuming X and y are your features and target variables respectively
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

- run cross validation for the train and test datasets for cv=10
- print the mean score for the cross validation using linear regressor (LinearRegressor())
- do the same two steps above for random forest regressor. For that you can import

from sklearn.ensemble import RandomForestRegressor

```
# cross valudation with LinearRegression
from sklearn.model_selection import cross_val_score
from sklearn.linear_model import LinearRegression
import numpy as np

# Initialize the model
linear_regressor = LinearRegression()

# Perform cross_validation
cv_scores = cross_val_score(linear_regressor, X_train, y_train, cv=10)

# Print the mean score
print("Mean cross_validation score for Linear Regression:", np.mean(cv_scores))
```

Mean cross-validation score for Linear Regression: 0.6986301933640767

```
# Cross-Validation with Random Forest Regressor
from sklearn.ensemble import RandomForestRegressor

# Initialize the model
random_forest_regressor = RandomForestRegressor(random_state=42)

# Perform cross-validation
cv_scores_rf = cross_val_score(random_forest_regressor, X_train, y_train, cv=10)

# Print the mean score
print("Mean cross-validation score for Random Forest Regressor:", np.mean(cv_scores_rf))

Mean cross-validation score for Random Forest Regressor: 0.8211604949722642
```

Boston Housing Data with Polynomials

- · Split the dataset into test and train datasets
- · Transform the dataset using polynomial transformation, imputer and scaler
- · Run cross validation
- · Plot the coefficients of the model

Step 1 & 2: Load the Boston Housing dataset and Split It

```
# We already did this in the above one
Step 3: Transform the Dataset
from sklearn.preprocessing import PolynomialFeatures, StandardScaler
from sklearn.impute import SimpleImputer
from sklearn.pipeline import make_pipeline
# Create a pipeline for transformations
pipeline = make_pipeline(
    SimpleImputer(strategy='mean'), # Impute missing values
    PolynomialFeatures(degree=2, include_bias=False), # Polynomial transformation
    StandardScaler() # Feature scaling
# Fit and transform the training data
X_train_transformed = pipeline.fit_transform(X_train)
# Transform the test data (no fitting!)
X_test_transformed = pipeline.transform(X_test)
Step 4: Run Cross-Validation
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import cross_val_score
import numpy as np
# Initialize the Linear Regression model
linear_model = LinearRegression()
# Perform cross-validation
cv_scores = cross_val_score(linear_model, X_train_transformed, y_train, cv=10)
# Print the mean cross-validation score
print("Mean cross-validation score:", np.mean(cv_scores))
    Mean cross-validation score: 0.8042073305749259
Step 5: Plot the Coefficients of the Model
```

```
import matplotlip.pyplot as plt
```

```
# Fit the model to the transformed training data
linear_model.fit(X_train_transformed, y_train)

# Get the coefficients
coefficients = linear_model.coef_

# Plot the coefficients
plt.figure(figsize=(10, 6))
plt.plot(coefficients, 'o')
plt.title('Coefficients of the Linear Model')
plt.xlabel('Coefficient Index')
plt.ylabel('Coefficient Value')
plt.show()
```

