import numpy as np # --> Importing NumPy for numerical operations
import pandas as pd # --> Importing Pandas for data manipulation
import seaborn as sns # --> Importing Seaborn for data visualization
import matplotlib.pyplot as plt # --> Importing Matplotlib for plotting graphs
from sklearn.model_selection import train_test_split # --> Importing function to split data
from sklearn.preprocessing import StandardScaler # --> Importing scaler for normalization
from sklearn.linear_model import LinearRegression # --> Importing Linear Regression model
from sklearn.metrics import mean_squared_error, mean_absolute_error, r2_score # -->
Importing evaluation metrics

import keras # --> Importing Keras for building neural networks

from keras.models import Sequential # --> Importing Sequential model from Keras

from keras.layers import Dense # --> Importing Dense layer for neural network

import plotly.graph_objects as go # --> Importing Plotly for interactive plots

url = "https://lib.stat.cmu.edu/datasets/boston" # --> Boston housing dataset URL
raw_df = pd.read_csv(url, sep="\s+", skiprows=22, header=None) # --> Reading dataset from URL

data = np.hstack([raw_df.values[::2, :], raw_df.values[1::2, :2]]) # --> Combining feature rows target = raw_df.values[1::2, 2] # --> Extracting target variable (PRICE)

feature_names = ['CRIM', 'ZN', 'INDUS', 'CHAS', 'NOX', 'RM', 'AGE', 'DIS', # --> Defining column names

'RAD', 'TAX', 'PTRATIO', 'B', 'LSTAT']

data = pd.DataFrame(data, columns=feature_names) # --> Creating DataFrame with features data['PRICE'] = target # --> Adding target column to DataFrame

data.head() # --> Displaying first few rows of dataset

print("First look at the data:") # --> Printing title
print(data.head()) # --> Printing first few rows
print("\nData shape:", data.shape) # --> Printing shape of dataset
print("\nNull values check:") # --> Checking for null values

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print(data.isnull().sum()) # --> Summing null values in each column
print("\nData statistics:") # --> Descriptive statistics of dataset
print(data.describe()) # --> Showing summary statistics
sns.set(style="whitegrid") # --> Setting Seaborn style
plt.figure(figsize=(12, 6)) # --> Setting figure size
sns.distplot(data['PRICE']) # --> Plotting distribution of house prices
plt.title('Distribution of House Prices') # --> Title of the plot
plt.show() # --> Displaying the plot
data.head(n=10) # --> Viewing first 10 rows of the dataset
print(data.shape) # --> Printing shape of dataset
data.isnull().sum() # --> Checking for null values
data.describe() # --> Showing dataset statistics
import seaborn as sns # --> Re-importing Seaborn (already imported above)
sns.distplot(data.PRICE) # --> Plotting price distribution again
sns.boxplot(data.PRICE) # --> Drawing boxplot for target variable
correlation = data.corr() # --> Calculating correlation matrix
correlation.loc['PRICE'] # --> Viewing correlation with PRICE
import matplotlib.pyplot as plt # --> Re-importing matplotlib (already imported above)
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fig, axes = plt.subplots(figsize=(15, 12)) # --> Setting up heatmap plot size

sns.heatmap(correlation, square=True, annot=True) # --> Plotting heatmap with annotations

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plt.figure(figsize=(20, 5)) # --> Plotting figure for scatter plots
features = ['LSTAT', 'RM', 'PTRATIO'] # --> Features for scatter plot
for i, col in enumerate(features): # --> Looping through selected features
  plt.subplot(1, len(features), i+1) # --> Creating subplots
 x = data[col] # --> Selecting x values
 y = data.PRICE # --> Setting y as target variable
  plt.scatter(x, y, marker='o') # --> Plotting scatter points
  plt.title("Variation in House prices") # --> Plot title
  plt.xlabel(col) # --> X-axis label
  plt.ylabel('House prices in $1000') # --> Y-axis label
X = data.iloc[:, :-1] # --> Selecting all columns except target as features
y = data.PRICE # --> Selecting target variable
from sklearn.model_selection import train_test_split # --> Re-importing train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42) # -->
Splitting dataset
mean = X_train.mean(axis=0) # --> Calculating feature means
std = X_train.std(axis=0) # --> Calculating feature standard deviations
X_train = (X_train - mean) / std # --> Normalizing training features
X_test = (X_test - mean) / std # --> Normalizing test features
from sklearn.linear_model import LinearRegression # --> Re-importing Linear Regression
regressor = LinearRegression() # --> Creating Linear Regression model
regressor.fit(X_train, y_train) # --> Training the Linear Regression model
y_pred = regressor.predict(X_test) # --> Making predictions on test set
from sklearn.metrics import mean_squared_error # --> Re-importing MSE
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rmse = (np.sqrt(mean_squared_error(y_test, y_pred))) # --> Calculating RMSE
print(rmse) # --> Printing RMSE
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from sklearn.metrics import r2_score # --> Re-importing R2 Score r2 = r2_score(y_test, y_pred) # --> Calculating R2 score print(r2) # --> Printing R2 score

from sklearn.preprocessing import StandardScaler # --> Re-importing StandardScaler sc = StandardScaler() # --> Creating StandardScaler object

X_train = sc.fit_transform(X_train) # --> Scaling training data

X_test = sc.transform(X_test) # --> Scaling test data

import keras # --> Re-importing Keras

from keras.layers import Dense, Activation, Dropout # --> Importing Keras layers from keras.models import Sequential # --> Re-importing Sequential model

model = Sequential() # --> Initializing Sequential model

model.add(Dense(128, activation='relu', input_dim=13)) # --> Adding input layer with 128 neurons

model.add(Dense(64, activation='relu')) # --> Adding hidden layer with 64 neurons model.add(Dense(32, activation='relu')) # --> Adding hidden layer with 32 neurons model.add(Dense(16, activation='relu')) # --> Adding hidden layer with 16 neurons model.add(Dense(1)) # --> Adding output layer with 1 neuron

model.compile(optimizer='adam', loss='mean_squared_error', metrics=['mae']) # --> Compiling model

history = model.fit(X_train, y_train, epochs=100, validation_split=0.05) # --> Training model with validation

from plotly.subplots import make_subplots # --> Importing subplots from Plotly import plotly.graph_objects as go # --> Re-importing Plotly

fig = go.Figure() # --> Initializing figure

fig.add_trace(go.Scattergl(y=history.history['loss'], name='Train')) # --> Adding training loss plot fig.add_trace(go.Scattergl(y=history.history['val_loss'], name='Valid')) # --> Adding validation loss plot

fig.update_layout(height=500, width=700, xaxis_title='Epoch', yaxis_title='Loss') # --> Updating plot layout

fig.show() # --> Showing the plot

fig = go.Figure() # --> Creating new figure

fig.add_trace(go.Scattergl(y=history.history['mae'], name='Train')) # --> Adding training MAE plot fig.add_trace(go.Scattergl(y=history.history['val_mae'], name='Valid')) # --> Adding validation MAE plot

fig.update_layout(height=500, width=700, xaxis_title='Epoch', yaxis_title='Mean Absolute Error') # --> Layout update

fig.show() # --> Showing the plot

y_pred = model.predict(X_test) # --> Predicting test data with neural network

mse_nn, mae_nn = model.evaluate(X_test, y_test) # --> Evaluating model performance

print('Mean squared error on test data: ', mse_nn) # --> Printing neural network MSE

print('Mean absolute error on test data: ', mae_nn) # --> Printing neural network MAE

from sklearn.metrics import mean_absolute_error # --> Re-importing MAE

lr_model = LinearRegression() # --> Creating Linear Regression model again
lr_model.fit(X_train, y_train) # --> Training the model
y_pred_lr = lr_model.predict(X_test) # --> Predicting test set
mse_lr = mean_squared_error(y_test, y_pred_lr) # --> Calculating MSE
mae_lr = mean_absolute_error(y_test, y_pred_lr) # --> Calculating MAE

print('Mean squared error on test data: ', mse_lr) # --> Printing MSE
print('Mean absolute error on test data: ', mae_lr) # --> Printing MAE

from sklearn.metrics import r2_score # --> Re-importing R2 Score again
r2 = r2_score(y_test, y_pred) # --> Recalculating R2 score
print(r2) # --> Printing R2

from sklearn.metrics import mean_squared_error # --> Re-importing MSE again rmse = (np.sqrt(mean_squared_error(y_test, y_pred))) # --> Recalculating RMSE print(rmse) # --> Printing RMSE

import sklearn # --> Importing sklearn for preprocessing

new_data = sklearn.preprocessing.StandardScaler().fit_transform(([[0.1, 10.0, 5.0, 0, 0.4, 6.0, 50, 6.0, 1, 400, 20, 300, 10]])) # --> Scaling new input data

prediction = model.predict(new_data) # --> Predicting price for new data

print("Predicted house price:", prediction) # --> Printing predicted price