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import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import OneHotEncoder
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline
from sklearn.metrics import r2_score, mean_squared_error
import numpy as np
# Data as a dictionary
data = {
       'Area (sqft)': [1000, 1500, 1200, 1800, 1100, 1400, 2000, 1700, 1300, 2200, 1600, 1450, 950, 1850, 1250],
       'Bedrooms': [2, 3, 2, 4, 2, 3, 4, 3, 2, 5, 3, 3, 2, 4, 2],
       'Location': ['Chennai', 'Bangalore', 'Pune', 'Chennai', 'Che
       'Price (Lakh)': [50, 75, 60, 90, 55, 70, 105, 85, 65, 120, 80, 72, 48, 95, 62]
def predict_house_price():
       Trains a Linear Regression model to predict house prices and provides an
      interactive way to get predictions for new house details.
      try:
              # 1. Load the housing dataset from the dictionary into a DataFrame
              df = pd.DataFrame(data)
              print("Dataset loaded successfully.")
              print("First 5 rows of the dataset:")
              print(df.head())
              print("-" * 30)
             # 2. Preprocessing and feature engineering
              # Separate features (X) and target (y)
             X = df.drop('Price (Lakh)', axis=1)
             y = df['Price (Lakh)']
              # Define which columns are numeric and which are categorical
              numeric_features = ['Area (sqft)', 'Bedrooms']
              categorical_features = ['Location']
              # Create a preprocessing pipeline
              # The ColumnTransformer applies different transformations to different columns
              preprocessor = ColumnTransformer(
                     transformers=[
                            ('cat', OneHotEncoder(), categorical_features)
                     remainder='passthrough'
              )
              # 3. 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)
              print(f"Data \ split \ into \ training \ (\{len(X\_train)\} \ samples) \ and \ testing \ (\{len(X\_test)\} \ samples).")
              print("-" * 30)
              # 4. Create and train the model using a pipeline
              # A pipeline chains together the preprocessor and the model, making the workflow cleaner
              model = Pipeline(steps=[
                     ('preprocessor', preprocessor),
                     ('regressor', LinearRegression())
              1)
              print("Training the Linear Regression model...")
              model.fit(X_train, y_train)
              print("Model training complete.")
              print("-" * 30)
              # 5. Evaluate the model on the test set
              y pred = model.predict(X test)
              r2 = r2_score(y_test, y_pred)
              mse = mean_squared_error(y_test, y_pred)
              print(f"Model Evaluation:")
              print(f"R-squared (R2) Score: {r2:.2f}")
              print(f"Mean Squared Error (MSE): {mse:.2f}")
              print("-" * 30)
```

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# 6. Predict house price based on user input
       print("Ready to predict house prices for new listings.")
       print("Example Input: Area: 1300, Bedrooms: 3, Location: Chennai")
       # Create a new DataFrame for user input
       new_data = pd.DataFrame([{
           'Area (sqft)': 1300,
           'Bedrooms': 3,
           'Location': 'Chennai'
       }])
       predicted_price = model.predict(new_data)
       print(f"\nPrediction for Area: 1300, Bedrooms: 3, Location: Chennai")
       print(f"Predicted House Price: {predicted_price[0]:.2f} Lakh")
   except FileNotFoundError:
       print(f"Error: 'house_data.csv' not found. Please create the file with the sample data.")
   except Exception as e:
       print(f"An error occurred: {e}")
# Run the prediction function
if __name__ == "__main__":
   predict_house_price()

→ Dataset loaded successfully.

    First 5 rows of the dataset:
       Area (sqft) Bedrooms Location Price (Lakh)
                     2
              1000
                               Chennai
    1
              1500
                          3 Bangalore
                       2 Pune
4 Chennai
              1200
                                                   60
    2
    3
              1800
                                                   90
             1100
                        2 Bangalore
                                                   55
    Data split into training (12 samples) and testing (3 samples).
    Training the Linear Regression model...
    Model training complete.
    Model Evaluation:
    R-squared (R2) Score: 0.98
    Mean Squared Error (MSE): 15.03
    Ready to predict house prices for new listings.
    Example Input: Area: 1300, Bedrooms: 3, Location: Chennai
    Prediction for Area: 1300, Bedrooms: 3, Location: Chennai
    Predicted House Price: 65.96 Lakh
```