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# Step 1: Import required libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import mean_squared_error, r2_score
# Step 2: Load the dataset
# If you just created the file in the same session:
df = pd.read_csv('weather_data_500.csv')
# If you're uploading the file manually:
# from google.colab import files
# uploaded = files.upload()
# df = pd.read_csv('weather_data_500.csv')
# Step 3: Display basic info
print("Dataset preview:")
display(df.head())
print("\nDataset summary:")
print(df.describe())
# Step 4: Visualize relationships
sns.pairplot(df)
plt.suptitle('Feature Relationships', y=1.02)
plt.show()
# Step 5: Split into features and target
X = df[['humidity', 'wind_speed', 'pressure']]
y = df['temperature']
# Step 6: Train-test split
X train, X test, y train, y test = train test split(X, y, test size=0.2, random state=42)
# Step 7: Train Linear Regression Model
model = LinearRegression()
model.fit(X_train, y_train)
# Step 8: Make Predictions
y_pred = model.predict(X_test)
# Step 9: Evaluate the model
mse = mean_squared_error(y_test, y_pred)
r2 = r2_score(y_test, y_pred)
print(f"\n Model Evaluation:")
print(f"Mean Squared Error (MSE): {mse:.2f}")
print(f"R2 Score: {r2:.2f}")
# Step 10: Plot Predictions
plt.figure(figsize=(8, 5))
plt.scatter(y_test, y_pred, color='blue', alpha=0.6)
plt.plot([y.min(), y.max()], [y.min(), y.max()], '--', color='red')
plt.xlabel("Actual Temperature")
plt.ylabel("Predicted Temperature")
plt.title("Actual vs Predicted Temperature")
plt.grid(True)
plt.show()
```

## → Dataset preview:

	humidity	wind_speed	pressure	temperature
0	52.472407	13.963234	985.736299	130.911913
1	87.042858	10.721927	1016.061581	165.289022
2	73.919637	6.190552	1044.200396	152.851528
3	65.919509	16.275900	1032.239115	144.527302
4	39.361118	13.694623	1038.557698	141.983482

## Dataset summary:

	humidity	wind_speed	pressure	temperature
count	500.000000	500.000000	500.000000	500.000000
mean	59.913703	9.639028	1013.992440	146.410427
std	17.921305	5.709869	25.261399	9.603409
min	30.303695	0.092640	970.419898	125.266381
25%	44.476781	4.581985	990.504384	139.244636
50%	60.789825	9.436431	1015.877759	146.219637
75%	75.367493	14.526736	1036.074219	153.929314
max	89.577888	19.994353	1054.950167	170.249103

## Feature Relationships

