```
# 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
import pandas as pd
import numpy as np
# Set random seed for reproducibility
np.random.seed(42)
# Generate 500 rows of synthetic weather data
data = pd.DataFrame({
    'humidity': np.random.uniform(30, 90, 500),
    'wind_speed': np.random.uniform(0, 20, 500),
    'pressure': np.random.uniform(970, 1055, 500)
})
# Generate temperature using a linear combination of other features
# Add some noise for realism
data['temperature'] = (
    0.45 * data['humidity']
    - 0.25 * data['wind_speed']
    + 0.12 * data['pressure']
    + np.random.normal(0, 3, 500)
)
# Save the dataset to CSV
data.to_csv('weather_data_500.csv', index=False)
print(" File 'weather_data_500.csv' has been created.")
File 'weather_data_500.csv' has been created.
# Step 2: Load the dataset
# If you just created the file in the same session:
df = pd.read_csv('weather_data_500.csv')
# Step 3: Display basic info
print("Dataset preview:")
display(df.head())
→ 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
         39.361118
                     13.694623 1038.557698
                                             141.983482
print("\nDataset summary:")
print(df.describe())
\overline{\rightarrow}
     Dataset summary:
              humidity wind_speed
                                       pressure temperature
            500.000000
                        500.000000
                                     500.000000
                                                   500.000000
     count
             59.913703
                          9.639028 1013.992440
                                                  146.410427
     mean
             17.921305
                          5.709869
                                      25.261399
                                                     9.603409
     std
```

```
min
        30.303695
                     0.092640
                                970.419898
                                              125.266381
25%
                                              139.244636
        44.476781
                     4.581985
                                990.504384
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
```

# Step 4: Visualize relationships
sns.pairplot(df)
plt.suptitle('Feature Relationships', y=1.02)
plt.show()



## Feature Relationships 90 80 70 humidity 60 50 40 30 20 15 wind\_speed 1040 1020 1000 980 170 160 temperature 150 140 130 60 80 10 980 1000 1020 15 20 1040 130 140 150 160 170 humidity wind\_speed pressure temperature

```
# 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)
```

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# Step 7: Train Linear Regression Model
model = LinearRegression()
model.fit(X_train, y_train)
      ▼ LinearRegression ① ?
      LinearRegression()
# 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}")
\overline{\Rightarrow}

    Model Evaluation:

     Mean Squared Error (MSE): 10.24
     R<sup>2</sup> Score: 0.87
# Step 10: Plot Predictions
plt.figure(figsize=(8, 5))
plt.scatter(y_test, y_pred, color='blue', alpha=0.6)
\verb|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()
\overline{\Rightarrow}
                                       Actual vs Predicted Temperature
         170
         160
```

