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import numpy as np
import matplotlib.pyplot as plt
from sklearn.linear model import LinearRegression
from sklearn.preprocessing import PolynomialFeatures
from sklearn.metrics import r2 score
from sklearn.model selection import train test split
# Generate synthetic non-linear data
np.random.seed(0)
X = np.sort(5 * np.random.rand(100, 1), axis=0)
y = np.sin(X).ravel() + np.random.normal(0, 0.2, X.shape[0]) # Non-
linear function with noise
# Split data
X train, X test, y train, y test = train test split(X, y, test size=0.2,
random state=42)
# Linear Regression
lin reg = LinearRegression()
lin reg.fit(X train, y train)
y pred lin = lin reg.predict(X test)
# Polynomial Regression (degree = 4)
poly = PolynomialFeatures(degree=4)
X_train_poly = poly.fit transform(X train)
X test poly = poly.transform(X test)
poly reg = LinearRegression()
poly reg.fit(X train poly, y train)
y pred poly = poly reg.predict(X test poly)
# Evaluation
print("Linear Regression R<sup>2</sup> Score:", r2_score(y_test, y_pred_lin))
print("Polynomial Regression R2 Score:", r2 score(y test, y pred poly))
# Visualization
X \text{ range} = \text{np.linspace}(0, 5, 100).\text{reshape}(-1, 1)
y range lin = lin reg.predict(X range)
y range poly = poly reg.predict(poly.transform(X range))
plt.scatter(X, y, color='blue', label='Actual Data')
plt.plot(X_range, y_range_lin, color='red', label='Linear Regression')
plt.plot(X range, y range poly, color='green', label='Polynomial
Regression (deg=4)')
plt.title("Comparison of Linear vs Polynomial Regression")
plt.xlabel("X")
plt.ylabel("y")
plt.legend()
plt.show()
```