

## **Polynomial Regression**

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In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.linear model import LinearRegression
from sklearn.preprocessing import PolynomialFeatures
from sklearn.metrics import mean squared error, r2 score
# Sample dataset: Hours Studied vs Test Score (non-linear pattern)
X = np.array([1, 2, 3, 4, 5, 6, 7, 8, 9, 10]).reshape(-1, 1)
y = np.array([35, 37, 50, 55, 65, 68, 80, 82, 90, 95])
# --- Step 1: Transform features to polynomial features ---
degree = 2 # quadratic
poly = PolynomialFeatures(degree=degree)
X poly = poly.fit transform(X) # adds X^2, X^1, and bias term
# --- Step 2: Fit Linear Regression on polynomial features ---
model = LinearRegression()
model.fit(X poly, y)
# --- Step 3: Predict ---
y pred = model.predict(X poly)
# --- Step 4: Evaluation ---
mse = mean squared error(y, y pred)
r2 = r2 \ score(y, y \ pred)
print(f"Mean Squared Error: {mse:.3f}")
print(f"R^2 Score: {r2:.3f}")
print(f"Intercept: {model.intercept :.3f}")
print(f"Coefficients: {model.coef }")
# --- Step 5: Plot ---
plt.scatter(X, y, color='blue', label='Actual Data')
plt.plot(X, y pred, color='red', linewidth=2, label=f'Polynomial Regression (d
plt.xlabel("Hours Studied")
plt.ylabel("Test Score")
plt.title("Polynomial Regression")
plt.legend()
plt.show()
```

Mean Squared Error: 4.221 R^2 Score: 0.990 Intercept: 24.483 Coefficients: [ 0. 8.4219697 -0.13257576]



