

Polynomial Regression – Simplified Theory

1. Introduction

- **Polynomial Regression** is a type of regression that **extends Linear Regression** by adding **non-linear relationships** between the independent variable(s) and the dependent variable.
 - It allows a model to **fit curves** to the data instead of just a straight line.
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2. What is Polynomial Regression?

- Standard **Linear Regression** equation:

$$y = \beta_0 + \beta_1 x + \epsilon$$

This assumes a **straight-line relationship** between x and y.

- In **Polynomial Regression**, we include **higher-degree terms** of x:

$$y = \beta_0 + \beta_1 x + \beta_2 x^2 + \beta_3 x^3 + \dots + \beta_n x^n + \epsilon$$

Where:

- x^2, x^3, \dots, x^n = **polynomial terms** that capture **curvature** in data.
 - $\beta_0, \beta_1, \dots, \beta_n$ = model coefficients.
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3. Why Polynomial Regression?

- Many real-world relationships are **not perfectly linear**.
 - **Example:**
 1. Predicting **growth of a plant** with time may follow a **curved pattern**.
 2. Predicting **sales revenue** based on advertising spend may have **diminishing returns**.
 - Linear regression cannot capture these patterns, but polynomial regression can.
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4. How It Works

- The data is **transformed** by creating **additional features**:
 - From a single variable x , we create:
 - x^2, x^3, \dots, x^n
- We then apply **Linear Regression** on these new features:

$$\hat{y} = \beta_0 + \beta_1 x + \beta_2 x^2 + \dots + \beta_n x^n$$

Though the relationship is non-linear, the model remains **linear in coefficients**, making it solvable using linear regression techniques.

5. Degree of Polynomial

- **Degree 1**: Linear regression (straight line).
- **Degree 2**: Quadratic regression (parabolic curve).
- **Degree 3+**: More complex curves.

Choosing the degree:

- **Too low**: Underfitting (model too simple).
 - **Too high**: Overfitting (model too complex, memorizing noise).
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6. Advantages

- Captures **non-linear patterns** in data.
 - Simple to implement using linear regression techniques.
 - Works well when **relationship between variables is curved**.
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7. Disadvantages

- Higher degrees can lead to:
 1. **Overfitting**, especially with small datasets.
 2. Poor performance on unseen data.
 - **Extrapolation risk**: Predictions outside data range can behave unpredictably.
 - **Multicollinearity**: Polynomial terms can be highly correlated.
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➤ Key Takeaway

- Polynomial regression:
 1. **Extends linear regression** by adding powers of features.
 2. Allows fitting **non-linear relationships** while keeping a **linear model structure**.
 - **Formula**:
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$$y = \beta_0 + \beta_1 x + \beta_2 x^2 + \cdots + \beta_n x^n$$
