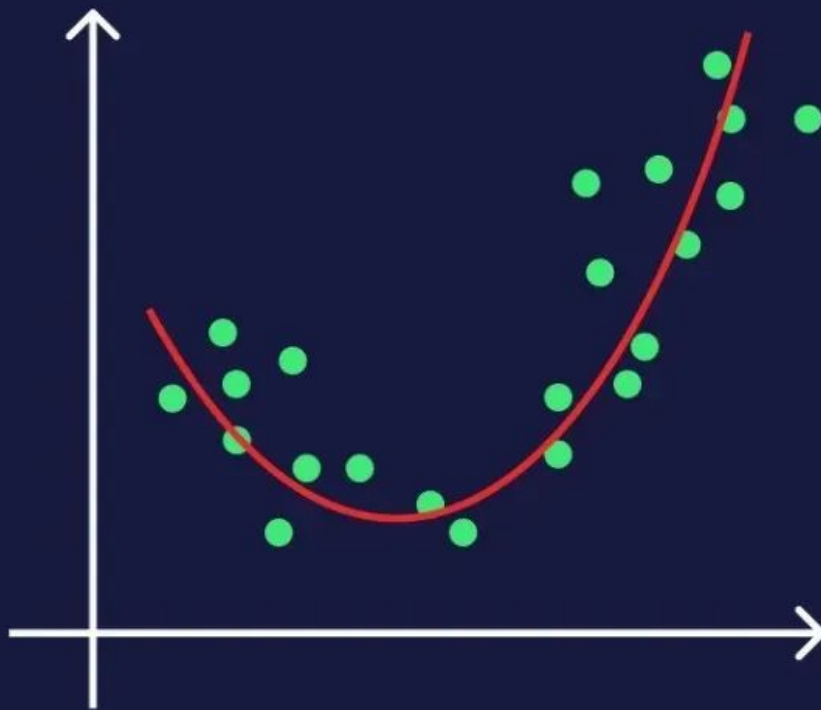


# Polynomial Regression in ML



## What is Polynomial Regression?

A regression procedure known as polynomial regression models the relationship between the independent variable ( $x$ ) and dependent variable ( $y$ ) as an  $n$ th degree polynomial.

- ✎ By making the necessary adjustments, a linear regression model is transformed into a polynomial regression model.



## Polynomial Regression Algorithm:

The basic goal of any regression study is to represent an independent variable's expression in terms of the expected value of a dependent variable.

## Polynomial Regression Equation:

$$y = \beta_0 + \beta_1 x + \beta_2 x^2 + \beta_3 x^3 + \cdots + \beta_n x^n + \varepsilon.$$



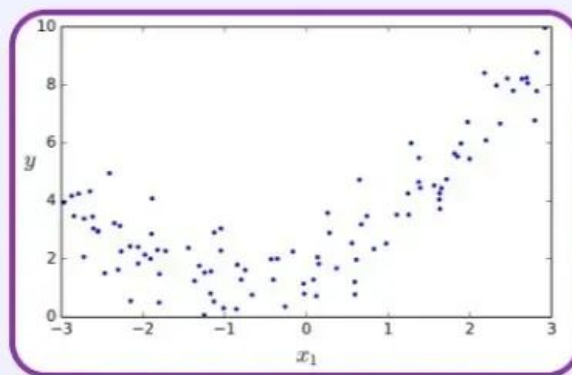
## How it works?

Using a straightforward quadratic equation, let's create some nonlinear data.

## Example:

```
m = 100
X = 6 * np.random.rand(m, 1) - 3
y = 0.5 * X**2 + X + 2 +
    np.random.randn(m, 1)
```

## Result:



Now, we should fit a model that passes through the datapoints.

Adding the square (2nd-degree polynomial) of each feature in the training set as new features

```
>>> from sklearn.preprocessing import PolynomialFeatures
>>> poly_features = PolynomialFeatures(degree=2,
>>> include_bias=False)
>>> X_poly = poly_features.fit_transform(X)
>>> X[0]
array([-0.75275929])
>>> X_poly[0]
array([-0.75275929,  0.56664654])
```



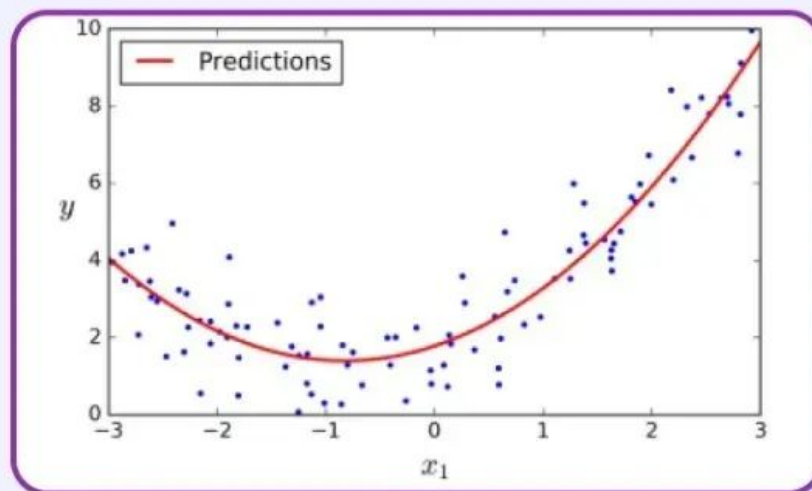


The original feature of  $X$  as well as its square are now both present in  $X_{poly}$ .

## Example:

```
>>> lin_reg = LinearRegression()  
>>> lin_reg.fit(X_poly, y)  
>>> lin_reg.intercept_, lin_reg.coef_  
(array([ 1.78134581]), array([[ 0.93366893,  0.56456263]]))
```

## Result:



PolynomialFeatures(degree=d)  
transforms an array containing n  
features into an array containing  
 $(n+d)!/n!d!$  features.

**Original model:**

$$y = 0.5x^2 + 1.0x + 2.0 + \text{Gaussian noise.}$$

**Predicted Model:**

$$y = 0.56x^2 + 0.93x + 1.78$$

This has predicted very very closely.



**Did you find it useful ?**

Let us know in the comments

**FOLLOW FOR MORE**

