Polynomial Regression (Handwriting Assignment)

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October 9, 2023

Introduction

In the mid-term project, we will look at a polynomial regression algorithm which can be used to fit non-linear data by using a polynomial function. The polynomial Regression is a form of regression analysis in which the relationship between the independent variable x and the dependent variable y is modeled as an nth degree polynomial in x.

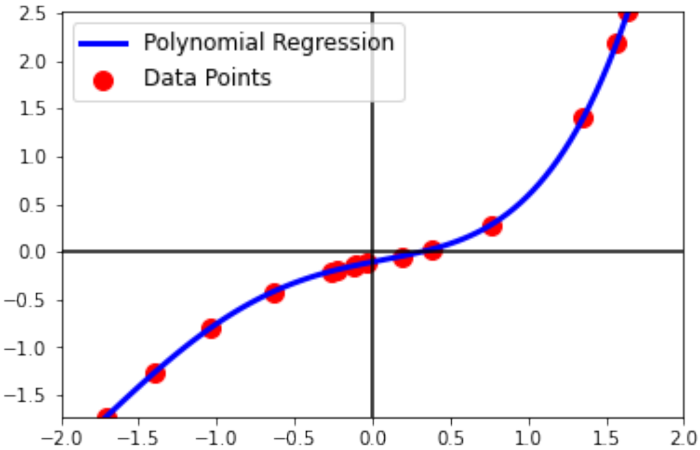


Figure 1: Example of Polynomial Regression

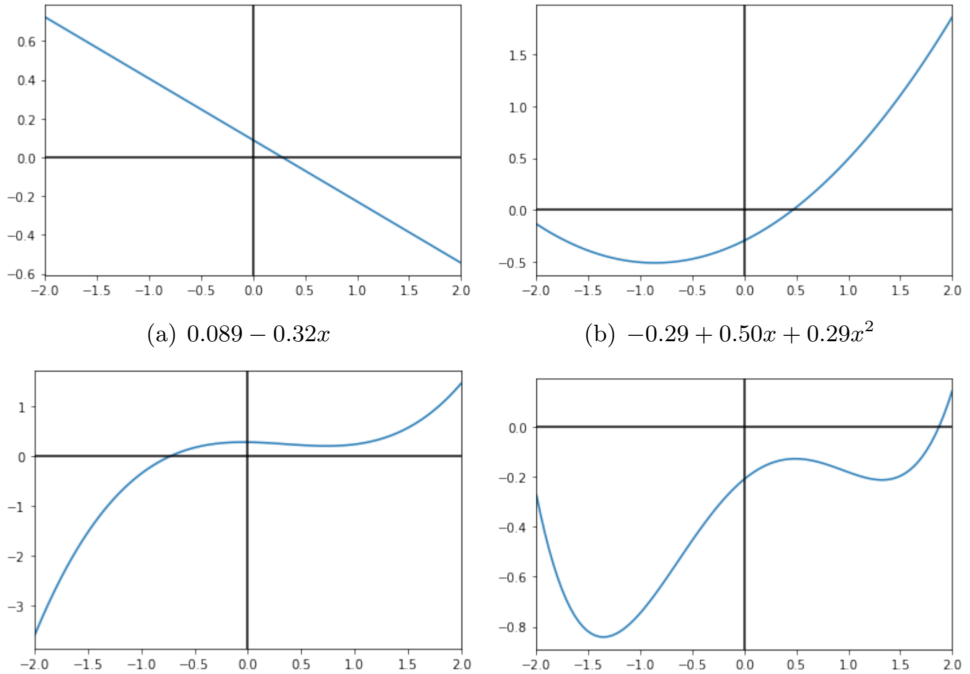
First, what is a regression? we can find a definition from the book as follows: *Regression analysis is a form of predictive modelling technique which investigates the relationship between a dependent and independent variable.* Actually, this definition is a bookish definition, in simple terms the regression can be defined as *finding a function that best explain data which consists of input and output pairs.* Let assume that we have 100 data points,

$$(x_1, y_1),\ (x_2, y_2),\ (x_3, y_3),\ \cdots (x_{98}, y_{98}),\ (x_{99}, y_{99}),\ (x_{100}, y_{100}).$$

The goal of regression is to find a function $\hat{f}$ such that

$$\hat{f}(x_1) = y_1,\ \hat{f}(x_2) = y_2,\ \hat{f}(x_3) = y_3,\ \cdots,\ \hat{f}(x_{99}) = y_{100},\ \hat{f}(x_{100}) = y_{100}.$$

This is the simplest definition of the regression problem. Note that many details about regression analysis are omitted here, but, you will learn more rigorous definition in other courses such as



(a) $0.089 - 0.32x$ (b) $-0.29 + 0.50x + 0.29x^2$

(c) $0.28 - 0.03x - 0.34x^2 + 0.32x^3$ (d) $-0.21 + 0.34x - 0.35x^2 - 0.06x^3 + 0.1x^4$

Figure 2: Examples of polynomial functions

machine learning or statistics. Then, the polynomial regression is the regression framework that employs the polynomial function to fit the data.

So, what is the polynomial function? I guess you may remember, from high school, the following functions:

$$\begin{aligned}
\text{Degree of } 0 &: f(x) = w_0 \\
\text{Degree of } 1 &: f(x) = w_1 \cdot x + w_0 \\
\text{Degree of } 2 &: f(x) = w_2 \cdot x^2 + w_1 \cdot x + w_0 \\
\text{Degree of } 3 &: f(x) = w_3 \cdot x^3 + w_2 \cdot x^2 + w_1 \cdot x + w_0 \\
&\vdots \\
\text{Degree of } d &: f(x) = \sum_{i=0}^{d} w_i \cdot x^i,
\end{aligned}$$

where $w_0, w_1, \cdots, w_d$ are a coefficient of polynomial and d is called a degree of a polynomial. So, we can determine a polynomial function $f(x)$ by deciding its degree d and corresponding coefficients $\{w_0, w_1, \cdots, w_d\}$. Figure 2 illustrates some examples of polynomial functions.

Then, the polynomial regression is a regression problem to find the best polynomial function to fit the given data points. Especially, the polynomial function is determined by coefficients (let just assume that d is fixed). We can restate the polynomial regression as *finding coefficients of polynomials such that, for all data point,* (x_i, y_i), $y_i = \hat{f}(x_i)$ *holds* (if we have noise free data). Figure 1 shows the example of polynomial regression. In the following problems, you have to study how to compute the coefficients of the polynomial to fit the data points.

Problems

1. (80 pt. in total)

Assume that we have n data points, $(x_1, y_1), (x_2, y_2), \cdots, (x_n, y_n)$. Let the degree of polynomial be d. Then, we want to find $w_0, w_1, w_2, \cdots, w_d$ of the polynomial such that

$$
\begin{aligned}
\hat{f}(x_1) &= w_0 + w_1 x_1 + w_2 x_1^2 + \cdots + w_d x_1^d = y_1, \\
\hat{f}(x_2) &= w_0 + w_1 x_2 + w_2 x_2^2 + \cdots + w_d x_2^d = y_2, \\
\hat{f}(x_3) &= w_0 + w_1 x_3 + w_2 x_3^2 + \cdots + w_d x_3^d = y_3, \\
\hat{f}(x_4) &= w_0 + w_1 x_4 + w_2 x_4^2 + \cdots + w_d x_4^d = y_4, \\
\hat{f}(x_5) &= w_0 + w_1 x_5 + w_2 x_5^2 + \cdots + w_d x_5^d = y_5, \\
&\vdots \\
\hat{f}(x_n) &= w_0 + w_1 x_n + w_2 x_n^2 + \cdots + w_d x_n^d = y_n.
\end{aligned}
$$

Now, we reformulate the equations into the vector and matrix form. First, let $\mathbf{w} = [w_0, w_1, \cdots, w_d]^T$ and $\mathbf{y} = [y_1, y_2, \cdots, y_n]^T$. Then, the above equations can be rewritten as

$$
\hat{f}(x_1) = [1, x_1, x_1^2, x_1^3, \cdots, x_1^d] \cdot \begin{bmatrix} w_0 \\ w_1 \\ w_2 \\ w_3 \\ \vdots \\ w_d \end{bmatrix} = [1, x_1, x_1^2, x_1^3, \cdots, x_1^d]\mathbf{w} = y_1
$$

Similarly, we have,

$$
\begin{aligned}
[1, x_2, x_2^2, x_2^3, \cdots, x_2^d]\mathbf{w} &= y_2, \\
[1, x_3, x_3^2, x_3^3, \cdots, x_3^d]\mathbf{w} &= y_3, \\
[1, x_4, x_4^2, x_4^3, \cdots, x_4^d]\mathbf{w} &= y_4, \\
[1, x_5, x_5^2, x_5^3, \cdots, x_5^d]\mathbf{w} &= y_5, \\
&\vdots \\
[1, x_n, x_n^2, x_n^3, \cdots, x_n^d]\mathbf{w} &= y_n.
\end{aligned}
$$

Then, all equations can be written as the form of linear equation,

$$
A\mathbf{w} = \mathbf{y},
$$

where A is the stack of $[1, x_i, x_i^2, x_i^3, \cdots, x_i^d]$ for $i = 1, \cdots, n$. Under this setting, answer the following questions.

1-(a) What is the size of vector w and y? (10pt)

Size of w : $d+1$

Size of y : n

1-(b) What is the size of matrix A? Write A. (10pt)

A is $n \times (d+1)$ matrix. $A = \begin{bmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^d \\ 1 & x_2 & x_2^2 & \cdots & x_2^d \\ 1 & x_3 & x_3^3 & \cdots & x_3^d \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & \cdots & x_n^d \end{bmatrix}$

1-(c) Let $d+1=n$, then, A becomes a square matrix. Compute the determinant of A. (40pt in total, Derivation: 30pt, Answer: 10pt)

$(-x_1)$를 곱해 더하기

$$\det(A) = \begin{vmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \cdots & x_2^{n-1} \\ 1 & x_3 & x_3^3 & \cdots & x_3^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & \cdots & x_n^{n-1} \end{vmatrix}$$

$\det(A) = \det(A^T)$ 이므로 기본행연산과 마찬가지로 기본열 연산을 통해 determinant를 구할수있다.

$$= \begin{vmatrix} 1 & \overset{0}{\cancel{x_1 - x_1}} & \overset{0}{\cancel{x_1^2 - x_1^2}} & \cdots & \overset{0}{\cancel{x_1^{n-1} - x_1^{n-1}}} \\ 1 & x_2 - x_1 & x_2^2 - x_1 x_2 & \cdots & x_2^{n-1} - x_1 x_2^{n-2} \\ 1 & x_3 - x_1 & x_3^2 - x_1 x_3 & \cdots & x_3^{n-1} - x_1 x_3^{n-2} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n - x_1 & x_n^2 - x_1 x_n & \cdots & x_n^{n-1} - x_1 x_n^{n-2} \end{vmatrix} = \begin{vmatrix} (x_2 - x_1) & x_2(x_2 - x_1) & \cdots & x_2^{n-2}(x_2 - x_1) \\ (x_3 - x_1) & x_3(x_3 - x_1) & \cdots & x_3^{n-2}(x_3 - x_1) \\ \vdots & & & \\ (x_n - x_1) & x_n(x_n - x_1) & \cdots & x_n^{n-2}(x_n - x_1) \end{vmatrix}$$

$(-x_2)$를 곱해 더하기

$$= (x_2 - x_1)(x_3 - x_1)\cdots(x_n - x_1) \begin{vmatrix} 1 & x_2 & x_2^2 & \cdots & x_2^{n-2} \\ 1 & x_3 & x_3^2 & \cdots & x_3^{n-2} \\ \vdots & & & & \\ 1 & x_n & x_n^2 & \cdots & x_n^{n-2} \end{vmatrix} = (x_2 - x_1)(x_3 - x_1)\cdots(x_n - x_1) \begin{vmatrix} 1 & 0 & 0 & \cdots & 0 \\ 1 & x_3 - x_2 & x_3^2 - x_2 x_3 & \cdots & x_3^{n-2} - x_2 x_3^{n-3} \\ \vdots & \vdots & \vdots & & \vdots \\ 1 & x_n - x_2 & x_n^2 - x_2 x_n & \cdots & x_n^{n-2} - x_2 x_n^{n-3} \end{vmatrix}$$

$$= (x_2 - x_1)(x_3 - x_1)\cdots(x_n - x_1) \begin{vmatrix} (x_3 - x_2) & x_3(x_3 - x_2) & \cdots & x_3^{n-3}(x_3 - x_2) \\ (x_4 - x_2) & x_4(x_4 - x_2) & \cdots & x_4^{n-3}(x_4 - x_2) \\ \vdots & & & \\ (x_n - x_2) & x_n(x_n - x_2) & \cdots & x_n^{n-3}(x_n - x_2) \end{vmatrix}$$

| | 안의 정사각행렬의 크기가 점점 주는것을 알 수있다.

$$= (x_2 - x_1)(x_3 - x_1)\cdots(x_n - x_1)(x_3 - x_2)(x_4 - x_2)\cdots(x_n - x_2) \begin{vmatrix} 1 & x_3 & \cdots & x_3^{n-3} \\ \vdots & & & \vdots \\ 1 & x_n & \cdots & x_n^{n-3} \end{vmatrix},$$

| | 안의 행렬이 2x2의 크기가 될때까지 위의 과정을 반복하자.

$$(\sim)(\sim)\cdots(\sim) \begin{vmatrix} 1 & x_{n-1} \\ 1 & x_n \end{vmatrix} = (\sim)(\sim)\cdots(\sim)(x_n - x_{n-1})$$

$$= (x_n - x_1)(x_{n-1} - x_1)\cdots(x_2 - x_1)(x_n - x_2)(x_{n-1} - x_2)\cdots(x_3 - x_2)\cdots(x_n - x_{n-2})(x_{n-1} - x_{n-2})(x_n - x_{n-1})$$

$$= \prod_{1 \le i < j \le n} (x_j - x_i)$$

1-(d) What is the condition that makes the determinant of A non-zero? (10pt)

j와 i 가 같지 않을때 ($1 \le i < j \le n$) x_j와 x_i 또한 같지 않아야 한다.

1-(e) Assume that the determinant of A is non-zero, then, what is the solution of linear equation, $A\mathrm{w} = \mathrm{y}$, with respect to w? (10pt)

$\det(A) \neq 0$ 이므로 A는 invertible 하다.

$A^{-1} A w = A^{-1} y$, $w = A^{-1} y$

$\therefore w = A^{-1} y$

2. (20pt)

Suppose that $n > d + 1$. Then, we cannot compute the inverse of A since A is not a square matrix. In this case, how can we solve the linear equation $A\mathbf{w} = \mathbf{y}$?

Singular Value Decomposition을 이용하자.

$A = U\Sigma V^T$로 특잇값 분해할수 있다.

여기서

U는 $n \times n$ 크기의 직교 행렬이다. U의 각 열은 A의 열공간을 나타낸다.

Σ는 $n \times (d+1)$ 크기의 대각 행렬이다. 이 행렬의 대각 원소들은 Singular Value이며, 크기순으로 정렬되어 있다.

V^T는 $(d+1) \times (d+1)$ 크기의 직교행렬이다. V^T의 각행은 A의 행공간을 나타낸다.

y를 U^T로 변환해 $U^T y$를 얻는다.

Σ^{-1}을 곱해 $\Sigma^{-1} U^T y$를 얻고 V를 곱해 $V\Sigma^{-1}U^T y$를 얻는다.

$W = V\Sigma^{-1}U^T y$

이렇게 해를 구할수 있다.