

## 1 Polynomial Interpolation

Given  $n$  distinct points, we can find a unique degree  $n - 1$  polynomial that passes through these points. Let the polynomial  $p$  be

$$p(x) = a_0 + a_1x + a_2x^2 + \cdots + a_{n-1}x^{n-1}.$$

Let the  $n$  points be

$$p(x_1) = y_1, p(x_2) = y_2, \cdots, p(x_n) = y_n,$$

where  $x_1 \neq x_2 \neq \cdots \neq x_n$ .

We can construct a matrix-vector equation as follows to recover the polynomial  $p$ .

$$\underbrace{\begin{bmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \cdots & x_2^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & \cdots & x_n^{n-1} \end{bmatrix}}_A \underbrace{\begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{bmatrix}}_{\vec{a}} = \underbrace{\begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}}_{\vec{y}}$$

We can solve for the  $a$  values by setting:

$$\vec{a} = A^{-1}\vec{y}$$

Note that the matrix  $A$  is known as a Vandermonde matrix whose determinant is given by

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

Since  $x_1 \neq x_2 \neq \cdots \neq x_n$ , the determinant is non-zero and  $A$  is always invertible.

## 2 Polynomial Regression

Sometimes we may want to fit our data to a polynomial with an order less than  $n - 1$ . If we fit the data to a polynomial of order  $m < n$  we get:

$$p(x) = a_0 + a_1x + a_2x^2 + \cdots + a_{m-1}x^{m-1}$$

Now when we construct the matrix-vector equation to recover polynomial  $p$ , we get:

$$\underbrace{\begin{bmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^{m-1} \\ 1 & x_2 & x_2^2 & \cdots & x_2^{m-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_m & x_m^2 & \cdots & x_m^{m-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & \cdots & x_n^{m-1} \end{bmatrix}}_A \underbrace{\begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_{m-1} \end{bmatrix}}_{\vec{a}} = \underbrace{\begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \\ \vdots \\ y_n \end{bmatrix}}_{\vec{y}}$$

With this matrix equation, we have  $n$  equations with  $m$  unknowns, which means our system is over-defined (since  $m < n$ ). One way to find the best fitting  $a$  values for this polynomial is to use least-squares, where you set:

$$\vec{a} = (A^T A)^{-1} A^T \vec{y}$$

### 1. Interpolation Example

Use polynomial interpolation to find the polynomial that passes through the points  $(1, 5)$ ,  $(2, 15)$  and  $(3, 33)$

### 2. Regression Example

Using least-squares, find the best-fit quadratic equation for the data set:  $(-2, 28)$ ,  $(-1, -14)$ ,  $(0, 0)$ ,  $(1, -42)$ , and  $(2, 56)$ .

### 3. Minimum Norm Polynomial Interpolation

We have two data points:  $(0, 0)$  and  $(1, 1)$ .

- (a) Find a linear fit curve for the two data points.
- (b) Find the second order polynomial for which the coefficients have the smallest norm. Compare the norm of the result to the first order polynomial found in part (a). Note that a first order polynomial is also a (degenerate) second order polynomial.