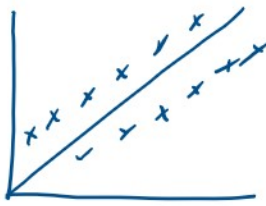


# Polynomial Fit

April 12, 2020 12:30 AM

## Polynomial Fit



$n=12$

$m=2$

→ polynomial approx. of  $n$  datapoints  
with a single polynomial of degree  $m-1$   
(if  $m=n$ , → long range interpolation)

→ For  $m < n$ : following data points are given:  
 $(x_1, y_1), \dots, (x_n, y_n)$  → look for LS solution

- Fit polynomial  $m-1$ :  $p(x) = a_1 x^{m-1} + a_2 x^{m-2} + \dots + a_{m-1} x + a_m$

→ For  $p(x_j) = y_j$  and  $m=n$ ,  $\forall a = y$  where:

-  $a = [a_1, \dots, a_m]^T$  (coeff. of  $p(x)$ )

-  $y = [y_1, \dots, y_n]^T$  (data values)

Matrix  $V$  has Vandermonde Matrix

→ For  $m < n$ , Vandermonde-like Matrix

$$V = \begin{bmatrix} x_1^{m-1} & x_1^{m-2} & \dots & x_1 \\ x_2^{m-1} & x_2^{m-2} & \dots & x_2 \\ \vdots & \vdots & \ddots & \vdots \\ x_n^{m-1} & x_n^{m-2} & \dots & x_n \end{bmatrix} \quad \because n > m, \text{ this matrix is tall} \\ \text{(more rows than columns)}$$

$\therefore$  Columns of  $V$  are ltr. dependent as long as  $x_i \neq x_j$  when  $i \neq j$

Corresponding least square equation  $V^T V a = V^T y$  has unique solution

given by  $a_{LS} = (V^T V)^{-1} V^T y$

## Special Case:

- Best fitting (linear regression)  $m=2$ , polynomial:  $p(x) = a_1 x + a_2$  → polynomial deg. 1

$$V = \begin{bmatrix} x_1 & 1 \\ x_2 & 1 \\ x_3 & 1 \\ \vdots & \vdots \\ x_n & 1 \end{bmatrix}, \quad V^T V = \begin{bmatrix} \sum_{i=1}^n x_i^2 & \sum_{i=1}^n x_i \\ \sum_{i=1}^n x_i & n \end{bmatrix}, \quad V^T y = \begin{bmatrix} \sum_{i=1}^n x_i y_i \\ \sum_{i=1}^n y_i \end{bmatrix}$$

- If  $x_i$ 's distinct,  $V^T V \rightarrow$  invertible and LS solution given by:

$$a_{LS} = \begin{bmatrix} \sum_{i=1}^n x_i^2 & \sum_{i=1}^n x_i \\ \sum_{i=1}^n x_i & n \end{bmatrix}^{-1} \begin{bmatrix} \sum_{i=1}^n x_i y_i \\ \sum_{i=1}^n y_i \end{bmatrix} \rightarrow \text{computable in MATLAB using data points}$$