

# Method of Least Squares

C&K 12.1 (pp. 495-501)

# Motivation

Surface tension  $S$  in a liquid is known to be a linear function of temperature  $T$ . For a particular liquid, measurements have been made of the surface tension at certain temperatures. The results were as follows:

$T$	0	10	20	30	40	80	90	95
$S$	68.0	67.1	66.4	65.6	64.6	61.8	61.0	60.0

How can the most probable values of the constants in the equation

$$S = aT + b$$

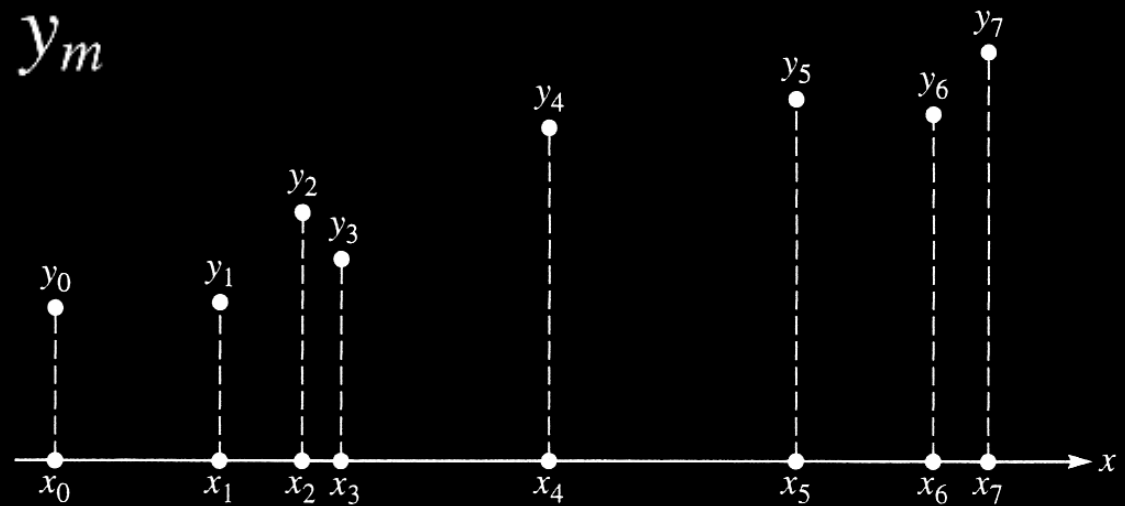
be determined? Methods for solving such problems are developed in this chapter.

(C&K p. 495)

# Linear Least Squares

$x$	$x_0$	$x_1$	$\cdots$	$x_m$
$y$	$y_0$	$y_1$	$\cdots$	$y_m$

**FIGURE 12.1**  
Experimental  
data



What straight line best fits the data?

$$y = ax + b$$

# Minimizing sum of squared errors

Sum of squared errors:

$$\varphi(a, b) = \sum_{k=0}^m (ax_k + b - y_k)^2$$

Necessary conditions for minimum (from multivariate calculus):

$$\frac{\partial \varphi}{\partial a} = 0 \quad \frac{\partial \varphi}{\partial b} = 0$$

Which leads to the normal equations...

... leads to the Normal Equations

$$\begin{cases} \sum_{k=0}^m 2(ax_k + b - y_k)x_k = 0 \\ \sum_{k=0}^m 2(ax_k + b - y_k) = 0 \end{cases}$$

Or...

$$\begin{cases} \left( \sum_{k=0}^m x_k^2 \right) a + \left( \sum_{k=0}^m x_k \right) b = \sum_{k=0}^m y_k x_k \\ \left( \sum_{k=0}^m x_k \right) a + (m+1)b = \sum_{k=0}^m y_k \end{cases}$$

# Solution (ugly)

$$\begin{cases} \left( \sum_{k=0}^m x_k^2 \right) a + \left( \sum_{k=0}^m x_k \right) b = \sum_{k=0}^m y_k x_k \\ \left( \sum_{k=0}^m x_k \right) a + (m+1)b = \sum_{k=0}^m y_k \end{cases}$$

Simply notation:

$$p = \sum_{k=0}^n x_k \quad q = \sum_{k=0}^n y_k \quad r = \sum_{k=0}^n x_k y_k \quad s = \sum_{k=0}^n x_k^2$$

Yields: 
$$\begin{bmatrix} s & p \\ p & m+1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} r \\ q \end{bmatrix}$$

Solve simple 2x2 system using Cramer's rule:

$$a = \frac{1}{d} \text{Det} \begin{bmatrix} r & p \\ q & m+1 \end{bmatrix} = \frac{1}{d} [(m+1)r - pq]$$

$$b = \frac{1}{d} \text{Det} \begin{bmatrix} s & r \\ p & q \end{bmatrix} = \frac{1}{d} [sq - pr]$$

$$d = \text{Det} \begin{bmatrix} s & p \\ p & m+1 \end{bmatrix}$$

# Linear Least Squares

Another form of this result is

$$a = \frac{1}{d} \left[ (m + 1) \left( \sum_{k=0}^m x_k y_k \right) - \left( \sum_{k=0}^m x_k \right) \left( \sum_{k=0}^m y_k \right) \right]$$
$$b = \frac{1}{d} \left[ \left( \sum_{k=0}^m x_k^2 \right) \left( \sum_{k=0}^m y_k \right) - \left( \sum_{k=0}^m x_k \right) \left( \sum_{k=0}^m x_k y_k \right) \right]$$

where

$$d = (m + 1) \left( \sum_{k=0}^m x_k^2 \right) - \left( \sum_{k=0}^m x_k \right)^2$$

**EXAMPLE 1** As a concrete example, find the linear least-squares solution for the following table of values.

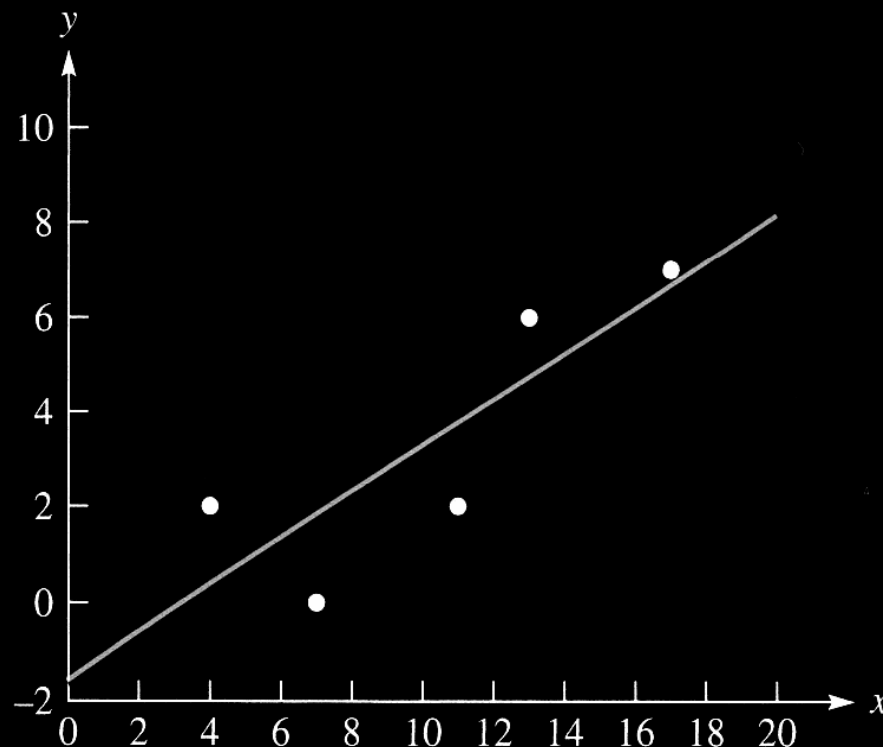
$x$	4	7	11	13	17
$y$	2	0	2	6	7

Plot the original data points and the line using a finer set of grid points.

**Solution** The equations in Algorithm 1 leads to this system of two equations:

$$\begin{cases} 644a + 52b = 227 \\ 52a + 5b = 17 \end{cases}$$

whose solution is  $a = 0.4864$  and  $b = -1.6589$ . By Equation (3), we obtain the value  $\varphi(a, b) = 10.7810$ . Figure 12.2 is a plot of the given data and the linear least squares straight line.



**FIGURE 12.2**  
Linear least  
squares



# Matrix View of Normal Eqns

Line constraints yield linear system:

$$Ax = y$$

where

$$\begin{bmatrix} x_1 & 1 \\ x_2 & 1 \\ x_3 & 1 \\ x_4 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix}$$

Minimizing sum of squared error (residual's two-norm squared):

$$\|Ax - b\|_2^2$$

Yields normal equations:

$$A^T Ax = A^T b$$

# Basis Function View (p. 500)

$$y = \sum_{j=0}^n c_j g_j(x)$$

Sum of squared errors:

$$\varphi(c_0, c_1, \dots, c_n) = \sum_{k=0}^m \left[ \sum_{j=0}^n c_j g_j(x_k) - y_k \right]^2$$

Matrix view... (usually much easier & more convenient)

# Example

## Nonpolynomial Example

The method of least squares is not restricted to linear (first-degree) polynomials or to any specific functional form. Suppose, for instance, that we want to fit a table of values  $(x_k, y_k)$ , where  $k = 0, 1, \dots, m$ , by a function of the form

$$y = a \ln x + b \cos x + ce^x$$

in the least-squares sense. The unknowns in this problem are the three coefficients  $a$ ,  $b$ , and  $c$ . We consider the function

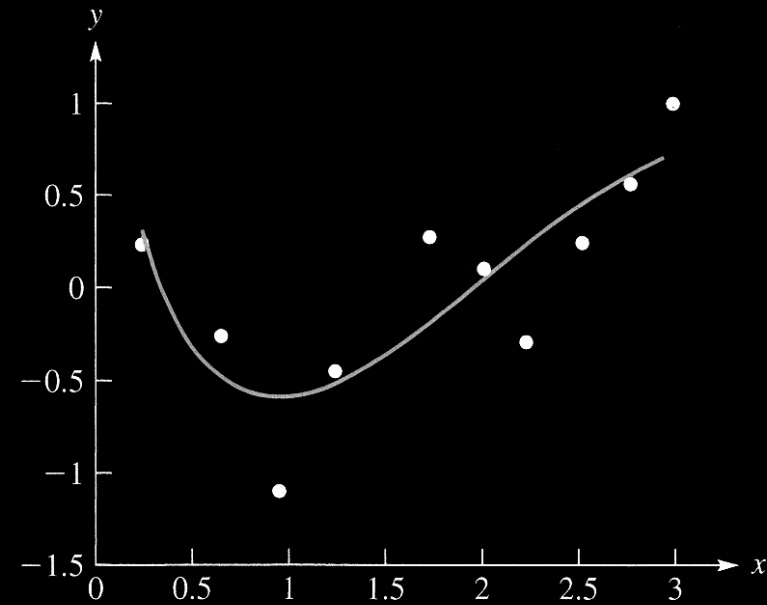
$$\varphi(a, b, c) = \sum_{k=0}^m (a \ln x_k + b \cos x_k + ce^{x_k} - y_k)^2$$

# Example (cont'd)

and set  $\partial\varphi/\partial a = 0$ ,  $\partial\varphi/\partial b = 0$ , and  $\partial\varphi/\partial c = 0$ . This results in the following three normal equations:

$$\begin{cases} a \sum_{k=0}^m (\ln x_k)^2 & + b \sum_{k=0}^m (\ln x_k)(\cos x_k) & + c \sum_{k=0}^m (\ln x_k)e^{x_k} & = \sum_{k=0}^m y_k \ln x_k \\ a \sum_{k=0}^m (\ln x_k)(\cos x_k) & + b \sum_{k=0}^m (\cos x_k)^2 & + c \sum_{k=0}^m (\cos x_k)e^{x_k} & = \sum_{k=0}^m y_k \cos x_k \\ a \sum_{k=0}^m (\ln x_k)e^{x_k} & + b \sum_{k=0}^m (\cos x_k)e^{x_k} & + c \sum_{k=0}^m (e^{x_k})^2 & = \sum_{k=0}^m y_k e^{x_k} \end{cases}$$

# Example (with data)



Fit a function of the form  $y = a \ln x + b \cos x + ce^x$  to the following table values:

$x$	0.24	0.65	0.95	1.24	1.73	2.01	2.23	2.52	2.77	2.99
$y$	0.23	-0.26	-1.10	-0.45	0.27	0.10	-0.29	0.24	0.56	1.00

Using the table and the equations above, we obtain the  $3 \times 3$  system

$$\begin{cases} 6.79410a - 5.34749b + 63.25889c = 1.61627 \\ -5.34749a + 5.10842b - 49.00859c = -2.38271 \\ 63.25889a - 49.00859b + 1002.50650c = 26.77277 \end{cases}$$

It has the solution  $a = -1.04103$ ,  $b = -1.26132$ , and  $c = 0.03073$ . So the curve

$$y = -1.04103 \ln x - 1.26132 \cos x + 0.03073e^x$$