

Recall that  $SL(2, 7)$  is the special linear group of  $2 \times 2$  matrices of determinant 1 over the integers modulo 7. Also,  $PSL(2, 7)$  is the factor group of  $SL(2, 7)$  by  $\left\{ \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix} \right\}$ , and  $PSL(2, 7)$  has order 168.

EXERCISE. Use a counting argument to prove  $PSL(2, 7)$  is simple.

#### References

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### A GENERALIZED PROBLEM OF LEAST SQUARES

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In regression analysis [1], the process of determining an interpolation function for a given set of data under the least squares criterion, one assumes the independent variables to be exact and not subject to error or variation. For instance, in elementary statistics one usually uses vertical deviations. In this context  $x$  is called the independent variable which is under control and assumed to be exact (occasionally, some textbooks [2] mention using horizontal deviations; in this situation  $x$  and  $y$  interchange roles). If one allows all the variables to reflect error, the problem of determining the "best-fitting" function for a given set of data becomes more complicated. In this paper I will consider this generalized least squares problem for the determination of an interpolation function in two variables in detail.

Consider a set of  $n$  points  $(x_1, y_1), \dots, (x_n, y_n)$ . Without loss of generality, one can assume that the  $n$  points are noncollinear. The generalized linear interpolation function for the data is the line

$$(1) \quad y - b - mx = 0,$$

such that the sum of squares of the perpendicular distances  $\sum d_j^2$  ( $j$  runs from 1 to  $n$ ) from the given  $n$  points to the line (1) is a minimum, where

$$d_j = |y_j - b - mx_j| / \sqrt{1 + m^2}.$$

The object is to determine the parameters  $b$  and  $m$  in (1) which minimize  $\sum d_j^2$ . First note that  $x_j$  and  $y_j$  are the fixed numbers which one has observed. Denote

$$f(b, m) = \sum d_j^2 = \frac{1}{1 + m^2} \sum (y_j - b - mx_j)^2.$$

$$(2) \quad \begin{aligned} \partial f / \partial b = 0 \text{ and } \partial f / \partial m = 0 \text{ give} \\ b = \bar{y} - m\bar{x}, \end{aligned}$$

and

$$(3) \quad Am^2 - Bm - A = 0,$$

where  $\bar{x}$  and  $\bar{y}$  are the arithmetic means of  $x_1, \dots, x_n$  and  $y_1, \dots, y_n$  respectively, and

$$A = \sum x_j y_j - n\bar{x}\bar{y} \text{ and } B = \sum y_j^2 - n\bar{y}^2 - \sum x_j^2 + n\bar{x}^2.$$

Let  $\hat{m}_1$  and  $\hat{m}_2$  be the roots of (3). If  $A \neq 0$ , then

$$\hat{m}_1 = (B + \sqrt{B^2 + 4A^2})/2A$$

and

$$\hat{m}_2 = (B - \sqrt{B^2 + 4A^2})/2A.$$

Note that  $\hat{m}_1$  has the same sign as  $A$ .

Substituting  $\hat{m}_1$  and  $\hat{m}_2$  into (2), one can find the corresponding values  $\hat{b}_1$  and  $\hat{b}_2$  of  $b$ . Thus the regression lines in (1) are determined, denote them by

$$(4) \quad L_1: y - \hat{b}_1 - \hat{m}_1 x = 0,$$

and

$$(5) \quad L_2: y - \hat{b}_2 - \hat{m}_2 x = 0.$$

Actually equation (3) is  $[\sum x_j^2 - \sum x_j^2] \cdot m^3 + Am^2 - Bm - A = 0$ .

If one views it as a cubic equation of  $m$ , then it has a third root  $\hat{m}_3 = \infty$ , and the corresponding regression line  $L_3$  will be

$$(6) \quad L_3: x - \bar{x} = 0.$$

The lines thus determined have the following properties:

*Property I.* All the lines  $L_1$ ,  $L_2$ , and  $L_3$  pass through the point  $(\bar{x}, \bar{y})$ , since  $(\bar{x}, \bar{y})$  satisfies the Equations (4), (5), and (6).

*Property II.*  $L_1$  and  $L_2$  are perpendicular, since

$$\hat{m}_1 \hat{m}_2 = -A/A = -1.$$

*Property III.*  $\Sigma d_j^2(2) \geq \Sigma d_j^2(3) \geq \Sigma d_j^2(1)$ ,

where  $\Sigma d_j^2(k)$  is the sum of square distances from the  $n$  points  $(x_1, y_1), \dots, (x_n, y_n)$  to the line  $L_k$  ( $k = 1, 2, 3$ ). Equalities occur if and only if the  $n$  points coincide, or the  $n$  points are symmetric about the point  $(\bar{x}, \bar{y})$  (in the latter case,  $n$  must be an even number).

*Proof.* First one proves that  $\Sigma d_j^2(3) \geq \Sigma d_j^2(1)$ . This is equivalent to showing

$$\Sigma (x_j - \bar{x})^2 - \frac{1}{1 + \hat{m}_1^2} \Sigma (y_j - \hat{b}_1 - \hat{m}_1 x_j)^2 \geq 0.$$

Now

$$\begin{aligned} (1 + \hat{m}_1^2) \Sigma (x_j - \bar{x})^2 - \Sigma (y_j - \hat{b}_1 - \hat{m}_1 x_j)^2 \\ &= (1 + \hat{m}_1^2) \Sigma (x_j - \bar{x})^2 - \{ \Sigma (y_j - \hat{m}_1 x_j)^2 - n \hat{b}_1^2 \} \\ &= -(\Sigma y_j^2 - n \bar{y}^2 - \Sigma x_j^2 + n \bar{x}^2) + 2 \hat{m}_1 (\Sigma x_j y_j - n \bar{x} \bar{y}) \\ &= -B + B + \sqrt{B^2 + 4A^2} \geq 0. \end{aligned}$$

Similarly for  $\Sigma d_j^2(2) \geq \Sigma d_j^2(3)$ .

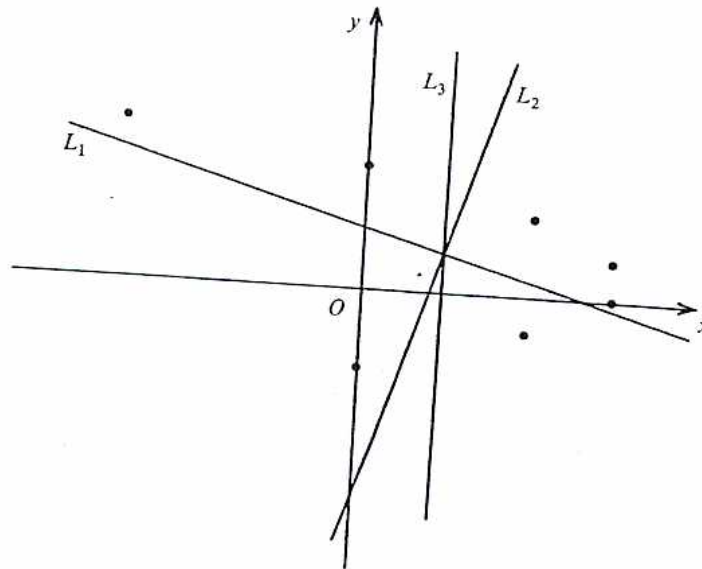
Property III says that the line  $L_1$  always minimizes  $\Sigma d_j^2$ .

**EXAMPLE.** In an experiment, let seven points  $(0, -2)$ ,  $(4, -1)$ ,  $(6, 0)$ ,  $(6, 1)$ ,  $(4, 2)$ ,  $(0, 3)$  and  $(-6, 4)$  be observed. Then  $n = 7$ ,  $\bar{x} = 2$ ,  $\bar{y} = 1$ ,  $A = -28 < 0$  and  $B = -84$ . From Eq. (3), one has

$$\begin{aligned} m^2 - 3m - 1 &= 0, \\ \hat{m}_1 &= (3 - \sqrt{13})/2 < 0, \end{aligned}$$

and

$$\hat{m}_2 = (3 + \sqrt{13})/2.$$



From Eq. (2) one obtains  $\hat{b}_1 = -2 + \sqrt{13}$  and  $\hat{b}_2 = -2 - \sqrt{13}$ ,  
The three lines

$$L_1: 2y - (3 - \sqrt{13})x + 2(2 - \sqrt{13}) = 0$$

$$L_2: 2y - (3 + \sqrt{13})x + 2(2 + \sqrt{13}) = 0$$

and

$$L_3: x - 2 = 0. \text{ (See figure.)}$$

The lines  $L_1$ ,  $L_2$ , and  $L_3$  give the values of  $\Sigma d_j^2$ , 19.52, 120.48 and 112 respectively.

References

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WHY THE PRODUCT TOPOLOGY?

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A student is often puzzled when he or she first encounters the product topology. The usual topology on an infinite product  $\Pi_{\alpha} X_{\alpha}$  of topological spaces  $X_{\alpha}$  has as a basis the sets of the form  $\Pi_{\alpha} U_{\alpha}$ , where each  $U_{\alpha}$  is open and  $U_{\alpha} = X_{\alpha}$  for all but finitely many values of  $\alpha$ . Why this last