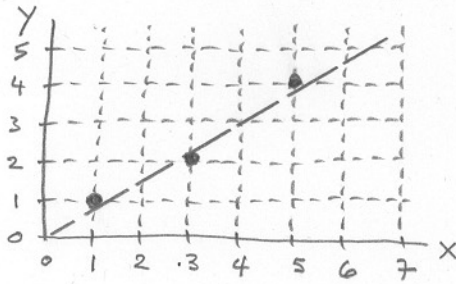


Example: Least Squares Line Fit or Regression using Method of Indirect Observations & Longhand or Brute Force Approach

CE506
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X	Y
1	1
3	2
5	4

x; constant.
y; observation

$n = 3$ total observations
 $n_0 = 2$ minimum observations to fix model
 $r = 1$ redundancy, extra observations

For the method of indirect observations you must carry $n_0 (= 2)$ unknowns, and you must write $n (= 3)$ condition equations.

Our equations will be of the form $y = mx + b$ (3 of them), and we must add a correction (residual) to each observation,

The two unknown parameters are m and b (*)

$$\begin{aligned} Y_1 + V_1 &= m X_1 + b & \Rightarrow V_1 &= m X_1 + b - Y_1 & \Rightarrow V_1 &= m \cdot 1 + b - 1 \\ Y_2 + V_2 &= m X_2 + b & \Rightarrow V_2 &= m X_2 + b - Y_2 & \Rightarrow V_2 &= m \cdot 3 + b - 2 \\ Y_3 + V_3 &= m X_3 + b & \Rightarrow V_3 &= m X_3 + b - Y_3 & \Rightarrow V_3 &= m \cdot 5 + b - 4 \end{aligned}$$

We must minimize the Least Squares Objective Function $\Phi = \sum_{i=1}^n V_i^2$
 In this case, $\Phi = V_1^2 + V_2^2 + V_3^2$. Plug in expressions for V_i , differentiate with respect to m, b the two unknowns, set equal to zero, and solve the normal equations.

$$\Phi = V_1^2 + V_2^2 + V_3^2 = (m + b - 1)^2 + (3m + b - 2)^2 + (5m + b - 4)^2 \rightarrow \text{minimum}$$

$$\frac{\partial \Phi}{\partial m} = 2(m + b - 1) + 2(3m + b - 2)3 + 2(5m + b - 4)5 = 0$$

$$\frac{\partial \Phi}{\partial b} = 2(m + b - 1) + 2(3m + b - 2) + 2(5m + b - 4) = 0$$

cancel the 2's, collect terms,

$$\begin{aligned} 35m + 9b &= 27 \\ 9m + 3b &= 7 \end{aligned} \Rightarrow \begin{bmatrix} 35 & 9 \\ 9 & 3 \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} = \begin{bmatrix} 27 \\ 7 \end{bmatrix}$$

$$\begin{bmatrix} \hat{m} \\ \hat{b} \end{bmatrix} = \begin{bmatrix} 35 & 9 \\ 9 & 3 \end{bmatrix}^{-1} \begin{bmatrix} 27 \\ 7 \end{bmatrix} = \begin{bmatrix} 0.7500 \\ 0.0833 \end{bmatrix}$$

Normal Equations (Symmetric)

Solve with matlab
 $N = [35 \ 9; 9 \ 3];$
 $t = [27; 7];$
 $\Delta = \text{inv}(N) * t$

plug into (*) to determine the V 's

$$V_1 = 0.7500 \times 1 + 0.0833 - 1 = -0.1667$$

$$V_2 = 0.7500 \times 3 + 0.0833 - 2 = 0.3333$$

$$V_3 = 0.7500 \times 5 + 0.0833 - 4 = -0.1667$$

corrected observations \Rightarrow

$$\begin{bmatrix} \hat{Y}_1 \\ \hat{Y}_2 \\ \hat{Y}_3 \end{bmatrix} = \begin{bmatrix} Y_1 + V_1 \\ Y_2 + V_2 \\ Y_3 + V_3 \end{bmatrix} = \begin{bmatrix} 0.8333 \\ 2.3333 \\ 3.8333 \end{bmatrix}$$