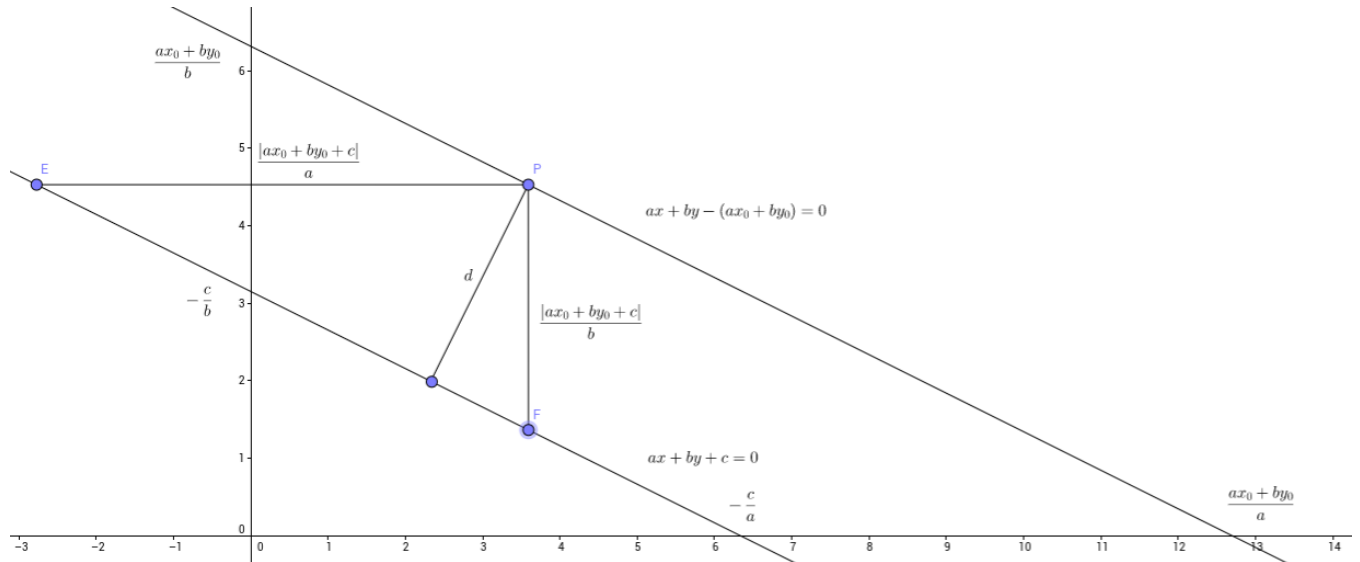


An overview of least squares line fitting is presented on the slide 13 of Lecture 4. Study it in detail and present the derivation with the following stages:

- The distance from a point to a line is the perpendicular distance of the point to the line.



$ax + by - (ax_0 + by_0) = 0$ . The legs of the right triangle EPF can be obtained by the differences between the interception of x and y-axes of the two lines. In the figure,

Since we are only interested in the magnitude, we have

The hypotenuse EF is calculated using Pythagorean theorem:

Denote  $d$  the distance of  $P$  to the line. Due to the area equality of the right triangle, we have:

$$d \cdot EF = EP \cdot PF \Rightarrow d|ax_0 + by_0 + c| \frac{\sqrt{a^2 + b^2}}{ab} = \frac{|ax_0 + by_0 + c|}{b} \frac{|ax_0 + by_0 + c|}{a}$$

$$\begin{aligned}
&\Rightarrow d |ax_0 + by_0 + c| \frac{\sqrt{a^2 + b^2}}{ab} = \frac{|ax_0 + by_0 + c|^2}{ab} \\
&\Rightarrow d |ax_0 + by_0 + c| \sqrt{a^2 + b^2} = |ax_0 + by_0 + c|^2 \\
&\Rightarrow d = \frac{|ax_0 + by_0 + c|^2}{|ax_0 + by_0 + c| \sqrt{a^2 + b^2}} = \frac{|ax_0 + by_0 + c|}{\sqrt{a^2 + b^2}} \text{ (proven)}
\end{aligned}$$

2) Thus, given  $n$  points  $(x_i, y_i)$ ,  $i = 1, \dots, n$ , the sum of squared distances between the points and the line is  $E = \sum_{i=1}^n (ax_i + by_i - d)^2$ . In order to find the minimum of  $E$ , compute the partial derivative  $\partial E / \partial d$ , set it to zero, and solve  $d$  in terms of  $a$  and  $b$ .

The partial derivative of the sum of squared differences with respect to  $d$  is:

$$\begin{aligned}
\frac{\partial E}{\partial d} &= \sum_{i=1}^n (ax_i + by_i - d)^2 \frac{\partial}{\partial d} = \sum_{i=1}^n (a^2 x_i^2 + b^2 y_i^2 + 2abx_i y_i - 2ax_i d - 2by_i d + d^2) \frac{\partial}{\partial d} \\
&= \sum_{i=1}^n (-2ax_i - 2by_i - 2d) = \sum_{i=1}^n -2(ax_i + by_i - d)
\end{aligned}$$

The minimum of  $E$  when the partial derivative is equal to 0:

$$\begin{aligned}
\frac{\partial E}{\partial d} &= \sum_{i=1}^n -2(ax_i + by_i - d) = 0 \Rightarrow \sum_{i=1}^n d - \sum_{i=1}^n ax_i - \sum_{i=1}^n by_i = 0 \\
&\Rightarrow nd - a \sum_{i=1}^n x_i - b \sum_{i=1}^n y_i = 0 \Rightarrow nd = a \sum_{i=1}^n x_i + b \sum_{i=1}^n y_i \\
&\Rightarrow d = a \frac{1}{n} \sum_{i=1}^n x_i + b \frac{1}{n} \sum_{i=1}^n y_i \Rightarrow d = a\bar{x} + b\bar{y} \text{ (answer)}
\end{aligned}$$

3) Substitute the expression obtained for  $d$  to the formula of  $E$ , and show that then  $E = (a \ b)U^T U(a \ b)^T$ , where matrix  $U$  depends on the point coordinates  $(x_i, y_i)$ .

Plugging in  $d$  from part (2) into the error formula:

$$E = \sum_{i=1}^n (ax_i + by_i - (a\bar{x} + b\bar{y}))^2 = \sum_{i=1}^n (a(x_i - \bar{x}) + b(y_i - \bar{y}))^2$$

$$\text{Let } U = \begin{bmatrix} x_1 - \bar{x} & y_1 - \bar{y} \\ \vdots & \vdots \\ x_n - \bar{x} & y_n - \bar{y} \end{bmatrix} \Rightarrow E = \left\| \begin{bmatrix} x_1 - \bar{x} & y_1 - \bar{y} \\ \vdots & \vdots \\ x_n - \bar{x} & y_n - \bar{y} \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} \right\|^2 = \left( U \begin{bmatrix} a \\ b \end{bmatrix} \right)^T \left( U \begin{bmatrix} a \\ b \end{bmatrix} \right)$$

$$\Rightarrow E = \left( U \begin{bmatrix} a \\ b \end{bmatrix} \right)^T \left( U \begin{bmatrix} a \\ b \end{bmatrix} \right) = [a \ b] U^T U \begin{bmatrix} a \\ b \end{bmatrix} \text{ (proven)}$$

4) Thus, the task is to minimize  $\|U(a \ b)^T\|$  under the constraint  $a^2 + b^2 = 1$ . The solution for  $(a \ b)^T$  is the eigenvector of  $U^T U$  corresponding to the smallest eigenvalue, and  $d$  can be solved thereafter using the expression obtained above in the stage two.

We can use the Lagrange multiplier technique to find the answer