Least squares Sample Subtitle

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Showing why least squares

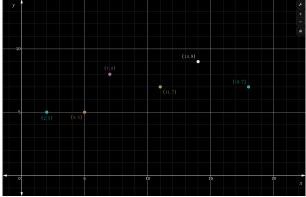
We have a variable y. We know that it's dependent of another variable x in some way. But we don't know how, exactly. So we go to the field and measure certain points. Those that we can reach. Six of them.

$$(2,5), (5,5), (7,8), (11,7), (14,9), (18,7)$$

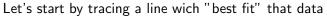
That is: at x = 2 we measure y = 5, at x = 5 we measure y = 5 again, but at x = 7 we got y = 8, and so on.

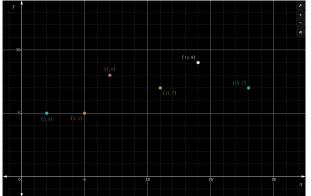
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Next step, obviously, is to plot these points.

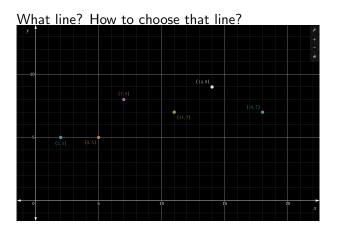


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A very common way to express a line is using the equation (a polinomial function)

$$y = mx + b$$

Using vector notation:

$$y = \begin{bmatrix} x & 1 \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix}$$

Or, switching terms:

$$\begin{bmatrix} x & 1 \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} = y$$

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Ok. Fine. That's it. This form

$$\begin{bmatrix} x & 1 \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} = y$$

will be a useful one because we know x and y in six points. For example take the first point (2,5)

$$\begin{bmatrix} 2 & 1 \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} = 5$$

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And yes, your guess is true. We can incorporate the second point (5,5) and get

$$\begin{bmatrix} 2 & 1 \\ 5 & 1 \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} = \begin{bmatrix} 5 \\ 5 \end{bmatrix}$$

...and the third point...and the fourth...

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Reaching the end this is the result

$$\begin{bmatrix} 2 & 1 \\ 5 & 1 \\ 7 & 1 \\ 11 & 1 \\ 14 & 1 \\ 18 & 1 \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} = \begin{bmatrix} 5 \\ 8 \\ 7 \\ 9 \\ 7 \end{bmatrix}$$

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Time to name things.

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Call A to the matrix

$$A = \begin{bmatrix} 2 & 1 \\ 5 & 1 \\ 7 & 1 \\ 11 & 1 \\ 14 & 1 \\ 18 & 1 \end{bmatrix}$$

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Call x to the column vector

$$x = \begin{bmatrix} m \\ b \end{bmatrix}$$

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Call b to the column vector

$$b = egin{bmatrix} 5 \ 5 \ 8 \ 7 \ 9 \ 7 \end{bmatrix}$$

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Thus, the equation

$$\begin{bmatrix} 2 & 1 \\ 5 & 1 \\ 7 & 1 \\ 11 & 1 \\ 14 & 1 \\ 18 & 1 \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} = \begin{bmatrix} 5 \\ 5 \\ 8 \\ 7 \\ 9 \\ 7 \end{bmatrix}$$

becomes

$$Ax = b$$

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In other words: if you want to know the values m and b in the equation y = mx + b of the line we are searching find the x such that Ax = b, but...

- ...that equation have no solution!...
- ...because the matrix A is tall, so Ax = b is over-determined, there are more equations (m) than variables to choose. There is no x such that the norm of the difference between Ax and b equals zero

$$||Ax - b|| \neq 0$$

We need the least squares approximate solution

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In the least squares approximate solution of Ax = b we, provided that $||Ax - b|| \neq 0$, select x that the *norm* of the *residual* r = Ax - b be minimal.

It's all here: http://vmls-book.stanford.edu/ and in many other sources. The least squares approximate solution is very known. Let's see it.

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All start with our equation Ax = b.

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Take A,

$$A = \begin{bmatrix} 2 & 1 \\ 5 & 1 \\ 7 & 1 \\ 11 & 1 \\ 14 & 1 \\ 18 & 1 \end{bmatrix}$$

transpose it.

$$A^{\mathsf{T}} = \begin{bmatrix} 2 & 5 & 7 & 11 & 14 & 18 \\ 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

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Multiply A^{T} by A

$$A^{\mathsf{T}}A = \begin{bmatrix} 2 & 5 & 7 & 11 & 14 & 18 \\ 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 5 & 1 \\ 7 & 1 \\ 11 & 1 \\ 14 & 1 \\ 18 & 1 \end{bmatrix} = \begin{bmatrix} 719 & 57 \\ 57 & 6 \end{bmatrix}$$

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Invert the product

$$(A^{\mathsf{T}}A)^{-1} = \begin{bmatrix} 719 & 57 \\ 57 & 6 \end{bmatrix}^{-1} = \begin{bmatrix} \frac{2}{355} & -\frac{19}{355} \\ -\frac{19}{355} & \frac{719}{1065} \end{bmatrix}$$

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Multiply inverse and transpose

$$(A^{\mathsf{T}}A)^{-1}A^{\mathsf{T}} = \begin{bmatrix} \frac{2}{355} & -\frac{19}{355} \\ -\frac{19}{355} & \frac{719}{1065} \end{bmatrix} \begin{bmatrix} 2 & 5 & 7 & 11 & 14 & 18 \\ 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} -\frac{3}{71} & -\frac{9}{355} & -\frac{1}{71} & \frac{3}{355} & \frac{9}{355} & \frac{17}{355} \\ \frac{121}{213} & \frac{434}{1065} & \frac{64}{213} & \frac{92}{1065} & -\frac{79}{1065} & -\frac{307}{1065} \end{bmatrix}$$

Let's call this product $(A^{T}A)^{-1}A^{T}$ the *pseudo-inverse*

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Finally multiply the pseudo-inverse by b

$$x = \begin{bmatrix} -\frac{3}{71} & -\frac{9}{355} & -\frac{1}{71} & \frac{3}{355} & \frac{9}{355} & \frac{17}{355} \\ \frac{121}{213} & \frac{434}{1065} & \frac{64}{213} & \frac{92}{1065} & -\frac{79}{1065} & -\frac{307}{1065} \end{bmatrix} \begin{bmatrix} 5\\5\\8\\7\\9\\7 \end{bmatrix} = \begin{bmatrix} \frac{61}{355}\\ \frac{5539}{1065} \end{bmatrix}$$

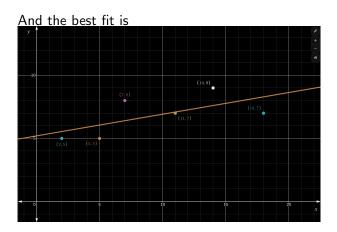
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$$x = \begin{bmatrix} m \\ b \end{bmatrix} = \begin{bmatrix} \frac{61}{355} \\ \frac{5539}{1065} \end{bmatrix} \text{ or } \begin{bmatrix} 0.171831 \\ 5.20094 \end{bmatrix}$$

and m = 0.171831 and b = 5.20094 and the line is

$$y = 0.171831x + 5.20094$$

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Recapitulation

You have a bunch of data points

You decide derive a polinomial function y = Ax + B from this points.

You build a matrix A and a vector b from the polinomial and the points.

The polinomial coefficients are a vector x so Ax = bThen $x = (A^TA)^{-1}A^Tb$

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The polinomial function

In the example we picked a line for regression.

A line in a plane is a polinomial function where y (the dependent variable, a real number) is a function of x (the independent variable, a real number)

$$y: \mathbb{R} \to \mathbb{R} = Ax + B$$