

Mathematics II

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1 / 14

L-19

1 Highlights of Lesson 19

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- Mean
- Standard deviation and variance
- Ordinary Least Squares (OLS)

2 / 14

L-19

You can find the last version of these course materials at

<https://github.com/mbujosab/MatematicasII/tree/main/Eng>

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1 / 14

L-19

2 Restriction in statistics and probability

Norm of constant vector “one” is 1

This fails using the dot product in \mathbb{R}^m ($m > 1$)

$$\|\mathbf{1}\|^2 = \langle \mathbf{1}, \mathbf{1} \rangle = \mathbf{1} \cdot \mathbf{1} = \sum_{i=1}^m 1 = m.$$

New scalar product in \mathbb{R}^m for statistics

$$\langle \mathbf{x}, \mathbf{y} \rangle_s = \frac{1}{m} (\mathbf{x} \cdot \mathbf{y})$$

$$(\text{so: } \|\mathbf{1}\|^2 = \frac{1}{m} (\mathbf{1} \cdot \mathbf{1}) = 1)$$

3 / 14

3 Mean

The mean μ_y is the scalar product of \mathbf{y} and $\mathbf{1}$

$$\mu_y = \frac{1}{m}(\mathbf{1} \cdot \mathbf{y}), \quad \text{so, } \mu_y = \frac{1}{m} \sum_{i=1}^m y_i$$

The mean μ_y is the *value* by which to multiply $\mathbf{1}$ to get the orthogonal projection of \mathbf{y} onto $\mathcal{L}([\mathbf{1};])$

μ_y : projection of $\mathbf{y} \in \mathbb{R}^m$ onto the line $\mathcal{L}([\mathbf{1};]) \subset \mathbb{R}^m$

$$\boxed{\mu_y = \mathbf{1}\hat{a}} \quad \text{and} \quad (\mathbf{y} - \mu_y) \perp \mathbf{1} \Rightarrow \frac{1}{m}(\mathbf{y} - \mu_y) \cdot \mathbf{1} = 0$$

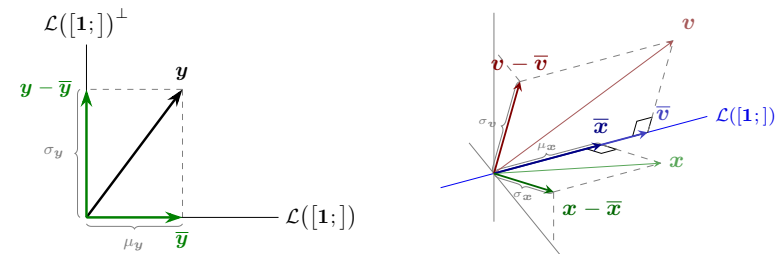
$$\frac{1}{m}(\mathbf{y} - \mathbf{1}\hat{a}) \cdot \mathbf{1} = 0 \Leftrightarrow \frac{1}{m}(\mathbf{y} \cdot \mathbf{1}) - \frac{1}{m}(\mathbf{1} \cdot \mathbf{1})\hat{a} = 0;$$

Therefore

$$\hat{a} = \frac{1}{m}(\mathbf{y} \cdot \mathbf{1}) = \mu_y$$

4 / 14

4 Mean

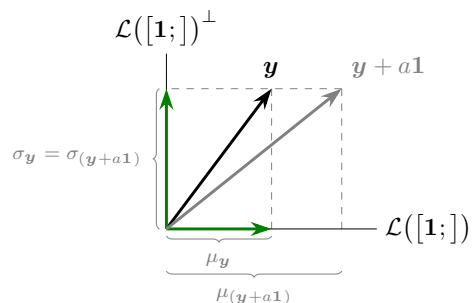


5 / 14

5 Standard deviation

$$\sigma_y = \|\mathbf{y} - \mu_y\|.$$

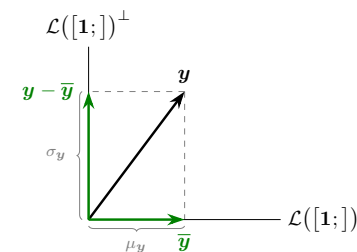
Adding a constant vector $a\mathbf{1}$ to \mathbf{y} does not change the standard deviation.



6 / 14

6 Variance and the Pythagorean theorem

$$\sigma_y^2 = \|\mathbf{y} - \mu_y\|^2 = \frac{1}{m}(\mathbf{y} - \mu_y) \cdot (\mathbf{y} - \mu_y) = \frac{1}{m} \sum_i (y_i - \mu_y)^2.$$

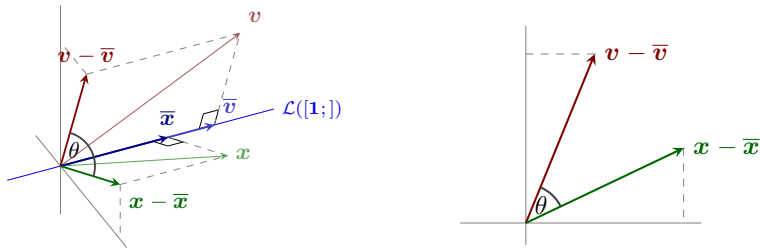


$$\sigma_y^2 = \|\mathbf{y} - \mu_y\|^2 = \|\mathbf{y}\|^2 - \|\mu_y\|^2 = \frac{1}{m}(\mathbf{y} \cdot \mathbf{y}) - \mu_y^2 = \frac{\sum_i y_i^2}{m} - \mu_y^2.$$

7 / 14

7 Covariance and correlation

$$\sigma_{xy} = \frac{1}{m}(\mathbf{x} - \boldsymbol{\mu}_x) \cdot (\mathbf{y} - \boldsymbol{\mu}_y);$$



$$\rho_{xy} = \frac{\frac{1}{m}(\mathbf{x} - \boldsymbol{\mu}_x) \cdot (\mathbf{y} - \boldsymbol{\mu}_y)}{\|(\mathbf{x} - \boldsymbol{\mu}_x)\| \cdot \|(\mathbf{y} - \boldsymbol{\mu}_y)\|} = \frac{\sigma_{xy}}{\sqrt{\sigma_x \sigma_y}} = \cos(\theta).$$

8 / 14

8 Ordinary Least Squares (OLS)

Let \mathbf{X} such that $\mathcal{L}([1;]) \subset \mathcal{C}(\mathbf{X})$.

$\hat{\mathbf{y}}$ is the orthogonal projection of $\mathbf{y} \in \mathbb{R}^m$ onto $\mathcal{C}(\mathbf{X})$

$$\hat{\mathbf{y}} = \mathbf{X}\hat{\boldsymbol{\beta}} \quad \text{and} \quad (\mathbf{y} - \hat{\mathbf{y}}) \perp \mathcal{C}(\mathbf{X}) \Rightarrow \frac{1}{m}\mathbf{X}^\top(\mathbf{y} - \hat{\mathbf{y}}) = \mathbf{0}$$

$$\frac{1}{m}\mathbf{X}^\top(\mathbf{y} - \mathbf{X}\hat{\boldsymbol{\beta}}) = \mathbf{0} \iff \frac{1}{m}\mathbf{X}^\top\mathbf{y} - \frac{1}{m}\mathbf{X}^\top\mathbf{X}\hat{\boldsymbol{\beta}} = \mathbf{0}.$$

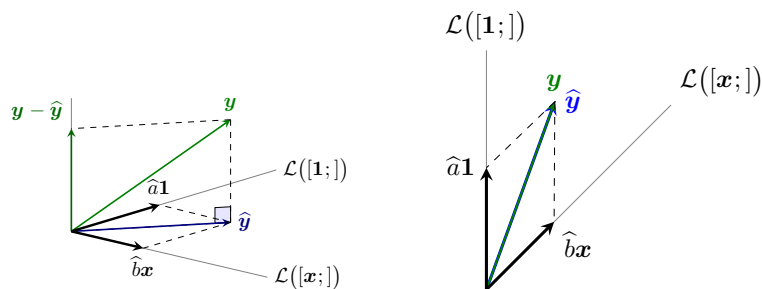
Therefore

$$\left(\frac{1}{m}\mathbf{X}^\top\mathbf{X}\right)\hat{\boldsymbol{\beta}} = \frac{1}{m}\mathbf{X}^\top\mathbf{y}.$$

9 / 14

9 Ordinary Least Squares (OLS)

If $\mathbf{X} = [1; x;]$ has rank 2.



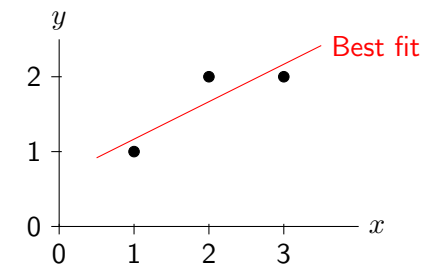
$$\left(\frac{1}{m}\mathbf{X}^\top\mathbf{X}\right)\begin{pmatrix} \hat{a} \\ \hat{b} \end{pmatrix} = \frac{1}{m}\mathbf{X}^\top\mathbf{y}.$$

10 / 14

10 Application: Least Squares (Fitting by a line)

"looking for the best fitting line $\hat{y} = \hat{a} + \hat{b}x$ "

Points (x, y) : $(1, 1)$; $(2, 2)$; $(3, 2)$



$$\begin{cases} a + 1b = 1 \\ a + 2b = 2 \\ a + 3b = 2 \end{cases} \rightarrow \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 3 \end{bmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \\ 2 \end{pmatrix} \quad (\mathbf{X}\boldsymbol{\beta} = \mathbf{y} \text{ No solution})$$

11 / 14

11 Application: Least Squares (Fitting by a line)

$$\mathbf{X}\boldsymbol{\beta} = \mathbf{y} \quad (\text{No solution}) \rightarrow \left(\frac{1}{m}\mathbf{X}^T\mathbf{X}\right)\hat{\boldsymbol{\beta}} = \frac{1}{m}\mathbf{X}^T\mathbf{y} \rightarrow \hat{\mathbf{y}} = \mathbf{X}\hat{\boldsymbol{\beta}}.$$

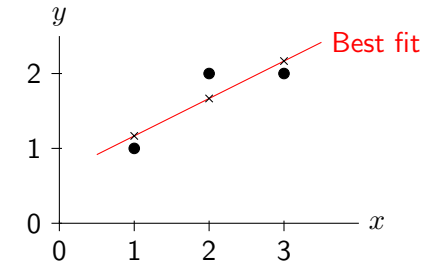
$$\begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 3 \end{bmatrix} \begin{pmatrix} \hat{a} \\ \hat{b} \end{pmatrix} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \end{bmatrix} \begin{pmatrix} 1 \\ 2 \\ 2 \end{pmatrix}$$

$$\begin{bmatrix} 3 & 6 \\ 6 & 14 \end{bmatrix} \begin{pmatrix} \hat{a} \\ \hat{b} \end{pmatrix} = \begin{pmatrix} 5 \\ 11 \end{pmatrix} \Rightarrow \hat{a} = \frac{2}{3}; \quad \hat{b} = \frac{1}{2}.$$

$$\text{Best solution: } \frac{2}{3} + \frac{1}{2}x$$

12 / 14

12 Application: Least Squares (Fitting by a line)



$$\hat{\mathbf{y}} = \mathbf{X}\hat{\boldsymbol{\beta}} = \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 3 \end{bmatrix} \begin{pmatrix} \hat{a} \\ \hat{b} \end{pmatrix} \rightarrow \hat{\mathbf{y}} = \begin{pmatrix} 7/6 \\ 10/6 \\ 13/6 \end{pmatrix} \rightarrow \hat{\mathbf{e}} = \begin{pmatrix} -1/6 \\ 2/6 \\ -1/6 \end{pmatrix}$$

$$\mathbf{y} = \hat{\mathbf{y}} + \hat{\mathbf{e}} \quad \text{and} \quad \begin{cases} \hat{\mathbf{e}} \cdot \hat{\mathbf{y}} = 0 \\ \hat{\mathbf{e}}\mathbf{X} = \mathbf{0} \end{cases}.$$

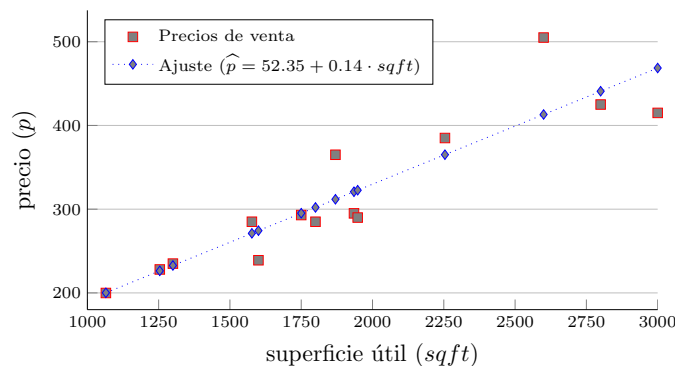
13 / 14

13 Application: Least Squares (Fitting by a line)

Selling price and living area of single family homes in University City community of San Diego, in 1990.

price = Sale price is in thousands of dollars

sqft = Square feet of living area (Ramanathan, 2002, pp. 78)



14 / 14

Questions of the Lecture 19

(L-19) QUESTION 1. With the measurements $\mathbf{y} = (0, 8, 8, 20,)$ at $\mathbf{x} = (0, 1, 3, 4,)$,

- Set up and solve the normal equations $\mathbf{A}^T\mathbf{A}\hat{\boldsymbol{\beta}} = \mathbf{A}^T\mathbf{y}$.
 - For the best straight line, find its four fits p_i and four errors e_i .
 - What is the value of the square of the norm of the error vector $\|\mathbf{e}\|^2 = e_1^2 + e_2^2 + e_3^2 + e_4^2$?
 - Draw the regression line
 - Change the measurements to $\mathbf{p} = (1, 5, 13, 17,)$ write down the four equations $\mathbf{A}\boldsymbol{\beta} = \mathbf{p}$. Find an exact solution to $\mathbf{A}\boldsymbol{\beta} = \mathbf{p}$
 - Check that $\mathbf{e} = \mathbf{y} - \mathbf{p} = (-1, 3, -5, 3,)$ is perpendicular to both columns of the same matrix \mathbf{A} .
 - What is the shortest distance $\|\mathbf{e}\|$ from \mathbf{y} to the column space of \mathbf{A} ?
- (Strang, 2003, exercise 1–3 from section 4.3.)

(L-19) QUESTION 2.

- Write down three equations $y = \alpha + \beta x$ given the data: $y = 7$ at $x = -1$, $y = 7$ at $x = 1$, and $y = 21$ at $x = 2$. Find the least squares solution $\hat{\boldsymbol{\beta}} = (\hat{\alpha}, \hat{\beta})$ and draw the closest line.
- Find the projection $\mathbf{p} = \mathbf{A}\hat{\boldsymbol{\beta}}$. This gives the three heights of the closest line. Show that the error vector is $\mathbf{e} = (2, -6, 4,)$. Why is $\mathbf{P}\mathbf{e} = \mathbf{0}$?

14 / 14

(L-19) **QUESTION 3.** Our measurements at times $t = 1, 2, 3$ are $b = 1, 4$, and b_3 . We want to fit those points by the nearest line $C + Dt$, using least squares.

- (a) Which value for b_3 will put the three measurements on a straight line? *Which line is it?* Will least squares choose that line if the third measurement is $b_3 = 9$? (Yes or no).
- (b) What is the linear system $\mathbf{A}\mathbf{x} = \mathbf{b}$ that would be solved exactly for $\mathbf{x} = (C, D)$ if the three points do lie on a line? Compute the projection matrix \mathbf{P} onto the column space of \mathbf{A} .
- (c) What is the rank of that projection matrix \mathbf{P} ? How is the column space of \mathbf{P} related to the column space of \mathbf{A} ? (You can answer with or without the entries of \mathbf{P} computed in (b).)
- (d) Suppose $b_3 = 1$. Write down the equation for the best least squares solution $\hat{\mathbf{x}}$, and show that the best straight line is horizontal.

Ramanathan, R. (2002). *Introductory Econometrics with applications*. South-Western, Mason, Ohio, fifth ed. ISBN 0-03-034186-8.

Strang, G. (2003). *Introduction to Linear Algebra*. Wellesley-Cambridge Press, Wellesley, Massachusetts. USA, third ed. ISBN 0-9614088-9-8.