

# Link 07/05/2020

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- Link alle due lezioni registrate:
  - [https://drive.google.com/open?id=1juVxhpCZDPCA\\_J5pDnL08f6\\_BILqAF1e](https://drive.google.com/open?id=1juVxhpCZDPCA_J5pDnL08f6_BILqAF1e)

## Es. 5.8 (5.19)

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**5.19. ★★** A student measures a quantity  $y$  many times and calculates his mean as  $\bar{y} = 23$  and his standard deviation as  $\sigma_y = 1$ . What fraction of his readings would you expect to find between

(a) 22 and 24?

(d) 21 and 23?

(b) 22.5 and 23.5?

(e) 24 and 25?

(c) 21 and 25?

Finally, (f) within what limits (equidistant on either side of the mean) would you expect to find 50% of his readings? (The necessary information for all parts of this question is in Figure 5.13. More detailed information on these kinds of probabilities is in Appendixes A and B.)

$\bar{y} = 23$     $\sigma_y = 1$     $t = 0.7$     $22.3 \div 23.7$    07/05/2020   (1)

a)  $22 \div 24$     $\bar{y} = 23$   
 $\downarrow$     $\downarrow$   
 $-1\sigma$     $+1\sigma$     $\bar{y} \pm 1\sigma$    68%

b)  $22.5 \div 23.5$     $(23) \pm 0.5$   
 $t = 1/2$    38%

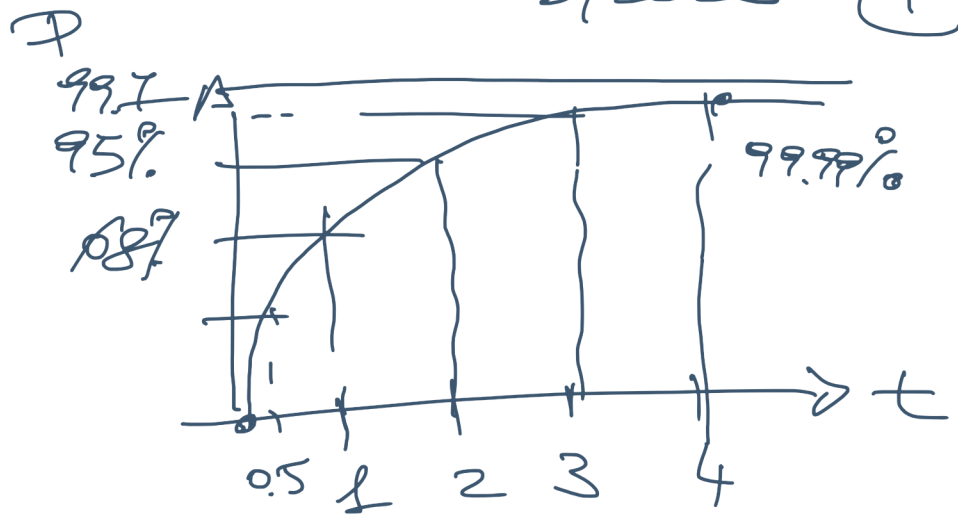
c)  $22 \div 25$     $(23) \pm 2$   
 $t = 2$     $\frac{95\%}{95.45\%}$

e)  $24 \div 25$     $(23)$     $1\sigma$     $2\sigma$

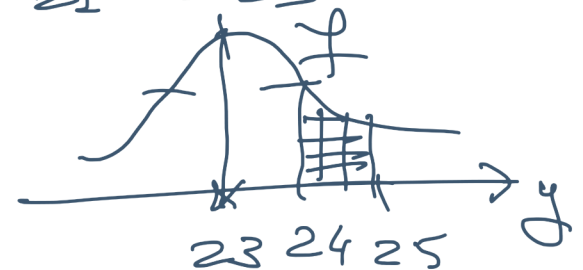
Probab.  $(23 \div 25) = 48\%$

Probab.  $(23 \div 24) = \frac{68\%}{2} = 34\%$

Probab.  $= 48\% - 34\% = 14\%$



d)  $21 \div 23$     $\bar{y} = 23$   
 $\Rightarrow 2\sigma$     $1\sigma$



$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$(x_1, \dots, x_n)$$

$$\bar{x} = \frac{\sum x_i}{n}$$

mit jeder  
Stimme di

$$\frac{\sum (x_i - \bar{x})^2}{n-1}$$

$x, y, z, \dots$

$$(x, \sigma_x) \quad (y, \sigma_y) \quad (z, \sigma_z)$$

$$q = f(x, y, z) \quad q, \sigma_q$$

$$q = x + A \quad x, \sigma_x$$

$$x = q - A \quad \Rightarrow \quad e^{-\frac{(q-A-x)^2}{2\sigma_x^2}} = e^{-\frac{(q-(x+A))^2}{2\sigma_x^2}}$$

$$q = x + A \quad \sigma_q = \sigma_x$$

$$Q = B \cdot X \quad X = Q$$

$$P \propto e^{-\frac{(Q - B \cdot X)^2}{2\sigma_x^2}} = P \frac{1}{B^2} \frac{1}{\sigma_x^2} (Q - B \cdot X)^2 = P \frac{1}{2(B\sigma_x)^2} (Q - B \cdot X)^2$$

$$Q = B \cdot X \quad \sigma_Q = B \sigma_x$$

$$Q = X + Y$$

$$P(Q) = P(X) \cdot P(Y) =$$

$$P \frac{1}{\sigma_x} e^{-\frac{(x-x)^2}{2\sigma_x^2}} \cdot P \frac{1}{\sigma_y} e^{-\frac{(y-y)^2}{2\sigma_y^2}} = P \frac{1}{\sigma_x \sigma_y} e^{-\left[ \frac{(x-x)^2}{2\sigma_x^2} + \frac{(y-y)^2}{2\sigma_y^2} \right]}$$

$$P \frac{1}{\sigma_x \sigma_y} e^{-\frac{(x+y)^2 - (x-y)^2}{2\sigma_x^2}}$$

$$X, Y = 0$$

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$$e^{-\left(\frac{x^2}{a^2} + \frac{y^2}{b^2}\right)} +$$

$$A = a^2$$

$$B = b^2$$

(4)

$$\frac{x^2}{A} + \frac{y^2}{B} = \frac{(x+y)^2}{A+B} + \frac{(Bx - Ay)^2}{AB(A+B)}$$

$$\frac{AB(x^2 + 2xy + y^2) + (B^2x^2 - 2ABxy + A^2y^2)}{AB(A+B)} =$$

$$AB(A+B)$$

$$= \cancel{ABx^2} + \cancel{2ABxy} + \cancel{AB y^2} + \cancel{B^2x^2} - \cancel{2ABxy} + \cancel{A^2y^2}$$

$$= \frac{AB(A+B)}{AB(A+B)} \left( x^2 + \frac{B}{A}x^2 + y^2 + \frac{A}{B}y^2 \right) = \frac{x^2 \cancel{(A+B)} + y^2 \cancel{(B+A)}}{A+B}$$

$$= \frac{x^2}{A} + \frac{y^2}{B}$$

$$A x^2 + B y^2 = \frac{(x+y)^2}{A+B} + \frac{(Bx - Ay)^2}{AB(A+B)}$$

(5)

$$P - \left( \frac{A x^2}{2} + \frac{B y^2}{2} \right)$$

$$A = \sigma_x^2 \quad B = \sigma_y^2$$

$$g = x + y$$

$$P(g) = P(x) \cdot P(y)$$

$$P - \left( \frac{A x^2}{2} + \frac{B y^2}{2} \right)$$

$$P(g) = P - \left[ \frac{(x+y)^2}{A+B} + \frac{z^2}{2} \right]$$

$$= P - \frac{(x+y)^2}{2(A+B)} - \frac{z^2}{2}$$

$$P - \frac{(x+y)^2}{2(A+B)} = P - \frac{z^2}{2}$$

$x + y = g$

$$\sigma_g^2 = A + B = \sigma_x^2 + \sigma_y^2$$

$$\sqrt{2\sigma^2}$$

$$f(x, y) \approx f(x_0, y_0) + \left. \frac{\partial f}{\partial x} \right|_{x_0, y_0} (x - x_0) + \left. \frac{\partial f}{\partial y} \right|_{x_0, y_0} (y - y_0) \quad (6)$$

A diagram showing the Taylor expansion formula. The terms  $f(x, y)$ ,  $f(x_0, y_0)$ ,  $(x - x_0)$ ,  $\left. \frac{\partial f}{\partial x} \right|_{x_0, y_0}$ , and  $(y - y_0)$  are circled. The word "Taylor" is written below the partial derivative terms. Arrows point from the words "numero" and "piccolo" to the points  $(x_0, y_0)$  and the partial derivative terms respectively.

$$A + B(x - x_0) + C(y - y_0)$$

$\sigma_x$   $\sigma_y$

$(x_0, y_0)$

$(x, y)$

$$\sigma_f^2 = B^2 \sigma_x^2 + C^2 \sigma_y^2$$

$$= \left. \left( \frac{\partial f}{\partial x} \right)^2 \right|_{x_0, y_0} \sigma_x^2 + \left. \left( \frac{\partial f}{\partial y} \right)^2 \right|_{x_0, y_0} \sigma_y^2$$



$$(x_1, \dots, x_n)$$

$$D_x$$

$$e^{-\frac{(x-\bar{x})^2}{2\sigma^2}}$$

(7)

$$x \neq \sigma$$

$$x = \frac{\sum_{i=1}^n x_i}{n}$$

$$D_x^2 = \frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n-1}$$

$$\bar{x} = \frac{\sum_{i=1}^n x_i}{n}$$

$$x_1, x_2, \dots, x_n$$

$$= f(x_1, \dots, x_n)$$

$$D_x^2 = \sum_{i=1}^n \left( \frac{\partial \bar{x}}{\partial x_i} \right)^2 \cdot D_{x_i}^2$$

$$\left( \frac{\partial \bar{x}}{\partial x_1} \right)^2 D_{x_1}^2 + \dots + \left( \frac{\partial \bar{x}}{\partial x_n} \right)^2 D_{x_n}^2$$

$$\frac{\partial \bar{x}}{\partial x_1} = \frac{1}{n} = \frac{\partial \bar{x}}{\partial x_2} = \dots = \frac{\partial \bar{x}}{\partial x_n}$$

$$= \frac{1}{n^2} D_{x_1}^2 + \frac{1}{n^2} D_{x_2}^2 + \dots + \frac{1}{n^2} D_{x_n}^2$$

$$\sigma_x = \frac{\sigma}{\sqrt{N}}$$

MAXIMUM  
 LIKELIHOOD

8

CRP. 7

$$X_A \pm \sigma_A$$

$$X_B \pm \sigma_B$$

$$\frac{X_A + X_B}{2}$$

$$\sigma_A \sigma_B$$

$$\begin{aligned}
 \phi(X_A, X_B) &= P(X_A) \cdot P(X_B) \\
 &= \frac{1}{\sigma_A \sigma_B} \cdot \frac{1}{\sigma_A \sigma_B} \cdot \exp\left[-\frac{(X_A - X_A)^2}{2\sigma_A^2} - \frac{(X_B - X_B)^2}{2\sigma_B^2}\right] \\
 &= \frac{1}{\sigma_A \sigma_B} \cdot \frac{1}{\sigma_A \sigma_B} \cdot \exp\left[-\frac{(X_A - X_B)^2}{2\sigma_A^2} - \frac{(X_B - X_A)^2}{2\sigma_B^2}\right]
 \end{aligned}$$

$$\chi^2 = \frac{(x_A - X)^2}{\sigma_A^2} + \frac{(x_B - X)^2}{\sigma_B^2} \quad (9)$$

CHI QUADRO

$$\frac{d\chi^2}{dX} = \cancel{\frac{2(x_A - X)}{\sigma_A^2}} + \cancel{\frac{2(x_B - X)}{\sigma_B^2}} = 0$$

$$\frac{x_A}{\sigma_A^2} - \frac{X}{\sigma_A^2} + \frac{x_B}{\sigma_B^2} - \frac{X}{\sigma_B^2} = 0$$

$$w_A = \frac{1}{\sigma_A^2}$$

$$w_B = \frac{1}{\sigma_B^2}$$

$$x_A w_A + x_B w_B = X (w_A + w_B)$$

$$X = \frac{x_A w_A + x_B w_B}{w_A + w_B}$$

(weight) per  
MEDIA PESATA

$$R_1 = 11 \pm 1 \rightarrow w_1 = 1$$

$$R_2 = 12 \pm 1 \rightarrow w_2 = 1$$

$$\bar{w} = \frac{19}{9} \approx 2$$

~~$$R_3 = 10 \pm 3 \rightarrow w = \frac{1}{3}$$~~

$$\bar{R} = \frac{\sum w_i R_i}{\sum w_i} = 11.4 \pm 0.7$$

$$R_{12} = 11.5 \pm 0.7$$