

## Basic Math Review

### Some Greek letters

Greek letters are frequently used to denote either specific population properties, to name specific statistical test or as mathematical operator.

Greek Letter	Phonetic	Usage
$\alpha$	<i>alpha</i>	error of first type
$\beta$	<i>beta</i>	error of second type / regression parameters
$\varepsilon$	<i>epsilon</i>	regression population error term
$\gamma$	<i>gamma</i>	parameter in some distributions and functions
$\lambda$	<i>lambda</i>	parameter in some distributions and functions
$\mu$	<i>mu</i>	expected population mean
$\pi$	<i>pi</i>	population probability in binomial distribution
$\rho$	<i>rho</i>	population correlation coefficient
$\sigma$	<i>sigma</i>	population standard deviation
$\chi$	<i>chi</i>	$\chi^2$ -test
$\Pi$	<i>capital pi</i>	multiplication symbol
$\Sigma$	<i>capital sigma</i>	summation symbol

## Standard Symbols and Definition

Operation	Meaning
$\frac{\text{Numerator}}{\text{Denominator}}$	ratio between the numerator and the denominator
$\times$ or $\cdot$ , and $\div$ or $/$ , $+$ , $-$	multiplication and division take precedence over addition and subtraction
$X < Y$	$X$ is less than $Y$
$X \leq Y$	$X$ is less or equal than $Y$
$X \pm Y$	$X$ plus minus $Y$ , i.e., the two values $X + Y$ and $X - Y$
$ X $	The absolute value $ X  = \begin{cases} X & \text{for } X \geq 0 \\ -X & \text{for } X < 0 \end{cases}$
$\frac{1}{X} = X^{-1}$	Reciprocal of $X$
$X^n$	The $n^{\text{th}}$ power of $X$ ; for $n$ being integer we get $x^n = \underbrace{x \cdot x \cdots x}_{n\text{-times}}$
$\sqrt{X} = X^{\frac{1}{2}}$	square root of $X$
$i \in \{1, 2, \dots, n\}$	$i$ is an element in the set $\{1, 2, \dots, n\}$ . It takes the values <b>1, 2, ... n</b> .

## Notation of Variables

A variable is denoted by a capital letter  $X$  while a lower case letter with a subscript index  $x_i$  relates to a specific observation. The index  $i$  ranges from  $1, 2, \dots, n$ . The number of observations in a variable is  $n$ . Therefore,  $X =$

$$\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}. \text{ For example, if } X \text{ has } n = 4 \text{ observation then } X = \begin{pmatrix} x_1 = 3 \\ x_2 = 2 \\ x_3 = 5 \\ x_4 = 4 \end{pmatrix}.$$

## Ranked Data

- Statisticians frequently work with an ascending sorted sequence of observations which is denoted by square

brackets  $X_{[ranked]} = \begin{pmatrix} x_{[1]} \\ x_{[2]} \\ \vdots \\ x_{[n]} \end{pmatrix}$ . For example,  $X_{[ranked]} = \begin{pmatrix} x_{[1]} = 2 \\ x_{[2]} = 3 \\ x_{[3]} = 4 \\ x_{[4]} = 5 \end{pmatrix}$  with  $x_{[1]} \leq x_{[2]} \leq x_{[3]} \leq x_{[4]}$ .

- Should two observations have the same rank, such as  $x_i = 5$  and  $x_j = 5$ , then the ranks  $[r]$  and  $[r + 1]$  will be assigned arbitrarily, that is,  $x_{[r]} = 5$  and  $x_{[r+1]} = 5$ , respectively.

## Basic Summation $\sum$ – Rules:

- $\sum_{i=1}^n x_i \equiv x_1 + x_2 + \dots + x_n$ . The lower index  $i = 1$  express the starting value of the summation sequence and the upper index  $n$  the value where the summation index  $i$  stops.
- more specifically  $\sum_{i=2}^5 x_i = x_2 + x_3 + x_4 + x_5$
- for a sum over a constant  $c$  we get  $\sum_{i=1}^n c = n \cdot c$
- for a mixture of a constant and a variable  $\sum_{i=1}^n c \cdot x_i = c \cdot \sum_{i=1}^n x_i$
- for an additive mixture of variables  $\sum_{i=1}^n (x_i + y_i) = \sum_{i=1}^n x_i + \sum_{i=1}^n y_i$
- Inequalities (so do not confuse either side of the expression; they lead to different results):

$$\sum_{i=1}^n x_i \cdot y_i \neq \sum_{i=1}^n x_i \cdot \sum_{i=1}^n y_i$$

$$\sum_{i=1}^n x_i^2 \neq \left( \sum_{i=1}^n x_i \right)^2$$


- Special rule used when we are dealing with ranks:  $\sum_{i=1}^n i = \frac{n}{2} \cdot (n+1)$
- Doubly index variables  $x_{ij}$  in a data table with  $I$  rows for the observations and  $J$  columns for the variables:

Let:

$$\begin{bmatrix} x_{11} & x_{12} & \cdots & x_{1j} & \cdots & x_{1J} \\ x_{21} & x_{22} & \cdots & x_{2j} & \cdots & x_{2J} \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\ x_{i1} & x_{i2} & \cdots & x_{ij} & \cdots & x_{iJ} \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\ x_{I1} & x_{I2} & \cdots & x_{Ij} & \cdots & x_{IJ} \end{bmatrix}$$

Then the  $i^{th}$  row sum is  $x_{i+} = \sum_{j=1}^J x_{ij}$  and the  $j^{th}$  column sum is  $x_{+j} = \sum_{i=1}^I x_{ij}$  and the total sum becomes

$$x_{++} = \sum_{i=1}^I \sum_{j=1}^J x_{ij} = \sum_{i=1}^I x_{i+} \text{ or } \sum_{j=1}^J x_{+j}$$

- Applications (not test relevant):
  - Rewrite:  $\sum_{i=1}^n (x_i + y_i)^2 = \sum_{i=1}^n (x_i^2 + 2 \cdot x_i \cdot y_i + y_i^2) = \sum_{i=1}^n x_i^2 + 2 \cdot \sum_{i=1}^n x_i \cdot y_i + \sum_{i=1}^n y_i^2$
- The  function for summations:
  - **sum()** calculated the sum over the elements of a vector
  - **rowSums()** calculates along the rows of a matrix a vector of row sums.
  - **colSums()** calculates along the columns of a matrix a vector of column sums.

## Finding the minimum of a quadratic function (**not test relevant**)

- In statistic, we encounter frequently the need to find an optimal value of an estimation function, such as for the central tendency.
- The minimum is found at that point, where the slope of the function is zero. The slope of a function is measured by the first derivative.
- Basic rules of derivatives:

$$\frac{\partial}{\partial x} a \cdot x^n = a \cdot n \cdot x^{n-1} \quad \text{Example: } \frac{\partial}{\partial x} 3 \cdot x^2 = 6 \cdot x$$

$$\frac{\partial}{\partial x} (f(x) + g(x)) = \frac{\partial}{\partial x} f(x) + \frac{\partial}{\partial x} g(x) \quad \text{Example: } \frac{\partial}{\partial x} (3 \cdot x^2 + 5 \cdot x^{-1}) = \underbrace{3 \cdot 2 \cdot x}_{=6} + \underbrace{5 \cdot -1}_{=-5} \cdot x^{-2}$$

- Example: The Arithmetic Mean

Which value of  $\theta$  (theta) minimizes the quadratic expression  $\sum_{i=1}^n (x_i - \theta)^2$ ?

Analysis Steps:

[a] Multiply the term under the square out:

$$f(\theta) = \sum_{i=1}^n (x_i - \theta)^2 = \sum_{i=1}^n x_i^2 - 2 \cdot \theta \cdot \sum_{i=1}^n x_i + n \cdot \theta^2$$

[b] Take the first derivative with regard to  $\theta$ , which is the slope of  $f(\theta)$  at  $\theta$ :

$$\frac{\partial}{\partial \theta} \left( \sum_{i=1}^n x_i^2 - 2 \cdot \theta \cdot \sum_{i=1}^n x_i + n \cdot \theta^2 \right) = -2 \cdot \sum_{i=1}^n x_i + 2 \cdot n \cdot \theta$$

[c] At its maximum or minimum the first derivative (that is, the slope) is zero.

Therefore set the expression to zero:

$$-2 \cdot \sum_{i=1}^n x_i + 2 \cdot n \cdot \theta = 0$$

[d] Solve the expression for the unknown parameter  $\theta$ :

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

$\Rightarrow$  This is the well-known arithmetic mean!!!

Example: The data values are  $x_i \in \{2, 5, 4, 6, 8\}$ .

Thus the function to be minimized with respect to  $\theta$  is

$$f(\theta) = (2 - \theta)^2 + (5 - \theta)^2 + (4 - \theta)^2 + (6 - \theta)^2 + (8 - \theta)^2$$

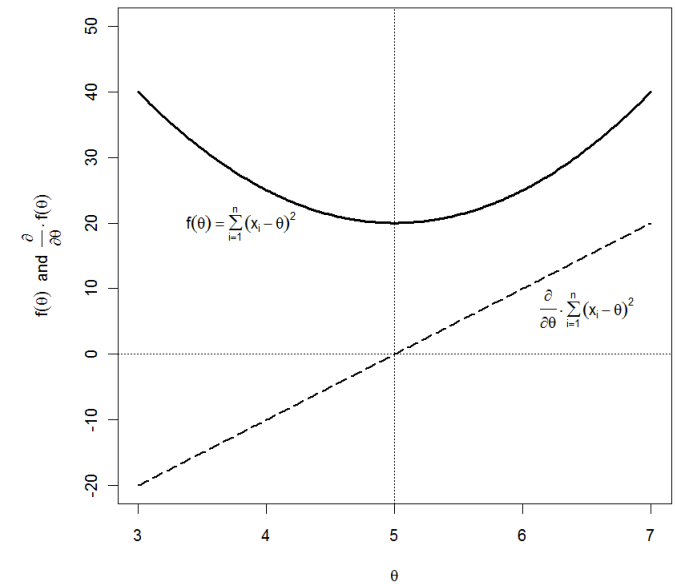
The solution is found at  $\theta \Rightarrow \bar{x} = 5$ .

## The exponential and logarithmic functions (not test relevant)

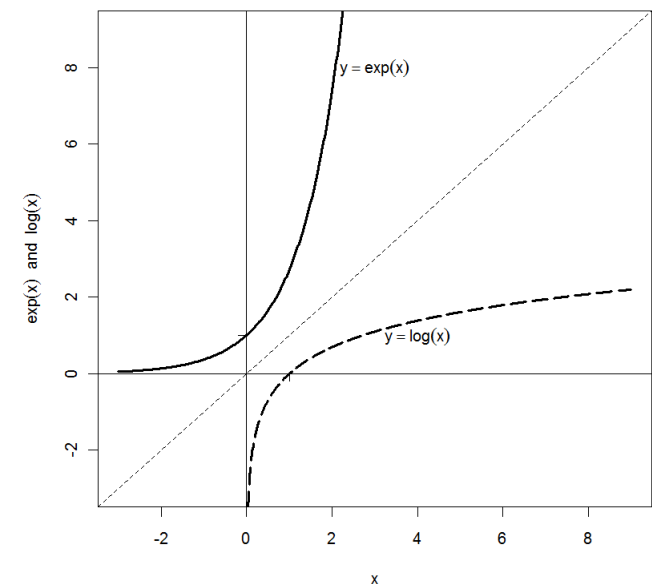
- Both functions are inversely related:  $x = \exp(\log(x))$  and  $x = \log(\exp(x))$ .

Note: the support (valid arguments for the variable  $x$ ) of the logarithmic function is limited from below by zero, that is,  $x \in ]0, \infty]$ .

Minimizing the Squared Differences Around  $\theta$  for  $x_i \in \{2, 4, 5, 6, 8\}$



Graph of the Functions  $\exp(x)$  and  $\log(x)$



- The both functions distort the constant distance units in the variable  $x$ .

E.g.,  $\Delta x_1 = 10 - 1 = 9$  and

$\Delta x_2 = 20 - 11 = 9$  but

$\log(10) - \log(1) = 2.302585$

but  $\log(20) - \log(11) = 0.597837$ , respectively.

- Basic rules:

- Logarithmic function:

$$\log(x \cdot y) = \log(x) + \log(y),$$

$$\log(x/y) = \log(x) - \log(y) \text{ and}$$

$$\log(x^y) = y \cdot \log(x)$$

- Exponential function:

$$\exp(x + y) = \exp(x) \cdot \exp(y),$$

$$\exp(x - y) = \exp(x)/\exp(y) \text{ and}$$

$$[\exp(x)]^y = \exp(x \cdot y)$$