

## DEFINITIONS

**POPULATION** : Entire aggregate of individuals or items from which SAMPLES are drawn.

**SAMPLE** : A set of individuals or items selected from a PARENT POPULATION so that properties or parametres of the population may be estimated.

**Sampling distribution of the means** : distribution of the mean of samples of particular size. **Sampling distribution of the variance** : distribution of the variance of samples.

**STATISTIC** : any function of sample data, containing no unknown parametres, such as mean, median, variance or standard deviation.

**Population's statistics** are generally indicated with Greek letters ( $\mu$ ,  $\sigma$ , etc.).

**Sample's statistics** are generally indicated with Latin letters (m, s, etc.).

**RANDOM VARIABLE** : a variable which takes values in certain range with probabilities specified by a PROBABILITY DENSITY FUNCTION or PROBABILITY MASS FUNCTION (ex. if we express head or tail as 0 or 1, the toss of a coin is a random variable).

**EXPECTED VALUE** : or **MEAN** of a random variable or a function of a variable  $E[X]$  or  $\bar{X}$  is

1.  $\sum_{i=1}^n x_i p(x_i)$  for a DISCRETE VARIABLE;
2.  $\int x f(x) dx$  for a CONTINUOUS VARIABLE.

It has the following properties:

$$E[\lambda X] = \lambda E[X]$$

$$E[X \pm Y] = E[X] \pm E[Y]$$

$$E[X^2] = E[X]^2 + \text{Var}(X)$$

$$E[XY] = E[X]E[Y] + \text{cov}(X, Y)$$

**VARIANCE** : a measure of dispersion

- *Theoretical variance*

$$\text{Var}(X) = \sum_{i=1}^n \frac{(x_i - \bar{x})^2}{n} = E[X^2] - E[X]^2$$

- *Sample variance*

$$s^2 = \frac{\sigma^2}{n}$$

**STANDARD DEVIATION** is the square root of the variation and has the same unit of the random variable

- *Theoretical standard deviation*

$$\sigma = \sqrt{\text{Var}(X)}$$

- *Sample standard deviation = Standard error*

$$s = \frac{\sigma}{\sqrt{n}}$$

**PROBABILITY MASS FUNCTION (PMF)** : the distribution of the probability of the different values of a discrete random variable X. If the variable X takes values  $x_1, \dots, x_n$  with probabilities  $p(x_1), \dots, p(x_n)$ , than:

1.  $\sum_i p(x_i) = 1$
2.  $p(x_i) \leq 0 \ \forall i$

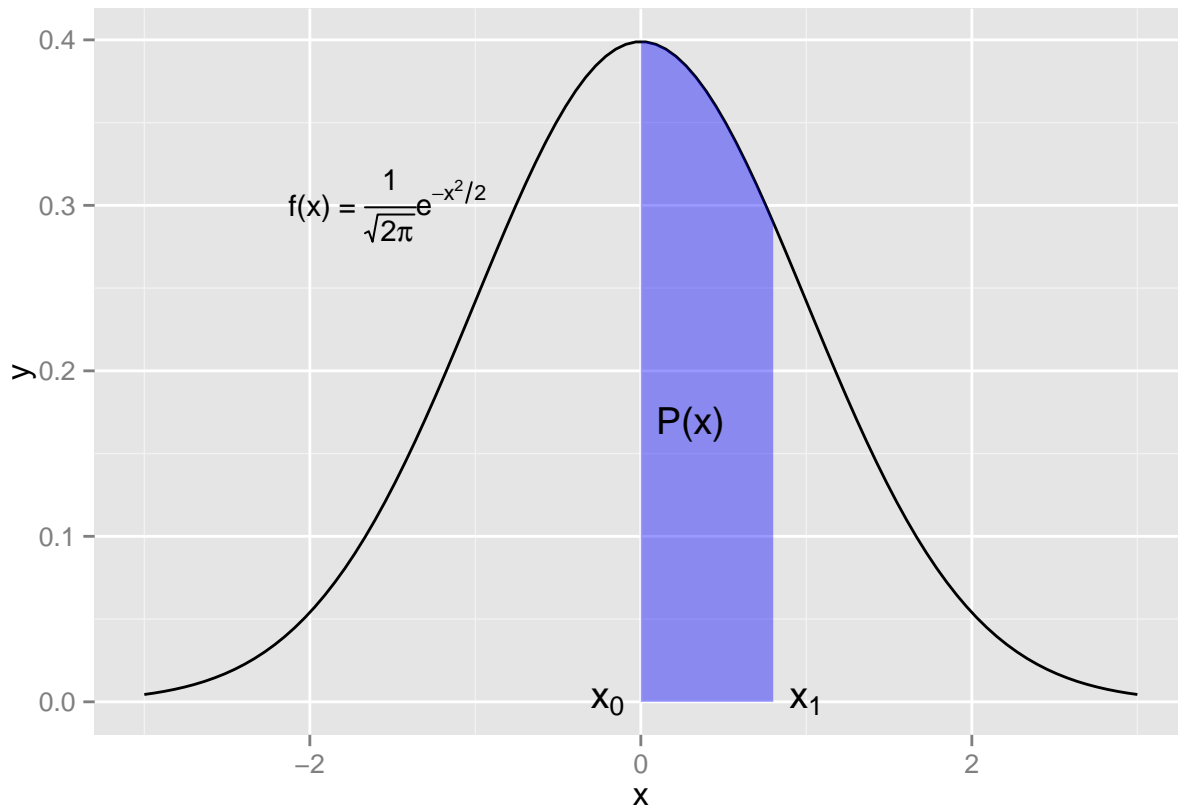
**PROBABILITY DENSITY FUNCTION (PDF)** : the function  $f(x)$  of a continuous variable X such that:

1. the probability that X lies between  $x_0$  and  $x_1$  is  $\int_{x_0}^{x_1} f(x)dx$
2. the cumulative probability of the whole range of x is equal to 1 :  $\int_x f(x)dx = 1$
3. the probability is always greater than 0 :  $f(x) \leq 0 \ \forall x$

**PROBABILITY** : a measure of the relative frequency or likelihood of occurrence of an event. Values are deived from a **theoretical distribution** or from **observations**.

- **P(x)** is the probability of the event x.
- $0 \leq P(x) \leq 1$
- **Discrete variables** :  $\frac{\text{number of required outcomes}}{\text{total number of possible outcomes}}$
- **Continuous variables** : The relevant area under the graph of its probability density function  $f(x)$

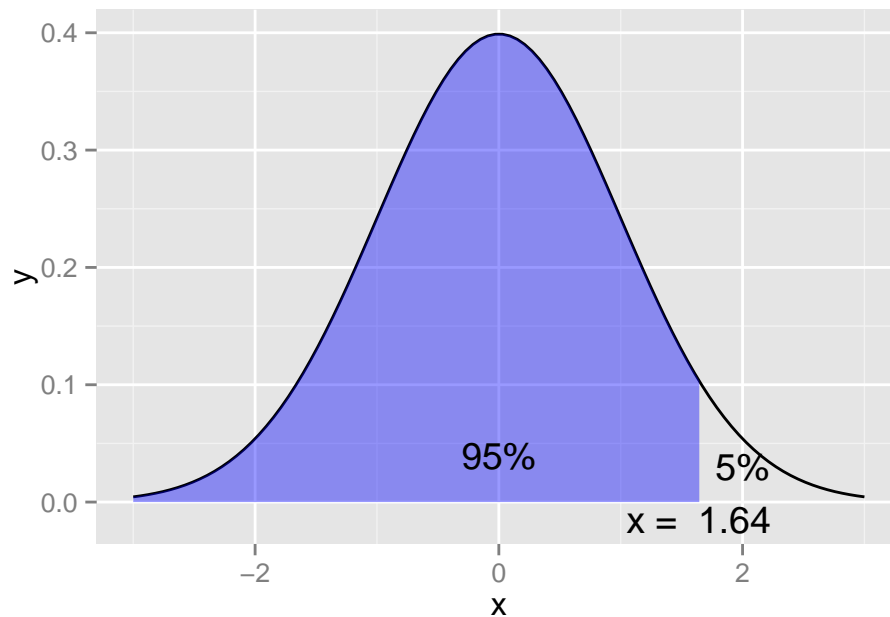
$$P(x) = \int_{x_0}^{x_1} f(x)dx$$



**QUANTILE** : general name for the values of a variable which divides its distribution into equal groups

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qnorm(.95)
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## [1] 1.644854
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**CUMULATIVE DISTRIBUTION FUNCTION (CDF)** : the function  $F(x)$  which gives the cumulative frequency

## DISTRIBUTIONS

**BERNOULLI DISTRIBUTION** : a type of binomial distribution when the random variables take only values 0 or 1 with probability  $p$  and  $1-p$ , respectively

- *Probability mass function* :  $P(X = x_1) = p^{x_1}(1-p)^{1-x_1}$  with  $X = [0,1]$
- *Expected value* :  $\mu = p$
- *Variance* :  $Var(X) = p(1-p)$

**BINOMIAL** : is obtained as the sum of a bunch of iid bernoulli random variables (ex. number of heads on a biased coin). Let  $x_1, \dots, X_n$  be an iid Bernoulli with probability  $p$ , then

$$X = \sum_i x_i = \sum_{i=1}^n x_i$$

is a *binomial random variable* with mass function

$$P(X = x_i) = \binom{n}{k} p^{x_i} (1-p)^{1-x_i}$$

**NORMAL DISTRIBUTION** Gaussian distribution with mean  $\mu$  and variance  $\sigma^2$

$$X \sim N(\mu, \sigma^2)$$

- *Probability Mass Function* :

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

**STANDARD NORMAL DISTRIBUTION** : is the specific case of a normal distribution with  $\mu = 0$  and  $\sigma = 1$

$$X \sim N(0, 1)$$

with probability mass function

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

