
Uni-variate (1D) random variables (complete description)

Random variable	X
Outcome (value)	$X = x$
Probability density function (<u>pdf</u>)	$f_X(x) = \lim_{dx \rightarrow 0} \frac{P(X \in [x, x+dx])}{dx}$ $\int_{-\infty}^{\infty} f_X(x) dx = 1, \quad f_X(x) \geq 0$ $P(X \in [a, b]) = \int_a^b f_X(x) dx$
Cumulative density function (<u>cdf</u>)	$F_X(x) = P(X \leq x) = \int_{-\infty}^x f_X(x) dx$ $F_X(-\infty) = 0, \quad F_X(\infty) = 1$ $P(X \in [a, b]) = F_X(b) - F_X(a)$
Domain and range	all random variables can be considered <u>continuous</u> with possible values $x \in R$, since <u>discrete</u> random variables can be modeled with impulse-functions in their pdf (and step-functions in their cdf) and since the pdf can equal zero in “impossible” x -regions

Population, i.e. theoretical, moments (summarizing, incomplete description)

<u>Population</u> mean (1'st order)	$\mu_X = E[X] = \int_{-\infty}^{\infty} x \cdot f_X(x) dx$ measures <u>location</u> (of probability mass)
<u>Population</u> variance (2'nd order)	$\sigma_X^2 = E[(X - \mu_X)^2] = \int_{-\infty}^{\infty} (x - \mu_X)^2 \cdot f_X(x) dx$ $\sigma_X^2 = E[X^2] - \mu_X^2$ measures <u>dispersion</u> (of probability mass)
<u>Population</u> standard deviation	$\sigma_X = \sqrt{\sigma_X^2}$

Descriptive statistics and sample moments

Sample (observed data values)	$X = \{x_1, x_2, \dots, x_n\}$
Sample size	n
<u>Sample</u> mean (average)	$\hat{\mu}_X = \bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$
<u>Sample</u> variance	$\hat{\sigma}_X^2 = S_X^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$
<u>Sample</u> standard deviation	$\hat{\sigma}_X = \sqrt{\hat{\sigma}_X^2}$
