

# Probability Theory v1.0.1\_20240904

- The principle of indifference in classical probability, such as when guessing a password  

```
if (nchar(input_password) != nchar(password)) print("Try Constant-Time Crypto")
```

- Aleatory or Epistemic. *Frequentist* or *Bayesian*.

*Probability* is the long-run frequency of repeated events.  $f(Y | \theta)$  is a function of  $Y$  with  $\theta$  fixed but unknown.

*Probability* is a degree of uncertainty about values.  $f(Y | \theta)$  is a function of  $\theta$  with  $Y$  fixed.

- Complete probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  and probability axioms (Kolmogorov, 1933)

sample space  $\{\omega_1, \dots, \omega_n\}$ ,  $\sigma$ -algebra of subsets of  $\Omega$  and event space  $\mathcal{F}$ , and probability measure  $\mathbb{P}(E) = |E| / |\Omega|$

$$\text{i) } \mathbb{P}(E) \geq 0, \forall E \in \mathcal{F} \quad \text{ii) } \mathbb{P}(\Omega) := 1 \quad \text{iii) } \mathbb{P}\left(\bigcup_{i=1}^{\infty} E_i\right) = \sum_{i=1}^{\infty} \mathbb{P}(E_i), \text{ where } E_i \cap E_j = \emptyset \text{ for } i \neq j$$

- Neither “random” nor “variable”. Which expression is more rigorous in its formulation? 🤖

$X(\omega) = s, s \in \{1,2,3,4,5,6\} \subset \mathbb{N}^+$  or  $X(\omega) = i \cdot \mathbf{1}_{S_i}(\omega), i \in \{1,2,3,4,5,6\}$

**random variable** and vector - measurable function  $X : \Omega \rightarrow R$  measurable space, usually  $R = \mathbb{R}^n$

- `geom_step()` is a non-decreasing and right-continuous function. 🔴

cumulative distribution function  $F_X(x) := \mathbb{P}(X \leq x)$ , e.g., the survival function  $S(t) := \mathbb{P}(T > t) = 1 - F_T(t)$

- Can a “zero-probability” event happen?  $\mathbb{P}(X = x) = 0, \forall x \in \mathbb{R}$  but  $\mathbb{P}(X \in \mathbb{R}) = 1$  almost surely.

discrete RV and probability mass function: Bernoulli, binomial, Poisson, hypergeometric, negative binomial, uniform

continuous RV and probability density function: normal, exponential, Cauchy, triangular, gamma, beta, Laplace,

Weibull, Pareto, Pearson, Dirichlet, Wishart, uniform

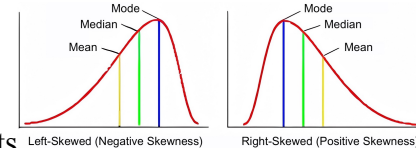
sampling distribution:  $\chi^2, F, t$ ; quantile function,  $p$ -value

- Which distributed RV has an undefined expected value?

$\text{Kurt}(X) \neq 3 \Rightarrow \text{non-normal}$

mean, median, and mode, variance and covariance, skewness, kurtosis, and moments

moment generating function, characteristic function, Fisher information, entropy, statistical distance



- Being **marginally** normally distributed and linearly uncorrelated does not necessarily imply independence!

- Note  $P$  and  $p$ .

$$\underbrace{\frac{P(\mathcal{M}_1 | d)}{P(\mathcal{M}_0 | d)}}_{\text{posterior odds}} = \underbrace{\frac{p(d | \mathcal{M}_1)}{p(d | \mathcal{M}_0)}}_{\text{Bayes factor } BF_{10}} \cdot \underbrace{\frac{P(\mathcal{M}_1)}{P(\mathcal{M}_0)}}_{\text{prior odds}} \quad (\text{Ly et al., 2016, p. 21})$$

We use the term probability loosely in the development to refer either to probability mass or to probability density, depending on whether the data are discrete or continuous (Rouder et al., 2012, p. 358-359).

Bayesian one-sample  $t$ -test:

$$JZSBF_{10}(t(n), n) = (2\pi)^{-\frac{1}{2}} h \left(1 + \frac{t^2}{n-1}\right)^{\frac{n}{2}} \int_0^\infty (1+ng)^{-\frac{1}{2}} \left(1 + \frac{t^2}{(1+ng)(n-1)}\right)^{-\frac{n}{2}} g^{-\frac{3}{2}} e^{-\frac{h^2}{2g}} dg$$

