

# Continuous and Discrete Random Variables

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# Agenda

- ▶ Continuous Random Variables
- ▶ Discrete Random Variables

# Continuous Random Variables

A continuous random variable (r.v.) can take on an uncountably infinite number of values.

Given a probability density function (pdf),  $f(y)$ , allows us to compute

$$P(a \leq Y \leq b) = \int_a^b f(y) dy.$$

Properties of continuous random r.v.'s:

- ▶  $\int f(y) dy = 1.$
- ▶ For any value  $y$ ,

$$P(Y = y) = \int_y^y f(y) dy = 0 \implies$$

$$P(y < Y) = P(y \leq Y).$$

# Discrete Random Variables

A discrete random variable has a countable number of possible values, where the associated probabilities are calculated for each possible value using a probability mass function (pmf).

A pmf is a function that calculates  $P(Y = y)$ , given each variable's parameters.

# Common Discrete distributions

- ▶ Bernoulli
- ▶ Geometric
- ▶ Negative Binomial
- ▶ Hypergeometric

# Common Continuous distributions

- ▶ Exponential
- ▶ Gamma
- ▶ Normal (Gaussian)
- ▶ Beta

## Beta distribution

Given  $a, b > 0$ , we write  $\theta \sim \text{Beta}(a, b)$  to mean that  $\theta$  has pdf

$$p(\theta) = \text{Beta}(\theta|a, b) = \frac{1}{B(a, b)} \theta^{a-1} (1 - \theta)^{b-1} \mathbb{1}(0 < \theta < 1),$$

i.e.,  $p(\theta) \propto \theta^{a-1} (1 - \theta)^{b-1}$  on the interval from 0 to 1.

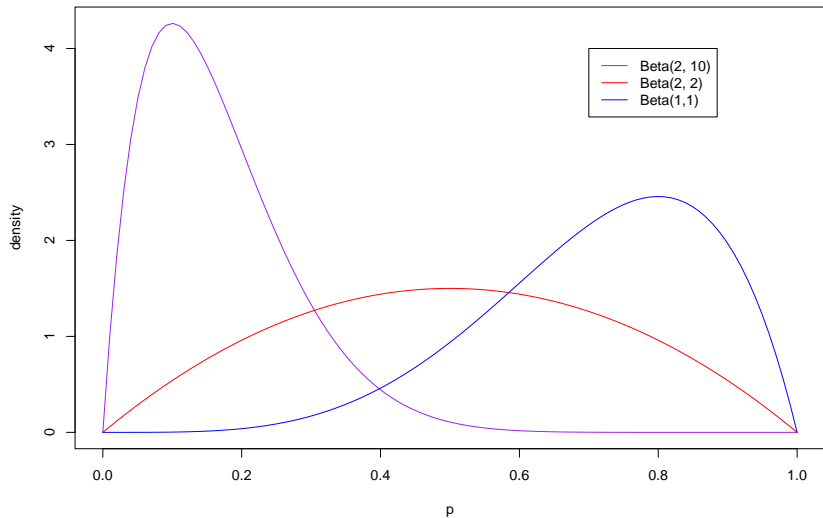
► Here,

$$B(a, b) = \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)}$$

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- Parameters  $a, b$  control the shape of the distribution.
- This distribution models random behavior of percentages/proportions.

# Beta distribution





## Beta distribution example

Suppose that a college models probabilities of student accepting admission via the Beta( $a, b$ ) distribution, where  $a, b > 0$  are fixed and known.

What is the probability that a randomly selected student has prob of accepting an offer larger than 80 percent, where  $a=4/3$  and  $b=2$ .

```
pbeta(0.8, shape1 = 4/3, shape2 = 2, lower.tail = FALSE)
```

```
## [1] 0.05930466
```