

# Chapter 1

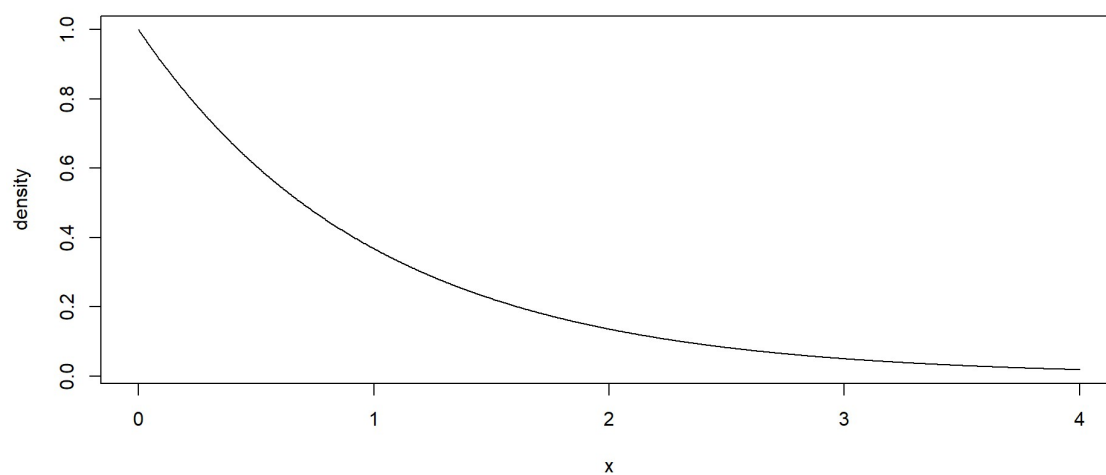
MathStat Preliminaries

## 1.1 Wait. Where are we going?

## Exponential pdf

If  $X \sim \text{Exponential}(\lambda)$ , then  $X$  has the pdf  
 $f(x) = \lambda \exp(-\lambda x) I_{[0, \infty]}(x)$ .

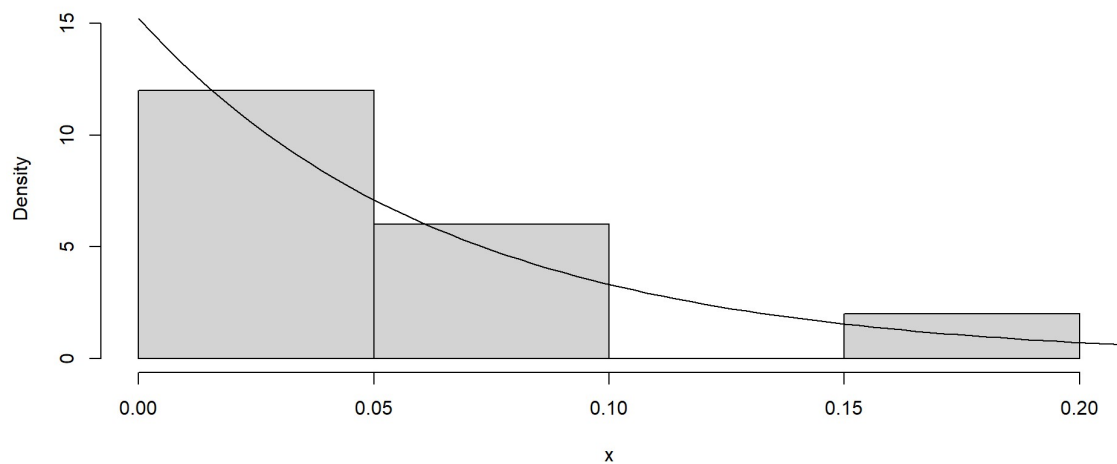
## Exponential pdf visualized



## Sample

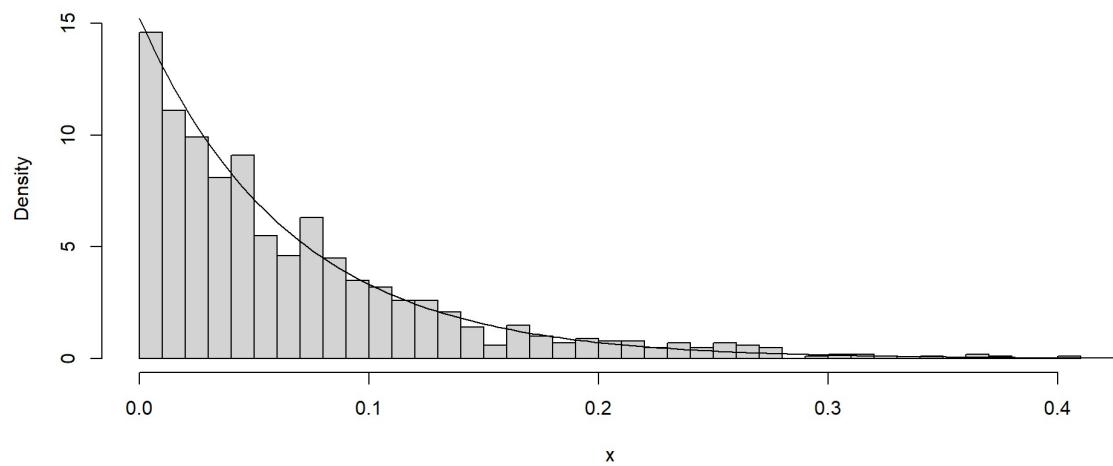
A sample of  $n$  realization of  $X$ , denoted  $X_1, X_2, \dots, X_n$ , can be used to approximate the distribution of  $X$ .

## Histogram for 10 realizations



A histogram of 10 realizations from an Exponential(15.2).

## Histogram for 1000 realizations



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A histogram of 1000 realizations from an Exponential(15.2).

### 1.1.1 A very special trick

## Approach

We can avoid doing actual integration if we can manipulate the integrand to look more like a pdf, which must integrate to 1 of the range of the random variable.

Example: Determine

$$\int_0^{\infty} 3 \exp(-2x) dx.$$

Special trick (cont)

## **1.2 Transformations of Random Variables**

### **1.2.1 The discrete case and the binomial distribution**

## The binomial pmf

$X \sim \text{Binomial}(n, p)$  has the pmf

$$f(x) = P(X = x) = \binom{n}{x} p^x (1 - p)^{n-x} I_{\{0,1,\dots,n\}}(x).$$

Determine the pmf of  $Y := n - X$ .





## 1.2.2 The Continuous Case and the Gamma Distribution

### Continuous Transformation PDF

Let  $X$  be a continuous random variable with pdf  $f_X(x)$ . Let  $Y = g(X)$ , where  $g$  is invertible and differentiable. Then the pdf for  $Y$  is

$$f_Y(y) = f_X(g^{-1}(y)) \left| \frac{d}{dy} g^{-1}(y) \right|.$$

**Proof**

**Proof (cont)**

**Proof (cont)**

**Proof (cont)**

### Example 1.2.2

A continuous random variable  $X \sim \text{Gamma}(\alpha, \beta)$  has pdf

$$f_X(x) = \frac{1}{\Gamma(\alpha)} \beta^\alpha x^{\alpha-1} \exp(-\beta x) I_{(0,\infty)}(x).$$

### Example 1.2.2 (cont)

Let  $Y := cX$  with  $c > 0$ . Determine the pdf of  $Y$ .

## Example 1.2.2 (cont)

### The shape parameter

The  $\alpha$  term is the “shape” parameter of the Gamma distribution.

The kernel (non-constant) part of the pdf of the Gamma distribution is

$$x^{\alpha-1} \exp(-\beta x) I_{(0,\infty)}(x).$$

## The shape parameter

The exponential function part is an unscaled exponential density.

- $\exp(-\beta x)$  dominates  $x^{\alpha-1}$  for large  $x$ .
- For small  $x$ , the  $x^{\alpha-1}$  term controls the shape.

## Shape parameter visualized

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The  $\alpha$  terms controls the shape of the Gamma distribution.

## The Gamma Function

The pdf for the gamma distribution is defined using the **gamma function**, denoted by  $\Gamma(\alpha)$ .

$$\Gamma(\alpha) = \int_0^{\infty} x^{\alpha-1} \exp(-x) dx.$$

## The Gamma Function (cont)

## The Gamma Function (cont)

## The Gamma Function (cont)



## 1.3 Bivariate Transformations

### Bivariate Transformation PDF

Suppose the  $X_1$  and  $X_2$  are jointly continuous random variables with pdf  $f_{X_1, X_2}(x_1, x_2)$ . Let  $Y_1 = g_1(X_1, X_2)$  and  $Y_2 = g_2(X_1, X_2)$ . For  $h_1$  and  $h_2$  differentiable, let

$X_1 = h_1(Y_1, Y_2)$  and  $X_2 = h_2(Y_1, Y_2)$ .

Then

$$f_{Y_1, Y_2}(y_1, y_2) = f_{X_1, X_2}(h_1(y_1, y_2), h_2(y_1, y_2)) |J|,$$

where  $|J|$  is the absolute value of the Jacobian.

## Bivariate transformation pdf continued

The absolute value of the Jacobian is

$$J = \det \begin{bmatrix} \frac{\partial x_1}{\partial y_1} & \frac{\partial x_1}{\partial y_2} \\ \frac{\partial x_2}{\partial y_1} & \frac{\partial x_2}{\partial y_2} \end{bmatrix}.$$

### Example 1.3.1

Let  $X_1, X_2 \stackrel{i.i.d.}{\sim} \text{Gamma}(\alpha, \beta)$

Determine the pdf of

$$Y = \frac{X_1}{X_1 + X_2}.$$

**Example 1.3.1 (cont)**

**Example 1.3.1 (cont)**

**Example 1.3.1 (cont)**

**Example 1.3.1 (cont)**

## The Beta Distribution

The continuous random variable  $X \sim \text{Beta}(a, b)$  had pdf

$$f(x) = \frac{1}{\mathcal{B}(a, b)} x^{a-1} (1-x)^{b-1} I_{(0,1)}(x),$$

for  $a, b > 0$ , where

$$\mathcal{B}(a, b) = \int_0^1 x^{a-1} (1-x)^{b-1} dx$$

denotes the **beta function**, which normalizes the kernel of the Beta distribution.

## The Beta Distribution (cont)

The Beta distribution is a flexible distribution for modeling a random variable between 0 and 1.

Note: The Uniform(0, 1) distribution is a special case of the Beta distribution with

## The Beta Distribution (cont)

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### The Beta Function

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

Proof:

## The Beta Function (cont)