# Other random variables



## Definition 1 (*Γ*-distribution)

A random variable is said to have a gamma distribution with parameter  $(\alpha, \lambda)$ ,  $\lambda > 0$ ,  $\alpha > 0$ , if its density function is given by

$$f(x) = \begin{cases} \frac{\lambda e^{-\lambda x} (\lambda x)^{\alpha - 1}}{\Gamma(\alpha)} = \frac{\lambda^{\alpha}}{\Gamma(\alpha)} e^{-\lambda x} x^{\alpha - 1} & x \ge 0\\ 0 & x < 0, \end{cases}$$

where  $\Gamma(\alpha)$ , the gamma function, is defined as

$$\Gamma(\alpha) = \int_0^\infty e^{-y} y^{\alpha-1} dy.$$

# Gamma function



#### Lemma 2

The gamma function has the following property:

 $\Gamma(\alpha) = (\alpha - 1)\Gamma(\alpha - 1).$ 

Specially, if  $\alpha = n$ , then

 $\Gamma(n) = (n-1)!.$ 

# Mean and variance



## **Proposition 3**

Let X be a gamma random variable with parameter  $(\alpha, \lambda)$ . Then

$$\mathbb{E}[X] = \frac{\alpha}{\lambda}, \quad \operatorname{Var}(X) = \frac{\alpha}{\lambda^2}.$$

# Proof.

For the expected value,

$$E[X] = \frac{1}{\Gamma(\alpha)} \int_0^\infty x \cdot \lambda e^{-\lambda x} (\lambda x)^{\alpha - 1} dx$$
$$= \frac{1}{\lambda \Gamma(\alpha)} \int_0^\infty \lambda e^{-\lambda x} (\lambda x)^\alpha dx$$
$$= \frac{\Gamma(\alpha + 1)}{\lambda \Gamma(\alpha)}$$
$$= \frac{\alpha}{\lambda}.$$

# The Cauchy distribution



## **Definition 4**

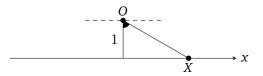
A random variable X is said to have a Cauchy distribution if its density is given by

$$f(x) = \frac{1}{\pi} \frac{1}{1+x^2}, \quad -\infty < x < \infty.$$





Suppose that a narrow-beam flashlight is spun around its center, which is located a unit distance from the *x*-axis. Consider the point X at which the beam intersects the *x*-axis when the flashlight has stopped spinning. (If the beam is not pointing toward the *x*-axis, repeat the experiment.) Find the pdf of X.



#### Solution.

Let  $\theta$  be the angle between the flashlight and the *y*-axis, which is uniformly distributed between  $-\pi/2$  and  $\pi/2$ . The distribution function of *X* is thus given by

$$F(x) = \mathbb{P}(X \le x)$$
  
=  $\mathbb{P}(\tan \theta \le x)$   
=  $\mathbb{P}(\theta \le \tan^{-1} x)$   
=  $\frac{1}{2} + \frac{1}{\pi} \tan^{-1} x.$ 

Therefore,

$$f(x) = \frac{d}{dx}F(x) = \frac{1}{\pi(1+x^2)}, \quad -\infty < x < \infty,$$

and hence, X has the Cauchy distribution.

# The Beta distribution



## **Definition 6**

A random variable is said to have a beta distribution if its density is given by

$$f(x) = \begin{cases} \frac{1}{B(a,b)} x^{a-1} (1-x)^{b-1} & 0 < x < 1\\ 0 & \text{otherwise} \end{cases}$$

#### where

$$B(a,b) = \int_0^1 x^{a-1} (1-x)^{b-1} dx = \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)}.$$

## Remark

The beta distribution can be used to model a random phenomenon whose set of possible values is [0, 1]. When a = b, then the beta distribution is symmetric about 1/2. When a = b = 1, then the beta distribution is identical to U(0, 1).

# Mean and variance



**Proposition 7** 

If  $X \sim \text{Beta}(a, b)$ , then

$$\mathbb{E}[X] = \frac{a}{a+b}, \quad \operatorname{Var}(X) = \frac{ab}{(a+b)^2(a+b+1)}.$$

- Functions of continuous random variables, where the function is monotone.
- Linear combination of random variables: general Uniform distribution, general normal distribution.
- Log-normal distribution.
- Chi-squared distribution.
- Cauchy distribution:  $\tan U$  where  $U \sim U(-\pi/2, \pi/2)$ .
- Inverse distribution function, and generating random numbers.

# Distribution of a function of a random variable



- Suppose that we have known the distribution of a continuous random variable *X*, ...
- **and we want to find the distribution of** g(X)
- Consider the following examples:

 $X \sim U(0, 1), \quad Y = X^n \sim ???$ 

$$X \sim f_X, \quad Y = X^2 \sim ???$$



## Example 8

Let  $X \sim U(0, 1)$ . Find the distribution of  $Y = X^n$ .

### Proof.

For  $0 \le y \le 1$ ,  $F_Y(y) = \mathbb{P}(Y \le y)$   $= \mathbb{P}(X^n \le y)$   $= \mathbb{P}(X \le y^{1/n})$   $= F_X(y^{1/n})$  $= y^{1/n}$ . Therefore, the pdf is given by

$$\begin{split} f_Y(y) &= \frac{d}{dy} F_Y(y) \\ &= \begin{cases} \frac{1}{n} y^{1/n-1} & 0 \leqslant y \leqslant 1 \\ 0 & \text{otherwise.} \end{cases} \end{split}$$



### **Proposition 9**

Let X be a continuous random variable having probability density function  $f_X$ . Suppose that g(x) is a strictly monotone (increasing or decreasing), differentiable function of x. Then the random variable Y = g(X) has a probability density function given by

$$f_Y(y) = \begin{cases} f_X[g^{-1}(y)] \left| \frac{d}{dy} g^{-1}(y) \right| & \text{if } y = g(x) \text{ for some } x \\ 0 & \text{if } y \neq g(x) \text{ for all } x \end{cases}$$

where  $g^{-1}(y)$  is the inverse function of g.



# Example 10 (The lognormal distribution)

If  $X \sim N(0, 1)$ , find the distribution of  $Y = e^X$ .

#### Solution.

Note that  $g(x) = e^x$  is an increasing function, and

$$g^{-1}(y) = \log y, \quad \frac{d}{dy}g^{-1}(y) = \frac{1}{y}, \quad y > 0.$$

Then, the pdf of Y is given by

$$f_Y(y) = f_X[g^{-1}(y)] \left| \frac{d}{dy} g^{-1}(y) \right| = \frac{1}{2\pi} \exp\left\{ -\frac{1}{2} (\log y)^2 \right\} \cdot \frac{1}{y}, \quad y > 0.$$

and equals to 0 otherwise.





If *X* is a continuous random variable with probability density  $f_X$ , find the distribution of  $Y = X^2$ .

## Theorem 12

Let X be a random variable with cdf F(x). Assume that F is strictly increasing and continuous, then

 $Y := F(X) \sim U(0, 1).$ 

## Proof.

In order to find the distribution of Y, it suffices to find the cdf of Y:  $F_Y(y) = \mathbb{P}(Y \leq y)$  for  $y \in \mathbb{R}$ . We only consider the case where  $y \in [0, 1]$ . (Why?)

If  $0 \leq y \leq 1$ , then

$$F_Y(y) = \mathbb{P}(Y \le y) = \mathbb{P}(F(X) \le y)$$
$$= \mathbb{P}(X \le F^{-1}(y)) = F(F^{-1}(y)) = y.$$

# **Inverse transformation**

# Inverse function $F^{-1}(\cdot)$



- The function *F* may be flat...
- and thus  $F^{-1}$  does not always exits, that is,

 $F^{-1}(\{u\})$  may not be unique.

Choose a respective from the points as its inverse.

## **Definition 13**

If *X* has the distribution function F(x). For any *u*, define

 $F^{-1}(u) = \inf\{x : F(x) \ge u\}.$ 

#### Theorem 14

Let F be an arbitrary distribution function, and let  $U \sim U(0, 1)$ . Then,

 $X = F^{-1}(U)$ 

has the distribution function F.



# Example 15

Note that the distribution function of  $\operatorname{Exp}(\lambda)$  is

 $F(x) = 1 - e^{-\lambda x}.$ 

Then,

$$F^{-1}(u) = -\frac{1}{\lambda}\log(1-u).$$

Therefore, if  $U \sim U(0, 1)$ , then

 $F^{-1}(U) \sim \operatorname{Exp}(\lambda).$ 

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## Assume that *X* is a discrete random variable with the pmf:

x	1	2	3	4
p(x)	0.1	0.2	0.3	0.4

Then, the distribution function of *X* is

$$F(x) = \begin{cases} 0 & x < 1\\ 0.1 & 1 \le x < 2\\ 0.3 & 2 \le x < 3\\ 0.6 & 3 \le x < 4\\ 1 & x \ge 4 \end{cases}$$

Find  $F^{-1}(u)$  for 0 < u < 1.