# Concepts

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#### Abstract

Useful concepts of probability and statistics.

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# 1 Laws

## 1.1 Law of total probability

It is common practice to compute the sum of total probability of all available options.

Thinking forwardly. Which means starting from the **reason** to the **result**.

Theorem 1.1. Law of total probability

For random variables A and B, we have

$$P(A) = \sum P(A|B_i) \cdot P(B_i), \forall B_i \in B$$

It is automatically accepted that all the  $B_is$  are all separable, and mutually exclusive with each other, which means

$$P(B_i, B_j) = P(B_i) \cdot P(B_j), i \neq j$$
  

$$P(B_i, B_i) = P(B_i)$$

It is a prior rule to be accepted, and we will accept it if not specified.

Thinking **backwardly**. If we have already known that A only has one option (noted as a), which is also inevitable (P(A = a) = 1).

Proposition 1.1. Sum of probability of every options is ONE

*Proof.* We have P(a) = 1 and  $P(a|B_i) = 1, \forall B_i \in B$ . Thus,

$$1 = \sum P(B_i), \forall B_i \in B$$

Since a can be something that naturally happens regardless of the choice of B, the proposition may not affected by the choice of a. Thus, the total probability of all options of B is 1.

### 2 Definitions

#### 2.1 Definition of e

The constant e is defined as the infinity integral of

$$e = \lim_{x \to \infty} (1 + \frac{1}{x})^x \tag{1}$$

the very existence of e is based on the fact that integrand function is monotone bounded.

**Proposition 2.1.** The function f(x) is monotone bounded

$$f(x) = \left(1 + \frac{1}{x}\right)^x, x \in \mathcal{R}$$

Proof. Use its discrete sequence version

$$g_n = (1 + \frac{1}{n})^n, n \in \mathcal{N}$$

by expanding it, we have

$$g_n = \sum_{i=0}^{n} (n, i) \cdot \frac{1}{n^i}$$

$$g_{n+1} = \sum_{i=0}^{n+1} (n+1, i) \cdot \frac{1}{(n+1)^i}$$

$$g_{n+1} = \sum_{i=0}^{n} \frac{(n+1, i)}{n+1} \cdot \frac{1}{n^i} + c$$

where c > 0. For every i, we have  $\frac{(n+1,i)}{n+1} > (n,i)$ . Thus, the  $g_n$  is **monotone increasing**.

To find a valid upper bound, we enlarge the  $g_n$ .

$$g_n < \sum_{i=0}^n \frac{1}{i!}$$

the enlarge is based on the idea of  $(n,i) < \frac{n^i}{i!}$ . Further,

$$g_n < c + \sum_{i=4}^{n} \frac{1}{2^i}$$

where constant c > 0. It uses the idea of  $2^i < i!, i \ge 4$ . Use the sum of geometric series, we can say the  $g_n$  is **upper bounded**.

Back to f(x), use **Squeeze Theorem** 

$$(1 + \frac{1}{n+1})^n < f(x) < (1 + \frac{1}{n})^{n+1}$$

where  $n = \lfloor x \rfloor$ . On the left hand, it equals to  $g_{n+1}/(1 + \frac{1}{n+1})$ ; On the right hand, it equals to  $g_n * (1 + \frac{1}{n})$ . Based on the analysis above, they are of the same value which is defined as e.

Hence proved.

# 2.2 Gaussian integral

The gaussian integral is the improper integral defined as

$$I = \int_{-\infty}^{\infty} e^{-x^2} dx \tag{2}$$

and the value equals to  $\sqrt{\pi}$ .

**Lemma 2.1.** The gaussian integral equals to  $\sqrt{\pi}$ 

$$\int_{-\infty}^{\infty} e^{-x^2} dx = \sqrt{\pi}$$

*Proof.* Use the square-shaped integral

$$I^2 = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-x^2} e^{-y^2} dx dy$$

Transform into polar coordinates

$$I^2 = \int_0^{2\pi} \int_0^\infty e^{-r^2} r dr d\theta$$

Evaluate the integral of r

$$\int_0^\infty e^{-r^2} r dr = \frac{1}{2} \int_0^\infty e^{-u} du = \frac{1}{2}$$

Thus

$$I^2 = \int_0^{2\pi} \frac{1}{2} d\theta = \pi$$

Hence proved.

### 2.3 Gamma function

The Gamma function is defined as

$$\Gamma(z) = \int_0^\infty t^{z-1} e^{-t} dt, Re(z) > 0$$
 (3)

The Gamma function has useful properties

$$\Gamma(1) = 1$$

$$\Gamma(z+1) = z \cdot \Gamma(z)$$

$$\Gamma(z) = 2 \int_0^\infty t^{2z-1} e^{-t^2} dt$$

Based on the gaussian integral in (2), we have

$$\Gamma(\frac{1}{2}) = \sqrt{\pi}$$

There are countless useful properties of Gamma function. The followings are from the website  $^{1}.$ 

#### 2.3.1 Euler's reflection formula

The Gamma function satisfies the reflection formula

$$\Gamma(x)\Gamma(1-x) = \frac{\pi}{\sin(x\pi)} \tag{4}$$

The proof can be found in  $^2$ . The formula can verify the value of  $\Gamma(\frac{1}{2})$ .

### 2.3.2 Euler's product representation

The Gamma function can be expressed as an infinite product

$$\Gamma(x) = \frac{1}{x} \prod_{n=1}^{\infty} (1 + \frac{1}{n})^x (1 + \frac{x}{n})^{-1}$$
 (5)

### 2.3.3 Weierstrass's product representation

The Gamma function can be expressed as another infinite product

$$\Gamma(x) = \frac{exp(-\gamma x)}{x} \prod_{n=1}^{\infty} (1 + \frac{x}{n})^{-1} exp(\frac{x}{n})$$
 (6)

where  $\gamma$  denotes the Euler-Mascheroni constant

$$\gamma = \lim_{n \to \infty} \sum_{k=1}^{n} \frac{1}{k} - \log(n) \tag{7}$$

This definition implies the reflection formula along with the Weierstrass product of sine and is equivalent with Euler's representation when the definition of the Euler-Mascheroni constant is substituted and rearranged.

#### 2.3.4 Riemann's reflection formula

The Gamma function plays a key role in the analytic continuation of the Riemann zeta function to the complex plane

$$\zeta(s) = 2^s \pi^{s-1} \sin(\frac{\pi s}{2}) \Gamma(1-s) \zeta(1-s)$$
(8)

This example shows how closely related the Gamma function is to other functions. Another similar relation with the zeta function is

$$\zeta(z)\Gamma(z) = \int_0^\infty \frac{t^{z-1}}{e^t - 1} dt$$

<sup>1</sup>https://math.wikia.org/wiki/Gamma\_function

<sup>2</sup>https://brilliant.org/discussions/thread/proof-of-euler-reflection-formula/

### 2.3.5 Other properties

The relationship between Beta function

$$\int_0^1 t^{n-1} (1-t)^{m-1} dt = \frac{\Gamma(n)\Gamma(m)}{\Gamma(n+m)}$$

Use  $t = sin^2\theta$  for convenience, we have

$$\int_{0}^{\frac{\pi}{2}} \cos^{2n-1}(\theta) \sin^{2m-1}(\theta) d\theta = \frac{\Gamma(n)\Gamma(m)}{2\Gamma(n+m)}$$

When  $n \in \mathcal{N}$ , we have

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

$$\Gamma(\frac{n}{2}) = \sqrt{\pi} \frac{(n-2)!!}{2^{(n-1)/2}}$$

where !! denotes the operator of double factorial <sup>3</sup>.

### 2.4 Sterling's formula

In short version, the sterling's formula gives an expression of n!

$$n! = n^n e^{-n} \sqrt{2\pi n} [1 + \mathcal{O}(\frac{1}{n})]$$
 (9)

*Proof.* Proof the Sterling's formula

$$n! = n^n e^{-n} \sqrt{2\pi n} [1 + \mathcal{O}(\frac{1}{n})]$$

Formulate new formula based on Gamma function

$$f(x) = \frac{\Gamma(x)e^x \sqrt{x}}{x^x} = 2 \int_0^\infty e^{x-u^2} (\frac{u}{\sqrt{x}})^{2x-1} du$$

Changing variables  $u = \sqrt{x} + v$ 

$$f(x) = 2 \int_{-\sqrt{x}}^{\infty} \phi_x(v) e^{-v^2} dv$$

where  $\phi_x(v) = e^{-2vx^{1/2}}(1+\frac{v}{\sqrt{x}})^{2x-1}, \forall v \geq -\sqrt{x}$ . Additionally, we define  $\phi_x(v) = 0, \forall v < -\sqrt{x}$ , one gets

$$f(x) = 2 \int_{-\infty}^{\infty} \phi_x(v) e^{-v^2} dv$$

Firstly, using the series expansion

$$log(1+\frac{v}{\sqrt{x}}) = \frac{v}{\sqrt{x}} - \frac{v^2}{2x} + \cdots$$

thus, for a fixed v

$$log\phi_x(v) = -v^2 + \mathcal{O}(\frac{1}{\sqrt{x}}), x \to \infty$$

 $<sup>^3 {\</sup>tt https://math.wikia.org/wiki/Double\_factorial}$ 

hence

$$\phi_x(v) = e^{-v^2 + \mathcal{O}(\frac{1}{\sqrt{x}})}$$
$$\lim_{x \to \infty} \phi_x(v) = e^{-v^2}$$

Use Lebesgue dominated convergence theorem we conclude

$$\lim_{x \to \infty} \frac{\Gamma(x)e^x \sqrt{x}}{x^x} = 2 \int_{-\infty}^{\infty} e^{-2v^2} dv = \sqrt{2\pi}$$

additionally, the approximation can be expressed as

$$\frac{\Gamma(x)e^x\sqrt{x}}{x^x} = \sqrt{2\pi} \cdot e^{\mathcal{O}(\frac{1}{\sqrt{x}})}$$

Use the condition of  $n = x, n \in \mathcal{N}$ , and  $\Gamma(n) = (n-1)!$ , we get

$$n! = n^n e^{-n} \sqrt{2\pi n} [1 + \mathcal{O}(\frac{1}{n})]$$

Hence proved.

# 3 Basic concepts

# 3.1 Expectation and Variation

For a random variable, it follows certain distribution

$$x \approx p(x), \forall 0 < p(x) < 1, \int_{x} p(x)dx = 1$$

its expectation and variation are computed as

$$\mathcal{E}(x) = \int_{x} x \cdot p(x) dx$$
$$\mathcal{V}(x) = \int_{x} (x - \mathcal{E}(x))^{2} \cdot p(x) dx$$

additionally, the variation can be computed as

$$\mathcal{V}(x) = \mathcal{E}(x^2) - \mathcal{E}^2(x)$$