

Probability Calculus

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CENTRAL LIMIT THEOREM – EXAMPLES

SOME INTERESTING DISTRIBUTIONS

Plan for Today

- Central Limit Theorem – examples of applications
- Some interesting distributions



Central Limit Theorem – reminder

1. Classical version:

Let X_1, X_2, \dots be identically distributed independent random variables, such that $\mathbb{E}X_1^2 < \infty$. If by $m = \mathbb{E}X_1$ we denote the mean, and by $\sigma^2 = \text{Var}X_1$ the variance of this distribution, then for any $t \in \mathbb{R}$, we have that

$$\mathbb{P}\left(\frac{X_1 + X_2 + \dots + X_n - nm}{\sigma\sqrt{n}} \leq t\right) \xrightarrow{n \rightarrow \infty} \Phi(t),$$

where $\Phi(t) = \int_{-\infty}^t \frac{1}{\sqrt{2\pi}} \exp(-x^2/2) dx$

is the CDF of the standard normal distribution.

also:

$$\mathbb{P}\left(s \leq \frac{X_1 + X_2 + \dots + X_n - nm}{\sigma\sqrt{n}}\right) \xrightarrow{n \rightarrow \infty} 1 - \Phi(s)$$



$$\mathbb{P}\left(s \leq \frac{X_1 + X_2 + \dots + X_n - nm}{\sigma\sqrt{n}} \leq t\right) \xrightarrow{n \rightarrow \infty} \Phi(t) - \Phi(s)$$

De Moivre-Laplace Theorem – reminder

2. Theorem:

Let X_1, X_2, \dots be a sequence of independent identically distributed random variables, such that $\mathbb{P}(X_n = 1) = p = 1 - \mathbb{P}(X_n = 0)$.

Then, we have that for any $s < t$,

$$\mathbb{P} \left(s \leq \frac{X_1 + X_2 + \dots + X_n - np}{\sqrt{np(1-p)}} \leq t \right) \xrightarrow{n \rightarrow \infty} \Phi(t) - \Phi(s).$$

each inequality (both in the CLT and in dML) may be changed to strict without consequences



Central Limit Theorem

3. Examples

- *boys and girls*
- how many students should be accepted?
- aggregate errors
- confidence intervals



Some useful distributions

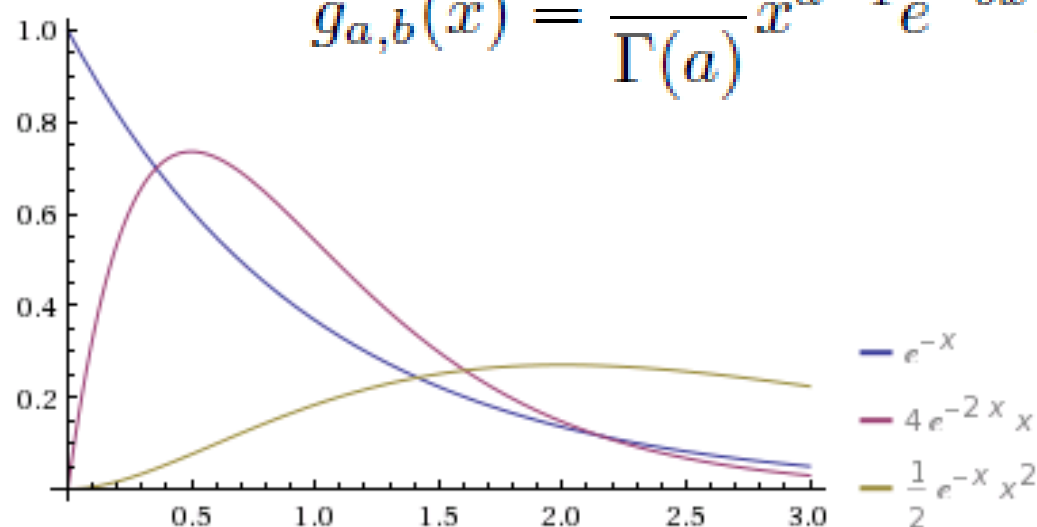
used in statistics and economics when modeling or as a result of operations performed on other distributions



Gamma Distribution $\Gamma(a,b)$, $a,b>0$

$$\Gamma(a) = \int_0^{\infty} t^{a-1} e^{-t} dt \quad \Gamma(n) = (n-1)! \text{ dla } n = 1, 2, \dots$$

$$g_{a,b}(x) = \frac{b^a}{\Gamma(a)} x^{a-1} e^{-bx} 1_{(0,\infty)}(x)$$



$$\mathbb{E}X = a/b$$

$$\text{Var}X = a/b^2$$



Gamma Distribution properties

- for $a=1$ – exponential distribution $\exp(b)$
- for integer a – Erlang Distribution
- $\Gamma(n/2, 1/2)$ – **chi-squared distribution** $\chi^2(n)$

Theorem:

A sum of independent random variables from gamma distributions $\Gamma(a_i, b)$ has a $\Gamma(\sum a_i, b)$ distribution



$$g_{a,b}(x) = \frac{b^a}{\Gamma(a)} x^{a-1} e^{-bx} 1_{(0,\infty)}(x)$$

How the Gamma Distribution is used

In econometrics: duration of events

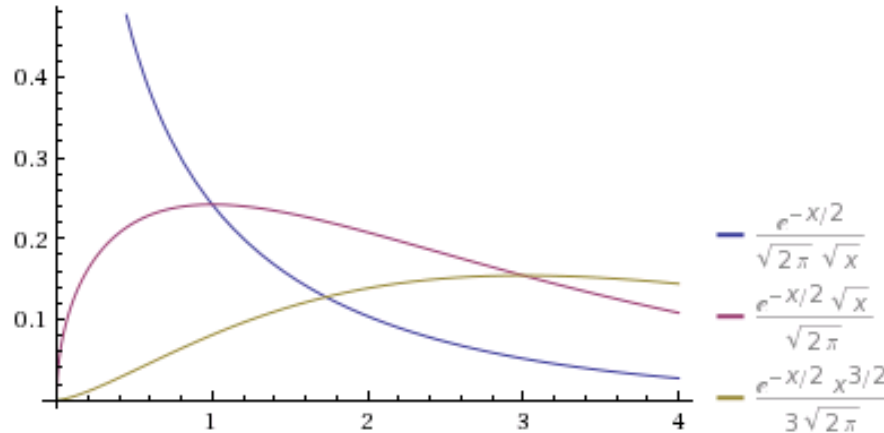
In statistics: bayesian statistics.

....the chi-squared distribution

Describes the time until the a th event in a Poisson proces in which there is on average $1/b$ events per unit of time



Chi-squared Distribution $\chi^2(n)$



Theorem: sum of squares of n IIN random variables (independent identically $N(0,1)$ distributed) has a $\chi^2(n)$ distribution

$$\mathbb{E}X = n, \quad \text{Var}X = 2n$$

The distribution of the sample mean and variance


Theorem:

Let X_1, X_2, \dots, X_n be independent identically distributed random variables from a standard normal distribution and let \bar{X}, s^2 be the sample mean and variance, respectively:

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

$$s^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2.$$

*Then $\sqrt{n}\bar{X}$ has a $\mathcal{N}(0, 1)$, distribution,
— ns^2 has a χ_{n-1}^2 distribution*

 *and \bar{X}, s^2 are independent.*

Chi-squared distribution uses

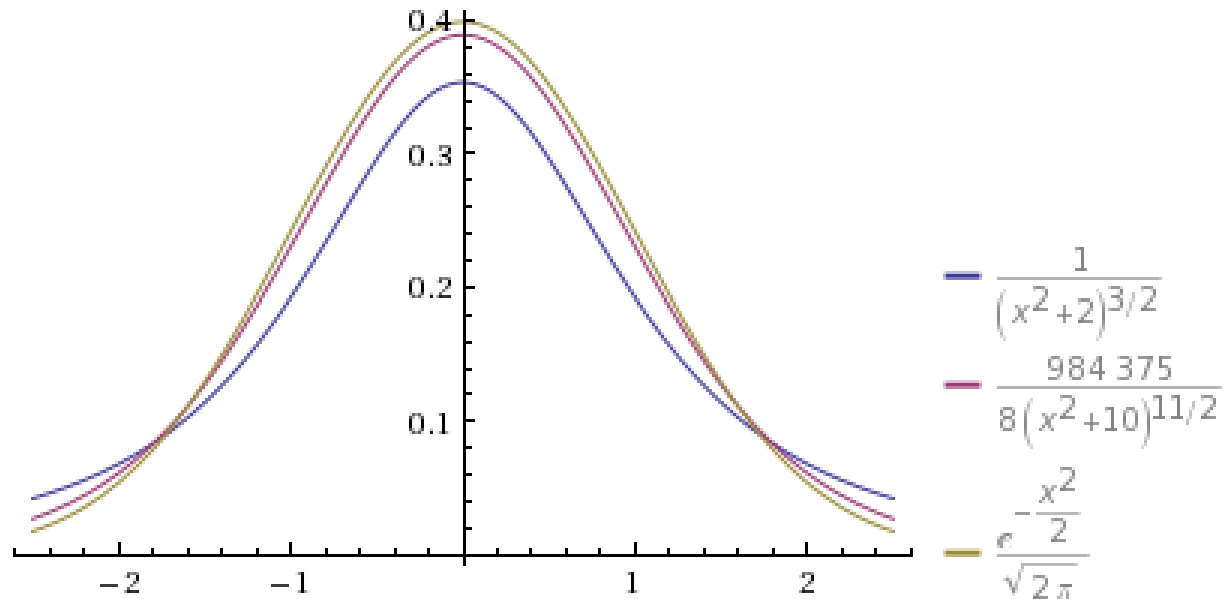
Used very frequently in statistics: appears in „standard” models when testing hypotheses, constructing confidence intervals, as a „component” of other distributions.



***t*-Student Distribution $t(n)$, $n=1,2,\dots$**

$n^{1/2}X/Y^{1/2}$ for independent X and Y , $X \sim N(0,1)$, $Y \sim \chi^2(n)$

$$g_n(x) = \frac{1}{\sqrt{n\pi}} \cdot \frac{\Gamma(\frac{1}{2}(n+1))}{\Gamma(\frac{1}{2}n)} \left(1 + \frac{x^2}{n}\right)^{-\frac{1}{2}(n+1)}, \quad n = 1, 2, \dots$$



$$\mathbb{E}X = 0 \quad n > 1 \quad \text{Var}X = n/(n-2) \quad n > 2$$

***t*-Student distribution uses**

Statistics: regression analysis, hypothesis testing, construction of confidence intervals

Econometrics: as an alternative for the normal distribution in cases where fat tails are needed

for large n : almost $N(0, 1)$.



***F*-Snedecor $F(d_1, d_2)$, $d_1, d_2 = 1, 2, \dots$**

X has a $F(d_1, d_2)$ distribution, if $X = (Y_1/d_1)/(Y_2/d_2)$,
where Y_i are independent, $\chi^2(d_i)$

$$g_{d_1, d_2}(x) = \frac{\left(\frac{d_1 x}{d_1 x + d_2}\right)^{d_1/2} \left(1 - \frac{d_1 x}{d_1 x + d_2}\right)^{d_2/2}}{x B(d_1/2, d_2/2)} 1_{(0, \infty)}(x)$$

$$B(a, b) = \int_0^1 t^{a-1} (1-t)^{b-1} dt = \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)}, \quad a, b > 0$$

$$\mathbb{E}X = d_2 / (d_2 - 2) \quad d_2 > 2$$

$$\text{Var}X = \frac{2 d_2^2 (d_1 + d_2 - 2)}{d_1 (d_2 - 2)^2 (d_2 - 4)} \quad d_2 > 4$$

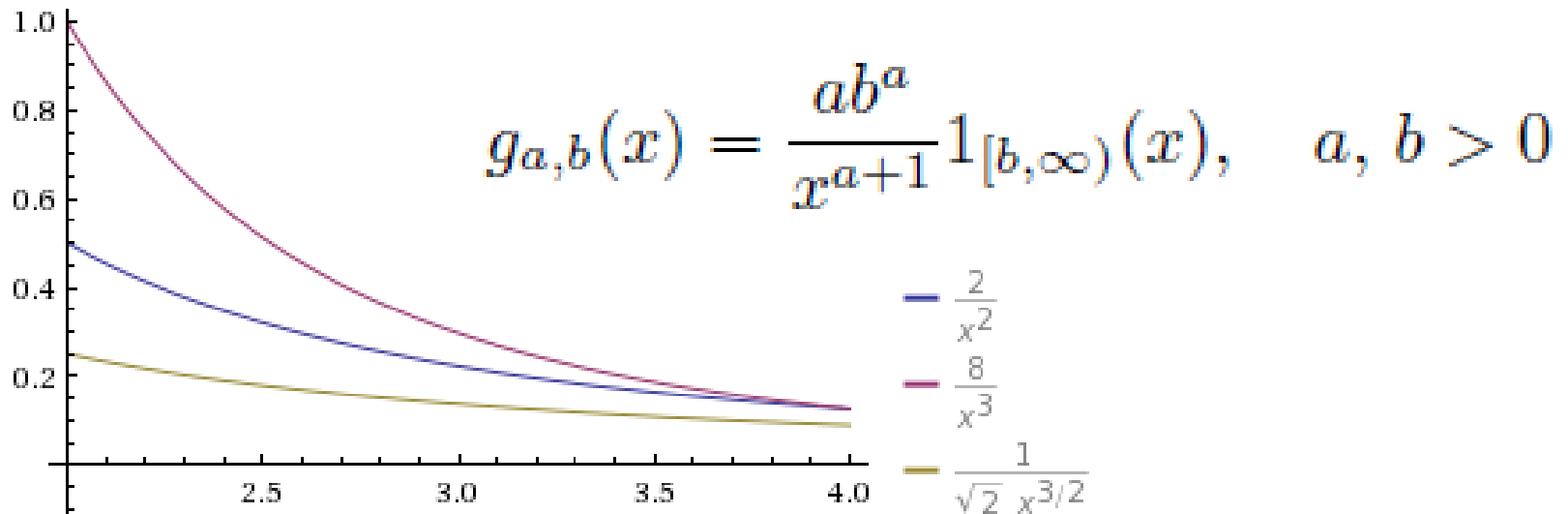


***F* distribution uses**

Statistics: testing hypotheses



Pareto distribution



$$\mathbb{E}X = ab/(a - 1) \text{ dla } a > 1$$

$$\text{Var}X = ab^2/[(a - 1)^2(a - 2)] \text{ dla } a > 2.$$



Pareto distribution uses

Pareto described the distribution of wealth/income among various groups in a population

(property: larger part of wealth accumulated by a smaller part of the population

→ Pareto 80-20 rule, which corresponds to a value $\alpha > 1$)

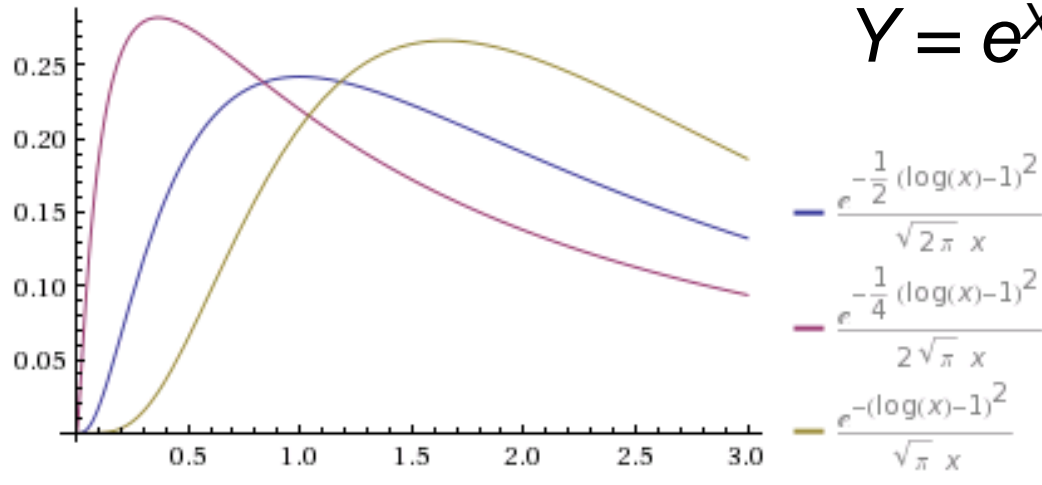
Used not only for income and wealth, but also in finance, insurance, actuarial



Lognormal distribution $L(m, \sigma^2)$, where $m \in \mathbb{R}$, $\sigma > 0$

$$g(x) = \frac{1}{\sqrt{2\pi\sigma x}} \exp\left(-\frac{(\ln x - m)^2}{2\sigma^2}\right) 1_{(0, \infty)}(x)$$

$$Y = e^X, \text{ where } X \sim N(m, \sigma^2)$$



$$\mathbb{E}Y = \exp\left(m + \frac{1}{2}\sigma^2\right)$$

$$\text{Var}Y = (\exp(\sigma^2) - 1) \cdot \exp(2m + \sigma^2)$$

Theorem: A product of lognormal variables has a lognormal distribution



Lognormal distribution uses

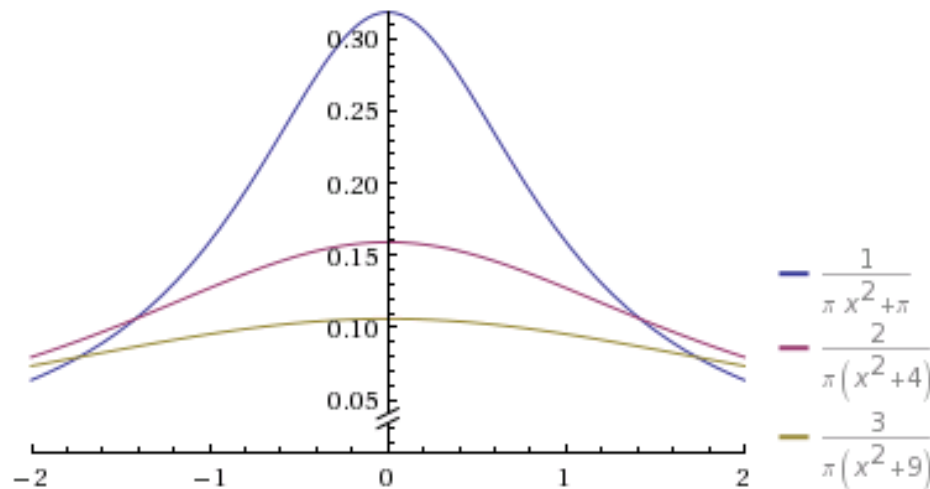
In most cases where we model positive random variables of multiplicative character (the CLT may be used for logarithms)

For modeling income (apart from the most rich).
In finance (eg. Black-Scholes model)



Cauchy distribution $\text{Cau}(a, m)$, where $a > 0$, $m \in \mathbb{R}$

$$g(x) = \frac{a}{\pi((x - m)^2 + a^2)}$$



Does not have an expected value nor higher moments.

The LLN nor the CLT may not be applied!



Cauchy distribution properties

Theorem: an average of identical independent Cauchy random variables is a Cauchy variable
→ conclusions based on the mean are worthless

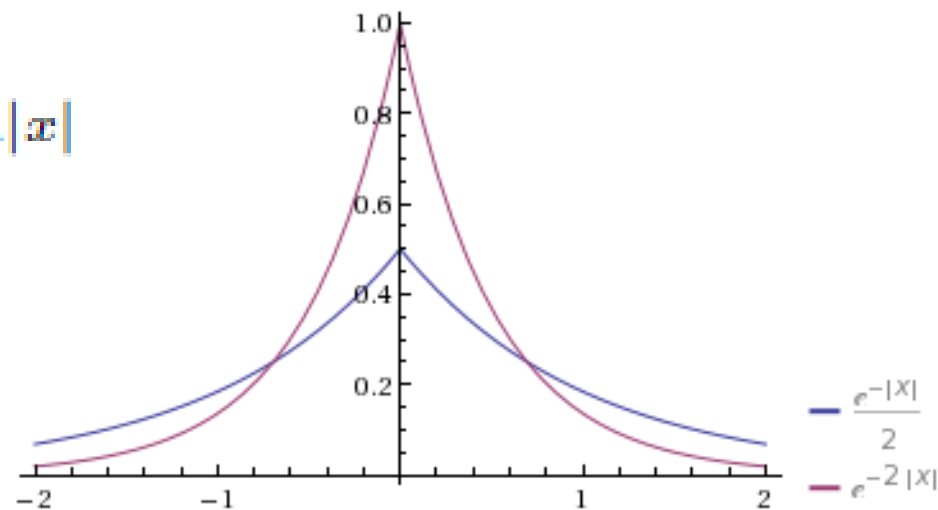
A ratio of two IIN random variables has a Cauchy distribution

$\text{Cau}(1,0)$ is equivalent to the t-Student distribution with one degree of freedom (1).



Two-sided exponential distribution (Laplace distribution) with parameter $\lambda > 0$

$$g(x) = \frac{1}{2} \lambda e^{-\lambda|x|}$$



$$\mathbb{E}X = 0, \text{Var}X = 2/\lambda^2$$



Weibull distribution

A different generalization of the exp. distr.

$$g(x) = \alpha\beta^{-\alpha}x^{\alpha-1}e^{-(x/\beta)^\alpha}1_{(0,\infty)}(x), \quad \alpha, \beta > 0.$$

$$\mathbb{E}X = \beta\Gamma\left(1 + \frac{1}{\alpha}\right)$$

$$\text{Var}X = \beta^2\left[\Gamma\left(1 + \frac{2}{\alpha}\right) - \left(\Gamma\left(1 + \frac{1}{\alpha}\right)\right)^2\right]$$

For modeling duration; depending on the value of α : decreasing, constant, increasing incidence. Actuarial sciences.



Survey

The last quiz...

<https://forms.gle/c4dC773k5uDTWzMD7>

