

Συνάριθμη Γάμμα

Η συνάριθμη $F: \mathbb{R}^+ \rightarrow \mathbb{R}^+$ ορίζεται ως εξής:

$$\Gamma(x) = \int_0^{+\infty} t^{x-1} \cdot e^{-t} dt \quad (I)$$

$x > 0$

ΙΔΙΟΤΗΤΕΣ ΣΥΝΑΡΙΘΜΗΣ ΓΑΜΜΑ:

$$(i): \Gamma(x) = \Gamma(x-1) \cdot (x-1)$$

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

$$(iii): \Gamma(1/2) = \sqrt{\pi}$$

$$(iv): \Gamma(n + \frac{1}{2}) = \sqrt{\pi} \cdot \prod_{j=0}^{n-1} (j + \frac{1}{2}) \quad (II)$$

Γάμμα Distribution: The Gamma distribution can be parameterized in terms of a scale parameter (α) and an inverse scale parameter (β), also called rate parameter

$$X \sim \Gamma(\alpha, \beta) \Leftrightarrow$$

$$f_X(x) = \frac{x^{\alpha-1} \cdot e^{-\beta x} \cdot \beta^\alpha}{\Gamma(\alpha)} \quad (III)$$

$x > 0$
 $\alpha > 0$
 $\beta > 0$

• Expected Value:

[#2]

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

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

Consider the following transformation:

$$u = \beta x \Rightarrow x = \frac{1}{\beta} u \Rightarrow dx = \frac{1}{\beta} du \quad (2)$$

$$E[X] = \frac{\beta^\alpha}{\Gamma(\alpha)} \int_0^{+\infty} \left(\frac{u}{\beta}\right)^\alpha \cdot e^{-u} \cdot \frac{du}{\beta} \rightarrow$$

$$E[X] = \frac{\beta^\alpha}{\beta^{\alpha+1} \Gamma(\alpha)} \int_0^{+\infty} u^\alpha \cdot e^{-u} \cdot du \quad (3) \rightarrow$$

$$E[X] = \frac{1}{\beta \Gamma(\alpha)} \cdot \int_0^{+\infty} u^\alpha \cdot e^{-u} \cdot du = \frac{\Gamma(\alpha+1)}{\beta \cdot \Gamma(\alpha)} \quad (4)$$

(*) We know that: $\Gamma(\alpha+1) = \alpha \Gamma(\alpha)$

$$E[X] = \frac{\alpha \Gamma(\alpha)}{\beta \Gamma(\alpha)} \Rightarrow$$

$$E[X] = \frac{\alpha}{\beta} \quad (5)$$

• Variance: Estimating the variance requires computing $E[X^2]$.

$$E[X^2] = \int_0^{+\infty} x^2 \cdot f_X(x) dx = \int_0^{+\infty} x^2 \cdot \frac{x^{\alpha-1} \cdot e^{-\beta x} \cdot \alpha}{\Gamma(\alpha)} dx \Rightarrow$$

$$E[X^2] = \frac{\beta^\alpha}{\Gamma(\alpha)} \cdot \int_0^{+\infty} x^{\alpha+1} \cdot e^{-\beta x} dx \quad (1)$$

Consider the same transformation:

$$u = \beta x \Rightarrow x = \frac{1}{\beta} u \Rightarrow dx = \frac{1}{\beta} du$$

$$E[X^2] = \frac{\beta^\alpha}{\Gamma(\alpha)} \cdot \int_0^{+\infty} \left(\frac{u}{\beta}\right)^{\alpha+1} \cdot e^{-u} \cdot \frac{du}{\beta} \rightarrow$$

$$E[X^2] = \frac{\beta^\alpha}{\beta^{\alpha+2} \Gamma(\alpha)} \int_0^{+\infty} u^{\alpha+1} \cdot e^{-u} du \rightarrow$$

$$E[X^2] = \frac{\Gamma(\alpha+2)}{\beta^2 \cdot \Gamma(\alpha)} \xrightarrow{\Gamma(\alpha+2) = (\alpha+1)\Gamma(\alpha+1) =} \\ = (\alpha+1) \cdot \alpha \cdot \Gamma(\alpha)$$

$$E[X^2] = \frac{\alpha(\alpha+1) \Gamma(\alpha)}{\beta^2 \Gamma(\alpha)} \rightarrow$$

$$E[X^2] = \frac{\alpha(\alpha+1)}{\beta^2} \quad (?)$$

[#4] Finally, the variance will be given as:

$$\text{Var}[X] = E[X^2] - E[X]^2 \Rightarrow$$

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

$$\text{Var}[X] = \frac{\alpha^2 + \alpha - \alpha^2}{\beta^2} \Rightarrow$$

$$\boxed{\text{Var}[X] = \frac{\alpha}{\beta^2} \quad (8)}$$

Moment Generating Function :

- The moment generating function $M_X(t)$ for a random variable X is defined as:

$$M_X(t) = E[e^{tX}] = \int_{-\infty}^{+\infty} e^{tx} f_X(x) dx \quad (A)$$

where $f_X(x)$ is the probability density function of X , given that the expectation exists for t in some neighborhood of 0.

- The moment generating function (MGF) (Moment-generating Function) is so called because it can be used to find the moments of the underlying distribution. Considering the Taylor series expansion of e^{tX} :

$$e^{tX} = 1 + tX + \frac{t^2 X^2}{2!} + \frac{t^3 X^3}{3!} + \dots + \frac{t^n X^n}{n!} + \dots \quad (B)$$

Thus,

$$E[e^{tX}] = 1 + tE[X] + \frac{t^2}{2!} E[X^2] + \frac{t^3}{3!} E[X^3] + \dots + \frac{t^n}{n!} E[X^n] + \dots \quad (C)$$

- The n -th term $M_n = E[X^n]$ identifies the n -th moment of X .

- The last formula suggests that:

$$M_X(t) = \sum_{n=0}^{\infty} \frac{t^n}{n!} M_n \quad (\text{A}) \text{ where } M_n = E[X^n] \quad (\text{E})$$

- The previous formulation suggests that if we are in position of acquiring the power series expansion of $M_X(t)$ in the following way:

$$M_X(t) = \sum_{n=0}^{\infty} a_n t^n \quad (\text{Z})$$

then by equating the respective terms of the two power series, we get that:

$$M_n \cdot \frac{t^n}{n!} = a_n t^n \Rightarrow M_n = n! a_n \quad (\text{H})$$

- TUTORIAL:** If random variable X has mgf $M_X(t)$,

then $E[X^n] = M_X^{(n)}(0)$ where

$$M_X^{(n)}(0) = \left. \frac{d^n}{dt^n} M_X(t) \right|_{t=0} \quad (\text{D})$$

ΑΣΚΗΣΗ: Να βρεται η ροηγεννησια συνδρομη πιας Τ.Μ. που αναλογει στην μαργαρινην Γαρβα.

$$M_X(t) = 2 \quad \text{οπως} \quad X \sim \text{Γαρβα}(a, \beta).$$

ΙΝΙ: Γνωριζουμε ότι: $f_X(x; a, \beta) = \frac{x^{a-1} \cdot e^{-\beta x} \cdot \beta^a}{\Gamma(a)}, \quad x > 0.$

Στη μετωπηα με την οριοθετηση της ροηγεννησιας συνδρομης θα ισχυει

οτι:
$$M_X(t) = E[e^{tX}] = \int_0^{+\infty} e^{tx} \cdot x^{a-1} \cdot e^{-\beta x} \cdot \frac{\beta^a}{\Gamma(a)} dx \quad (1)$$

Συνδιαλογισμοις των ευθεινων οπους έχουμε οτι:

$$M_X(t) = \frac{\beta^a}{\Gamma(a)} \cdot \int_0^{+\infty} x^{a-1} \cdot e^{-(\beta-t)x} \cdot dx \quad (2)$$

S.O.S The integral in Eq.(2) converges only when $\beta-t > 0$ or $t < \beta$.

Under these conditions, we may use the following transformation:

$$(\beta-t) \cdot x = u \Rightarrow x = \frac{u}{\beta-t} \Rightarrow dx = \frac{du}{\beta-t}$$

Eq.(2), then, provides:

$$M_X(t) = \frac{\beta^a}{\Gamma(a)} \int_0^{+\infty} \left(\frac{u}{\beta-t} \right)^{a-1} \cdot e^{-u} \cdot \frac{du}{\beta-t} \Rightarrow$$

$$M_X(t) = \frac{\beta^a}{(\beta-t)^a \Gamma(a)} \int_0^{+\infty} u^{a-1} \cdot e^{-u} du = \frac{\beta^a \Gamma(a)}{(\beta-t)^a \Gamma(a)} = \left(\frac{\beta}{\beta-t} \right)^a \quad (3)$$