

# Transformation of a bivariate $\Gamma$ -distribution

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## 1 Statement

Let  $X$  and  $Y$  be independent random variables  $\sim \Gamma(2, 5)$ ; we call  $f_X$  and  $f_Y$  their density function. Compute the density function of the variable  $T = X + Y$ .

## 2 Straight of the theory

$X$  and  $Y$  are random variables, that is they are  $X : \mathcal{S} \rightarrow \mathbb{R}^+$  and  $Y : \mathcal{S} \rightarrow \mathbb{R}^+$ . They are Gamma variables of parameters  $\alpha = 2$  and  $\beta = 5$  (see [Ano21]), thus, they are distributed in such a way that given a set of real-numbers  $E$ :

$$P(Y \in E) = P(X \in E) = \int_{t \in E} \frac{\beta^\alpha}{\Gamma(\alpha)} t^{\alpha-1} e^{-\beta t} dt = \int_{t \in E} \frac{5^2}{1} t e^{-5t} dt = \int_{t \in E} 25t e^{-5t} dt$$

Let us call  $f_X(t) = f_Y(t) = 25t e^{-5t}$ .

## 3 Calculation with a transformation

Introduce the transformation

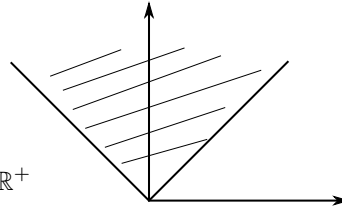
$$\begin{aligned} g : \mathbb{R}^+ \times \mathbb{R}^+ &\rightarrow O \\ (x, y) &\mapsto (x + y, x - y) \end{aligned}$$

This transformation is a bijection if  $O$  is defined to be

$$O = \{(a, b) \in \mathbb{R}^2 \mid a + b \geq 0 \text{ and } a - b \geq 0\}$$

Its inverse is:

$$\begin{aligned} g^{-1} : O &\rightarrow \mathbb{R}^+ \times \mathbb{R}^+ \\ (a, b) &\mapsto \left( \frac{1}{2}(a + b), \frac{1}{2}(a - b) \right) \end{aligned}$$



because  $g^{-1}(g(x, y)) = g^{-1}(x+y, x-y) = (\frac{1}{2}(x+y+x-y), \frac{1}{2}(x+y-x+y)) = (\frac{1}{2}2x, \frac{1}{2}2y) = (x, y)$

We apply the multivariate transformation theorem ([HMC20, 2.7]) and thus compute the Jacobian:

$$J_{g^{-1}} = \frac{\partial g_1^{-1}}{\partial a} \cdot \frac{\partial g_2^{-1}}{\partial b} - \frac{\partial g_2^{-1}}{\partial a} \cdot \frac{\partial g_1^{-1}}{\partial b} = \frac{1}{2} \cdot -\frac{1}{2} - \frac{1}{2} \cdot \frac{1}{2} = -\frac{1}{2}$$

The joint density of  $(X, Y)$  is  $f_{X,Y}(x, y) = f_X(x) \cdot f_Y(y)$  as the variables are independent.

The theorem thus states that the joint density of  $g(X, Y)$  is:

$$\begin{aligned} f_{A,B}(a, b) &= f_{X,Y}(g_1^{-1}(a, b), g_2^{-1}(a, b)) \cdot |J_{g^{-1}}| \\ &= |-\frac{1}{2}| \cdot f_X(\frac{1}{2}(a+b)) \cdot f_Y(\frac{1}{2}(a-b)) \\ &= \frac{1}{2} \cdot 25(\frac{1}{2}(a+b))e^{-\frac{5}{2}(a+b)} \cdot 25(\frac{1}{2}(a-b))e^{-\frac{5}{2}(a-b)} = \end{aligned}$$

then we can compute the density of the variable  $A$  which is  $X + Y$  as a marginal distribution:

$$f_A(a) = \int_{b \in \mathbb{R}^+} f_{A,B}(a, b) db = \int_{-a}^a \frac{1}{2} \cdot 25(\frac{1}{2}(a+b))e^{-\frac{5}{2}(a+b)} \cdot 25(\frac{1}{2}(a-b))e^{-\frac{5}{2}(a-b)} =$$

### 3.1 Result

The random variable  $T = X + Y$  is the same as the random variable  $A = g_1(X, Y)$ . Its density is given by:

$$\frac{625}{6}a^3e^{-5a}$$

as computed on WolframAlpha [inc20] with:

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integral between -a to a of (1/2)*25*(1/2)*(a+b)*e^(-2.5*(a+b))
* 25/2*(a-b)*e^(-2.5*(a-b))db
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## References

- [Ano21] Anonymous. Gamma distribution. web-page, February 2021. See [https://en.wikipedia.org/wiki/Gamma\\_distribution](https://en.wikipedia.org/wiki/Gamma_distribution).
- [HMC20] Robert Hogg, Joseph McKean, and Allen Craig. *Introduction to Mathematical Statistics*. Pearson Education, eighth edition (global) edition, 2020.
- [inc20] Wolfram Alpha inc. Wolfram alpha, 2020. See <https://wolframalpha.com>.