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## Gamma, Exponentials, Chi-squared, Normals and Convolutions

Thm 2. Assume  $X$  is gamma  $(\alpha, \lambda)$ ,  $Y$  is gamma  $(\alpha', \lambda)$ , and  $X$  and  $Y$  are independent. Then  $X+Y$  is gamma  $(\alpha+\alpha', \lambda)$ .

Proof. By scaling it suffices to consider  $\lambda=1$ .

[ Let  $x = x'/\lambda$ ,  $y = y'/\lambda$  where  $x'$  is gamma  $(\alpha, 1)$ ,  $y'$  is gamma  $(\alpha', 1)$  and  $x', y'$  are independent. If the result holds for  $x'+y'$ , it will follow for  $x+y = (x'+y')/\lambda$ . Check this. ]

Let  $a > 0$ .

$$\begin{aligned} \text{By Thm 1, } f_{X+Y}(a) &= \int_{-\infty}^{\infty} f_Y(a-x) f_X(x) dx \\ &= \int_{-\infty}^{\infty} \frac{(a-x)^{\alpha-1} e^{-(a-x)} \mathbb{1}_{\{a-x>0\}}}{\Gamma(\alpha)} \frac{x^{\alpha'-1} e^{-x} \mathbb{1}_{\{x>0\}}}{\Gamma(\alpha')} dx \end{aligned}$$

$$= \int_0^a \frac{(a-x)^{\alpha-1} x^{\alpha'-1} e^x e^{-x}}{\Gamma(\alpha) \Gamma(\alpha')} dx e^{-a}$$

$$= \int_0^a \frac{(1-\frac{x}{a})^{\alpha-1} (\frac{x}{a})^{\alpha'-1} \frac{dx}{a} a^{\alpha-1+\alpha'-1} a e^{-a}}{\Gamma(\alpha) \Gamma(\alpha')}$$

$$\left( w = \frac{x}{a}, dw = \frac{dx}{a} \right) = \underbrace{\left[ \int_0^1 \frac{(1-w)^{\alpha-1} w^{\alpha'-1} dw}{\Gamma(\alpha) \Gamma(\alpha')} \right]}_{\Gamma(\alpha+\alpha')} \underbrace{\frac{a^{\alpha+\alpha'-1} e^{-a}}{\Gamma(\alpha+\alpha')}}_{f(a)}$$

So  $f_{X+Y}(a) = C_{\alpha, \alpha'} f(a)$ , where  $f$  is the density of gamma  $(\alpha+\alpha', 1)$ .

$$\text{But } 1 = \int_{-\infty}^{\infty} f_{X+Y}(a) da = \int_{-\infty}^{\infty} C_{\alpha, \alpha'} f(a) da = C_{\alpha, \alpha'} \text{ so } f_{X+Y} = f. \\ \text{i.e. } X+Y \text{ is gamma } (\alpha+\alpha', 1) \quad \square$$

Corollary 1 If  $T_1, \dots, T_n$  are independent exponential ( $\lambda$ ) r.v.'s,

then  $S_n = T_1 + T_2 + \dots + T_n$  has a gamma( $n, \lambda$ ) distribution.

Proof  $T_1, \dots, T_n$  are independent gamma( $1, \lambda$ ) r.v.'s.

So by Thm 2,  $T_1 + T_2$  is gamma( $2, \lambda$ )

$(T_1 + T_2) + T_3$  is gamma( $3, \lambda$ )

and so iterating we get  $T_1 + \dots + T_n$  is gamma( $n, \lambda$ ),  $\square$

Corollary 2 If  $Z_1, \dots, Z_n$  are independent standard normal r.v.'s

then  $Y = Z_1^2 + Z_2^2 + \dots + Z_n^2$  has a gamma( $\frac{n}{2}, \frac{1}{2}$ ) distribution,

$$\text{i.e. } f_Y(y) = \frac{\left(\frac{y}{2}\right)^{n-1} e^{-y/2} \Gamma\left(\frac{n}{2}\right)}{\Gamma\left(\frac{n}{2}\right)} \mathbb{1}_{\{y > 0\}}$$

Remarks (i) Also say  $Y$  has a  $\chi^2$ -distribution (chi-squared distribution) with  $n$  degrees of freedom.

(ii) The  $\chi^2$ -distribution arises in statistical estimation of  $\mu$  and  $\sigma^2$  for normal populations.

Proof. It suffices to show (\*)  $Z_1^2$  is gamma( $\frac{1}{2}, \frac{1}{2}$ ).

Assume (ii). Then  $Z_1^2, \dots, Z_n^2$  are independent gamma( $\frac{1}{2}, \frac{1}{2}$ ) r.v.'s

By Theorem 3,  $Z_1^2 + \dots + Z_n^2$  is gamma( $\frac{n}{2}, \frac{1}{2}$ ) and we would be done.

To prove (i), recall we showed on Oct 31 that

$$f_{Z_1^2}(x) = \frac{x^{-1/2} e^{-x/2}}{\sqrt{2\pi}} \mathbb{1}_{\{x > 0\}}$$

The gamma( $\frac{1}{2}, \frac{1}{2}$ ) density is  $f(x) = \frac{\left(\frac{x}{2}\right)^{-1/2} e^{-x/2} \Gamma\left(\frac{1}{2}\right)}{\Gamma\left(\frac{1}{2}\right)} \mathbb{1}_{\{x > 0\}}$ .

$$= \frac{x^{-1/2} e^{-x/2}}{\sqrt{2} \Gamma\left(\frac{1}{2}\right)} \mathbb{1}_{\{x > 0\}}.$$

Since both  $f_{Z^2}$  and  $f$  must integrate to 1, it follows that  $f_{Z^2} = f$  (and  $P(Z^2) = \sqrt{\pi}$ ), and so (a) is proved.  $\square$

Thm 3. If  $X_1, X_2, \dots, X_n$  are independent normal r.v.'s, and  $X_i$  is  $N(\mu_i, \sigma_i^2)$ , then  $S_n = X_1 + X_2 + \dots + X_n$  is a normal r.v. with  $\mu = \sum_{i=1}^n \mu_i$ ,  $\sigma^2 = \sum_{i=1}^n \sigma_i^2$ .

Proof. We will give a proof based on moment-generating functions later. See p. 296 for a proof based on Theorem 1. To illustrate the idea we prove the result for  $X_1 + X_2$  where  $X_1, X_2$  are independent standard normals:

$$\begin{aligned} f_{X_1+X_2}(a) &= \int_{-\infty}^{\infty} f_{X_2}(a-x) f_{X_1}(x) dx \\ (1) \quad &= \int_{-\infty}^{\infty} \frac{e^{-\frac{1}{2}[(a-x)^2 + x^2]}}{\sqrt{2\pi} \sqrt{2\pi}} dx \end{aligned}$$

$$\text{Now } -\frac{1}{2}[(a-x)^2 + x^2] = -\frac{1}{2}[a^2 - 2ax + 2x^2] = -(x - \frac{a}{2})^2 - \frac{a^2}{4}$$

$$\therefore f_{X_1+X_2}(a) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-(x - \frac{a}{2})^2 - a^2/4} dx$$

$$= \frac{e^{-a^2/4}}{2\pi} \int_{-\infty}^{\infty} e^{-w^2} dw$$

$$w = x - \frac{a}{2}$$

$$= \frac{e^{-a^2/4}}{2\pi} \int_{-\infty}^{\infty} e^{-z^2/2} \frac{dz}{\sqrt{2}}$$

$$w = \frac{z}{\sqrt{2}} \quad dx = \frac{dz}{\sqrt{2}}$$

$$= \frac{e^{-a^2/4}}{2\pi} (\sqrt{2\pi}) / \sqrt{2}$$

$$= \frac{e^{-a^2/4}}{\sqrt{2\pi} \sqrt{2}} = \frac{e^{-a^2/2\sigma^2}}{\sqrt{2\pi} \sigma}$$

$$\text{where } \underline{\sigma^2 = 2}$$

$\therefore X_1 + X_2$  is  $N(0, 2)$ , as required.