一: Gamma 分布: 记 $X \sim \Gamma(\alpha, \lambda)$ (称参数为 α, β 的 Γ 分布),若

$$f(x) = \begin{cases} \frac{\lambda^{\alpha}}{\Gamma(\alpha)} x^{\alpha - 1} e^{-\lambda x}, & x > 0, \\ 0, & x \le 0. \end{cases}$$

显然,我们有

• $\Gamma(1,\lambda) = E(\lambda)$; 指数分布是 Gamma 分布的特例; 另外当 $\alpha = n \in N$, 也称 Gamma 分布为 Erlang 分布;

(称后者为自由度为1的 χ^2 分布),

事实上,我们有 $\forall y > 0$,

$$f_Y(y) = f_Z(\sqrt{y}) \frac{1}{2\sqrt{y}} + f_Z(-\sqrt{y}) \frac{1}{2\sqrt{y}} = \frac{1}{\sqrt{2\pi y}} e^{-\frac{y}{2}}$$

● 矩母函数: X 的矩母函数 $M_X(u)$, 并由此可求出EX,DX。

$$M_{X}(u) = E(e^{uX}) = \int_{0}^{\infty} e^{ux} \frac{\lambda^{\alpha}}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x} dx = \left(\frac{\lambda}{\lambda - u}\right)^{\alpha}, \forall u < \lambda$$

$$EX = M'_{X}(0) = \frac{\alpha}{\lambda},$$

$$EX^{2} = M''_{X}(0) = \frac{\alpha + \alpha^{2}}{\lambda^{2}} \Rightarrow DX = \frac{\alpha}{\lambda^{2}}$$

(直接计算如下:

$$EX^{n} = \int_{0}^{\infty} x^{n} \frac{\lambda^{\alpha}}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x} dx = \frac{\lambda^{\alpha}}{\Gamma(\alpha)} \int_{0}^{\infty} x^{n+\alpha-1} e^{-\lambda x} dx$$

$$= \frac{\lambda^{\alpha}}{\Gamma(\alpha)} \frac{\Gamma(n+\alpha)}{\lambda^{n+\alpha}} = \frac{\alpha(\alpha+1)\cdots(\alpha+n-1)}{\lambda^{n}}$$

● 可加性: 若 X_1, \dots, X_n 相互独立,且 $X_i \sim \Gamma(\alpha_i, \lambda)$,求 $\sum_{i=1}^n X_i$ 的分布。证明:

$$M_{\sum_{i=1}^{n} X_{i}}(u) = \prod_{i=1}^{n} M_{X_{i}}(u) = \prod_{i=1}^{n} \left(\frac{\lambda}{\lambda - u}\right)^{\alpha_{i}} = \left(\frac{\lambda}{\lambda - u}\right)^{\sum_{i=1}^{n} \alpha_{i}}, u < \lambda$$

$$\Rightarrow \sum_{i=1}^{n} X_{i} \sim \Gamma\left(\sum_{i=1}^{n} \alpha_{i}, \lambda\right)$$

★ 推论:

(1)
$$X_i^{i.i.d.} \sim E(\lambda) = \Gamma(1,\lambda)$$
, $\bigcup \sum_{i=1}^n X_i \sim \Gamma(n,\lambda)$;

(2)
$$X_i^{i.i.d.} \sim N(0,1)$$
,则 $\sum_{i=1}^n X_i^2 \sim \Gamma(\frac{n}{2},\frac{1}{2}) \triangleq \chi^2(n)$ (即自由度为 n 的 χ^2 分 布)。

Pf: (2) 的证明:
$$X_i \sim N(0,1) \Rightarrow X_i^2 \sim \chi^2(1) = \Gamma(\frac{1}{2}, \frac{1}{2})$$
,

从而
$$\sum_{i=1}^{n} X_i^2 \sim \Gamma(\frac{n}{2}, \frac{1}{2})$$
。

(3)
$$X_i^{i.i.d.} \sim U(0,1)$$
,则 $Y = -2\sum_{i=1}^n \ln X_i$ 是否为 Γ 分布,为什么?

Pf: (3) 的证明:

$$Y_i = -2 \ln X_i \Rightarrow x = e^{-\frac{y}{2}} \Rightarrow \frac{dx}{dy} = -\frac{1}{2} e^{-\frac{y}{2}}$$

$$f_{Y_i}(y) = f_{X_i}(e^{-\frac{y}{2}}) \left| \frac{dx}{dy} \right| = 1 \times \frac{1}{2} e^{-\frac{y}{2}}, y > 0$$

$$Y_i = -2 \ln X_i \sim E(\frac{1}{2}) = \Gamma(1, \frac{1}{2})$$

从而,
$$Y = -2\sum_{i=1}^{n} \ln X_i = \sum_{i=1}^{n} Y_i \sim \Gamma(n, \frac{1}{2}) = \chi^2(2n)$$

ullet $X_i \sim \Gamma(lpha_i, \lambda)$ 相互独立,证明 $U = X_1 + X_2$ 与 $V = \frac{X_1}{X_1 + X_2}$ 相互独立。

证明:
$$\begin{cases} U = \frac{X}{X+Y} \Rightarrow \begin{cases} x = uv \\ y = v(1-u) \end{cases} \\ \frac{\partial(x,y)}{\partial(u,v)} = \begin{vmatrix} v & u \\ -v & 1-u \end{vmatrix} = v \\ \frac{\partial}{\partial(x,y)} = \frac{\partial(x,y)}{\partial(x,y)} = \frac{\partial(x,y)}{\partial(x$$

即 $U \sim \mathrm{B}(\alpha_1, \alpha_2)$, $V \sim G(\alpha_1 + \alpha_2, \lambda)$ 且相互独立。

二: Beta 分布:

记
$$B(\alpha, \beta) = \int_{0}^{1} x^{\alpha-1} (1-x)^{\beta-1} dx, \, \alpha > 0, \beta > 0$$
, 称 $X \sim Beta(\alpha, \beta)$ 服从参数为

 α, β 的 Beta 分布,如果它的密度函数为

$$f(x) = \begin{cases} \frac{1}{B(\alpha, \beta)} x^{\alpha - 1} (1 - x)^{\beta - 1}, & 0 < x < 1, \\ 0, & otherwise. \end{cases}$$

$$B(\alpha, \beta) = \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha + \beta)}$$

显然: $Beta(\alpha,\beta) = U(0,1)$,

● Beta 分布的期望与方差

$$EX = \int_0^1 x \frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1} dx$$

$$= \frac{1}{B(\alpha, \beta)} \int_0^1 x^{\alpha} (1-x)^{\beta-1} dx$$

$$= \frac{B(\alpha+1, \beta)}{B(\alpha, \beta)} = \frac{\alpha}{\alpha+\beta}$$

$$EX^2 = \int_0^1 x^2 \frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1} dx$$

$$= \frac{1}{B(\alpha, \beta)} \int_0^1 x^{\alpha+1} (1-x)^{\beta-1} dx$$

$$= \frac{B(\alpha+2, \beta)}{B(\alpha, \beta)} = \frac{(\alpha+1)\alpha}{(\alpha+\beta+1)(\alpha+\beta)}$$

$$DX = \frac{(\alpha+1)\alpha}{(\alpha+\beta+1)(\alpha+\beta)} - (\frac{\alpha}{\alpha+\beta})^2 = \frac{\alpha\beta}{(\alpha+\beta+1)(\alpha+\beta)^2}$$

- 设 $Y_i \sim \Gamma(\alpha_i, \beta)$ 相互独立,试证 $V = \frac{Y_1}{Y_1 + Y_2} \sim B(\alpha_1, \alpha_2)$ 。
- Beta 分布与 F 分布 (统计相关):

*:
$$X \sim F(m,n)$$
, 则

$$EX = \frac{n}{n-2}, n > 2;$$

$$DX = 2(\frac{n}{n-2})^2 \frac{m+n-2}{m(n-4)}, n > 4$$

*:
$$X \sim F(m,n) \Rightarrow \frac{\frac{m}{n}X}{1+\frac{m}{n}X} \sim Beta(\frac{m}{2},\frac{n}{2}).$$

*:
$$X \sim Beta(\frac{m}{2}, \frac{n}{2}) \Rightarrow \frac{nX}{m(1-X)} \sim F(m, n)$$
.

三: Cauchy 分布:

记 $X \sim C(\lambda, \mu)$ (称参数为 λ, μ 的 Cauchy 分布),若

$$f(x) = \frac{\lambda}{\pi} \frac{1}{\lambda^2 + (x - \mu)^2}, x \in R \quad (数学期望不存在)$$

特征函数 $\varphi(\theta) = e^{i\mu\theta - \lambda|\theta|}$

* (可加性) $X \sim C(\lambda_1, \mu_1)$, $Y \sim C(\lambda_2, \mu_2)$ 独立,则

$$X + Y \sim C(\lambda_1 + \lambda_2, \mu_1 + \mu_2)$$

• $X \sim C(1,0)$ (标准 Cauchy 分布) $\Leftrightarrow \frac{1}{X} \sim C(1,0)$

证明: 记
$$Y = \frac{1}{X}$$
 , 反函数 $x^{-1}(y) = \frac{1}{y}$

$$f_Y(y) = f_X(x^{-1}(y)) \left| \frac{dx^{-1}(y)}{dy} \right| = \frac{1}{\pi} \frac{1}{1 + (\frac{1}{y})^2} \left| -\frac{1}{y^2} \right| = \frac{1}{\pi} \frac{1}{1 + y^2}, y \in R$$

• $X, Y \sim N(0, \sigma^2)$ 独立,则 $\frac{X}{Y} \sim C(1, 0)$

证明: 设
$$\begin{cases} U = \frac{X}{Y} \Rightarrow \begin{cases} x = uv & \frac{\partial(x, y)}{\partial(u, v)} = \begin{vmatrix} v & u \\ 0 & 1 \end{vmatrix} = v \end{cases}$$

$$f_{U,V}(u,v) = f_{X,Y}(uv,v) |v| = \frac{1}{2\pi\sigma^2} e^{-\frac{(uv)^2 + v^2}{2\sigma^2}} |v| = \frac{1}{2\pi\sigma^2} e^{-\frac{(u^2 + 1)v^2}{2\sigma^2}} |v|, u, v \in \mathbb{R}$$

$$f_{U}(u) = \int_{-\infty}^{\infty} f_{U,V}(u,v) dv = \int_{-\infty}^{\infty} \frac{1}{2\pi\sigma^{2}} e^{-\frac{(u^{2}+1)v^{2}}{2\sigma^{2}}} |v| dv = 2\int_{0}^{\infty} \frac{1}{2\pi\sigma^{2}} e^{-\frac{(u^{2}+1)v^{2}}{2\sigma^{2}}} v dv = \frac{1}{\pi} \frac{1}{1+u^{2}}, u \in \mathbb{R}$$

•
$$X \sim U(-\frac{\pi}{2}, \frac{\pi}{2})$$
独立,则 $\tan X \sim C(1,0)$

证明: $记 Y = \tan X$, 反函数 $x^{-1}(y) = \arctan y$

$$f_Y(y) = f_X(x^{-1}(y)) \left| \frac{dx^{-1}(y)}{dy} \right| = \frac{1}{\pi} \frac{1}{1+y^2}, y \in R$$