

## 556: MATHEMATICAL STATISTICS I

### MULTIVARIATE TRANSFORMATIONS: THE CAUCHY DISTRIBUTION

The Cauchy distribution is a symmetric distribution on  $(-\infty, \infty)$  with pdf

$$f_X(x; \theta, \sigma) = \frac{1}{\pi} \frac{1}{\sigma} \cdot \frac{1}{1 + \left(\frac{x - \theta}{\sigma}\right)^2} = \frac{1}{\pi} \cdot \frac{\sigma}{\sigma^2 + (x - \theta)^2}$$

The standard case is  $\theta = 0, \sigma = 1$ .

The Cauchy distribution arises as the ratio of two independent Gaussian random variables. Suppose that  $X, Y \sim \text{Normal}(0, 1)$ . We then proceed by

- (i) defining the transformation  $U = X/Y$  and  $V = |Y|$ ,
- (ii) finding the joint pdf  $f_{U,V}(u, v)$ , and
- (iii) integrating out  $V$  to obtain the marginal pdf of  $U$ .

The mapping  $U = X/Y$  and  $V = |Y|$  is not 1-1: the two points  $(x, y)$  and  $(-x, -y)$  map to the same  $(u, v)$ . However, we may partition the support of  $(X, Y)$  into  $A_0, A_1, A_2$  such that the mapping from  $A_i$  to  $(U, V)$  is one-to-one.

1.  $A_0 = \{(X, Y) : Y = 0\}$ : we can ignore this case as  $P_Y[Y = 0] = 0$  when  $Y \sim \text{Normal}(0, 1)$ .
2.  $A_1 = \{(X, Y) : Y > 0\}$ : The mapping  $U = X/Y, V = |Y|$  is 1-1, and the inverse mappings are  $h_{11}(u, v) = uv, h_{21}(u, v) = v$ .
3.  $A_2 = \{(X, Y) : Y < 0\}$ : The mapping  $U = X/Y, V = |Y|$  is one-to-one, and the inverse mappings are  $h_{12}(u, v) = -uv, h_{22}(u, v) = -v$ .

In cases (ii) and (iii) we have the following Jacobians:

$$J_1 = \begin{vmatrix} \frac{\partial h_{11}(u, v)}{\partial u} & \frac{\partial h_{11}(u, v)}{\partial v} \\ \frac{\partial h_{21}(u, v)}{\partial u} & \frac{\partial h_{21}(u, v)}{\partial v} \end{vmatrix} = \begin{vmatrix} \frac{\partial(uv)}{\partial u} & \frac{\partial(uv)}{\partial v} \\ \frac{\partial v}{\partial u} & \frac{\partial v}{\partial v} \end{vmatrix} = \begin{vmatrix} v & u \\ 0 & 1 \end{vmatrix} = v$$

$$J_2 = \begin{vmatrix} \frac{\partial h_{12}(u, v)}{\partial u} & \frac{\partial h_{12}(u, v)}{\partial v} \\ \frac{\partial h_{22}(u, v)}{\partial u} & \frac{\partial h_{22}(u, v)}{\partial v} \end{vmatrix} = \begin{vmatrix} \frac{\partial(-uv)}{\partial u} & \frac{\partial(-uv)}{\partial v} \\ \frac{\partial(-v)}{\partial u} & \frac{\partial(-v)}{\partial v} \end{vmatrix} = \begin{vmatrix} -v & -u \\ 0 & -1 \end{vmatrix} = v$$

We have that

$$f_{X,Y}(x, y) = \frac{1}{\sqrt{2\pi}} \exp\{-x^2/2\} \frac{1}{\sqrt{2\pi}} \exp\{-y^2/2\} = \frac{1}{2\pi} \exp\left\{-\frac{(x^2 + y^2)}{2}\right\}$$

so therefore

$$\begin{aligned} f_{U,V}(u, v) &= \mathbb{1}_{A_1}(u, v) f_{X,Y}(h_{11}(u, v), h_{21}(u, v)) |J_1| + \mathbb{1}_{A_2}(u, v) f_{X,Y}(h_{12}(u, v), h_{22}(u, v)) |J_2| \\ &= \frac{\mathbb{1}_{A_1}(u, v)}{2\pi} \exp\left(-\frac{(uv)^2 + v^2}{2}\right) |v| + \frac{\mathbb{1}_{A_2}(u, v)}{2\pi} \exp\left(-\frac{(-uv)^2 + (-v)^2}{2}\right) |v| \\ &= \frac{v}{\pi} \exp\left(-\frac{v^2(u^2 + 1)}{2}\right), \quad u \in \mathbb{R}, v \in \mathbb{R}^+ \end{aligned}$$

and hence, on marginalization

$$\begin{aligned}
 f_U(u) &= \int_0^\infty \frac{v}{\pi} \exp\left\{-\frac{v^2(u^2+1)}{2}\right\} dv && \text{integrating out } v \\
 &= \int_0^\infty \frac{1}{2\pi} \exp\left\{-\frac{(u^2+1)}{2}z\right\} dz && \text{setting } z = v^2 \text{ and } dz = 2v dv \\
 &= \frac{1}{2\pi} \cdot \frac{2}{1+u^2} && \int_0^\infty \exp(-\alpha z) dz = \frac{1}{\alpha} \\
 &= \frac{1}{\pi} \cdot \frac{1}{1+u^2}
 \end{aligned}$$

The general *Cauchy*( $\theta, \sigma$ ) form is generated using a location-scale transform: that is, if  $Z \sim \text{Cauchy}(0, 1)$ , then

$$X = \sigma Z + \theta$$

has a *Cauchy*( $\theta, \sigma$ ) distribution.

The second (equivalent) construction of the standard Cauchy distribution is as a *scale mixture*. Suppose  $X$  and  $Y$  have a joint distribution specified as

$$\begin{aligned}
 Y &\sim \chi_1^2 \equiv \text{Gamma}(1/2, 1/2) \\
 X|Y = y &\sim \text{Normal}(0, y^{-1})
 \end{aligned}$$

that is, the variance of  $X$  given  $Y = y$  is  $1/y$ . Then we have that

$$\begin{aligned}
 f_X(x) &= \int_{-\infty}^\infty f_{X,Y}(x, y) dy = \int_0^\infty f_{X|Y}(x|y) f_Y(y) dy \\
 &= \int_0^\infty \frac{1}{\sqrt{2\pi}} y^{1/2} \exp\left\{-\frac{y}{2}x^2\right\} \frac{(1/2)^{1/2}}{\Gamma(1/2)} y^{-1/2} \exp\left\{-\frac{y}{2}\right\} dy \\
 &= \frac{1}{2\pi} \int_0^\infty \exp\left\{-\frac{y}{2}(1+x^2)\right\} dy \\
 &= \frac{1}{\pi} \frac{1}{1+x^2}
 \end{aligned}$$

as  $\Gamma(1/2) = \sqrt{\pi}$ .