

# The Distribution Function of $X_1$ Divided by $X_2$

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**Exercise.** Given  $(X_1, X_2)^T \sim \mathcal{N}(\mu, V)$  where

$$\mu = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, V = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

then what's the distribution function of  $\frac{X_1}{X_2}$ ?

Let's first recall the definition of multivariate normal distribution.

**Definition 1.** A random vector  $\mathbf{X} = (X_1, X_2, \dots, X_n)^T$  is called a normal random vector, which means  $\mathbf{X} \sim \mathcal{N}(\mu, V)$  for some  $\mu, V$ , if it has the probability density function for  $x \in \mathbb{R}^n$

$$f(x) = \frac{\sqrt{|V^{-1}|}}{(2\pi)^{\frac{n}{2}}} e^{-\frac{1}{2}(x-\mu)^T V^{-1}(x-\mu)}$$

It remains to check some important properties like  $\int_{x \in \mathbb{R}^n} f(x) dx = 1$  and so on. However, the following definition is equivalent to the above, but the proof of equivalence seems to be complicated.

**Definition 2.** A random vector  $\mathbf{X} = (X_1, X_2, \dots, X_n)^T$  is called a normal random vector if  $\sum_{i=1}^n r_i X_i$  is normal for any  $r_i \in \mathbb{R}$ .

Now let's get back to the exercise. Consider the following general theorem.

**Theorem 1.** Let  $(X, Y)^T$  be a random vector with probability density function  $f(x, y)$ . Then the probability density function of  $\frac{Y}{X}$  is given by

$$f_{Y/X}(z) = \int_{-\infty}^{+\infty} |x| f(x, xz) dx.$$

*Proof of theorem 1.* Let  $G = \{(x, y) : \frac{y}{x} \leq z\}$  for some  $z$ . And then the distribution function of  $\frac{Y}{X}$  is

$$\begin{aligned}
F_{Y/X}(z) &= P\{Y/X \leq z\} = \iint_G f(x, y) \, dx \, dy \\
&= \left( \int_{y/x \leq z, x < 0} + \int_{y/x \leq z, x > 0} \right) f(x, y) \, dx \, dy \\
&= \int_{-\infty}^0 \left[ \int_{xz}^{\infty} f(x, y) \, dy \right] dx + \int_0^{\infty} \left[ \int_{-\infty}^{xz} f(x, y) \, dy \right] dx \\
&= \int_{-\infty}^z \left[ \int_{-\infty}^{\infty} |x| f(x, xw) \right] dw
\end{aligned}$$

Since typing formulas in L<sup>A</sup>T<sub>E</sub>X is tough and despairing, I omit some intermediate step and give the final expression directly. By the definition of pdf, we conclude

$$f_{Y/X}(z) = \int_{-\infty}^{+\infty} |x| f(x, xz) \, dx.$$

□

Suppose  $(X_1, X_2)^T \sim \mathcal{N}(\mu, V)$  where

$$\mu = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, V = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}.$$

By definition 1, it has probability density function

$$f(x_1, x_2) = \frac{1}{2\pi} e^{-\frac{x_1^2 + x_2^2}{2}}$$

Applying theorem 1, all we need to do is computing

$$\begin{aligned}
f_{X_1/X_2}(z) &= \int_{-\infty}^{+\infty} |x| f(xz, x) \, dx \\
&= 2 \int_0^{+\infty} x f(xz, x) \, dx \\
&= \int_0^{+\infty} f(xz, x) \, dx^2 \\
&= \frac{1}{2\pi} \int_0^{+\infty} e^{-\frac{x^2(1+z^2)}{2}} \, dx^2 \\
&= -\frac{1}{\pi(1+z^2)} e^{-\frac{t(1+z^2)}{2}} \Big|_0^{+\infty} \\
&= \frac{1}{\pi(1+z^2)}
\end{aligned}$$

Thus, the distribution function of  $\frac{X_1}{X_2}$  is

$$F_{X_1/X_2}(x) = \int_{-\infty}^x \frac{1}{\pi(1+z^2)} \, dz = \frac{\arctan(x)}{\pi} + \frac{1}{2}$$