Theorem If X_1 and X_2 are independent standard normal random variables, then $Y = X_1/X_2$ has the standard Cauchy distribution.

Proof Let X_1 and X_2 be independent standard normal random variables. We can write their probability density functions as

$$f_{X_1}(x_1) = \frac{e^{-x_1^2/2}}{\sqrt{2\pi}}$$
 $-\infty < x_1 < \infty$

and

$$f_{X_2}(x_2) = \frac{e^{-x_2^2/2}}{\sqrt{2\pi}}$$
 $-\infty < x_2 < \infty.$

Since X_1 and X_2 are independent, the joint probability density function of X_1 and X_2 is

$$f_{X_1, X_2}(x_1, x_2) = \frac{e^{-(x_1^2 + x_2^2)/2}}{2\pi} - \infty < x_1 < \infty, -\infty < x_2 < \infty.$$

Consider the 2×2 transformation

$$Y_1 = g_1(X_1, X_2) = \frac{X_1}{X_2}$$
 and $Y_2 = g_2(X_1, X_2) = X_2$

which is a 1–1 transformation from $\mathcal{X} = \{(x_1, x_2) \mid -\infty < x_1 < \infty, -\infty < x_2 < \infty\}$ to $\mathcal{Y} = \{(y_1, y_2) \mid -\infty < y_1 < \infty, -\infty < y_2 < \infty\}$ with inverses

$$X_1 = g_1^{-1}(Y_1, Y_2) = Y_1 Y_2$$
 and $X_2 = g_2^{-1}(Y_1, Y_2) = Y_2$

and Jacobian

$$J = \left| \begin{array}{cc} Y_2 & Y_1 \\ 0 & 1 \end{array} \right| = Y_2.$$

Therefore, by the transformation technique, the joint probability density function of Y_1 and Y_2 is

$$f_{Y_1,Y_2}(y_1, y_2) = f_{X_1,X_2} \left(g_1^{-1}(y_1, y_2), g_2^{-1}(y_1, y_2) \right) |J|$$

$$= \frac{e^{-(y_1^2 y_2^2 + y_2^2)/2}}{2\pi} |y_2| - \infty < y_1 < \infty, -\infty < y_2 < \infty.$$

The probability density function of Y_1 is

$$f_{Y_1}(y_1) = \int_{-\infty}^{\infty} f_{Y_1, Y_2}(y_1, y_2) dy_2$$

$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} |y_2| e^{-y_2^2 (y_1^2 + 1)/2} dy_2$$

$$= \frac{1}{\pi (y_1^2 + 1)} - \infty < y_1 < \infty,$$

which is the probability density function of a standard Cauchy random variable.

APPL verification: The APPL statements

```
X1 := NormalRV(0, 1);
X2 := NormalRV(0, 1);
g := [[x -> 1 / x, x -> 1 / x], [-infinity, 0, infinity]];
Y := Transform(X2, g);
Product(X1, Y);
```

produce the probability density function of a Cauchy random variable.