**Theorem** If  $X_1$  and  $X_2$  are independent gamma $(1, \beta_i)$  random variables, for i = 1, 2, then  $X_1/X_2$  has the inverted beta distribution.

**Proof** Let  $X_1 \sim \text{gamma}(1, \beta_1)$  and  $X_2 \sim \text{gamma}(1, \beta_2)$  be independent random variables. We can write their probability density functions as

$$f_{X_1}(x_1) = \frac{x_1^{\beta_1 - 1} e^{-x_1}}{\Gamma(\beta_1)} \qquad x > 0$$

and

$$f_{X_2}(x_2) = \frac{x_2^{\beta_2 - 1} e^{-x_2}}{\Gamma(\beta_2)} \qquad x > 0$$

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

$$f_{X_1,X_2}(x_1,x_2) = \frac{x_1^{\beta_1-1} x_2^{\beta_2-1} e^{-x_1-x_2}}{\Gamma(\beta_1)\Gamma(\beta_2)} \qquad x_1 > 0, x_2 > 0.$$

Consider the  $2 \times 2$  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  $\mathcal{X} = \{(x_1, x_2) | x_1 > 0, x_2 > 0\}$  to  $\mathcal{Y} = \{(y_1, y_2) | y_1 > 0, y_2 > 0\}$  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

$$\begin{aligned} f_{Y_1,Y_2}(y_1,y_2) &= f_{X_1,X_2} \left( g_1^{-1}(y_1,y_2), g_2^{-1}(y_1,y_2) \right) |J| \\ &= \frac{y_1^{\beta_1-1} y_2^{\beta_1-1} y_2^{\beta_2-1} e^{-y_1 y_2 - y_2}}{\Gamma(\beta_1) \Gamma(\beta_2)} |y_2| \\ &= \frac{y_1^{\beta_1-1} y_2^{\beta_1+\beta_2-1} e^{-y_2(y_1+1)}}{\Gamma(\beta_1) \Gamma(\beta_2)} \qquad y_1 > 0, y_2 > 0. \end{aligned}$$

Using integration by parts, the probability density function of  $Y_1$  is

$$f_{Y_1}(y_1) = \int_0^\infty f_{Y_1,Y_2}(y_1, y_2) \, dy_2$$
  
=  $\frac{1}{\Gamma(\beta_1)\Gamma(\beta_2)} \int_0^\infty y_1^{\beta_1 - 1} y_2^{\beta_1 + \beta_2 - 1} e^{-y_2(y_1 + 1)} \, dy_2$   
=  $\frac{y_1^{\beta_1 - 1}(y_1 + 1)^{-(\beta_2 + \beta_1)}\Gamma(\beta_1 + \beta_2)}{\Gamma(\beta_1)\Gamma(\beta_2)}$   $y_1 > 0,$ 

which is the probability density function of an inverted  $beta(\beta_1, \beta_2)$  random variable.

**APPL verification:** The APPL statements

```
X1 := GammaRV(1, beta1);
X2 := GammaRV(1, beta2);
g := [[x -> 1 / x], [0, infinity]];
Y := Transform(X2, g);
Z := Product(X1, Y);
```

confirm the result.