**Theorem** The log logistic distribution has the scaling property. That is, if  $X \sim \log \log \operatorname{istic}(\lambda, \kappa)$  then Y = cX also has the log logistic distribution.

**Proof** Let the random variable X have the log logistic( $\lambda, \kappa$ ) distribution with probability density function

$$f(x) = \frac{\lambda \kappa (\lambda x)^{\kappa - 1}}{(1 + (\lambda x)^{\kappa})^2} \qquad x > 0.$$

Let c be a positive, real constant. The transformation Y = g(X) = cX is a 1-1 transformation from  $\mathcal{X} = \{x \mid x > 0\}$  to  $\mathcal{Y} = \{y \mid y > 0\}$  with inverse  $X = g^{-1}(Y) = Y/c$  and Jacobian  $\frac{dX}{dx} = 1$ 

$$\frac{dX}{dY} = \frac{1}{d}$$

Therefore, by the transformation technique, the probability density function of Y is

$$f_Y(y) = f_X(g^{-1}(y)) \left| \frac{dx}{dy} \right|$$
  
=  $\frac{\lambda \kappa (\lambda y/c)^{\kappa-1}}{(1 + (\lambda y/c)^{\kappa})^2} \left| \frac{1}{c} \right|$   
=  $\frac{(\lambda/c) \kappa (\lambda y/c)^{\kappa-1}}{(1 + (\lambda y/c)^{\kappa})^2}$   $y > 0,$ 

which is the probability density function of a log logistic  $(\lambda/c, \kappa)$  random variable.

**APPL verification:** The APPL statements

```
assume(c > 0);
X := LogLogisticRV(lambda, kappa);
g := [[x -> c * x], [0, infinity]];
Y := Transform(X, g);
```

yield the probability density function of a log logistic  $(\lambda/c, \kappa)$  random variable.