Transformações de Variáveis Aleatórias

Kempes J.

8 de junho de 2015

Tranformações de Variáveis Aleatórias Contínuas

Definições Iniciais

$$Y = g(X)$$

$$X = g^{-1}(Y)$$

$$F_{Y}(y) = \mathbb{P}\{Y \le y\}$$

$$\mathbb{P}\{g(X) \le y\} = \mathbb{P}\{X \le g^{-1}(Y)\}$$

$$\mathbb{P}\{X \le g^{-1}(Y)\} = F_{X}(g^{-1}(Y))$$

$$f_{Y}(y) = F'_{Y}(y) = \frac{d}{dy}F_{X}(g^{-1}(y)) = f_{X}(g^{-1}(y))\frac{d}{dy}g^{-1}(y)$$

$$f_{Y}(y) = F'_{Y}(y) = \frac{d}{dy}F_{X}(g^{-1}(y))$$

Expressões Finais

$$f_Y(y) = f_X(g^{-1}(y)) \frac{d}{dy} g^{-1}(y)$$

$$f_Y(y) = -f_X(g^{-1}(y)) \frac{d}{dy} g^{-1}(y)$$

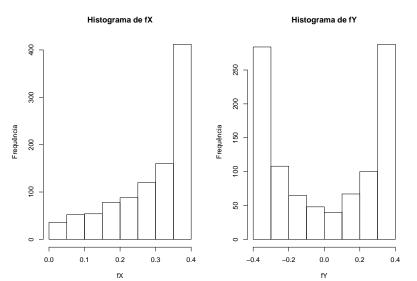
$$f_Y(y) = \left| f_X(g^{-1}(y)) \frac{d}{dy} g^{-1}(y) \right|$$

Resoluções $X \sim N(\mu, \sigma^2)$

$$f_X(x) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

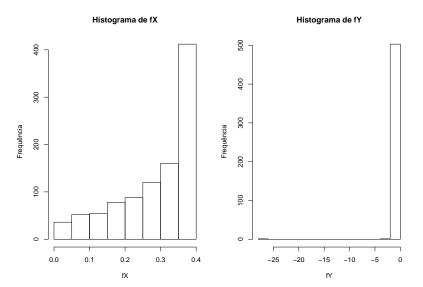
$$X \sim N(\mu, \sigma^2), \ Y = \frac{X-\mu}{\sigma} \ \text{com} \ X \sim N(0, 1)$$
 $Y = g(X) = \frac{X-\mu}{\sigma}, \ \text{para} \ \sigma \neq 0$
 $g^{-1}(Y) = X = Y\sigma + \mu$
 $\frac{d}{dy}g^{-1}(y) = \frac{d}{dy}(y\sigma + \mu) = \sigma$
 $f_Y(y) = f_X(g^{-1}(y))\frac{d}{dy}g^{-1}(y) = f_X(y\sigma + \mu)\sigma$
 $f_Y(y) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{(y\sigma + \mu - \mu)^2}{2\sigma^2}}\sigma = \left| \frac{1}{\sqrt{2\pi}}e^{-\frac{y^2}{2}} \right|$

$$Y = \frac{X - \mu}{\sigma}$$
 (Histogramas)



$$\begin{split} & X \sim \textit{N}(\mu, \sigma^2), \ Y = \frac{\sigma}{X - \mu} \ \text{com} \ X \sim \textit{N}(0, 1) \\ & Y = \textit{g}(X) = \frac{\sigma}{X - \mu}, \ \text{para} \ X \neq \mu \\ & \textit{g}^{-1}(Y) = X = \frac{\sigma}{Y} + \mu \\ & \frac{d}{dy} \textit{g}^{-1}(y) = \frac{d}{dy} (\frac{\sigma}{y} + \mu) = -\sigma y^{-2} \\ & f_Y(y) = f_X(\textit{g}^{-1}(y)) \frac{d}{dy} \textit{g}^{-1}(y) = f_X(\frac{\sigma}{y} + \mu)(-\sigma y^{-2}) \\ & f_Y(y) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(\frac{\sigma}{y} + \mu - \mu)^2}{2\sigma^2}} (-\sigma y^{-2}) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2}(\frac{1}{y})^2} (-\sigma y^{-2}) \\ & f_Y(y) = \ \left| -\frac{1}{y^2 \sqrt{2\pi}} e^{-\frac{1}{2y^2}} \right| \end{split}$$

$$Y = \frac{\sigma}{X - \mu}$$
 (Histogramas)



$$X \sim N(\mu, \sigma^2), Y = \left(\frac{X - \mu}{\sigma}\right)^3 \text{ com } X \sim N(0, 1)$$

$$Y = g(X) = \left(\frac{X - \mu}{\sigma}\right)^3 \text{ para } \sigma \neq 0$$

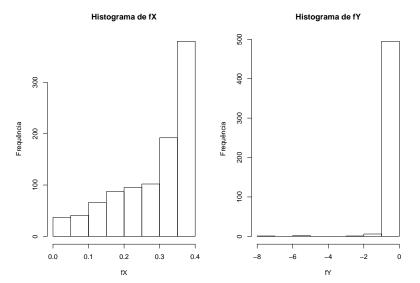
$$g^{-1}(Y) = X = \sigma Y^{\frac{1}{3}} + \mu$$

$$\frac{d}{dy}g^{-1}(y) = \frac{d}{dy}(\sigma y^{\frac{1}{3}} + \mu) = \frac{1}{3}\sigma y^{-\frac{2}{3}}$$

$$f_Y(y) = f_X(g^{-1}(y))\frac{d}{dy}g^{-1}(y) = f_X(\sigma y^{\frac{1}{3}} + \mu)(\frac{1}{3}\sigma y^{-\frac{2}{3}})$$

$$f_Y(y) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{(\sigma y^{\frac{1}{3}} + \mu - \mu)^2}{2\sigma^2}}(\frac{1}{3}\sigma y^{-\frac{2}{3}}) = y^{-\frac{2}{3}}\frac{1}{3\sqrt{2\pi}}e^{-\frac{1}{2}y^{\frac{2}{3}}}$$

$$Y = (\frac{X-\mu}{\sigma})^3$$
 (Histogramas)



$X \sim \varepsilon(1)$, transformar em $Y = X^p$, para $p \neq 0$

$$f_X(x) = \lambda e^{-\lambda x}$$

$$Y = g(X) = X^p$$

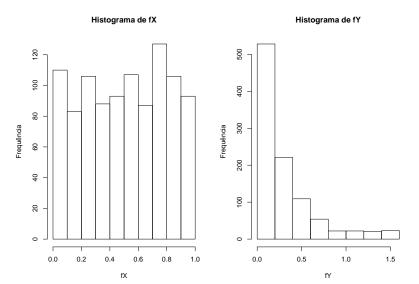
$$g^{-1}(Y) = X = Y^{\frac{1}{p}}$$

$$\frac{d}{dy}g^{-1}(y) = \frac{d}{dy}y^{\frac{1}{p}} = \frac{1}{p}y^{\frac{1-p}{p}}$$

$$f_Y(y) = -f_X(g^{-1}(y))\frac{d}{dy}g^{-1}(y) = -f_X(y^{\frac{1}{p}})\frac{1}{p}y^{\frac{1-p}{p}}$$

$$f_Y(y) = -(\lambda e^{-\lambda y^{\frac{1}{p}}})\frac{1}{p}y^{\frac{1-p}{p}}$$

$Y = X^p$ (Histogramas)



$X \sim U(0,1)$, transformar em Y = aX + b

$$f_X(x) = \frac{1}{b-a} \mathbb{I}_{[a,b]}(x)$$

$$Y = g(X) = aX + b$$

$$g^{-1}(Y) = X = \frac{a}{Y} - b$$

$$\frac{d}{dy} g^{-1}(y) = \frac{d}{dy} (\frac{a}{y} - b) = -ay^{-2}$$

$$f_Y(y) = f_X(g^{-1}(y)) \frac{d}{dy} g^{-1}(y) = f_X(\frac{a}{y} - b)(-ay^{-2})$$

$$f_Y(y) = \frac{1}{b-a} (-ay^{-2}) = -\frac{a}{v^2(b-a)}$$

Y = aX + b(Histogramas)

