

What about PDFs?

Theorem: If X is continuous and $Y = g(X)$

and if

① g is invertible

② g^{-1} is differentiable

then

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

pf. Case 1: g increasing

Our previous CDF theorem said

$$F_Y(y) = F_X(g^{-1}(y))$$

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

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

\downarrow g increasing
 g^{-1} increasing
 so $\frac{dg^{-1}}{dy} > 0$

Case 2: g decreasing

Case 2: y decreasing

By our previous CDF theorem

$$F_Y(y) = 1 - F_X(g^{-1}(y))$$

and so

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

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

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

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

$\rightarrow g$ decreasing

$\rightarrow g^{-1}$ decreasing

$$\rightarrow \frac{dg^{-1}}{dy} < 0$$

$$-(-5) = |5|$$

Ex. Let $X \sim \text{Gamma}(k, \lambda)$

$$\checkmark f_X(x) = \frac{\lambda e^{-\lambda x} (\lambda x)^{k-1}}{\Gamma(k)} \quad \text{for } x > 0$$

$$\text{Let } Y = \frac{1}{X} \text{ i.e. } y = g(x) = \frac{1}{x} \Leftrightarrow y = 1/x \Rightarrow x = \frac{1}{y} = \checkmark g^{-1}(y)$$

$$\text{So } \frac{dg^{-1}}{dy} = -\frac{1}{y^2}$$

Our PDF theorem says

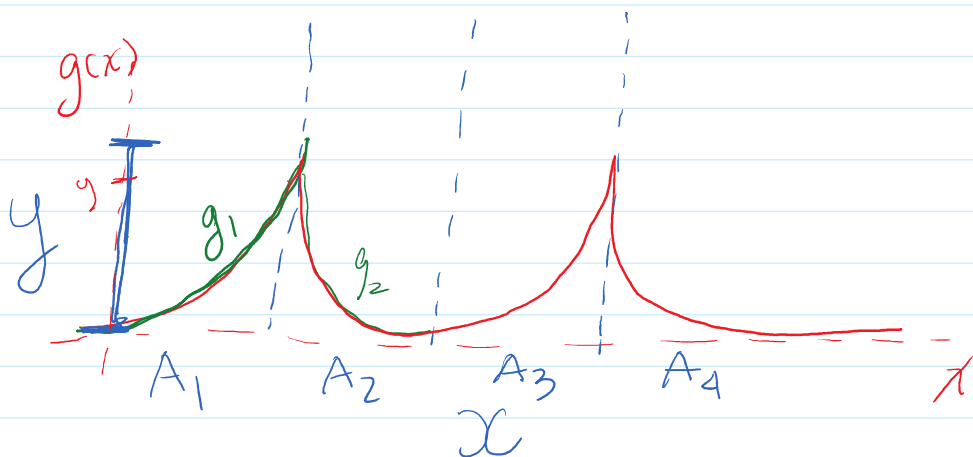
$$f_Y(y) = f_X(g^{-1}(y)) \left| \frac{dg^{-1}}{dy} \right|$$
$$= f_X\left(\frac{1}{y}\right) \left| -\frac{1}{y^2} \right|$$

$$= \frac{\lambda e^{-\frac{\lambda}{y}} \left(\frac{\lambda}{y}\right)^{k-1}}{\Gamma(k)} \frac{1}{y^2} \quad \text{for } y > 0$$

↑ called the Inverse Gamma dist.

What about non-invertible g ?

Theorem: Let X is a continuous RV w/ support \mathcal{X} and for $i=1, \dots, K$ let A_i partition \mathcal{X}



Let g_i to be g restricted to A_i .

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If ① my prev. theorem applies on each part of the partition separately

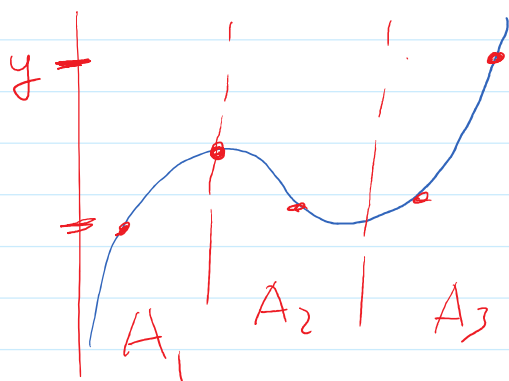
$$\left[\begin{array}{l} g_i \text{ invertible on } A_i \\ g_i^{-1} \text{ is differentiable} \end{array} \right]$$

② The image of A_i under each g_i is the same.

[all g_i have same range Y]

then

$$f_Y(y) = \sum_{i=1}^K f_X(g_i^{-1}(y)) \left| \frac{dg_i^{-1}}{dy} \right| \quad \text{for } y \in Y$$



Ex. Chi-squared Distribution

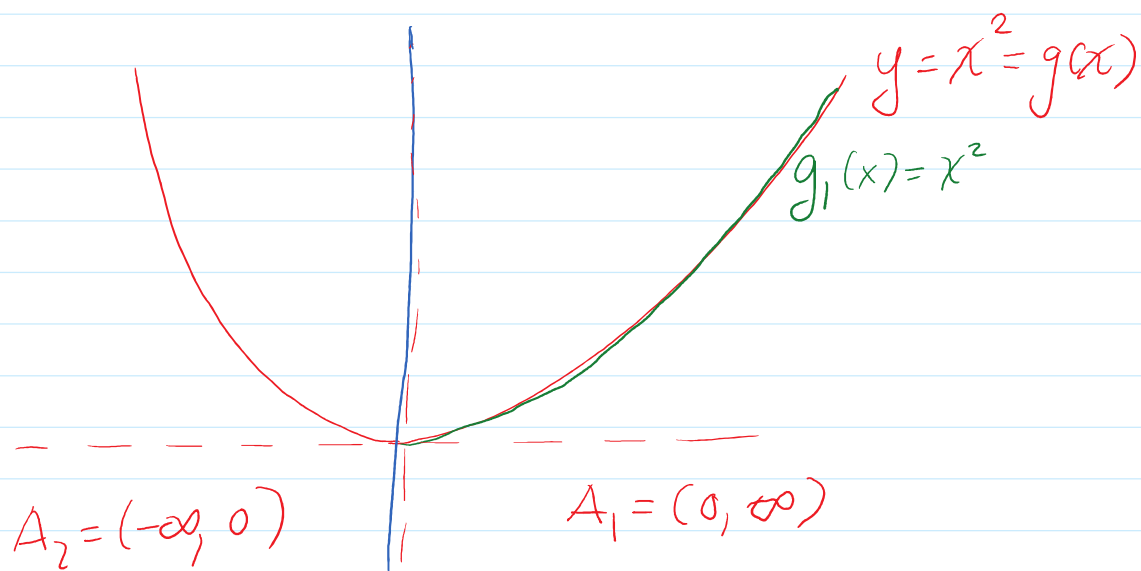
Ex, Chi-squared Distribution

If $X \sim N(0,1)$ and $Y = X^2$

then we say Y has a chi-sq. dist.
w/ one degree of freedom,

denoted $Y \sim \chi^2(1)$

What is the PDF of a Chi-sq?



$$A_1 = (0, \infty), g_1(x) = x^2, g_1^{-1}(y) = \sqrt{y}; \frac{dg_1^{-1}}{dy} = \frac{1}{2\sqrt{y}}$$

$$A_2 = (-\infty, 0), g_2(x) = x^2, g_2^{-1}(y) = -\sqrt{y}; \frac{dg_2^{-1}}{dy} = \frac{-1}{2\sqrt{y}}$$

$$f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2} \text{ for } x \in \mathbb{R}$$

$$f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2} \quad \text{for } x \in \mathbb{R}$$

$$f_Y(y) = f_X(g_1^{-1}(y)) \left| \frac{dg_1^{-1}}{dy} \right| + f_X(g_2^{-1}(y)) \left| \frac{dg_2^{-1}}{dy} \right|$$

$$= f_X(\sqrt{y}) \left| \frac{1}{2\sqrt{y}} \right| + f_X(-\sqrt{y}) \left| \frac{-1}{2\sqrt{y}} \right|$$

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

$$= \frac{1}{\sqrt{2\pi}} \frac{1}{2\sqrt{y}} (e^{-y/2} + e^{-y/2})$$

$$f_Y(y) = \frac{1}{\sqrt{2\pi}} \frac{1}{\sqrt{y}} e^{-y/2} \quad \text{for } y > 0$$

↑ PDF of a $\chi^2(1)$

Probability Integral Transformation

If X is a continuous RV w/ CDF F_X

$$F_X(X) \sim U(0,1).$$

pf: Assume F_X is | CDF of a $U(0,1)$

pf. Assume F_X is strictly increasing then F_X^{-1} exists.

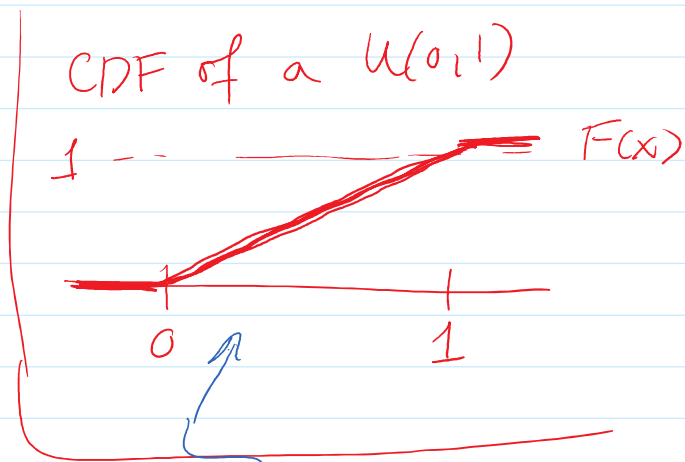
Our CDF theorem says

$$Y = F_X(X) = g(X)$$

then for $0 < y < 1$

$$F_Y(y) = F_X(g^{-1}(y)) = F_X(F_X^{-1}(y)) = y$$

So the CDF of Y is the CDF of a $U(0,1)$ and so $Y \sim U(0,1)$.



Converse:

$$g(X) \sim U(0,1) \iff g = F_X$$

Know how to generate $U \sim U(0,1)$

Want: generate some RV w/ cdf F_X

$$\text{let } Z = F_X^{-1}(U)$$

$$\text{then } F_Z(z) = P(Z \leq z) = P(F_X^{-1}(U) \leq z)$$

CDF of $U(0,1)$ is $F(x)=x$

then $F_Z(z) = P(Z \leq z) = P(F_X^{-1}(U) \leq z)$
 $= P(U \leq F_X(z))$
 $= F_X(z)$

So Z follows dist w/ CDF F_X .

Algorithm: Want $Z \sim F_X$

① $U \sim U(0,1)$

② $Z = F_X^{-1}(U)$

↑ this has correct dist.

Ex. Want $X \sim \text{Exp}(1)$

CDF of $\text{Exp}(1)$ is $F(x) = 1 - e^{-x}$

$$y = 1 - e^{-x}$$

$$1 - y = e^{-x}$$

$$\log(1 - y) = -x \Rightarrow$$

$$x = -\log(1 - y) = F_X^{-1}(y)$$

Bivariate RVs

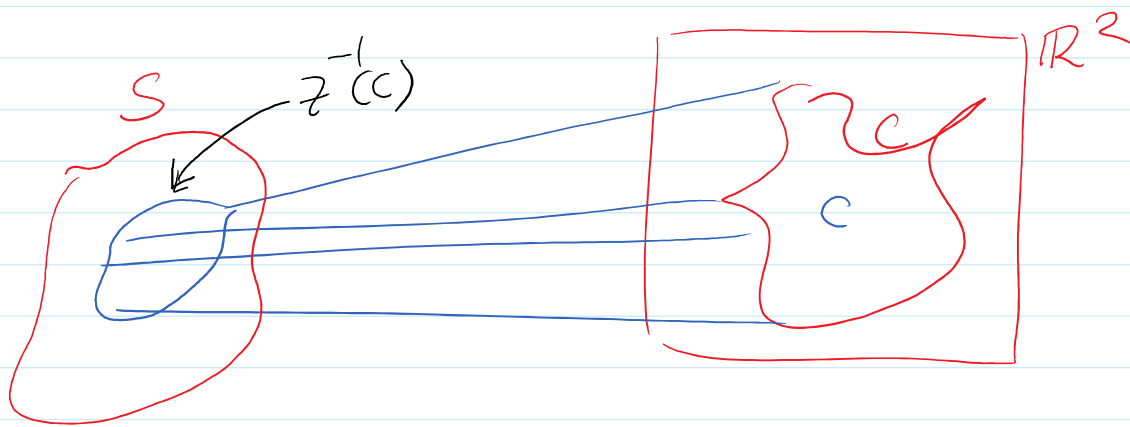
If $X: S \rightarrow \mathbb{R}$ and $Y: S \rightarrow \mathbb{R}$

then $Z = (X, Y)$ is called a bivariate RV

so $Z: S \rightarrow \mathbb{R}^2$ so that $Z(s) = (X(s), Y(s))$

Say: $P(Z \in C) = P((X, Y) \in C)$ $\uparrow C \subset \mathbb{R}^2$

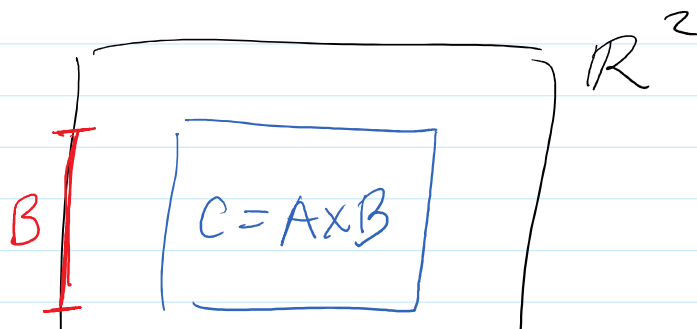
$$= P(Z^{-1}(C))$$



Often, $C = A \times B$ where $A, B \subset \mathbb{R}$

Write

$P((X, Y) \in C)$
or be lazy



$$P(X \in A, Y \in B)$$

↑ "and"

