

## Lecture 15

Theorem: If  $X$  is a cts RV and

$$Y = g(X) \text{ and}$$

- ①  $g$  is invertible
- ②  $g^{-1}$  is differentiable

then

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

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pf. Case 1:  $g$  increasing  
prev. CDF thm said

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

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

Case 2:  $g$  decreasing

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

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


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

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

$$Y = 1/X$$

$$g(x) = 1/x = y \Rightarrow x = 1/y = g^{-1}(y)$$

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

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

$$= f_x\left(\frac{1}{y}\right) \left| -\frac{1}{y^2} \right|$$

$$f_Y(y) = \frac{\lambda e^{-\lambda/y} (\lambda/y)^{k-1}}{\Gamma(k)} \frac{1}{y^2}, \quad y > 0$$

↑ Inverse Gamma dist

What about non-invertible  $g$ ?

Theorem: Let  $X$  be cts w/ support  $\mathcal{X}$   
and for  $i=1, \dots, K$  let  $A_i$  partition  $\mathcal{X}$



Let  $g_i$  be  $g$  restricted to  $A_i$ . Then if

① prev. thm applies to each  $g_i$

② The image of  $A_i$  under  $g_i$  is the same for all  $i$

then

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

for  $y \in Y$

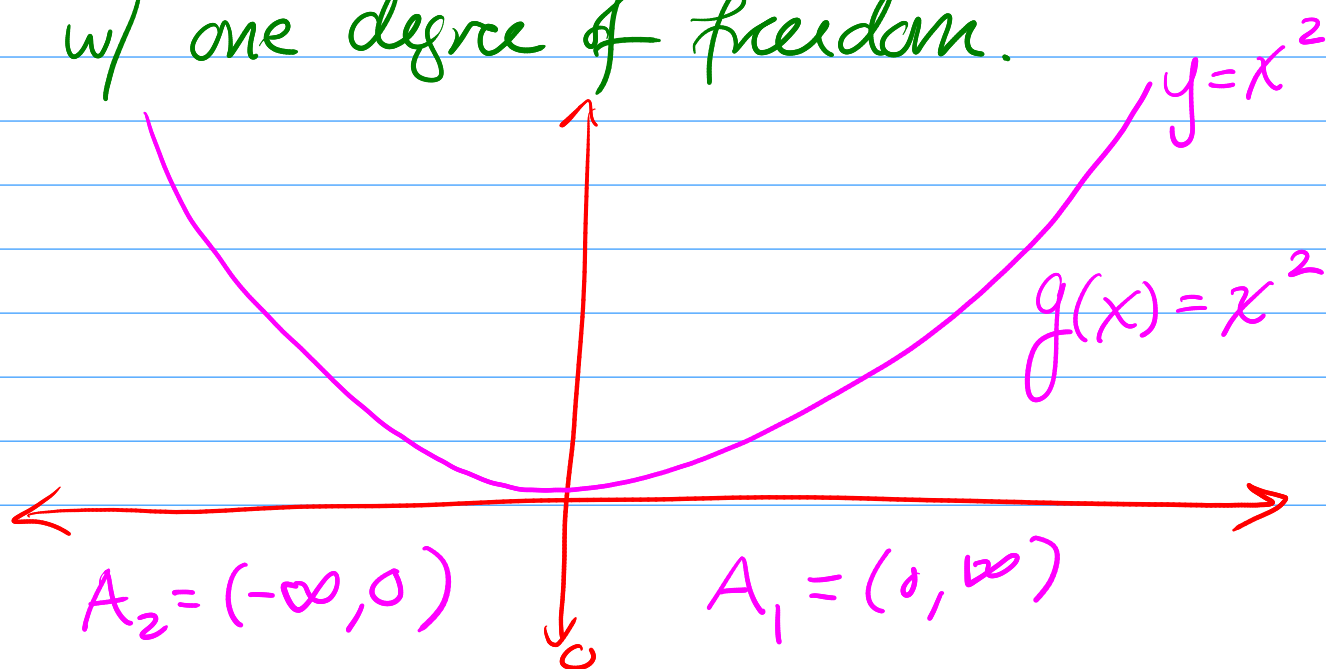
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Ex. Chi-Squared Dist.

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

then  $Y$  has a Chi-Sq. dist.

w/ one degree of freedom.



$$A_1 = (0, \infty), \quad g_1(x) = x^2$$

$$g_1^{-1}(y) = \sqrt{y} \Rightarrow \frac{dg_1^{-1}}{dy} = \frac{1}{2\sqrt{y}}$$

$$A_2 = (-\infty, 0), \quad g_2(x) = x^2$$

$$g_2^{-1}(x) = -\sqrt{y} \Rightarrow \frac{dg_2^{-1}}{dy} = \frac{-1}{2\sqrt{y}}$$

$$f_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}x^2\right)$$

$$\begin{aligned} f_Y(y) &= f_X(g_1^{-1}(y)) \left| \frac{dg_1^{-1}}{dy} \right| + f_X(g_2^{-1}(y)) \left| \frac{dg_2^{-1}}{dy} \right| \\ &= \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}(\sqrt{y})^2\right) \left| \frac{1}{2\sqrt{y}} \right| + \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}(-\sqrt{y})^2\right) \left| \frac{-1}{2\sqrt{y}} \right| \end{aligned}$$

$$= \frac{2}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}y\right) \frac{1}{2\sqrt{y}}, \quad y > 0$$

$$= \frac{1}{\sqrt{2\pi}} \frac{1}{\sqrt{y}} \exp\left(-\frac{1}{2}y\right), \quad y > 0$$

## Theorem: Prob. Integral Transf.

If  $X$  is cts w/ CDF  $F_X$  then

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

pf.  $F_X = g$  is strictly increasing,

So  $F_X^{-1}$  exists,  
 $g^{-1} = F_X^{-1}$

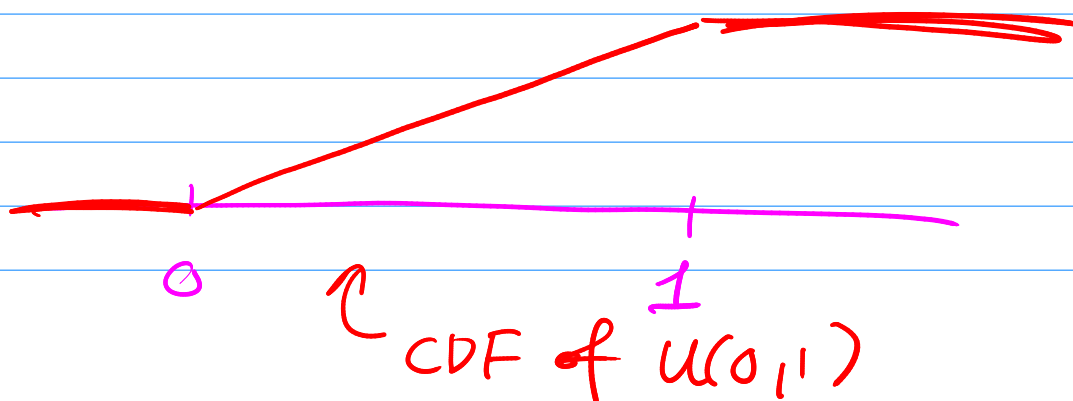
Our CDF theorem says if

$Y = g(X)$ ,  $g$  inc, then  $F_Y(y) = F_X(g^{-1}(y))$

So

$$F_Y(y) = F_X(F_X^{-1}(y)) = y \leftarrow \text{CDF of } U(0,1)$$

for  $0 < y < 1$



Know: how to generate  $U(0,1)$

Want: generate RV w/ CDF  $F_X$

let  $U \sim U(0,1)$

and  $Z = F_X^{-1}(U)$

then  $Z \sim F_X$ .

This works?

$$F_Z(z) = P(Z \leq z)$$

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

$$= P(U \leq F_X(z))$$

$$= F_U(F_X(z))$$

$$\hookrightarrow = F_X(z)$$

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

CDF of  $\text{Exp}(1)$  is  $F_X(x) = 1 - e^{-x}, x > 0$

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

## Bivariate RVs

If  $X: S \rightarrow \mathbb{R}$  and  $Y: S \rightarrow \mathbb{R}$   
then  $Z = (X, Y)$  is called a  
bivariate RV.

$$Z: S \rightarrow \mathbb{R}^2 \text{ s.t. } Z(\omega) = (X(\omega), Y(\omega)).$$

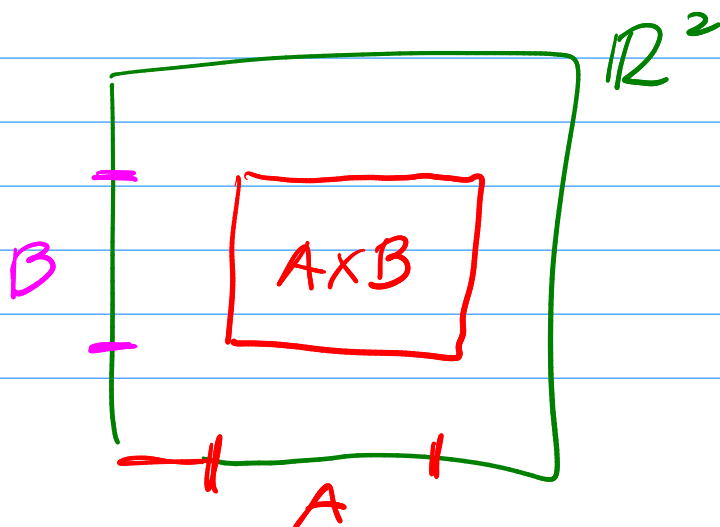
$$\text{Say: } P(Z \in C) \quad C \subset \mathbb{R}^2$$
$$= P((X, Y) \in C)$$



$$\text{Often, } C = A \times B$$

$$P((X, Y) \in C)$$

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





Ex. Flip a coin 3 times.

$$X = \begin{cases} 0, & \text{last flip T} \\ 1, & \text{last flip H} \end{cases}$$

$Y = \# \text{ heads among 3 flips}$

$$Z = (X, Y)$$

$\omega \in S$	$Z(\omega) \in \mathbb{R}^2$
H H H	(1, 3)
H H T	(0, 2)
H T H	(1, 2)
H T T	(0, 1)
T H H	(1, 2)
T H T	(0, 1)
T T H	(1, 1)
T T T	(0, 0)

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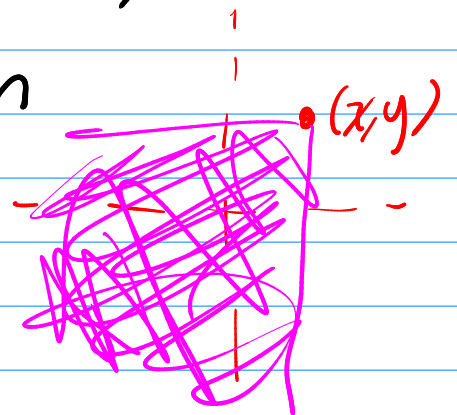
Defn: Bivariate CDF (Joint CDF)

The joint CDF is a function

$$F: \mathbb{R}^2 \rightarrow \mathbb{R}$$

defined for  $(x, y)$  as

$$F(x, y) = P(X \leq x, Y \leq y)$$



# Properties of Joint CDF

①  $F(x, y) \geq 0$

②  $\lim_{x, y \rightarrow \infty} F(x, y) = 1$

③  $\lim_{x \rightarrow -\infty} F(x, y) = 0$

$\lim_{y \rightarrow -\infty} F(x, y) = 0$

④  $F$  is non-decreasing and right-cts in each argument.

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