

This lecture :

- ① Know something about  $X$
- ② What do I know about  $Y = g(X)$ ?

## Discrete RVs (PMFs)

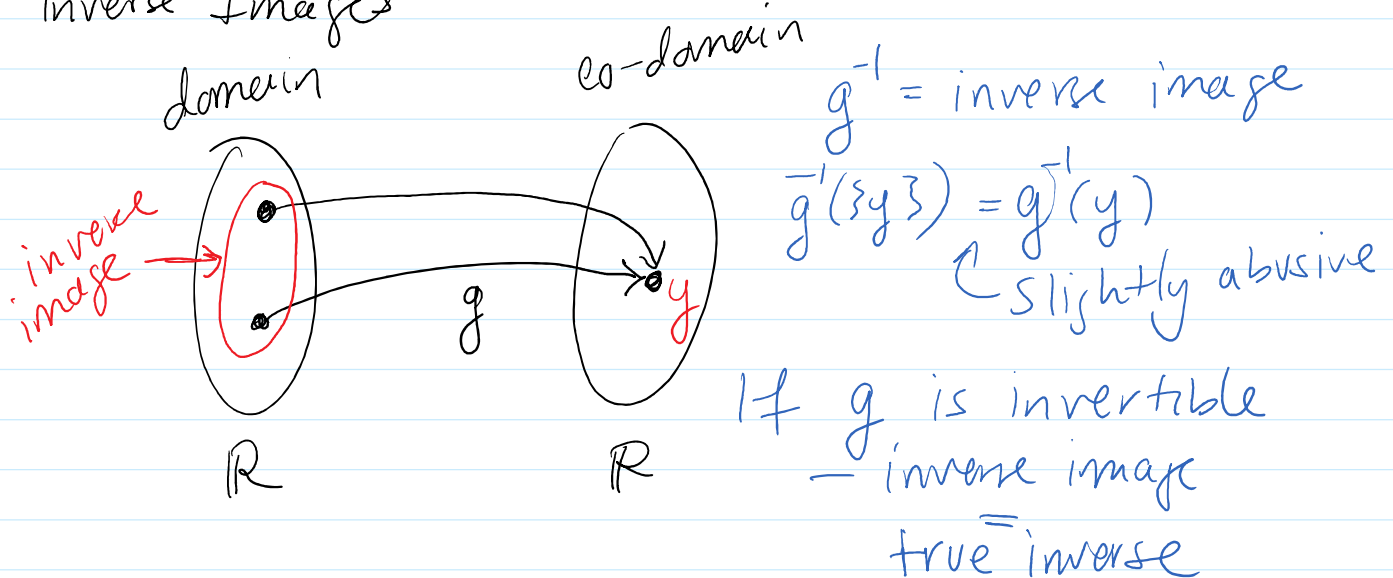
Let  $X$  be discrete and

Know  $f_X$  PMF of  $X$

let  $Y = g(X)$ .

What is  $f_Y$ ? PMF of  $Y$ .

Inverse Images



notice:

$$\begin{aligned}
 \underline{f_Y(y)} &= P(Y=y) = P(g(X)=y) \\
 &\quad \text{if } g \text{ is invertible} \\
 &= P(X = g^{-1}(y)) \\
 &= \underline{f_X(g^{-1}(y))}
 \end{aligned}$$

What if  $g$  not invertible

$$\begin{aligned}
 f_Y(y) &= P(Y=y) = P(g(X)=y) \\
 &= P(X \in \underbrace{g^{-1}(y)}_A)
 \end{aligned}$$

$$= \sum_{x \in g^{-1}(y)} f_X(x)$$

$\rightarrow x : g(x) = y$

$$\begin{aligned}
 P(X \in A) \\
 &= \sum_{x \in A} f_X(x)
 \end{aligned}$$

Theorem: If  $X$  discrete and  $Y=g(X)$

$$f_Y(y) = \sum_{x \in g^{-1}(y)} f_X(x)$$

Ex. Let  $X \sim \text{Bin}(n, p)$

↑ # of H in  $n$  indep coin  
flips each w/ prob- $p$  of H.

let  $Y = n - X \leftarrow$  # tails

$$y = g(x) = n - x \Leftrightarrow x = n - y$$

so  $g^{-1}(y) = n - y$

$$f_Y(y) = \sum_{x \in g^{-1}(y)} f_X(x) = \sum_{x=n-y} f_X(x)$$

$$= f_X(n-y)$$

$$f_X(x) = \binom{n}{x} p^x (1-p)^{n-x} \text{ for } x=0, 1, \dots, n$$

$$= \binom{n}{n-y} p^{n-y} (1-p)^{\cancel{n-(n-y)} \rightarrow y} \text{ for } y=0, 1, \dots, n$$

$q = 1-p$

$$\binom{n}{y} = \frac{n!}{y!(n-y)!} = \binom{n}{n-y}$$

$$f_Y(y) = \binom{n}{y} q^y (1-q)^{n-y} \text{ for } y=0, \dots, n$$

Bin( $n, q$ )

so  $\boxed{Y \sim \text{Bin}(n, 1-p)}$

So  $Y \sim \text{Bin}(n, 1-p)$

lets look at continuous RVs (CDFs).

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

① if  $g$  is increasing then

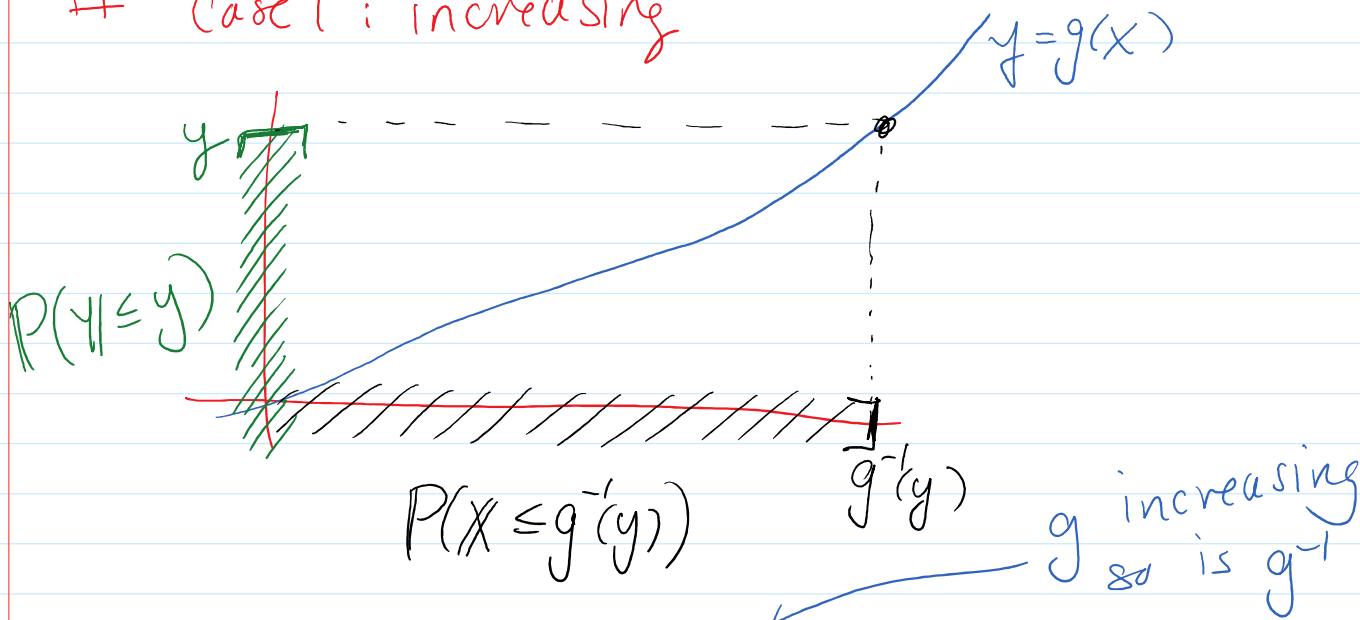
$$F_Y(y) = F_X(g^{-1}(y))$$

invertible

② if  $g$  is decreasing then

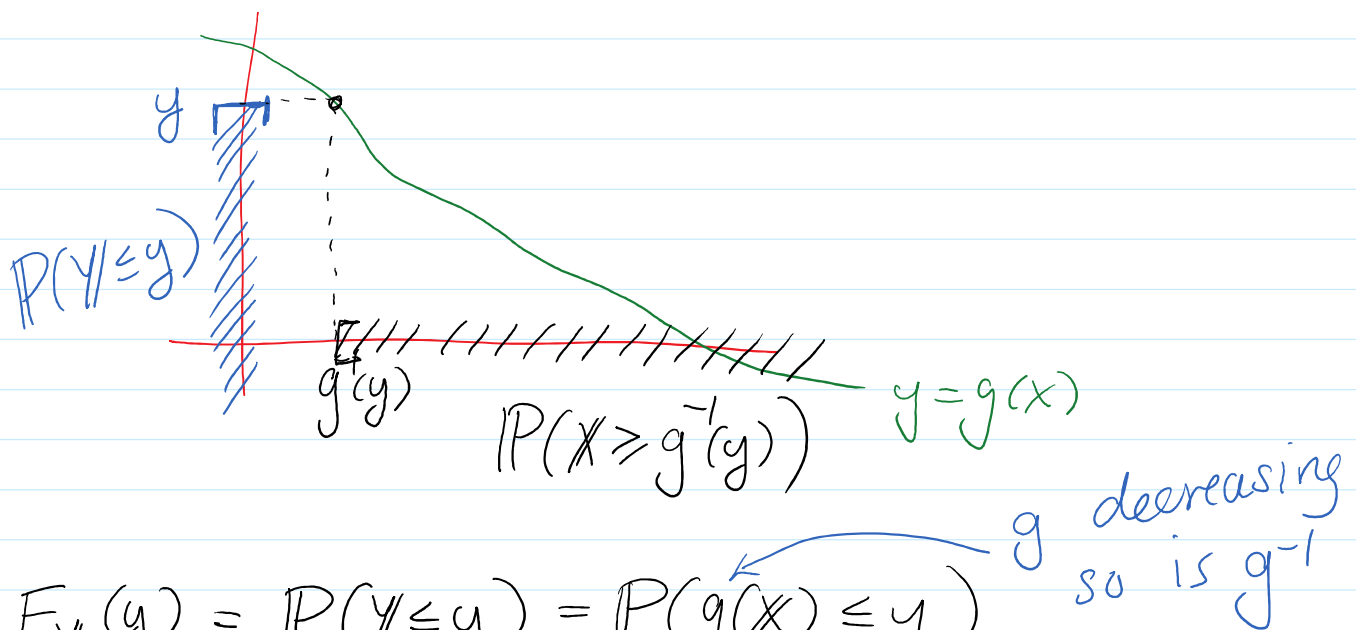
$$F_Y(y) = 1 - F_X(g^{-1}(y))$$

Prf. Case 1: increasing



$$\begin{aligned}
 F_Y(y) &= P(Y \leq y) = P(g(X) \leq y) \\
 &= P(X \leq g^{-1}(y)) \\
 &= F_X(g^{-1}(y))
 \end{aligned}$$

part 2:  $g$  decreasing



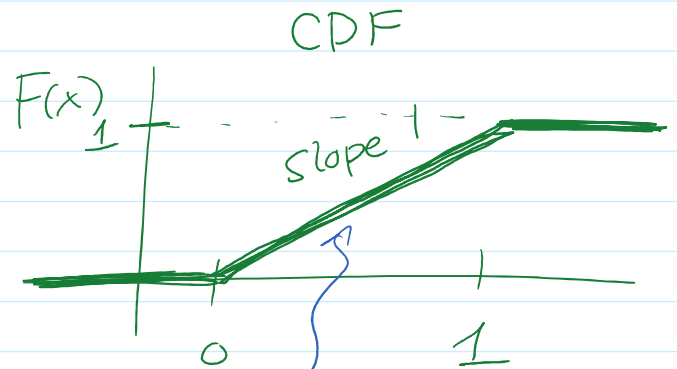
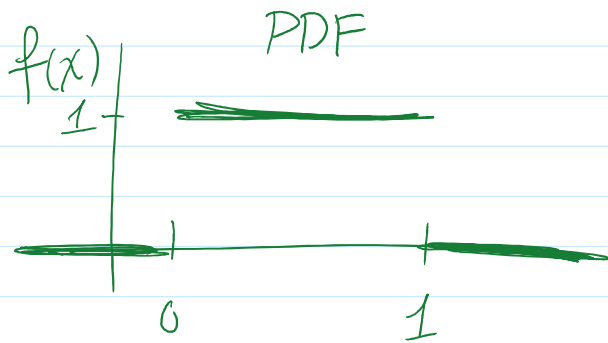
$$\begin{aligned}
 F_Y(y) &= P(Y \leq y) = P(g(X) \leq y) \\
 &= P(X \geq g^{-1}(y)) \\
 &= 1 - P(X \leq g^{-1}(y)) \\
 &= 1 - F_X(g^{-1}(y))
 \end{aligned}$$

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

$f_X(x)$

PDF

CDF



Let  $Y = -\log X$

$$y = g(x) = -\log(x)$$

$$\Leftrightarrow -y = \log x$$

$$\Leftrightarrow e^{-y} = x = g^{-1}(y)$$

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

(from above)

$$= 1 - F_X(e^{-y})$$

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

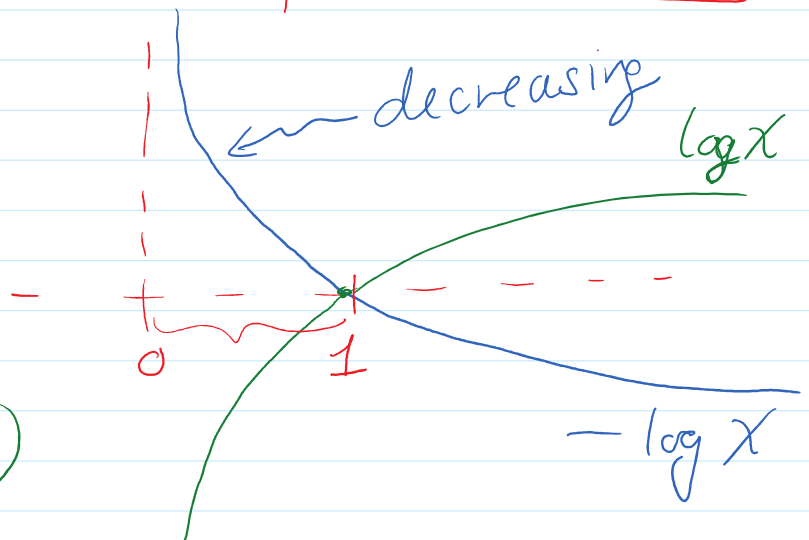
CDF of  
 $\text{Exp}(1)$

CDF for  $\text{Exp}(\lambda)$   
 $1 - e^{-\lambda x}$

So  $Y \sim \text{Exp}(1)$

$$F_X(x) = x$$

for  $0 < x < 1$



$$\left. \begin{array}{l} y > 0 \\ \Leftrightarrow -y < 0 \\ e^{-y} < e^0 = 1 \end{array} \right\} 0 < e^{-y} < 1$$

What about PDFs?

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

and

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

then

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

pf. Case 1:  $g$  increasing  
our prev. result says  $\rightarrow g'$  increasing so  $\frac{dg^{-1}}{dy} > 0$

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

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

Case 2:  $g$  decreasing  $\rightarrow g'$  decreasing so  $\frac{dg^{-1}}{dy} < 0$   
prev. result:  $F_Y(y) = 1 - F_X(g^{-1}(y))$

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

$$\left( \frac{f_X(g^{-1}(y)) \left| \frac{dy}{dx} \right|}{|-5| = -(-5)} \right)$$

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

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

Let  $Y = 1/X$

$$\text{so } g(x) = 1/x = y \Leftrightarrow x = g^{-1}(y) = 1/y$$

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

so

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

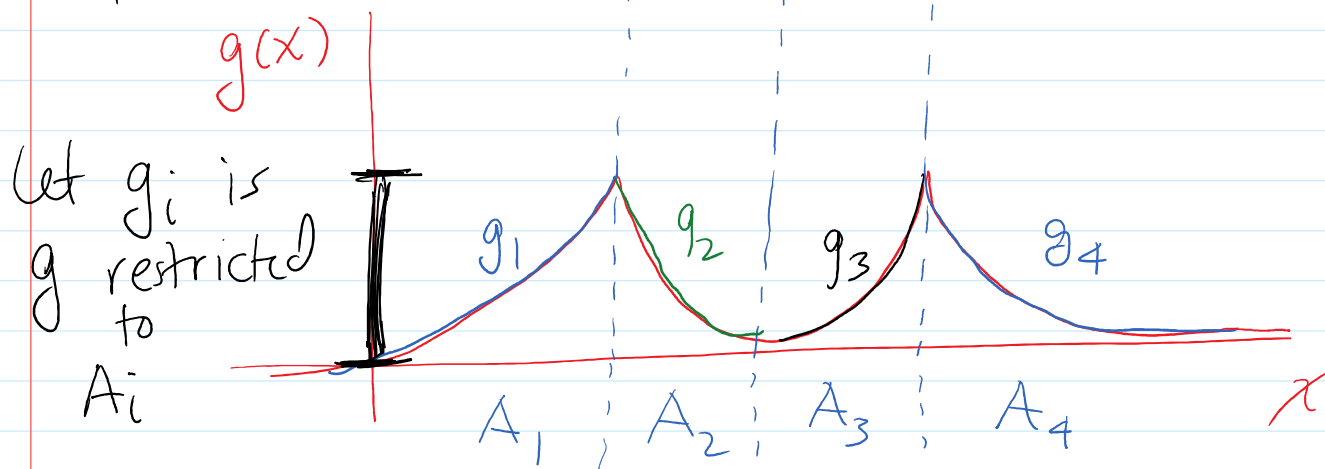
$$= \frac{\lambda e^{-\lambda/y} (\lambda/y)^{a-1}}{\Gamma(a)} \frac{1}{y^2} \quad \text{for } y > 0$$

↑ called Inverse Gamma distribution

What about non-invertible  $g$ ?



Theorem: Let  $X$  be continuous w/ support  $\mathcal{X}$   
 and for  $i=1, \dots, K$  let  $A_1, \dots, A_K$   
 partition  $\mathcal{X}$



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

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

and

② the image of  $A_i$  under  $g_i$  is the same  $\forall i=1, \dots, K$

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