

$$\boxed{Y = g(X)} \quad g: \mathbb{R} \rightarrow \mathbb{R}$$

really mean: $Y = g \circ X$

$$X: S \rightarrow \mathbb{R}$$

$$Y: S' \rightarrow \mathbb{R}$$

Formalisms:

$$\xrightarrow{ACR} P(Y \in A) = P((g \circ X) \in A) = P((g \circ X)^{-1}(A)) \\ = P(X^{-1}(g^{-1}(A)))$$

$$\boxed{P(Y \in A) = P(g(X) \in A) = P(X \in g^{-1}(A))}$$

This lecture: Know something about X
 what can I infer about $\underbrace{g(X)}_Y$?

How can I get the pmf of Y from the pmf of X ?
 → discrete

① discrete X

② Know $f_X \leftarrow$ pmf of X

(2) Know f_X pmf of X

(3) want: f_Y pmf of Y

notice:

$$f_Y(y) = P(Y=y) = P(g(X)=y)$$

$$= P(X \in g^{-1}(y))$$

g^{-1} is the
inverse
image

aside:

$$P(X \in A) = \sum_{x \in A} f_X(x)$$

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

Theorem:

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

$$P(Y=y) = \sum_{\substack{\text{all } x \text{ where} \\ g(x)=y}} P(X=x)$$

intuition.

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

let $Y = n - X$ alt. $Y = g(X)$

= # failures where $g(x) = n - x$

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

$$x = n-y \text{ so } \underline{g^{-1}(y) = n-y}$$

} aside : get g^{-1}

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

$\sum_{x=n-y} f_x(x)$

a bit over kill

Bin. pmf

$$f_x(x) = \binom{n}{x} p^x (1-p)^{n-x}$$

$$= \binom{n}{n-y} p^{n-y} (1-p)^{n-(n-y)}$$

$$= \binom{n}{y} p^{n-y} (1-p)^y$$

$$\begin{aligned} \binom{n}{n-y} &= \frac{n!}{(n-y)! (n-(n-y))!} \\ &= \frac{n}{(n-y)y!} = \binom{n}{y} \end{aligned}$$

$$q = 1-p$$

$$= \boxed{\binom{n}{y} q^y (1-q)^{n-y}}$$

pmf of a $\text{Bin}(n, q)$

$$f_y(y) =$$

↑

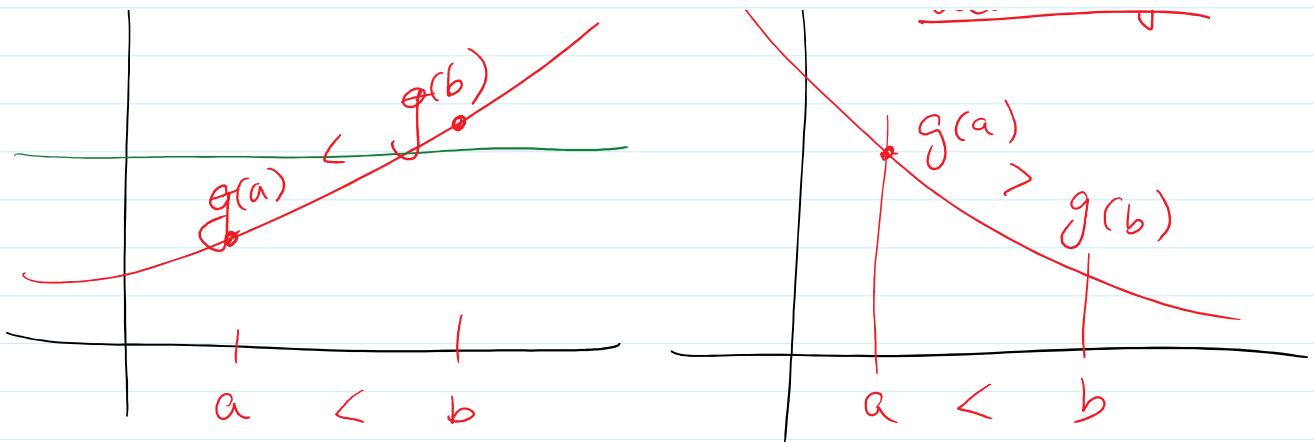
So $\boxed{Y \sim \text{Bin}(n, 1-p)}$

Defn: a function is monotone if it is either increasing or decreasing

increasing

a(b)

decreasing



Theorem: If $Y = g(X)$ then the CDF of Y is
 X is continuous then

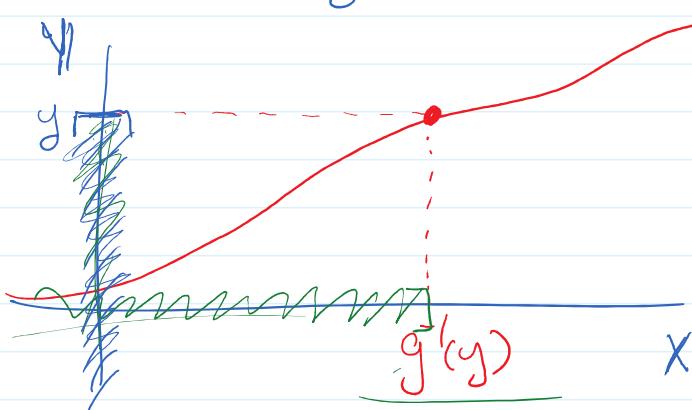
① If g is increasing:

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

② If g is decreasing:

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

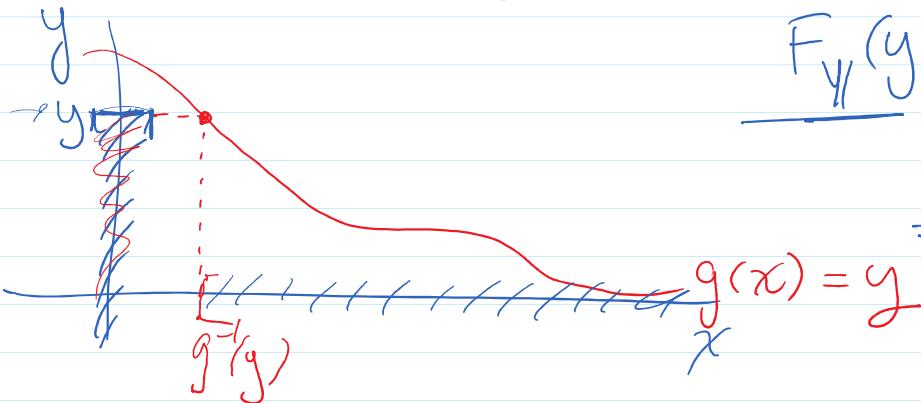
Case 1 pf. g increasing



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

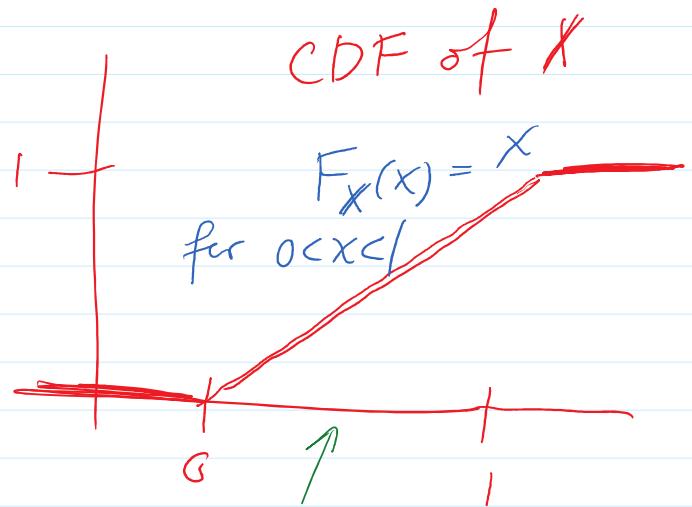
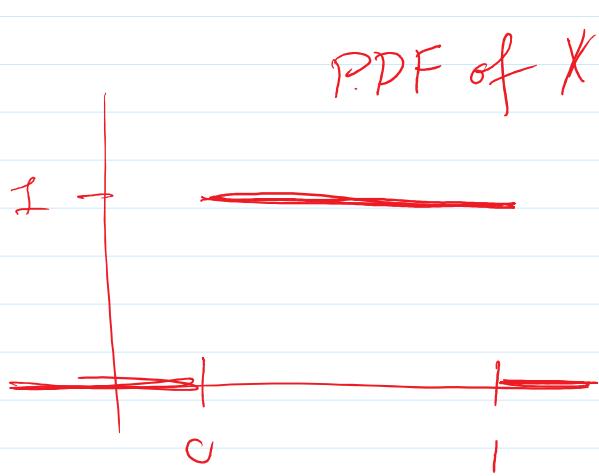
$$= \underline{1_X(J(y))}$$

Case 2: decreasing



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

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

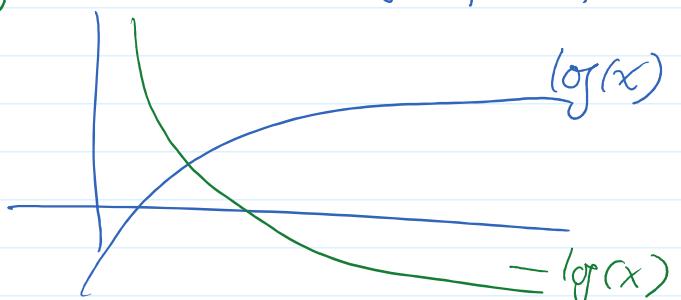


Let $Y = -\log(X)$. $g(x) = -\underline{\log(x)}$.

decreasing funct α

What's the CDF of Y ?

$$\begin{aligned} F_Y(y) &= 1 - F_X(g^{-1}(y)) \\ &= 1 - F_X(e^{-y}) \\ &= 1 - e^{-y} \quad 0 < e^{-y} < 1 \end{aligned}$$



aside: $y = -\log(x)$
 $-y = \underline{\log(x)}$

$$0 < e^x < 1 \quad \text{for } y > 0$$

$\rightarrow 1 - e^{-y}$

$$-y = \begin{cases} \ln(x) \\ -y = g^{-1}(y) \end{cases}$$

Shared $F_Y(y) = 1 - e^{-y}$ Notice: CDF of
an $\text{Exp}(1)$
then $Y \sim \text{Exp}(1)$.

What about pdf?

$$Y = g(X)$$

- ① X continuous
- ② g continuous, monotone
- ③ g^{-1} is continuously differentiable

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

for $y \in Y$

↑
Support of Y

Pf. Case 1: g increasing

$$\begin{aligned} f_Y(y) &= \frac{d}{dy} F_Y(y) = \frac{d}{dy} [F_X(g^{-1}(y))] \\ &= f_X(g^{-1}(y)) \left| \frac{dg^{-1}}{dy} \right|_{>0} \end{aligned}$$

Case 2: g decreasing

$1 \leftarrow > 0$

chain rule.

$$\begin{aligned} f_Y(y) &= \frac{d}{dy} F_Y(y) = \frac{d}{dy} [1 - F_X(g^{-1}(y))] \\ &= -f_X(g^{-1}(y)) \underbrace{\frac{dg^{-1}}{dy}}_{< 0} \\ &= f_X(g^{-1}(y)) \left| \frac{dg^{-1}}{dy} \right| \end{aligned}$$

Ex. $X \sim \text{Gamma}(n, \beta)$

$\uparrow n$ an integer

$$P(n) = (n-1)!$$

$$\text{So } f_X(x) = \frac{1}{(n-1)! \beta^n} x^{n-1 - \frac{x}{\beta}} e^{-\frac{x}{\beta}} \text{ for } x > 0$$

Consider $Y = \frac{1}{X}$ equiv. $g(x) = \frac{1}{x}$

- g monotone
 g^{-1} cts. differentiable

$$y = \frac{1}{x}$$

$$\text{then } x = \frac{1}{y} = g^{-1}(y)$$

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

Theorem:

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

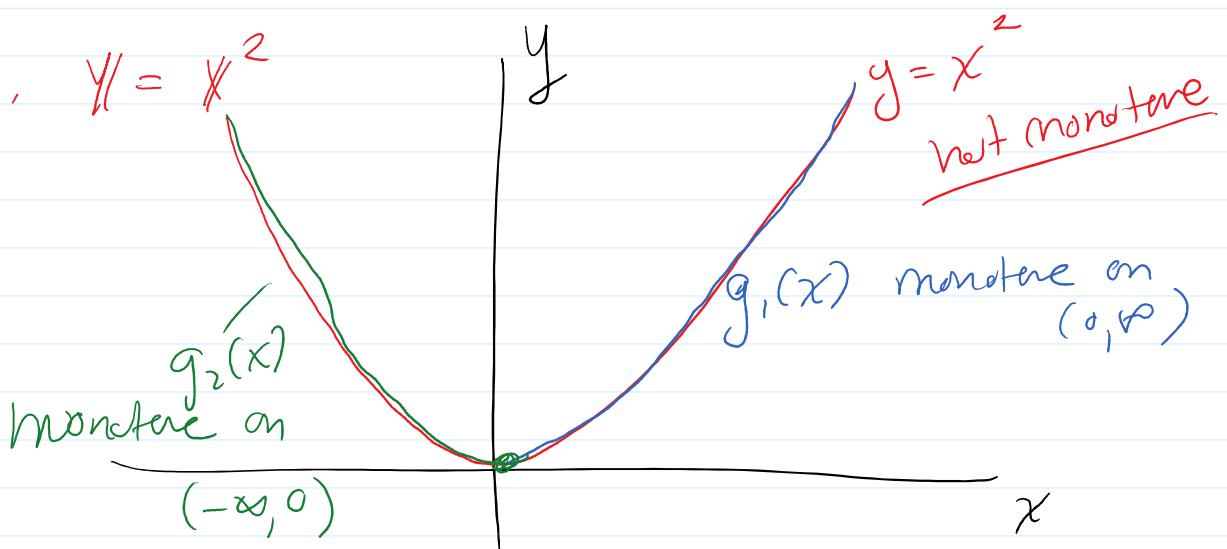
$$\begin{aligned}
 f_Y(y) &= f_X(g^{-1}(y)) \left| \frac{dy}{dx} \right| \\
 &= f_X\left(\frac{1}{y}\right) \left| -\frac{1}{y^2} \right| \\
 &= \frac{1}{(n-1)! \beta^n} \left(\frac{1}{y}\right)^{n-1} e^{-\frac{1}{y\beta}} \frac{1}{y^2} \\
 &= \begin{cases} \frac{1}{(n-1)! \beta^n} \left(\frac{1}{y}\right)^{n+1} e^{-\frac{1}{y\beta}} & \text{for } y > 0 \end{cases} \\
 &\quad \text{if } x > 0 \text{ then } \frac{1}{x} > 0
 \end{aligned}$$

$Y \sim \text{InverseGamma}(n, \beta)$.

What if g isn't monotone?

OK, as long as we can segment g into monotone chunks.

Ex. $Y = X^2$



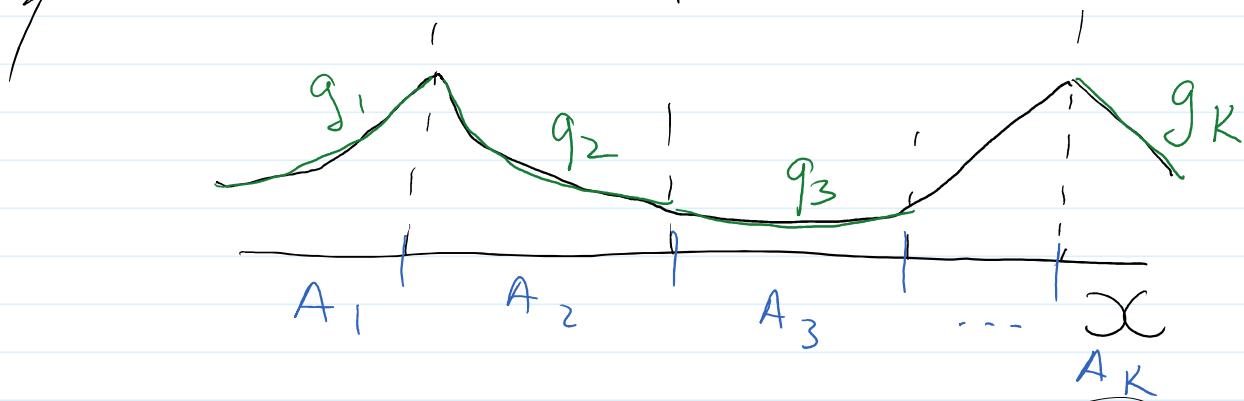
Theorem: Piecewise Monotone Transformation

X continuous has support \mathcal{X}

Let A_1, A_2, \dots, A_K partition \mathcal{X}

let g_i be the function g on A_i for $i=1, \dots, K$

① If each of these g_i satisfy the conditions of the previous theorem on A_i



② Image of A_i under g_i is Y

then if $Y = g(X)$

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

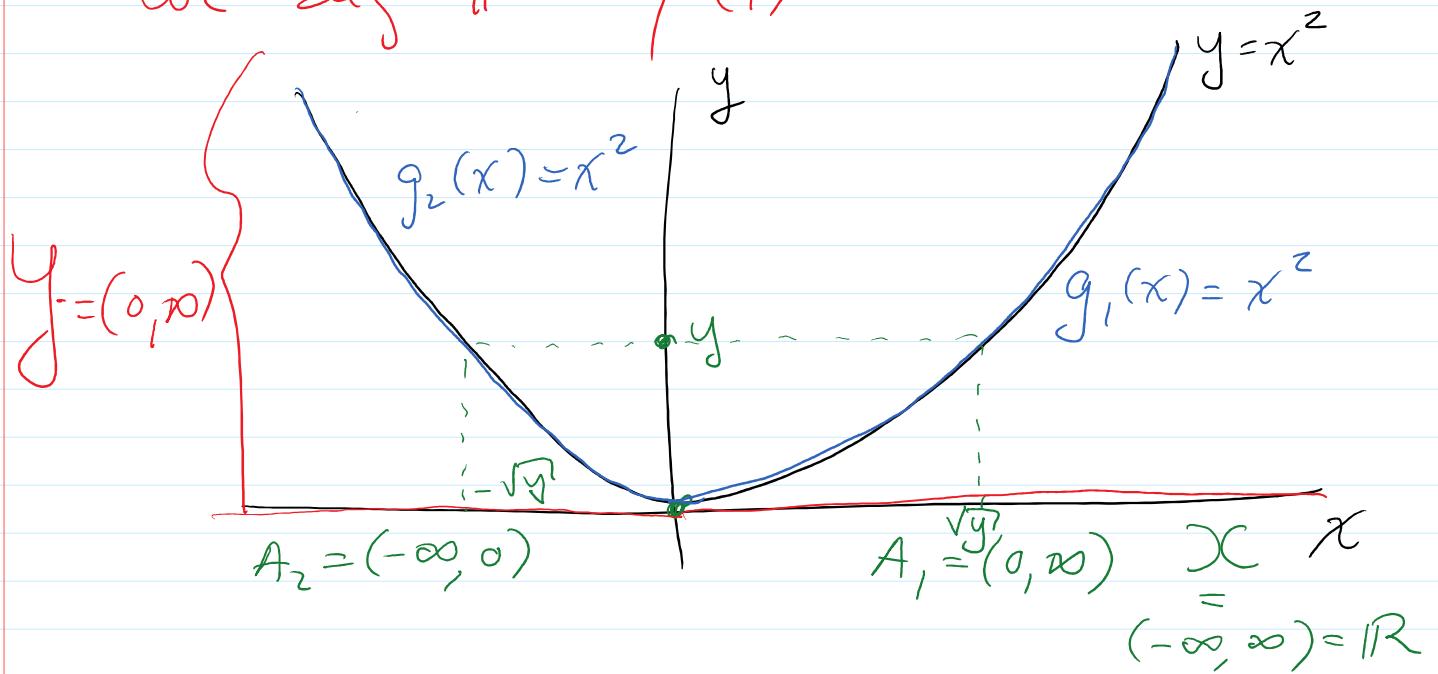
Ex. Chi-Sq.

(let $X \sim N(0,1)$ and $Y = X^2$ ($g(x) = x^2$))

we say $Y \sim \chi^2(1)$

$$Y = X^2$$

we say " - " / " (1)



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

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

$$\text{Need: } \frac{dg_1^{-1}}{dy} = \frac{1}{2\sqrt{y}} \quad \text{and} \quad \frac{dg_2^{-1}}{dy} = -\frac{1}{2\sqrt{y}}$$

$$\text{Recall: } f_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}x^2\right) \text{ for } x \in \mathbb{R}$$

$$f_Y(y) = \sum_{i=1}^2 f_X(g_i^{-1}(y)) \left| \frac{dg_i^{-1}}{dy} \right|$$

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

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

$$\begin{aligned}
 &= \frac{\exp(-\frac{1}{2}y)}{\sqrt{2\pi}} / \left| \frac{\exp(-\frac{1}{2}y)}{\sqrt{2\pi}} + \frac{\exp(\frac{1}{2}y)}{\sqrt{2\pi}} \right| \\
 &= \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}y\right) \frac{1}{2\sqrt{y}} + \frac{1}{\sqrt{2\pi}} \exp\left(\frac{1}{2}y\right) \frac{1}{2\sqrt{y}} \\
 &= \boxed{\frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}y\right) \frac{1}{\sqrt{y}}} \quad \text{pdf of a } \mathcal{X}^{(1)} \\
 &\quad \text{if } Y = X^2 \text{ when } X \sim N(0, 1)
 \end{aligned}$$

Theorem: Probability Integral Transformation

If X has CDF F_X and X is continuous
(so is F_X)

then

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

Pf: Assume F_X is (strictly) increasing

The F_X is invertible.

$$Y = F_X(X) = \underline{g(X)} \quad \text{where } \underline{g = F_X}$$

then

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

Recall the CDF for $U \sim U(0, 1)$



Recall the CDF for $U \sim U(0,1)$

$$F_U(u) = \begin{cases} 0 & u < 0 \\ u & 0 \leq u \leq 1 \\ 1 & u > 1 \end{cases}$$

CDF of a
 $U(0,1)$

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

Generating RVs.

Want \underline{X} , a r.v. w/ a CDF F_X

Know $\boxed{Y = F_X(X) \sim U(0,1)}$

$$\underline{F_X^{-1}(Y)} = F_X^{-1}(F_X(\underline{X})) = \underline{X} \leftarrow \text{have CDF } F_X$$

Algo: ① generate $Y \sim U(0,1)$

② plug Y into desired CDF inverse

③ $X = F_X^{-1}(Y) \sim$ as I want.