

Lecture-22

X is discrete, $h(x)$ \rightarrow $y = h(x)$

discrete $\xrightarrow{\text{cont}}$ discrete

X is continuous, $h(x)$ discontinuous

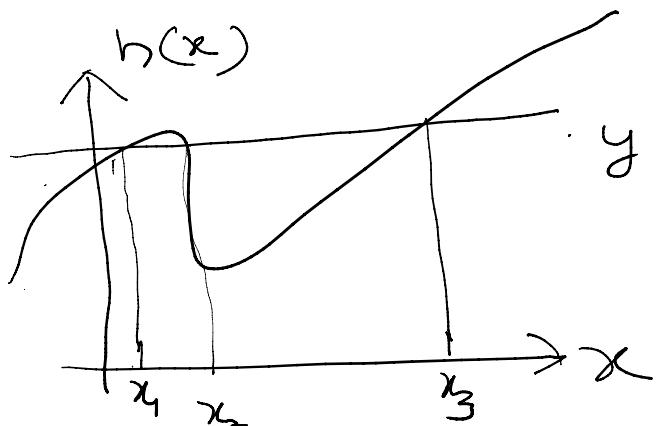
$y = h(x)$ discrete

X : Continuous, h : Continuous

$\{Y \leq y\}$ corresponds

to the event

$\{X \leq x_1 \text{ and } x_2 \leq X \leq x_3\}$



Thus

$$\begin{aligned} P\{Y \leq y\} &= P\{X \text{ values yielding} \\ &\quad Y \leq y\} \\ &= P\{X \mid Y \leq y\} \end{aligned}$$

$$\Rightarrow F_Y(y) = P\{Y \leq y\} = P\{X \mid Y \leq y\}$$

$$= \int_{\{x | y \leq x\}} f_x(x) dx$$

$$F_y(y) = \int_{\{x | y \leq x\}} f_x(x) dx$$

If $F_y(y)$ is differentiable, then

$$f_y(y) = \frac{d}{dy} \int_{\{x | y \leq x\}} f_x(x) dx$$

Ex: Let $X \sim \exp(5)$

Find p.d.f if it exists for the random variable $Y = \min\{X, 10\}$.

X : Continuous $y = h(x) = \min\{x, 10\}$
 h : Continuous

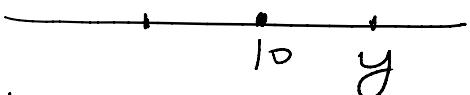
$$R_X = [0, \infty)$$

$$R_y = [0, 10]$$

Note that for $y \in \mathbb{R}$

$$\underline{\{Y \leq y\}} = \{X \leq y\} \cup \{10 \leq y\}$$

Note that if



$10 > y$, then $\{10 \leq y\} = \emptyset$

and if $y \geq 10$ then $\{10 \leq y\} = S$.

$$\Rightarrow \{Y \leq y\} = \begin{cases} X \leq y, & \text{if } y < 10 \\ S, & \text{if } y \geq 10 \end{cases}$$

$$\Rightarrow P\{Y \leq y\} = \begin{cases} P\{X \leq y\} & \text{if } y < 10 \\ P(S) & \text{if } y \geq 10 \end{cases}$$

i.e.

$$F_Y(y) = \begin{cases} F_X(y), & \text{if } y < 10 \\ , & \text{if } y \geq 10 \end{cases}$$

$$= \begin{cases} 0, & y < 0 \\ 1 - e^{-5y} & \text{if } 0 \leq y < 10 \\ 1 & \text{if } y \geq 10 \end{cases}$$

Hence, $F_Y(y)$ is discontinuous at $y = 10$
hence $Y = \min\{X, 10\}$ does not have
pd.f.

Ex. Let $X \sim U[0, 1]$.
Find the pd f. of r.v. $Y = X^2$ if it exists.

Ex. Let $X \sim U[0, 1]$.

Find the pd f. of r.v. $Y = X^2$ if it exists.

X : continuous, $Y = h(x) = x^2$ continuous
$$Y = X^2$$

Since $X \sim U[0, 1]$, so

$$F_X(x) = P(X \leq x)$$

$$= \begin{cases} 0, & \text{if } x < 0 \\ x, & \text{if } 0 \leq x \leq 1 \\ 1, & \text{if } x > 1 \end{cases}$$

For $y \in \mathbb{R}$,

$$F_Y(y) = P(Y \leq y) = P\{X^2 \leq y\}$$

$$= \begin{cases} 0 & \text{if } y < 0 \\ P(-\sqrt{y} \leq X \leq \sqrt{y}) & \text{if } y \geq 0 \end{cases}$$

$$= \begin{cases} 0 & \text{if } y < 0 \\ \frac{F_x(\sqrt{y}) - F_x(-\sqrt{y})}{2} & \text{if } y \geq 0 \end{cases}$$

$$= \begin{cases} 0, & \text{if } y < 0 \\ F_x(\sqrt{y}), & \text{if } y \geq 0 \end{cases}$$

$$= \begin{cases} 0, & \text{if } y < 0 \\ \sqrt{y} & \text{if } 0 \leq y \leq 1 \\ 1 & \text{if } y > 1 \end{cases}$$

CDF of $y = x^2$ is continuous on \mathbb{R} and differentiable everywhere except the point $y = 0, 1$. Also the derivative of CDF $F_y(y)$ is continuous everywhere except at $y = 0, 1$. Hence

$$f_y(y) = \begin{cases} 0, & y \leq 0 \\ \frac{1}{2\sqrt{y}}, & 0 < y < 1 \\ 0, & y \geq 1 \end{cases}$$

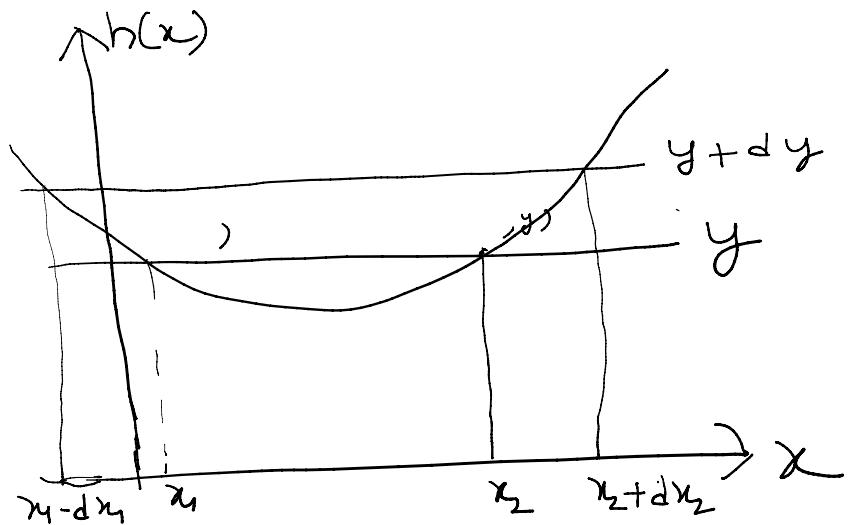
Thm. Suppose X is a random variable with pdf. $f_X(x)$. and $h: I \rightarrow \mathbb{R}$

be a function s.t. $Y = h(X)$ and Y has a p.d.f. Here

$$f_Y(y) = \sum_i \frac{f_X(x_i)}{|h'(x_i)|}$$

where x_1, x_2, \dots are the real roots of $y = h(x)$.

Boosht To avoid the generalities, we assume that the funcⁿ $y = h(x)$ has 2 roots say x_1 and x_2 i.e. $h(x_1) = h(x_2) = y$.



We know that

$$f_Y(y) |dy| = P \left\{ y \leq Y \leq y + |dy| \right\}$$

$$\begin{aligned} f_X(x) &\simeq f_X(a) \\ &= P \left\{ a - \frac{\epsilon}{2} < X < a + \frac{\epsilon}{2} \right\} \end{aligned}$$

$$= P\left\{ \underbrace{y \leq h(x) \leq y + |dy|}_{+ |dy|} \right\} = P\left\{ \overbrace{a - \frac{\epsilon}{2} < x < a + \frac{\epsilon}{2}}^2 \right\}$$

$$= P\left\{ x_1 - |dx_1| \leq x \leq x_1 \right\}$$

$$+ P\left\{ x_2 \leq x \leq x_2 + |dx_2| \right\}$$

$$= f_x(x_1) |dx_1| + f_x(x_2) |dx_2|$$

$$\rightarrow f_y(y) = f_x(x_1) \frac{|dx_1|}{|dy|} + f_x(x_2) \frac{|dx_2|}{|dy|}$$

$$= \frac{f_x(x_1)}{\left| \frac{dy}{dx} \right|_{x=x_1}} + \frac{f_x(x_2)}{\left| \frac{dy}{dx} \right|_{x=x_2}}$$

$$= \frac{f_x(x_1)}{|h'(x_1)|} + \frac{f_x(x_2)}{|h'(x_2)|}$$

In general

$$h(x_i) = y, \quad i = 1, 2, \dots$$

then

$$f_y(y) = \sum \frac{f_x(x_i)}{|h'(x_i)|}$$

$$f_y(y) = \sum_i \frac{f_x(x_i)}{|h'(x_i)|}$$

Ex 1 Let X be a cont. r.v. with
pd. f. $f_X(x)$.

$$Y = aX + b, \text{ where } a \neq 0, \\ b \in \mathbb{R}.$$

$$f_Y(y) = ?$$

$$y = h(x) = ax + b$$

then $x_1 = \frac{y-b}{a}$ is the only one
real root of $y = h(x)$.

$$f_Y(y) = \frac{f_X(x_1)}{|h'(x_1)|} \quad h'(x) = a \\ h'(x_1) = a$$

$$= \frac{f_X\left(\frac{y-b}{a}\right)}{|a|}$$

$$= \frac{1}{|a|} f_X\left(\frac{y-b}{a}\right)$$

$$\text{if } X \sim N(\mu, \sigma^2), \quad Y = aX + b$$

$$f_Y(y) = \frac{1}{|a|} f_X\left(\frac{y-b}{a}\right)$$

$$= \frac{1}{|a|} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2} \left\{ \frac{y-b}{a} - \mu \right\}^2}$$

$$= \frac{1}{a \sqrt{2\pi}} e^{-\frac{1}{2} \left\{ \frac{y - (a\mu + b)}{a} \right\}^2}$$

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

$$Y \sim N(a\mu + b, a^2 \sigma^2).$$