

10.5 Function of Continuous Random Variables: SISO

Reconsider the derived random variable $Y = g(X)$.

$$x \rightarrow \boxed{g(\cdot)} \rightarrow Y = g(x)$$

Recall that we can find $\mathbb{E}Y$ easily by (22):

$$\mathbb{E}Y = \mathbb{E}[g(X)] = \int_{\mathbb{R}} g(x) f_X(x) dx.$$

Ex $Y = X^2$
 $g(\cdot) = (\cdot)^2$
 $Y = 4 \mid X = -1.5$
 $g(\cdot) = 4 \mid \cdot = -1.5$
 $\mathbb{E}[Y] = \mathbb{E}[X^2] = \int_{-\infty}^{\infty} x^2 f_X(x) dx$

However, there are cases when we have to evaluate probability directly involving the random variable Y or find $f_Y(y)$ directly.

Recall that for discrete random variables, it is easy to find $p_Y(y)$ by adding all $p_X(x)$ over all x such that $g(x) = y$:

$$p_Y(y) = \sum_{x:g(x)=y} p_X(x). \quad (23)$$

For continuous random variables, it turns out that we can't⁴⁷ simply integrate or add the pdf of X to get the pdf of Y .

10.76. For $Y = g(X)$, if you want to find $f_Y(y)$, the following **two-step procedure** will always work and is easy to remember:

(a) Find the cdf $F_Y(y) = P[Y \leq y] = P[g(X) \leq y]$

(b) Compute the pdf from the cdf by "finding the derivative"
 $f_Y(y) = \frac{d}{dy} F_Y(y)$ (as described in 10.22).

Example 10.77. Suppose $X \sim \mathcal{E}(\lambda)$. Let $Y = 5X$. Find $f_Y(y)$.

Before doing this, check that Y is cont. by checking that the solutions for $y=g(x)$ is always (at most) countable at every y value.

Method 1

Step ①: $F_Y(y) = P[Y \leq y] = P[5X \leq y] = P[X \leq \frac{y}{5}] = F_X(\frac{y}{5})$

by defn. of being a cdf

$Y = 5X$

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

Step ②: $f_Y(y) = \frac{d}{dy} F_Y(y) = \frac{d}{dy} (F_X(\frac{y}{5})) = \frac{1}{5} f_X(\frac{y}{5}) = \frac{1}{5} \times \begin{cases} \lambda e^{-\lambda \frac{y}{5}}, & \frac{y}{5} > 0, \\ 0, & \text{otherwise} \end{cases}$

$= \begin{cases} \frac{\lambda}{5} e^{-\frac{\lambda}{5} y}, & y > 0, \\ 0, & \text{otherwise} \end{cases}$

chain rule



⁴⁷When you applied Equation (23) to continuous random variables, what you would get is $0 = 0$, which is true but not interesting nor useful.

Method 2

$Y = 5X \Rightarrow a = 5, b = 0$

Observe that $Y \sim \mathcal{E}(\frac{\lambda}{5})$

$f_Y(y) = \frac{1}{|a|} f_X(\frac{y-b}{a}) = \frac{1}{5} f_X(\frac{y}{5})$

(Affine)

10.78. **Linear Transformation:** Suppose $Y = aX + b$. Then, the cdf of Y is given by

step ①

$$F_Y(y) = P[Y \leq y] = P[aX + b \leq y] = \begin{cases} P\left[X \leq \frac{y-b}{a}\right], & a > 0, \\ P\left[X \geq \frac{y-b}{a}\right], & a < 0. \end{cases}$$

Now, by definition, we know that

$$P\left[X \leq \frac{y-b}{a}\right] = F_X\left(\frac{y-b}{a}\right),$$

and

$$\begin{aligned} P\left[X \geq \frac{y-b}{a}\right] &= P\left[X > \frac{y-b}{a}\right] + P\left[X = \frac{y-b}{a}\right] \\ &= 1 - F_X\left(\frac{y-b}{a}\right) + P\left[X = \frac{y-b}{a}\right]. \end{aligned}$$

For **continuous random variable**, $P\left[X = \frac{y-b}{a}\right] = 0$. Hence,

$$F_Y(y) = \begin{cases} F_X\left(\frac{y-b}{a}\right), & a > 0, \\ 1 - F_X\left(\frac{y-b}{a}\right), & a < 0. \end{cases}$$

step 2:

Finally, fundamental theorem of calculus and chain rule gives

$$f_Y(y) = \frac{d}{dy}F_Y(y) = \begin{cases} \frac{1}{a}f_X\left(\frac{y-b}{a}\right), & a > 0, \\ -\frac{1}{a}f_X\left(\frac{y-b}{a}\right), & a < 0. \end{cases}$$

Note that we can further simplify the final formula by using the $|\cdot|$ function:

$$f_Y(y) = \frac{1}{|a|}f_X\left(\frac{y-b}{a}\right), \quad a \neq 0. \quad (24)$$

$y = ax + b$
 $x = \frac{y-b}{a}$

Graphically, to get the plots of f_Y , we compress f_X horizontally by a factor of a , scale it vertically by a factor of $1/|a|$, and shift it to the right by b .

Of course, if $a = 0$, then we get the uninteresting degenerated random variable $Y \equiv b$.

$Y = aX + b$ where $a = 0$, X is cont.



$$P[Y=y] = \begin{cases} 0, & y \neq b, \\ \underbrace{P[X \in (-\infty, \infty)]}_1, & y = b, \end{cases}$$

$$f_x(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{x-m}{\sigma}\right)^2}$$

10.79. Suppose $X \sim \mathcal{N}(m, \sigma^2)$ and $Y = aX + b$ for some constants a and b . Then, we can use (24) to show that $Y \sim \mathcal{N}(am + b, a^2\sigma^2)$.

$$\begin{aligned} Y = aX + b &\Rightarrow f_Y(y) = \frac{1}{|a|} f_X\left(\frac{y-b}{a}\right) \\ &= \frac{1}{\sqrt{2\pi}\sigma|a|} e^{-\frac{1}{2}\left(\frac{y-b-m}{\sigma}\right)^2} \\ &= \frac{1}{\sqrt{2\pi}\sigma|a|} e^{-\frac{1}{2}\left(\frac{y-(am+b)}{|a|\sigma}\right)^2} \end{aligned}$$

$$Y \sim \mathcal{N}(am+b, \sigma^2 a^2) \quad \sigma_Y = \sigma|a|$$

Special cases: ([10.57] p. 144)

(a) $Z \sim \mathcal{N}(0,1) \Rightarrow X = \sigma Z + m \sim \mathcal{N}(m, \sigma^2)$
 $a = \sigma, b = m$

$a = 1/\sigma, b = -m/\sigma$

(b) $X \sim \mathcal{N}(m, \sigma^2) \Rightarrow Z = \frac{X-m}{\sigma} \sim \mathcal{N}(0,1)$

From earlier section (1),

$X \sim \mathcal{N}(m, \sigma^2)$

$\mathbb{E}X = m, \text{Var} X = \sigma^2$

$Y = aX + b$

$\mathbb{E}Y = a\mathbb{E}X + b = am + b$

$\text{Var} Y = a^2 \text{Var} X = a^2 \sigma^2$

$\sigma_Y = |a|\sigma$

Example 10.80. Amplitude modulation in certain communication systems can be accomplished using various nonlinear devices such as a semiconductor diode. Suppose we model the nonlinear device by the function $Y = X^2$. If the input X is a continuous random variable, find the density of the output $Y = X^2$.

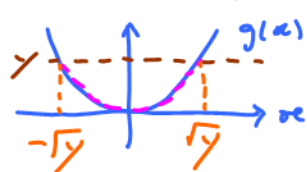
Remark: The "square" func. is a very important func. in EC. For example, you need it for power calculation.

① $F_Y(y) \equiv P[Y \leq y] = P[X^2 \leq y]$

①.1 First note that $Y = X^2 \geq 0$. Therefore, for $y < 0$, $F_Y(y) = 0$.
 (When you are working on this type of question, always try to bring out this easy case first.)

Ex. $F_Y(-3) = P[Y \leq -3] = P[X^2 \leq -3] = 0$.

①.2 For any $y > 0$, $F_Y(y) = P[X^2 \leq y]$



$= P[-\sqrt{y} \leq X \leq \sqrt{y}] = F_X(\sqrt{y}) - F_X(-\sqrt{y})$

X is a cont. RV

Ex. $P[X^2 \leq 4] =$

$P[-2 \leq X \leq 2]$

①.3 For $y = 0$, $F_Y(0) = P[X^2 \leq 0] = P[X = 0] = 0$

$$F_Y(y) = \begin{cases} F_X(\sqrt{y}) - F_X(-\sqrt{y}), & y > 0, \\ 0, & y \leq 0. \end{cases}$$

step ② $\Rightarrow \frac{d}{dy} f_Y(y) = \begin{cases} \frac{f_X(\sqrt{y})}{2\sqrt{y}} + \frac{f_X(-\sqrt{y})}{2\sqrt{y}}, & y > 0 \\ 0, & y \leq 0 \end{cases}$

add the case when $y = 0$.

(This is OK if we know Y is a cont. RV)

To check that a RV Y is a cont. RV, there are two important techniques:

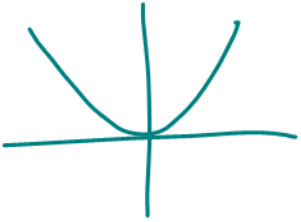
- ① Check that the cdf $F_Y(y)$ is a continuous function (for all y)
(no jump)
- ② Check that $P[Y=y] = 0$ (for any y).

Ex. Consider $Y = X^2$ when X is a continuous RV

$$P[Y=y] = P[X^2=y] = \begin{cases} P[X=0], & y=0, \\ P[X=\sqrt{y}] + P[X=-\sqrt{y}] & y>0, \\ 0 & y<0, \end{cases}$$

\uparrow
 $0+0$

There is no X that satisfies $X^2 = \text{negative number}$.



Because $P[Y=y] = 0$ for all y , we conclude that Y is a cont. RV.

Ex. consider $Y = g(X)$ when X is a cont. RV

Is Y a continuous RV? and

Yes.

(The number of solutions for the eq. $y = g(x)$ is finite.)



Ex. $Y = \cos(X)$ when X is a cont. RV

Is Y a continuous RV?

Yes

(The number of solutions for the eq. $y = \cos(x)$ is countable

$$P[Y=0] = P[\cos(X)=0] = P[X = \frac{\pi}{2} + k\pi, k=0, \pm 1, \pm 2, \dots]$$

$$= \sum_{k=-\infty}^{\infty} P[X = \frac{\pi}{2} + k\pi] = 0$$

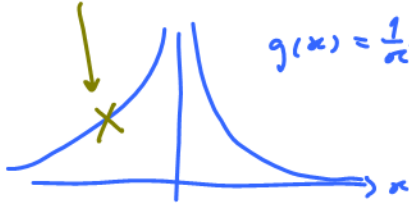


For each y value, the collection of x values that satisfy $y = g(x)$ is at most countable.

$\Rightarrow Y$ is a cont. RV.

In particular, one \rightarrow two solutions when $y > 0$ if we consider only $x > 0$ no solution when $y \leq 0$

Don't have to consider this side because $X \sim \mathcal{E}(\lambda)$ is always > 0 .



Example 10.81. Suppose $X \sim \mathcal{E}(\lambda)$. Let $Y = \frac{1}{X^2}$. Find $f_Y(y)$.

step ① $F_Y(y) \equiv P[Y \leq y] = P[\frac{1}{X^2} \leq y]$

①.1 Note that $\frac{1}{X^2} > 0 \Rightarrow F_Y(y) = 0$ for $y \leq 0$

①.2 For $y > 0$,

$$F_Y(y) = P[\frac{1}{X^2} \leq y] = P[X^2 \geq \frac{1}{y}] = P[X > \frac{1}{\sqrt{y}}]$$

$$= 1 - P[X \leq \frac{1}{\sqrt{y}}] = 1 - F_X(\frac{1}{\sqrt{y}}) + P[X < -\frac{1}{\sqrt{y}}]$$

for $X \sim \mathcal{E}$

①.1 + ①.2: $F_Y(y) = \begin{cases} 1 - F_X(\frac{1}{\sqrt{y}}), & y > 0, \\ 0, & \text{otherwise} \end{cases}$

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x}, & x > 0, \\ 0, & \text{otherwise} \end{cases}$$

$$f_X(\frac{1}{\sqrt{y}}) = \begin{cases} \lambda e^{-\lambda/\sqrt{y}}, & \frac{1}{\sqrt{y}} > 0, \\ 0, & \text{otherwise} \end{cases}$$

any $y > 0$ would satisfies this

Exercise 10.82 (F2011). Suppose X is uniformly distributed on the interval $(1, 2)$. ($X \sim \mathcal{U}(1, 2)$.) Let $Y = \frac{1}{X^2}$.

(a) Find $f_Y(y)$.

(b) Find $\mathbb{E}Y$.

step ② $\frac{d}{dy} \frac{1}{\sqrt{y}} = \frac{d}{dy} y^{-\frac{1}{2}} = -\frac{1}{2} y^{-\frac{3}{2}}$

$$f_Y(y) = \begin{cases} + f_X(\frac{1}{\sqrt{y}}) \left(+ \frac{1}{2} y^{-\frac{3}{2}} \right), & y > 0, \\ 0, & \text{otherwise.} \end{cases}$$

Exercise 10.83 (F2011). Consider the function

$$g(x) = \begin{cases} x, & x \geq 0 \\ -x, & x < 0. \end{cases}$$

$$= \begin{cases} \lambda e^{-\frac{\lambda}{\sqrt{y}}} \times \frac{1}{2} y^{-\frac{3}{2}}, & y > 0, \\ 0, & \text{otherwise.} \end{cases}$$

Suppose $Y = g(X)$, where $X \sim \mathcal{U}(-2, 2)$.

Remark: The function g operates like a **full-wave rectifier** in that if a positive input voltage X is applied, the output is $Y = X$, while if a negative input voltage X is applied, the output is $Y = -X$.

(a) Find $\mathbb{E}Y$.

(b) Plot the cdf of Y .

(c) Find the pdf of Y

Definition \equiv (minimal) support is countable $\equiv P[X = \kappa] = 0 \quad \forall \kappa$
 $\equiv \exists$ a countable set S \uparrow such that $P[X \in S] = 1$. there exists
 $\equiv F_X(\kappa)$ is a staircase func. $\equiv f_X(\kappa)$ is real-valued (no δ function)
 $\equiv F_X(\kappa)$ is a continuous function

	Discrete	Continuous
$P[X \in B] =$	$\sum_{x \in B} p_X(x)$	$\int_B f_X(x) dx$
$P[X = x] =$	$p_X(x) = F(x) - F(x^-)$	0
Interval prob.	$P^X((a, b]) = F(b) - F(a)$ $P^X([a, b]) = F(b) - F(a^-)$ $P^X([a, b)) = F(b^-) - F(a^-)$ $P^X((a, b)) = F(b^-) - F(a)$	$P^X([a, b]) = P^X((a, b]) = P^X([a, b)) = P^X((a, b))$ $= \int_a^b f_X(x) dx = F(b) - F(a)$
$\mathbb{E}X =$	$\sum_x x p_X(x)$	$\int_{-\infty}^{+\infty} x f_X(x) dx$
For $Y = g(X)$,	$p_Y(y) = \sum_{x: g(x)=y} p_X(x)$	$f_Y(y) = \frac{d}{dy} P[g(X) \leq y]$. Alternatively, $f_Y(y) = \sum_k \frac{f_X(x_k)}{ g'(x_k) }$, x_k are the real-valued roots of the equation $y = g(x)$.
For $Y = g(X)$, $P[Y \in B] =$	$\sum_{x: g(x) \in B} p_X(x)$	$\int_{\{x: g(x) \in B\}} f_X(x) dx$
$\mathbb{E}[g(X)] =$	$\sum_x g(x) p_X(x)$	$\int_{-\infty}^{+\infty} g(x) f_X(x) dx$
$\mathbb{E}[X^2] =$	$\sum_x x^2 p_X(x)$	$\int_{-\infty}^{+\infty} x^2 f_X(x) dx$
$\text{Var } X =$	$\sum_x (x - \mathbb{E}X)^2 p_X(x)$	$\int_{-\infty}^{+\infty} (x - \mathbb{E}X)^2 f_X(x) dx$

Table 7: Important Formulas for Discrete and Continuous Random Variables