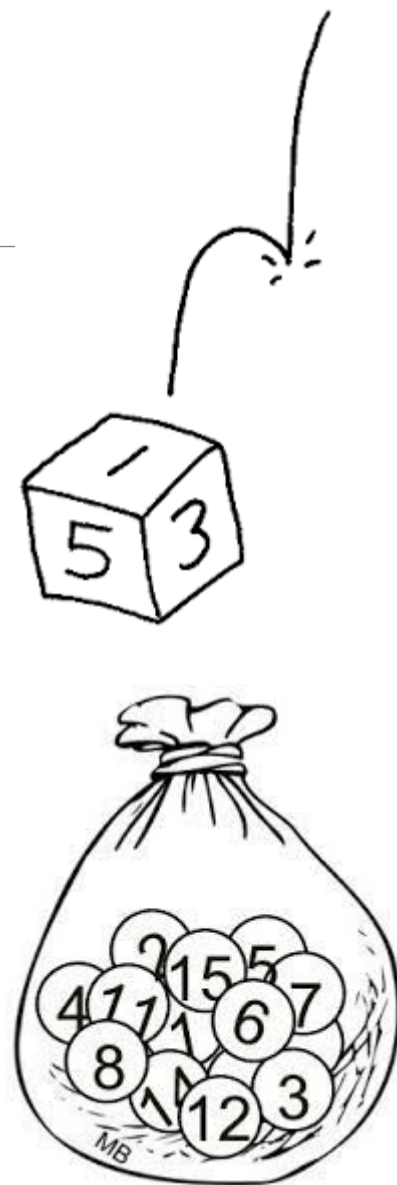


Lecture 09

- Independent Random Variables
- PDF of Functions of R.V.s



Independent discrete R.V.s

Recall the definition of independent events E and F : $P(EF) = P(E)P(F)$

Two discrete random variables X and Y are **independent** if:

for all x, y :

$$F(x, y) = F_X(x) \cdot F_Y(y)$$

$$P(X = x_i, Y = y_j) = P(X = x_i) \cdot P(Y = y_j)$$

$$p_{X,Y}(x_i, y_j) = p_X(x_i) \cdot p_Y(y_j)$$

$$p_{ij} = p_{i\cdot} \cdot p_{\cdot j}$$

Different notation,
same meaning:

- **Intuition:** knowing value of X tells us nothing about the distribution of Y (and vice versa).
- If two variables are not independent, they are **dependent**.

2-Dimensional

$$\begin{array}{ccc}
 F(x, y) & \xrightleftharpoons[f(x, y) = \frac{\partial^2 F(x, y)}{\partial x \partial y}]{} & f(x, y) \\
 \uparrow & & \uparrow \\
 F(x, y) = \int_{-\infty}^y \left[\int_{-\infty}^x f(u, v) du \right] dv & &
 \end{array}$$

$$\begin{array}{l}
 F(x, y) \\
 = F_X(x) \cdot F_Y(y) \\
 (X, Y \text{ are} \\
 \text{independent})
 \end{array}$$

$$\begin{array}{l}
 \text{Let } y \rightarrow +\infty, \\
 F(x, +\infty) = F_X(x) \\
 \text{Let } x \rightarrow +\infty, \\
 F(+\infty, y) = F_Y(y)
 \end{array}$$

$$\begin{array}{l}
 f(x, y) \\
 = f_X(x) \cdot f_Y(y) \\
 (X, Y \text{ are} \\
 \text{independent})
 \end{array}$$

$$\begin{array}{l}
 f_X(x) = \int_{-\infty}^{+\infty} f(x, y) dy \\
 f_Y(y) = \int_{-\infty}^{+\infty} f(x, y) dx
 \end{array}$$

$$\begin{array}{ccc}
 F_X(x), F_Y(y) & \xrightleftharpoons[F(x) = \int_{-\infty}^x f(t) dt]{} & f_X(x), f_Y(y) \\
 \uparrow & & \uparrow \\
 f(x) = \frac{dF(x)}{dx} & &
 \end{array}$$

1-Dimensional

Independent continuous R.V.s

Two continuous random variables X and Y are **independent** if:

for all x, y :

$$F(x, y) = F_X(x) \cdot F_Y(y)$$

$$f(x, y) = f_X(x) \cdot f_Y(y)$$

- Independent: $f(x, y) = f_X(x) \cdot f_Y(y)$ for all x, y
(独立)
- Uncorrelated: $E(XY) = E(X)E(Y)$
(不相关)
- Orthogonal: $E(XY) = 0$
(正交)













to be covered in a few weeks.

Simple dice rolling example

Let: D_1 and D_2 be the outcomes of two rolls

$S = D_1 + D_2$ be the sum of two rolls

- Each roll of a 6-sided die is an independent trial.
- Random variables D_1 and D_2 are independent.

						
	2	3	4	5	6	7
	3	4	5	6	7	8
	4	5	6	7	8	9
	5	6	7	8	9	10
	6	7	8	9	10	11
	7	8	9	10	11	12

1. Are events $(D_1 = 1)$ and $S = 7$ independent?



$$P(D_1 = 1) \cdot P(S = 7) = \frac{1}{6} \cdot \frac{1}{6} = \frac{1}{36} = p(D_1 = 1, S = 7)$$

2. Are events $(D_1 = 1)$ and $S = 5$ independent?



$$P(D_1 = 1) \cdot P(S = 5) = \frac{1}{6} \cdot \frac{1}{9} = \frac{1}{54} \neq p(D_1 = 1, S = 5)$$

3. Are random variables D_1 and S independent?

All events $X = x, Y = y$ must be independent!

Quick check

Ex. Assume X and Y are independent, fill the following table.

$X \backslash Y$	y_1	y_2	y_3	$p_{i\cdot}$
x_1		$1/8$		
x_2	$1/8$			
$p_{\cdot j}$	$1/6$			1

Quick check

Ex. Assume X and Y are independent, fill the following table.

$X \backslash Y$	y_1	y_2	y_3	$p_{i\cdot}$
x_1	$1/24$	$1/8$	$1/12$	$1/4$
x_2	$1/8$	$3/8$	$1/4$	$3/4$
$p_{\cdot j}$	$1/6$	$1/2$	$1/3$	1

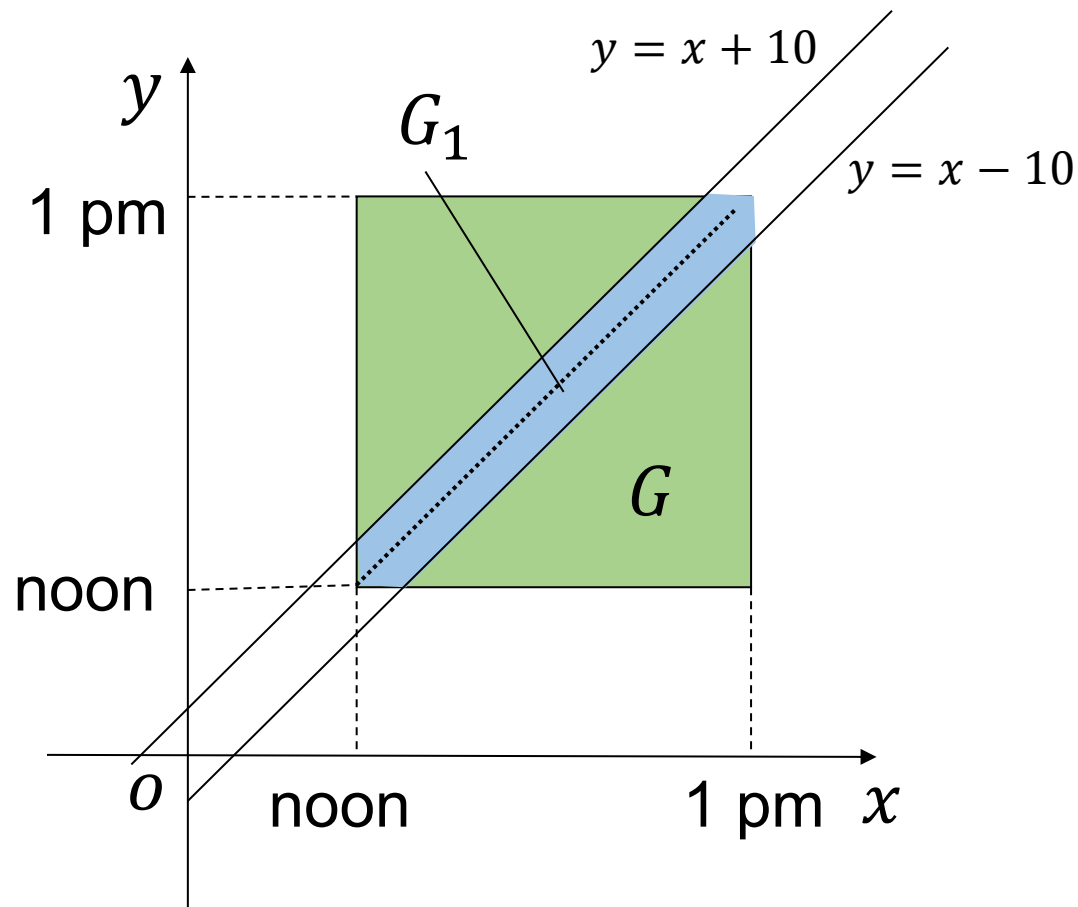
(Ex. 2c P249) A man and a woman decide to meet at a train station. If each of them independently arrives at a time uniformly distributed between 12 noon and 1 p.m., find the probability that the first to arrive has to wait longer than 10 minutes.

Sol. X and Y denote the time past 12 noon that the man and the woman arrive. $X, Y \sim U(0,60)$.

(Ex. 2c P249) A man and a woman decide to meet at a train station. If each of them independently arrives at a time uniformly distributed between 12 noon and 1 p.m., find the probability that the first to arrive has to wait longer than 10 minutes.

Sol. X and Y denote the time past 12 noon that the man and the woman arrive. $X, Y \sim U(0, 60)$.

$$\begin{aligned}
 P\{X + 10 < Y\} + P\{Y + 10 < X\} &\stackrel{\text{symmetric}}{=} 2 \cdot P\{X + 10 < Y\} \\
 &= 2 \iint_{x+10 < y} f(x, y) dx dy \stackrel{\text{independent}}{=} 2 \iint_{x+10 < y} f_X(x) f_Y(y) dx dy \\
 &= 2 \int_{10}^{60} \int_0^{y-10} \left(\frac{1}{60}\right)^2 dx dy = \frac{2}{60^2} \int_{10}^{60} (y - 10) dy = \frac{25}{36}
 \end{aligned}$$



$$\text{Area of } G = 2 \left[\frac{1}{2} \times \left(1 - \frac{1}{6} \right)^2 \right] = \frac{25}{36}$$

Ex. Given the joint PDF of R.V.s X and Y

$$1) \quad f(x, y) = \begin{cases} 6e^{-2x}e^{-3y} & 0 < x < \infty, 0 < y < \infty \\ 0 & \text{otherwise} \end{cases}$$

$$2) \quad f(x, y) = \begin{cases} 24xy & 0 < x < 1, 0 < y < 1, 0 < x + y < 1 \\ 0 & \text{otherwise} \end{cases}$$

Evaluate the independency of X and Y . Hint: $\int e^{ax} dx = \frac{1}{a} e^{ax}$

Ex. Given the joint PDF of R.V.s X and Y

$$1) \quad f(x, y) = \begin{cases} 6e^{-2x}e^{-3y} & 0 < x < \infty, 0 < y < \infty \\ 0 & \text{otherwise} \end{cases}$$

Evaluate the independency of X and Y . Hint: $\int e^{ax} dx = \frac{1}{a} e^{ax}$

Sol. Yes, since $f(x, y) = f_X(x) \cdot f_Y(y)$.

$$\begin{aligned} f_X(x) &= \int_0^{\infty} f(x, y) dy \\ &= 6e^{-2x} \int_0^{\infty} e^{-3y} dy = 2 \cdot e^{-2x}, 0 < x < \infty, \end{aligned}$$

$$\begin{aligned} f_Y(y) &= \int_0^{\infty} f(x, y) dx \\ &= 6e^{-3y} \int_0^{\infty} e^{-2x} dx = 3 \cdot e^{-3y}, 0 < y < \infty. \end{aligned}$$

Ex. Given the joint PDF of R.V.s X and Y

$$2) \quad f(x, y) = \begin{cases} 24xy & 0 < x < 1, 0 < y < 1, 0 < x + y < 1 \\ 0 & \text{otherwise} \end{cases}$$

Evaluate the independency of X and Y .

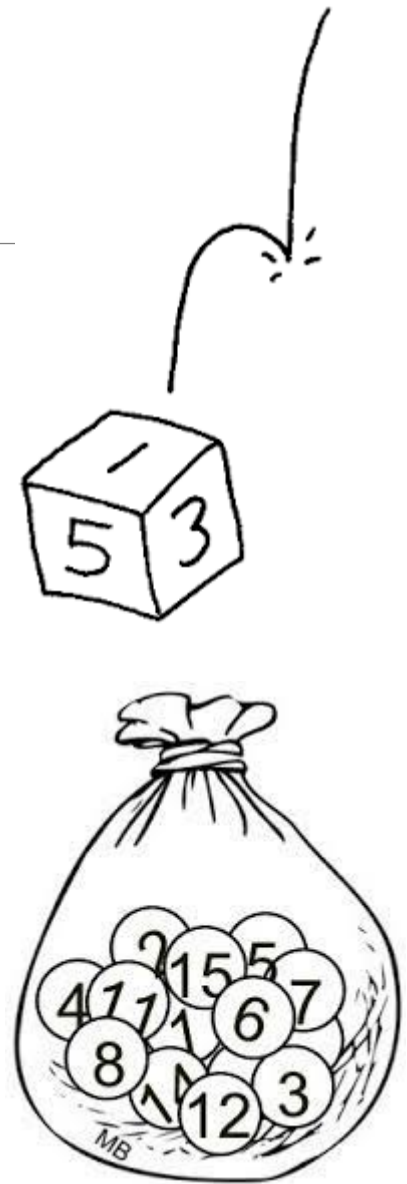
Sol. No, the defined ranges of x and y are not separable.

The values of x and y are dependent.

For example, if $x = 0.9999$, $P(y > 0.0001) = 0$

Lecture 09

- Independent Random Variables
- PDF of Functions of R.V.s



Review: PDF/PMF of a function of a R.V.

Known: PDF/PMF of X ; Want: PDF/PMF of $Y = g(x)$.

How do we measure temperature?

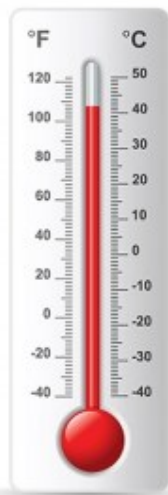
Height of the mercury bar (x)
 \Rightarrow Temperature (y)



Cold



Comfort



Hot

$$Y = g(X) = a \cdot X + b$$

Known: PDF of X ; Want: PDF of Y .

What if we consider the relationship between multiple R.V.s

\Rightarrow PDF/PMF of Functions of R.V.s
 $Z = g(X, Y)$

PMF of functions of discrete R.V.s

Ex. Given the PMF of (X, Y)

$X \backslash Y$	0	1	2
-1	0.2	0.3	0.1
2	0.1	0.1	0.2

Find the PMF of (1) $Z = X + Y$, (2) $Z = XY$.

Sol.

(X, Y)	$(-1, 0)$	$(-1, 1)$	$(-1, 2)$	$(2, 0)$	$(2, 1)$	$(2, 2)$
$Z = X + Y$	-1	0	1	2	3	4
$Z = XY$	0	-1	-2	0	2	4
$p(z)$	0.2	0.3	0.1	0.1	0.1	0.2

$Z = XY$	-2	-1	0	2	4
$p(z)$	0.1	0.3	0.3	0.1	0.2

PDF of functions of continuous R.V.s

Assume the PDF of (X, Y) is $f(x, y)$, find the PDF of $Z = g(X, Y)$.

$$\begin{aligned}F_Z(z) &= P\{Z \leq z\} \\&= P\{g(X, Y) \leq z\} \\&= P\{(X, Y) \in D_Z\} \\&= \iint_{D_Z} f(x, y) dx dy\end{aligned}$$

$$f_Z(z) = F'_Z(z)$$

Check P224 or slide06 for the single variable case $Y = g(X)$ (monotonically increasing):

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

$$\begin{aligned}F_Y(y) &= P\{Y \leq y\} = P\{g(x) \leq y\} \\&= P\{X \in D_y\} = \int_{D_y} f(x) dx\end{aligned}$$

PDF of functions of continuous R.V.s

PDF of $Z = X + Y$

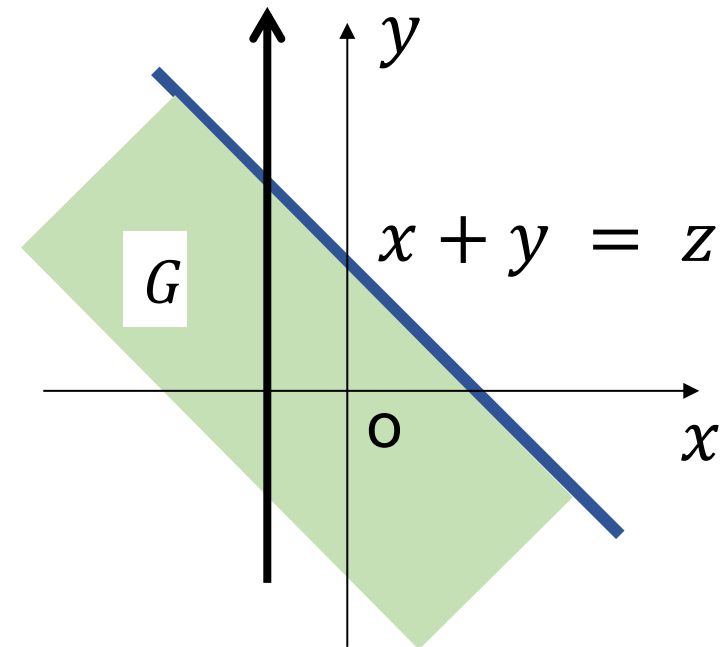
$$F_Z(z) = \iint_{D_Z} f(x, y) dx dy = \iint_{x+y \leq z} f(x, y) dx dy$$

$$= \int_{-\infty}^{+\infty} dx \int_{-\infty}^{z-x} f(x, y) dy$$

$$\begin{aligned} & \xrightarrow{y = u - x} \int_{-\infty}^{+\infty} dx \int_{-\infty}^z f(x, u - x) du \\ &= \int_{-\infty}^z \int_{-\infty}^{+\infty} f(x, u - x) dx du \end{aligned}$$

$$= \int_{-\infty}^z \int_{-\infty}^{+\infty} f(x, u - x) dx du$$

$$f_Z(z) = \int_{-\infty}^{+\infty} f(x, z - x) dx$$



Note: The derivation process is for your reference only.

PDF of functions of continuous R.V.s

PDF of $Z = X + Y$ is given as

$$f_Z(z) = \int_{-\infty}^{+\infty} f(x, z - x) dx$$

$$f_Z(z) = \int_{-\infty}^{+\infty} f(z - y, y) dy$$

If X and Y are independent

$$f_Z(z) = \int_{-\infty}^{+\infty} f_X(z - y) f_Y(y) dy$$

$$f_Z(z) = \int_{-\infty}^{+\infty} f_X(x) f_Y(z - x) dx$$

Ex. Let $X \sim \mathcal{N}(0,1)$, $Y \sim \mathcal{N}(0,1)$. X and Y are independent, find the PDF of $Z = X + Y$.

Sol. $f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}, \quad f_Y(y) = \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}}, (-\infty < x, y < +\infty)$

→ $f_Z(z) = \int_{-\infty}^{+\infty} f_X(x) f_Y(z-x) dx$

$$= \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} \frac{1}{\sqrt{2\pi}} e^{-\frac{(z-x)^2}{2}} dx = \frac{1}{2\pi} e^{-\frac{z^2}{4}} \int_{-\infty}^{+\infty} e^{-\left(x-\frac{z}{2}\right)^2} dx$$

→ $f_Z(z) = \frac{1}{2\pi} e^{-\frac{z^2}{4}} \int_{-\infty}^{+\infty} e^{-t^2} dt$ ← $t = x - \frac{z}{2}$

$$= \frac{1}{\sqrt{2}\sqrt{2\pi}} e^{-\frac{z^2}{4}} (-\infty < z < +\infty)$$

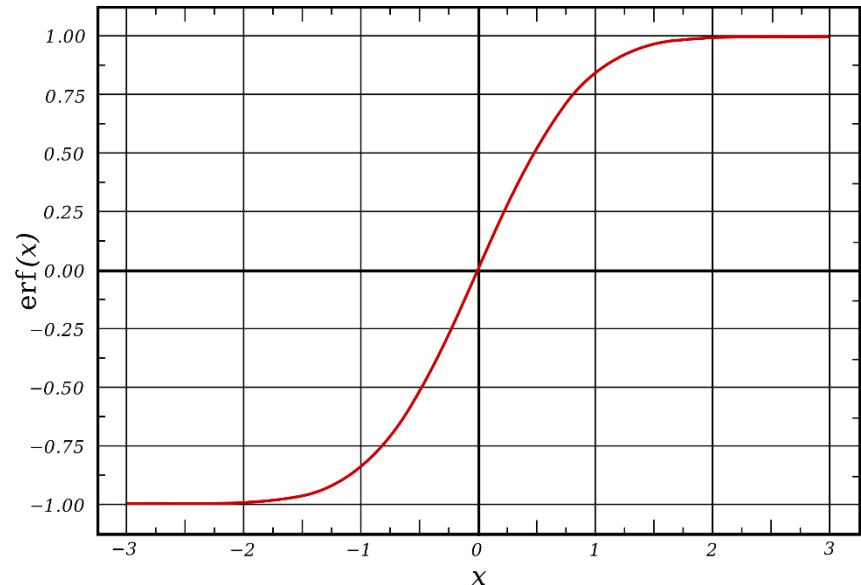
→ $Z = X + Y \sim \mathcal{N}(0,2)$

About error function, erf

$$\int e^{-ax^2} dx = \frac{\sqrt{\pi}}{2\sqrt{a}} \operatorname{erf}(\sqrt{a}x)$$

\Rightarrow

$$\int e^{-t^2} dt = \frac{\sqrt{\pi}}{2} \operatorname{erf}(x)$$



In mathematics, the **error function** (also called the **Gauss error function**), often denoted by **erf**, is a complex function of a complex variable defined as:

$$\operatorname{erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x e^{-t^2} dt$$

-----[Wikipedia](#)

- If $X \sim \mathcal{N}(\mu_1, \sigma_1^2)$, $Y \sim \mathcal{N}(\mu_2, \sigma_2^2)$ and X, Y are independent, then $X + Y$ follows normal distribution

$$X + Y \sim \mathcal{N}(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)$$

- If $X_i \sim \mathcal{N}(\mu_i, \sigma_i^2)$, ($i = 1, 2, \dots, n$) and X_i are independent, then

$$\sum_{i=1}^n X_i \sim \mathcal{N}\left(\sum_{i=1}^n \mu_i, \sum_{i=1}^n \sigma_i^2\right)$$

- The linear combination of finite number of independent normal variables follows normal distribution

$$\sum_{i=1}^n c_i X_i \sim \mathcal{N}\left(\sum_{i=1}^n c_i \mu_i, \sum_{i=1}^n c_i^2 \sigma_i^2\right)$$

Review: Linear transformation of Normal R.V.s

Let $X \sim \mathcal{N}(\mu, \sigma^2)$ with CDF $P(X \leq x) = F(x)$. Linear transformations of X are also Normal.

If $Y = aX + b$, then $Y \sim \mathcal{N}(a\mu + b, a^2\sigma^2)$

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

$$= P\{aX + b \leq y\}$$

$$= P\left\{X \leq \frac{y-b}{a}\right\}$$

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

Differentiation

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

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

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

$$Y \sim \mathcal{N}(a\mu + b, a^2\sigma^2)$$

PDF/CDF of functions of continuous R.V.s

$\max\{X, Y\}$ and $\min\{X, Y\}$, where X, Y are independent.

$$\begin{aligned} F_{\max}(z) &= P\{\max\{X, Y\} \leq z\} = P\{X \leq z, Y \leq z\} \\ &= P\{X \leq z\} \cdot P\{Y \leq z\} \end{aligned}$$

$$F_{\max}(z) = F_X(z) \cdot F_Y(z)$$

$$\begin{aligned} F_{\min}(z) &= P\{\min\{X, Y\} \leq z\} = 1 - P\{\min\{X, Y\} > z\} \\ &= 1 - P\{X > z, Y > z\} = 1 - P\{X > z\} \cdot P\{Y > z\} \end{aligned}$$

$$F_{\min}(z) = 1 - [1 - F_X(z)] \cdot [1 - F_Y(z)]$$

For X_1, \dots, X_n , n i.i.d. R.V.s with CDF $F(x)$, we have

$$F_{\max}(z) = [F(z)]^n, F_{\min}(z) = 1 - [1 - F(z)]^n$$

Ex. If the system L consists of two sub-systems with lifetime X and Y following PDFs

$$f_X(x) = \begin{cases} \alpha e^{-\alpha x}, & x > 0, \\ 0, & x \leq 0. \end{cases} \quad f_Y(y) = \begin{cases} \beta e^{-\beta y}, & y > 0, \\ 0, & y \leq 0. \end{cases}$$

where $\alpha > 0, \beta > 0, \alpha \neq \beta$. Find the PDFs of system lifetimes under the following connections.

(1) series. (2) parallel. (3) backup.

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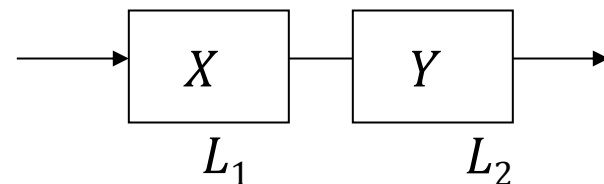
$$f_Z(z) = \int_{-\infty}^{+\infty} f(x, z-x) dx, Z = X + Y$$

$$F_{\text{max}}(z) = F_X(z) \cdot F_Y(z)$$

$$F_{\text{min}}(z) = 1 - [1 - F_X(z)] \cdot [1 - F_Y(z)]$$

Sol.

(1) Under series connection, we have $Z = \min\{X, Y\}$



$$F_{\min}(z) = 1 - [1 - F_X(z)] \cdot [1 - F_Y(z)]$$

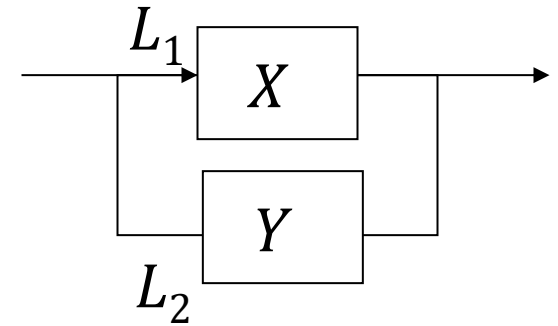
$$F_X(x) = \begin{cases} 1 - e^{-\alpha x}, & x > 0, \\ 0, & x \leq 0. \end{cases} \quad F_Y(y) = \begin{cases} 1 - e^{-\beta y}, & y > 0, \\ 0, & y \leq 0. \end{cases}$$

$$F_Z(z) = \begin{cases} 1 - e^{-(\alpha+\beta)z}, & z > 0, \\ 0, & z \leq 0. \end{cases}$$

$$f_Z(z) = F'_Z(z) = \begin{cases} (\alpha + \beta)e^{-(\alpha+\beta)z}, & z > 0, \\ 0, & z \leq 0. \end{cases}$$

(2) Under parallel connection, we have $Z = \max\{X, Y\}$.

$$F_{\max}(z) = F_X(z) \cdot F_Y(z)$$



$$F_X(x) = \begin{cases} 1 - e^{-\alpha x}, & x > 0, \\ 0, & x \leq 0. \end{cases} \quad F_Y(y) = \begin{cases} 1 - e^{-\beta y}, & y > 0, \\ 0, & y \leq 0. \end{cases}$$

$$F_Z(z) = \begin{cases} (1 - e^{-\alpha z})(1 - e^{-\beta z}), & z > 0 \\ 0, & z \leq 0 \end{cases}$$

$$f_Z(z) = \begin{cases} \alpha e^{-\alpha z} + \beta e^{-\beta z} - (\alpha + \beta)e^{-(\alpha + \beta)z}, & z > 0 \\ 0, & z \leq 0 \end{cases}$$

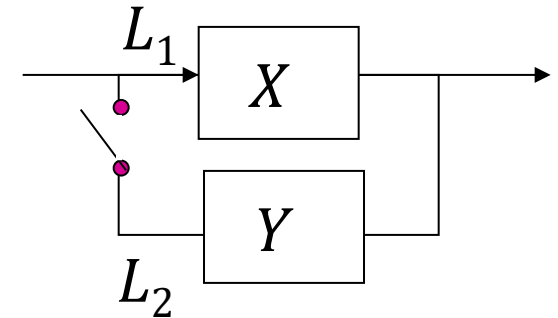
(3) Under backup connection, we have $Z = X + Y$.

$$f_Z(z) = \int_{-\infty}^{+\infty} f_X(x) \cdot f_Y(z - x) dx$$

$$f_X(x) = \begin{cases} \alpha e^{-\alpha x}, & x > 0, \\ 0, & x \leq 0. \end{cases}$$

$$f_Y(y) = \begin{cases} \beta e^{-\beta y}, & y > 0, \\ 0, & y \leq 0. \end{cases}$$

$$f_Z(z) = \begin{cases} \int_{-\infty}^{+\infty} \alpha e^{-\alpha x} \cdot \beta e^{-\beta(z-x)} dx, & 0 < x < z, \\ 0, & \text{otherwise} \end{cases}$$



Quiz 4

Date: 4-Nov-2024

Scope: Multi-dimensional R.V.s

Open-book exam, with fill-in-the-blank and multiple-choice questions.

Time: ~45 mins

Case study

The Task of Text Classification

CS 124: From Languages to Information
<https://web.stanford.edu/class/cs124/>