

Lecture 13

(19 September 2024)

Some problems and In-class queries

Q) Let F_1 and F_2 be two CDFs such that $F_1(x) < F_2(x)$ for all $x \in \mathbb{R}$. Assume that F_1 and F_2 are continuous and strictly increasing. Show that there exists RVs X_1 and X_2 with respective CDFs F_1 and F_2 defined on the same probability space such that $X_1 > X_2$.

Wrong Solution: Let $X_1 \not\geq X_2$

$$\Rightarrow X_1 \leq X_2$$

[This step is not correct]

$$F_{x_2}(x) = F_2(x)$$

$$= P(x_2 \leq x)$$

$$\leq P(x_1 \leq x)$$

$$= F_{x_1}(x) = F_1(x)$$

$\therefore F_2(x) \geq F_1(x)$ — a contradiction

$\therefore F_1(x) < F_2(x) \quad \forall x \Rightarrow x_1 > x_2$.

[of course this is wrong again]

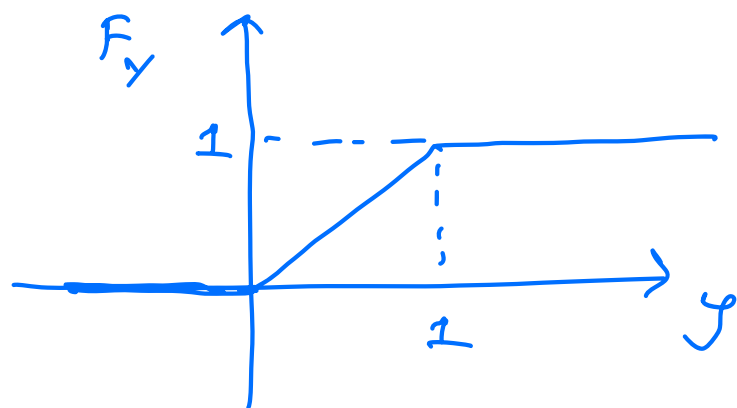
The negation of $x_1 > x_2$ i.e.,
 $x_1(\omega) > x_2(\omega) \quad \forall \omega \in \Omega$ is

$$\exists \omega \text{ s.t. } x_1(\omega) \leq x_2(\omega)$$

but not

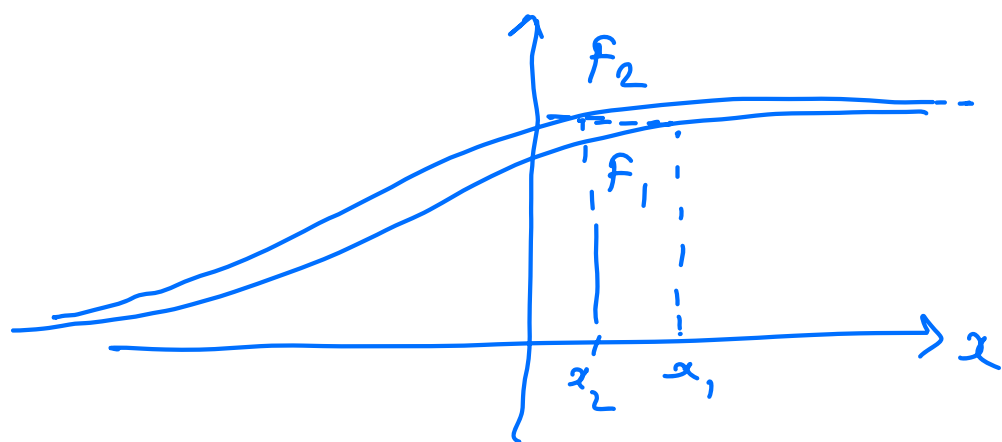
$$x_1(\omega) \leq x_2(\omega) \quad \forall \omega \in \Omega.$$

Correct solution Consider a RV y with the following cdf,



$$F_y(y) = \begin{cases} 0 & y < 0 \\ y & 0 \leq y < 1 \\ 1 & y \geq 1 \end{cases}$$

Given two cdfs F_1 & F_2 which are continuous and strictly increasing,



$$F_1(x_1) = F_2(x_2) = y \\ \Rightarrow x_1 > x_2,$$

Construct RV x_1 & x_2 as

$$x_1 = F_1^{-1}(y) \quad x_2 = F_2^{-1}(y)$$

We claim that cdf of $F_1^{-1}(y)$ is F_1 ,

$$\begin{aligned}
 P(x_i \leq x_i) &= P(F_i^{-1}(y) \leq x_i) \\
 &= P(y \leq F_i(x_i)) \\
 &= F_i(x_i) \quad , \quad i = \underline{1, 2} .
 \end{aligned}$$

$$F_1(x) < F_2(x) \quad \forall x \Rightarrow F_1^{-1}(y) > F_2^{-1}(y) \quad \forall y \in [0, 1]$$

$$\text{Suppose } x_1 = F_1^{-1}(y) \leq F_2^{-1}(y) = x_2 .$$

$$y = F_2(x_2) > F_1(x_2)$$

$$> F_1(x_1)$$

$$= y \rightarrow \text{A contradiction,}$$

$$\therefore x_1 = F_1^{-1}(y) > F_2^{-1}(y) = x_2 ,$$

Q) In class we defined conditional variance as

$$\text{Var}(X|Y=y) = E[(X - E[X|Y=y])^2 | Y=y]$$

The query is "why is the conditioning $Y=y$ required after the square"?

Not required i.e., the expression

$$E[(X - E[X|Y=y])^2]$$

is a well-defined quantity and it is a real number.

However if we define variance in this way (without $Y=y$) then we will not have

$$\text{Var}(X|Y=y) = E[X^2 | Y=y] - E[X|Y=y]^2$$

analogous to $\text{Var}(X) = E[X^2] - E[X]^2$.

$\therefore \text{Var}(X|Y=y) = E[(X - E[X|Y=y])^2]$ is opt.

Q) What is the intuition for the law of total variance?

$$\text{Var}(x) = E[\text{var}(x|y)] + \text{var}(E[x|y]).$$

Before we look at the intuition, let us first recall the definitions of $E[x|y]$ and $\text{var}(x|y)$.

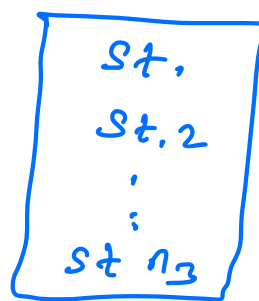
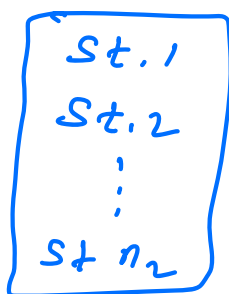
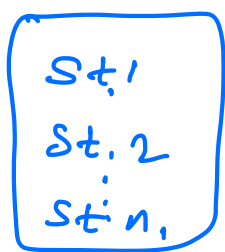
$$E[x|y=y] = \sum_x x p_{x|y}(x|y)$$

$$\text{Let } \phi(y) = E[x|y=y].$$

$$E[x|y] = \phi(y) = \begin{cases} E[x|y=y_1] & \text{w.p. } p_y(y_1) \\ E[x|y=y_2] & \text{w.p. } p_y(y_2) \\ \vdots \end{cases}$$

$$\text{Var}(x|y) = \psi(y) = \begin{cases} \text{var}(x|y=y_1) & \text{w.p. } p_y(y_1) \\ \text{var}(x|y=y_2) & \text{w.p. } p_y(y_2) \\ \vdots \end{cases}$$

We understand the intuition through an example. Consider students in three sections as below.



Section 1 Section 2 Section 3

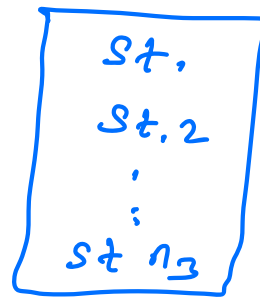
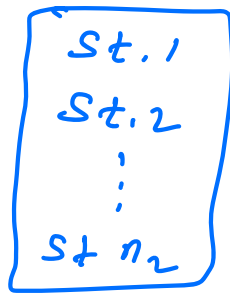
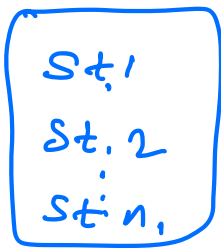
Let X = quiz score of a random student

Y = Section of random student

$Y \in \{1, 2, 3\}$, $X|Y=i$; takes n_i values.

$$\text{Var}(X) = E[\text{Var}(X|Y)] + \text{Var}(E[X|Y]).$$

$\text{Var}(X|Y=i)$ is the variance of quiz scores within section i .

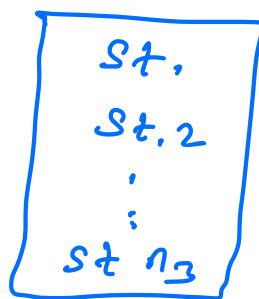
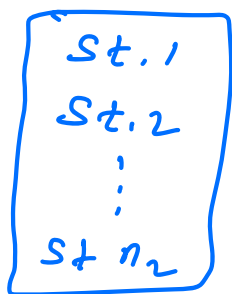
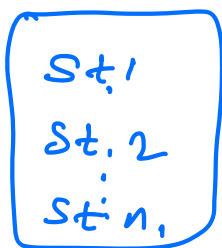


$$\text{Var}(X|Y=1) \quad \text{Var}(X|Y=2) \quad \text{Var}(X|Y=3)$$

$E[\text{Var}(X|Y)]$ is the average of these variances

$$E[\text{Var}(X|Y)] = P_Y(1) \text{Var}(X|Y=1) + P_Y(2) \text{Var}(X|Y=2) + P_Y(3) \text{Var}(X|Y=3)$$

However this does not capture how the quiz scores vary across the sections. This is exactly captured by the second term in the law of total variance.



$$E[X|Y=1]$$

$$E[X|Y=2]$$

$$E[X|Y=3]$$



average quiz scores
in each section

$\text{Var}(E[X|Y])$ is variance of
these expected values.

$$\text{Var}(E[X|Y]) = \sum_j (E[X|Y=j] - E[X])^2 p_Y(j)$$

$$\text{Thus } \text{Var}(X) = E[\text{Var}(X|Y)] + \text{Var}(E[X|Y])$$

= Avg. of variances within individual sections

+

variance of averages between sections

Recap of Last Lecture:

A RV X is said to be continuous if its CDF can be expressed as

$$F_X(x) = \int_{-\infty}^x f_X(u) du \quad x \in \mathbb{R}$$

for some integrable function $f: \mathbb{R} \rightarrow [0, \infty)$ called the probability density function (PDF),

$$f_X(x) = F_X'(x)$$

$$= \lim_{\Delta x \rightarrow 0} \frac{F_X(x + \Delta x) - F_X(x)}{\Delta x}$$

$$P(X \in B) = \int_B f_X(x) dx$$

$$\int_{-\infty}^{\infty} f_X(x) dx = 1$$

$$E[X] = \int_{-\infty}^{\infty} x f_X(x) dx$$

If $x \geq 0$ then

$$E[X] = \int_0^{\infty} P(X > x) dx$$

$$E[g(X)] = \int_{-\infty}^{\infty} g(x) f_X(x) dx$$

Uniform RV:

$$f_X(x) = \begin{cases} 1/(b-a) & a \leq x \leq b \\ 0 & \text{o.w.} \end{cases}$$

Exponential RV: $f_X(x) = \lambda e^{-\lambda x}$, $x \geq 0$,

Exercise. Show that for an exponential RV

$$P(X > s+t | X > s) = P(X > t)$$

for $s, t \geq 0$.