

23/12 LECTURE 27: EXP. OF CTS R.V.

X cts r.v. $\Leftrightarrow \exists$ a pdf f_X $f_X: \mathbb{R} \rightarrow [0, \infty)$ p.c.f.s, $\int_{-\infty}^{\infty} f_X(x) dx = 1$
 $\Leftrightarrow \exists$ a ^{cts} DF $F_X: \mathbb{R} \rightarrow [0, 1]$ $\exists F_X'$ exists except at fin. many pts.

$$P(X \in (a, b]) = \int_a^b f_X(x) dx = F_X(b) - F_X(a) \quad -\infty \leq a < b < \infty$$

THM 27.1 Let X be a cts r.v. $\Rightarrow F_X$ is a strictly \uparrow cts fn.
 Then (1) $F_X^{-1}(U) \stackrel{d}{=} X$ ($U \stackrel{d}{=} U(0,1)$ & F_X^{-1} is well-defined on $(0,1)$)
 (2) $F_X(X) \stackrel{d}{=} U$.

Proof: (1) F_X is strictly \uparrow & so F_X^{-1} is well-defined on $(0,1)$
 & also strictly increasing (ex.)
 $x \in \mathbb{R}$, $P(F_X^{-1}(U) \leq x) \stackrel{\Downarrow}{=} P(U \leq F_X(x)) = F_X(x)$

so X & $F_X^{-1}(U)$ have same CDFs i.e., $X \stackrel{d}{=} F_X^{-1}(U)$
 (equal in distribution)

(2) Similarly, $x \in (0,1)$

$$P(F_X(X) \leq x) \stackrel{\uparrow F_X \text{ strictly } \uparrow}{=} P(X \leq F_X^{-1}(x)) = F_X(F_X^{-1}(x)) \stackrel{\leftarrow F_X \text{ is strictly } \uparrow}{=} x = P(U \leq x)$$

$$\Rightarrow F_X(X) \stackrel{d}{=} U$$

Applications: output of RAND is U . To generate X , you take $F_X^{-1}(\text{RAND})$.

Qn: Suppose X is a discrete r.v. (For eg. X has pmf p on $\{1, \dots, n\}$)
 so there $h: \mathbb{R} \rightarrow \mathbb{R} \Rightarrow h(U) \stackrel{d}{=} X$?

Defn
27.2

X be cts r.v. We define mean / Expectation of X as

$$E[X] := \int_{-\infty}^{\infty} x f_X(x) dx \quad \text{if} \quad \int_{-\infty}^{\infty} |x| f_X(x) dx < \infty.$$

Else undefined or $E[X]$ doesn't exist.

$$[X \text{ Discrete, } E[X] = \sum_x x p_X(x) \quad \text{if} \quad \sum_x |x| p_X(x) < \infty]$$

Note $x \mapsto x f_X(x)$ is p.cts.

Ex
27.3

$$(1) \quad U \triangleq U(0,1). \quad E[X] = \int_0^1 x dx = \frac{1}{2}.$$

$$[\text{check } \int_{-\infty}^{\infty} |x| f_X(x) dx = \int_0^1 x dx < \infty]$$

$$(2) \quad X \triangleq \text{EXP}(\lambda) \quad E[X] = \lambda \int_0^{\infty} x e^{-\lambda x} dx \stackrel{\text{IBP}}{=} \frac{1}{\lambda}.$$

check $E[X]$ exists

$$(3) \quad X \triangleq N(0,1)$$

$$\text{Check } \int_{-\infty}^{\infty} |x| f_X(x) dx < \infty.$$

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

$$\left[\underset{f_X \text{ even}}{=} 2 \int_0^{\infty} x f_X(x) dx \quad ; \quad f_X(x) \leq \frac{C}{x^3} \quad \forall x > 0 \right]$$

$$= \int_0^{\infty} x f_X(x) dx + \int_{-\infty}^0 x f_X(x) dx$$

$$= \int_0^{\infty} x f_X(x) dx + \int_0^{\infty} x f_X(-x) dx \quad \xrightarrow{\text{Cov } x \mapsto -x}$$

$$= \int_0^{\infty} x f_X(x) dx - \int_0^{\infty} x f_X(x) dx$$

$$= \int_0^{\infty} x f_X(x) dx - \int_0^{\infty} x f_X(x) dx$$

$$= 0$$

(since f_X is even i.e., $f_X(x) = f_X(-x)$)

• If f_X is even & $\int_{-\infty}^{\infty} x f_X(x) dx < \infty$ then $E[X] = 0$.
(Same proof as above)

• If $\int_{-\infty}^{\infty} |x| f_X(x) dx < \infty$ then $\lim_{\substack{a \rightarrow -\infty \\ b \rightarrow \infty}} \int_a^b x f_X(x) dx = \int_{-\infty}^{\infty} x f_X(x) dx$

If $\sum_{n=1}^{\infty} |a_n| < \infty$ then $\lim_{n \rightarrow \infty} \sum_{l=1}^n a_l$ exists $(a_n = (-1)^n, n \geq 0)$

Ex
27.4

$X \stackrel{d}{=} \text{Cauchy}$

$$f_X(x) = \frac{1}{\pi} \frac{1}{1+x^2}$$

f_X is even. $\int_{-\infty}^{\infty} |x| f_X(x) dx = \frac{2}{\pi} \int_0^{\infty} \frac{x}{1+x^2} dx = \infty!$
check.

So $E[X]$ doesn't exist.

(Compare with $\sum_{n=1}^{\infty} \frac{1}{n^2}$)

Def
27.5

Let X be cts r.v. & $h: \mathbb{R} \rightarrow \mathbb{R}$ p-cts.

$$E[h(X)] = \int_{-\infty}^{\infty} h(x) f_X(x) dx \quad \text{if } \int_{-\infty}^{\infty} |h(x)| f_X(x) dx < \infty.$$

else undefined / $E[h(X)]$ doesn't exist.

VARIANCE. $VAR[X] := E[(X - \mu_X)^2]$ if $E[X^2]$ exists

where $\mu_X := E[X]$.

k^{th} MOMENT - $E[X^k]$, $k=1, 2, \dots$ if it exists.

THM
27.6

(1) If $E[X]$ exists then $\forall a, b \in \mathbb{R}$

$$E[ax + b] = aE[X] + b$$

(2) If $h_1 \leq h_2$ and $E[h_i(X)]$ exists

$$\text{then } E[h_1(X)] \leq E[h_2(X)]$$

(Monotonicity)

(3) if $E[X^2]$ exists

$$\text{VAR}[X] = E[X^2] - \mu_X^2$$

(4) if $E[X^2]$ exists, then $\forall a, b \in \mathbb{R}$.

$$\text{VAR}[aX+b] = a^2 \text{VAR}[X]$$

Proof: (1) To check $\int_{-\infty}^{\infty} |ax+b| f_X(x) dx < \infty$

observe $|ax+b| \leq |a||x| + |b|$. ('triangle ineq')

$$\begin{aligned} \int_{-\infty}^{\infty} |ax+b| f_X(x) dx &\leq \int_{-\infty}^{\infty} (|a||x| + |b|) f_X(x) dx \\ &\quad \text{(mon. for non-neg. integrals)} \leftarrow \\ &\quad \text{(linearity of integrals)} \leq |a| \int_{-\infty}^{\infty} |x| f_X(x) dx + |b| \int_{-\infty}^{\infty} f_X(x) dx \end{aligned}$$

$$\begin{aligned} \text{Since } E[X] \text{ exists } &\leftarrow < \infty & + |b| < \infty. \\ & & \rightarrow f_X \text{ is pdf} \end{aligned}$$

so $E[aX+b]$ exists.

$$\text{by defn } E[aX+b] = \int_{-\infty}^{\infty} (ax+b) f_X(x) dx$$

$$\text{(lin. of integrals)} = a \int_{-\infty}^{\infty} x f_X(x) dx + b \int_{-\infty}^{\infty} f_X(x) dx$$

$$\text{(defn of } E[X] \text{ \& } f_X \text{ is a pdf)} = a E[X] + b$$

(2) Ex.

$$(3) \text{VAR}[X] = \int_{-\infty}^{\infty} (x - \mu_X)^2 f_X(x) dx \quad (\text{defn})$$

$$\begin{aligned} &= \int_{-\infty}^{\infty} x^2 f_X(x) dx - 2\mu_X \int_{-\infty}^{\infty} x f_X(x) dx + \mu_X^2 \int_{-\infty}^{\infty} f_X(x) dx \\ &\quad \text{(linearity of integral)} \end{aligned}$$

$$= E[X^2] - 2\mu_X^2 + \mu_X^2 \quad (\text{by defns})$$

$$= E[X^2] - \mu_x^2.$$

(4) Let $Y = aX + b$.

From (1), $\mu_Y = a\mu_X + b$

$$\int_{-\infty}^{\infty} (ax+b)^2 f_X(x) dx \leq a^2 \int_{-\infty}^{\infty} x^2 f_X(x) dx + 2|b| \int_{-\infty}^{\infty} |x| f_X(x) dx + b^2$$

$< \infty$ as $E[X^2], E[X]$ exist.

so $E[(aX+b)^2]$ exists.

$$\Rightarrow \text{VAR}[aX+b] = \int_{-\infty}^{\infty} (ax+b - a\mu_X - b)^2 f_X(x) dx$$

(defn & $\mu_Y = a\mu_X + b$)

$$= a^2 \int_{-\infty}^{\infty} (x - \mu_X)^2 f_X(x) dx = a^2 \text{VAR}[X]. \quad \square$$

Note:

$$Y = aX + b$$

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

(1) of thm \rightarrow

$$E[aX+b] = E[Y]$$

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

$$= \int_{-\infty}^{\infty} (ax+b) f_X(x) dx$$

(change $\frac{y-b}{a} \rightarrow x$; $dy = a dx$)

So defn. of $E[h(X)]$ is consistent in the simple case of $h(x) = ax + b$!!!

[for more gen. case. say $Y = h(X)$
then $E[Y] = E[h(X)]$ follows by a gen. Cov]

Eg
27.7

$$(1) \quad U \quad E[U] = \frac{1}{2}$$

$$k \geq 1 \quad E[U^k] = \int_0^1 x^k dx = \frac{1}{k+1}$$

$$VAR[U] = E[U^2] - (E[U])^2 = \frac{1}{3} - \frac{1}{4} = \frac{1}{12}$$

$$(2) \quad X \stackrel{d}{=} U(b, b+a) \quad f_X(x) = \begin{cases} \frac{1}{a} & x \in (b, b+a) \\ 0 & \text{elsewhere} \end{cases} \quad \begin{matrix} a > 0 \\ b \in \mathbb{R} \end{matrix}$$

From previous class.

$$X \stackrel{d}{=} Y, \quad Y = \underline{a}U + b$$

$$E[X] = E[Y] \stackrel{\text{Thm 27.6(4)}}{=} \underset{\text{Thm 27.6(1)}}{a} E[U] + b = \frac{a}{2} + b$$

$$VAR[X] \stackrel{\text{Thm 27.6(4)}}{=} a^2 VAR[U] = \frac{a^2}{12}$$

$$(3) \quad X \stackrel{d}{=} \text{EXP}(\lambda) \quad E[X] = \frac{1}{\lambda}$$

$$E[X^2] = \lambda \int_0^{\infty} x^2 e^{-\lambda x} dx = \frac{2}{\lambda^2}$$

$$VAR[X] = \frac{2}{\lambda^2} - \left(\frac{1}{\lambda}\right)^2 = \frac{1}{\lambda^2}$$

$$(4) \quad Z \stackrel{d}{=} N(0,1) \quad X \stackrel{d}{=} N(\mu, \sigma) \quad \sigma > 0$$

Last class, if $Y = \sigma Z + \mu$, then $Y \stackrel{d}{=} X$.

$$E[X] = E[Y] = \sigma E[Z] + \mu = \mu$$

$$VAR[X] = \sigma^2 VAR[Z]$$

$$VAR[Z] = E[Z^2] \quad (E[Z] = 0)$$

$$= \int_{-\infty}^{\infty} z^2 f_Z(z) dz$$

$$\begin{aligned}
&= \frac{2}{\sqrt{2\pi}} \int_0^{\infty} z^2 e^{-z^2/2} dz \\
&= \frac{1}{\sqrt{2\pi}} \int_0^{\infty} x^{1/2} e^{-x/2} dx \quad \leftarrow x = z^2 \quad dz = \frac{dx}{2\sqrt{x}} \\
&\stackrel{(?) }{=} \frac{1}{\sqrt{2\pi}} \Gamma(3/2) = 1.
\end{aligned}$$

$$\text{So } \text{VAR}[X] = \sigma^2 \text{VAR}[Z] = \sigma^2.$$

$$X \stackrel{d}{=} N(\mu, \sigma) ; \quad \mu = E[X], \quad \sigma^2 = \text{VAR}[X]$$

$$\sigma = \text{SD}[X] = \sqrt{\text{VAR}[X]}.$$

Exercises compute $E[X^k]$ in exp., $N(\mu, \sigma)$, ...

Relab $X_n \stackrel{d}{=} \text{Bin}(n, \frac{\lambda}{n}) \quad \lambda > 0 \quad E X_n = \lambda, \text{Var}[X_n] = \lambda(1 - \frac{\lambda}{n})$

$$k=0,1, \dots \quad P(X_n = k) \longrightarrow P(X = k) \quad X \stackrel{d}{=} \text{Poi}(\lambda)$$

↳ Poisson limit theorem / law of small numbers
"David Hand!"

$$S_n \stackrel{d}{=} \text{Bin}(n, p) \quad p \in (0, 1]$$

$$E[S_n] = np \quad \text{VAR}[S_n] = np(1-p).$$

$a, b \in \mathbb{R}$

$$P\left(a < \frac{S_n - np}{\sqrt{np(1-p)}} < b\right) \stackrel{d}{\longrightarrow} P\left(N \in (a, b)\right)$$

$\stackrel{\text{i.i.d.}}{\parallel} N(0, 1)$

De-Moivre Laplace central limit theorem!

Ans:

(1) Given a seq r.v. X_n , $E[X_n] \rightarrow \lambda$, $VAR[X_n] \rightarrow \lambda$
Can one show Poisson limit theorem?

(2) Given a seq r.v. S_n , $E[S_n] \rightarrow \infty$, $VAR[S_n] \rightarrow \infty$

is it true $P(a < \frac{S_n - E[S_n]}{\sqrt{VAR[S_n]}} < b) \rightarrow P(N \in (a, b))$
(CLT) ?

Try to show U !

THANK U