

11) Typically the Lebesgue integral is PM 49
the Riemann integral, and

$$\int_{-\infty}^{\infty} |f(x)| dx = \int_{-\infty}^{\infty} f(x) dx = 1 \quad \text{so the}$$

pdf $f(x)$ is integrable.

12] Recall that RVS X_1, \dots, X_K are ind see 14]
iff $P[X_1 \in B_1, \dots, X_K \in B_K] = \prod_{i=1}^K P(X_i \in B_i)$

for any $B_1, \dots, B_K \in \mathcal{B}(\mathbb{R})$.

iff $F(x_1, \dots, x_K) = F_{X_1}(x_1) \cdots F_{X_K}(x_K)$ for all
real x_1, \dots, x_K .

If X_1, \dots, X_K have ^{joint} pdf $f(x)$, then

X_1, \dots, X_K are ind iff $f(x) =$

$f_{X_1}(x_1) \cdots f_{X_K}(x_K)$ for all real (x_1, \dots, x_K) .

3.18 Product measure and Fubini's Th.

13] Let $(\Omega_1, \mathcal{F}_1, P_1)$ and $(\Omega_2, \mathcal{F}_2, P_2)$
be two prob spaces.

Let the Cartesian product = cross product 49.5

$$\Omega_1 \times \Omega_2 = \{(\omega_1, \omega_2) : \omega_1 \in \Omega_1, \omega_2 \in \Omega_2\}$$

The product of \mathcal{F}_1 and \mathcal{F}_2 , denoted by $\mathcal{F}_1 \times \mathcal{F}_2$,

is $\mathcal{F}_1 \times \mathcal{F}_2 = \sigma(A)$ where

$$A = \{ \underbrace{A_1 \times A_2}_{\text{cross products}} : A_1 \in \mathcal{F}_1, A_2 \in \mathcal{F}_2 \}.$$

(4) Th: There is a unique probability

$$P = P_1 \times P_2 \text{ such that } P(A_1 \times A_2) = P_1(A_1) P_2(A_2)$$

for all $A_1 \in \mathcal{F}_1$ and $A_2 \in \mathcal{F}_2$.

(5) $P = P_1 \times P_2$ is the product of P_1 and P_2 .

The product probability space is

$$(\Omega_1 \times \Omega_2, \mathcal{F}_1 \times \mathcal{F}_2, P_1 \times P_2)$$

(6) The above can be extended to
 $(\Omega_i, \mathcal{F}_i, P_i), i = 1, \dots, n$.

Denote $P_1 \times \dots \times P_n$ by $\prod_{i=1}^n P_i$.

50

ex) If $(\Omega_i, \mathcal{F}_i, P_i) = (\mathbb{R}, \mathcal{B}(\mathbb{R}), P_i)$

then the product space is $(\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n), \prod_{i=1}^n P_i)$.

If $(\Omega_i, \mathcal{F}_i, P_i) = (\mathbb{R}, \mathcal{B}(\mathbb{R}), P_{X_i})$

then the product space is $(\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n), \prod_{i=1}^n P_{X_i})$.

14) Let X_i be defined on $(\mathbb{R}, \mathcal{B}(\mathbb{R}), P_{X_i})$

X_1, \dots, X_n ind. Then

$$F_{\underline{X}}(\underline{t}) = P(X_1 \leq t_1, \dots, X_n \leq t_n) =$$

$$P((-\infty, t_1] \times \dots \times (-\infty, t_n]) = \prod_{i=1}^n P_{X_i}((-\infty, t_i])$$

$$= \prod_{i=1}^n F_{X_i}(t_i) \quad \text{where } P = \prod_{i=1}^n P_{X_i}.$$

Hence $(\Omega, \mathcal{F}, P) = (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n), \prod_{i=1}^n P_{X_i})$

is the prob space for $\underline{X} = (X_1, \dots, X_n)$.

$$\begin{aligned} P_{X_i}(B) &= P(X_i \in B) = P(\mathbb{R} \times \dots \times \mathbb{R} \times B \times \mathbb{R} \times \dots \times \mathbb{R}) \\ &= P_{X_1}(\mathbb{R}) \dots P_{X_{i-1}}(\mathbb{R}) P_{X_i}(B) P_{X_{i+1}}(\mathbb{R}) \dots P_{X_n}(\mathbb{R}). \end{aligned}$$

$$15] \text{ Let } \int f d\mu = \int f(x) d\mu(x).$$

$$\text{Then } \iint_{\Omega_1 \times \Omega_2} f(x_1, x_2) d[\mathbb{P}_1 \times \mathbb{P}_2(x_1, x_2)]$$

$$= \int_{\Omega_1} \left[\int_{\Omega_2} f(x_1, x_2) d\mathbb{P}_2(x_2) \right] d\mathbb{P}_1(x_1)$$

$$= \int_{\Omega_2} \left[\int_{\Omega_1} f(x_1, x_2) d\mathbb{P}_1(x_1) \right] d\mathbb{P}_2(x_2).$$

16] Fubini's Theorem; a) Assume $f \geq 0$.

Then $\int_{\Omega_2} f(x_1, x_2) d\mathbb{P}_2(x_2)$ is measurable $\sigma \mathcal{F}_1$,

$\int_{\Omega_1} f(x_1, x_2) d\mathbb{P}_1(x_1)$ is measurable $\sigma \mathcal{F}_2$,

and 15] holds.

b) Assume f is integrable wrt $\mathbb{P}_1 \times \mathbb{P}_2$.

Then $\int_{\Omega_2} f(x_1, x_2) d\mathbb{P}_2(x_2)$ is finite a.e and measurable $\sigma \mathcal{F}_1$ a.e,

$\int_{\Omega_1} f(x_1, x_2) d\mathbb{P}_1$ is finite a.e and measurable $\sigma \mathcal{F}_2$ a.e, and 15] holds.

17] 16a) = Tonelli's Th

p238] 8] If f is not nonnegative, compute one of the integrals in 16] with $|f|$ replacing f . If the result is finite, compute one of the ~~double~~^{iterated} integrals in 15] with f . (If the result is infinite, then f is not integrable wrt $P_1 \times P_2$.)

19] The exchange of integrals 15] holds if $f \geq 0$ or if $\int \int |f(x,y)| d[P_1 \times P_2] < \infty$.

20] Fubini's th allows the computation of $E[h(x_1, x_2)]$ using integrals in 15] if x_1 and x_2 are independent.

21] A Product measure μ satisfies

$$\mu\left(\prod_{i=1}^n A_i\right) = \prod_{i=1}^n \mu_i(A_i).$$

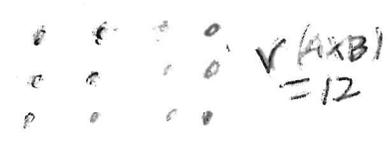
ex) i) $\mu = \prod_{i=1}^n P_{x_i}$ when the x_i 's are ind.

60.5/5

ii) Let λ be the Lebesgue measure on \mathbb{R}^2
and μ_L

Then $\lambda(A \times B) = \mu_L(A) \mu_L(B)$, $A, B \in \mathcal{B}(\mathbb{R})$.

iii) Let ν be the counting measure on \mathbb{Z}^2
 μ_c

$\mu_c(A) = \#$ of points in A . 

Then $\nu(A \times B) = \mu_c(A) \mu_c(B)$, $A, B \subseteq \mathbb{Z}$.

(counting measure μ_c is σ -finite if \mathcal{A} is countable, p158)

22) Fubini's theorem holds for product measures $\mu = \prod \mu_i$ if the μ_i are σ -finite.

23) Fubini's theorem for Lebesgue integrals:

Let $C = \{(x, y) : a \leq x \leq b, c \leq y \leq d\} = [a, b] \times [c, d]$.

Let $g(x, y)$ be measurable and Lebesgue integrable. Then

$$\iint_C g(x, y) dx dy = \int_c^d \left[\int_a^b g(x, y) dx \right] dy$$
$$= \int_a^b \left[\int_c^d g(x, y) dy \right] dx.$$

24] The result in 23] can be extended to where the limits of integration are infinite and to $n > 2$ integrals. Using $g(x, y) = h(x, y) f(x, y)$ where f is a pdf gives $E[h(x, y)]$.

Independent RVs

p267 25] RVs X_1, \dots, X_n are independent if

$$P(X_1 \in B_1, \dots, X_n \in B_n) = \prod_{i=1}^n P(X_i \in B_i)$$

for any $B_i \in \mathcal{B}(\mathbb{R})$. An infinite set of RVs is independent if each finite subset is ind.

26] X_1, \dots, X_n are ind

a) iff
$$F_{X_1, \dots, X_n}(t_1, \dots, t_n) = \prod_{i=1}^n F_{X_i}(t_i) \quad \forall t_1, \dots, t_n \in \mathbb{R}.$$

b) If pdfs exist, iff

$$f_{X_1, \dots, X_n}(t_1, \dots, t_n) = f_{X_1}(t_1) \cdots f_{X_n}(t_n) \quad \forall t_1, \dots, t_n \in \mathbb{R}.$$

c) If pmfs exist, iff

$$P_{X_1, \dots, X_n}(t_1, \dots, t_n) = P_{X_1}(t_1) \cdots P_{X_n}(t_n) \quad \forall t_1, \dots, t_n \in \mathbb{R}$$

52.5

Note: if the discrete RVs X_1, \dots, X_n take on values in a countable set \mathcal{F} , then X_1, \dots, X_n are ind

$$\text{iff } P_{X_1, \dots, X_n}(x_1, \dots, x_n) = \prod_{i=1}^n P_{X_i}(x_i) \quad \forall x_1, \dots, x_n \in \mathcal{F}.$$

P267
27] If X_1, \dots, X_n are ind and $\{\omega : \underbrace{X(\omega)} \in C\}$

$$\underline{X} = (X_1, \dots, X_n) \text{ has dist } P_{\underline{X}}(C) = P(\underline{X}^{-1}(C))$$

for $C \in \mathcal{B}(\mathbb{R}^n)$, then $P_{\underline{X}}(B_1 \times \dots \times B_n)$
cross product

$$= P(\underline{X} \in B_1 \times \dots \times B_n) = P(X_1 \in B_1, \dots, X_n \in B_n)$$

$$= \prod_{i=1}^n P(X_i \in B_i) = \prod_{i=1}^n P_{X_i}(B_i).$$

Thus $P_{\underline{X}}$ is a product probability.

P284
28] $E[g_1(X_1) \dots g_n(X_n)] = E[g_1(X_1)] \dots E[g_n(X_n)]$

if g_i is a function of X_i alone, X_1, \dots, X_n are ind, and the expected values exist.

29] Suppose $X \perp\!\!\!\perp Y$. Then for $t \in \mathbb{R}$,

$$F_{X+Y}(t) = \int F_X(t-y) dF_Y(y) = \int F_Y(t-x) dF_X(x)$$

$$\stackrel{\text{cdf}}{=} E_Y[F_X(t-y)] = E_X[F_Y(t-x)].$$

If X has pdf $f_X(x)$ and Y has pdf $g_Y(y)$,

then $\stackrel{\text{pdf}}{F_{X+Y}}(t) = \int_{-\infty}^{\infty} f_X(t-y) g_Y(y) dy = E_Y[F_X(t-y)]$

$$= \int_{-\infty}^{\infty} g_Y(t-x) f_X(x) dx = E_X[g_Y(t-x)]$$

Convolution formulas

Large Sample Theory

Approximating the dist of \bar{X}_n , CLT, WLLN
 SLLN, characteristic function $\frac{D}{L} \rightarrow \frac{L}{\text{mean}}$

→
 wpi
 ae
 as

p367 1] ^{know} Central Limit Theorem (CLT): 53.5

Let X_1, \dots, X_n be iid with mean

$E(X_i) = \mu$ and variance $V(X_i) = \sigma^2$.

Let the sample mean $\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$.

Then $\sqrt{n} (\bar{X}_n - \mu) \xrightarrow{D} N(0, \sigma^2)$.

Note: Hence $\sqrt{n} \left(\frac{\bar{X}_n - \mu}{\sigma} \right) = \underbrace{\frac{\bar{X}_n - \mu}{\sigma/\sqrt{n}}}_{\text{z score of } \bar{X}_n} = \sqrt{n} \left(\frac{\sum_{i=1}^n X_i - n\mu}{n\sigma} \right)$

$= \underbrace{\left(\frac{\sum_{i=1}^n X_i - n\mu}{n\sigma} \right)}_{\text{z score of } \sum_{i=1}^n X_i} \xrightarrow{D} N(0, 1)$.

Note: The CLT will be proved later.

§14, 25

2] ^{Def} If F_n and F are cdfs,

F_n converges weakly to F , written $F_n \xrightarrow{w} F$,

if $\lim_n F_n(x) = F(x)$ at every continuity

point of F .

p192
335

p338 3] know Def Let X_n and X be RVs with cdfs F_n and F .

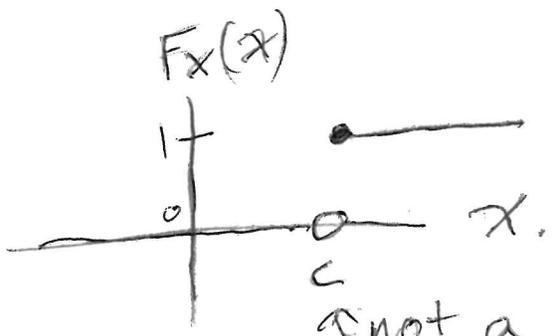
X_n converges in distribution to X or X_n converges in law to X , written $X_n \xrightarrow{D} X$ or $X_n \xrightarrow{L} X$,

iff F_n converges weakly to F :

$\lim_{n \rightarrow \infty} F_n(x) = F(x)$ at every continuity point of F .

4] X has a point mass at c if

$P(X=c) = 1$. Then $F_x(x) = \begin{cases} 1 & x \geq c \\ 0 & x < c \end{cases}$



Often $c = \mu$ or θ .

↑ not a continuity point of F_x

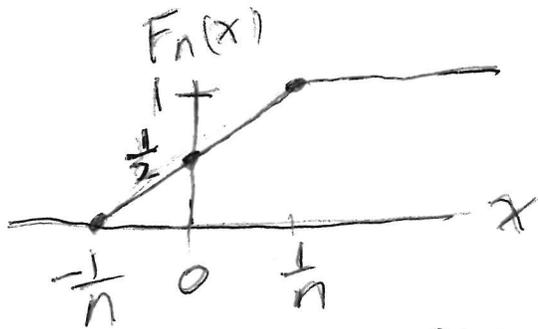
p339 5] $X_n \xrightarrow{D} X$ iff $\lim_{n \rightarrow \infty} P[X_n \leq x] = P[X \leq x]$

for every x such that $P[X=x] = 0$.

no jump in the cdf at x

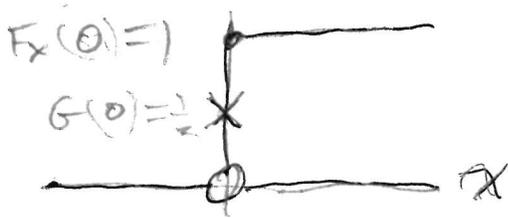
ex] suppose $X_n \sim U(-\frac{1}{n}, \frac{1}{n})$. Then

$$F_n(x) = \begin{cases} 0 & x \leq -\frac{1}{n} \\ \frac{nx}{2} + \frac{1}{2} & -\frac{1}{n} \leq x \leq \frac{1}{n} \\ 1 & x \geq \frac{1}{n} \end{cases} \quad F_n(0) = \frac{1}{2} \forall n$$



$$F_n(x) \rightarrow \begin{cases} 0 & x < 0 \\ \frac{1}{2} & x = 0 \\ 1 & x > 0 \end{cases} = G(x)$$

$F_X(x)$ and $G(x)$



$$\text{so } F_n(x) \rightarrow F_X(x) = \begin{cases} 0 & x < 0 \\ 1 & x \geq 0 \end{cases}$$

↑
pointmass at 0

for each continuity point of X .

Hence $X_n \xrightarrow{D} X$. The discontinuity

point of $F_X(x)$ is at $x=0$, and

$\lim_{n \rightarrow \infty} F_n(0)$ is unimportant for convergence in distribution (the limit need not exist and need not be equal to $F_X(0)$).

↑
discontinuity point
of $F_X(x)$

ex] $X_n \sim U(0, n)$. Then $F_n(x) = \begin{cases} 0 & x \leq 0 \\ \frac{x}{n} & 0 \leq x \leq n \\ 1 & x \geq n \end{cases}$

$\lim_{n \rightarrow \infty} F_n(x) = 0 \equiv H(x) \forall x \in \mathbb{R}$.

X_n does not converge in dist to a RV X .

since $H(x) \equiv 0$ is not a cdf and can't be changed to a cdf by modifying a countable number of discontinuity points.

6] If $X_n \xrightarrow{D} X$, then X is the limiting distribution or asymptotic distribution of X_n . Note that the distribution of X does not depend on n .
& sample size

7] For the CLT, $X \sim N(0, \sigma^2)$.

ex] If X_1, \dots, X_n are iid $U(0, 1)$, find the limiting distribution of $\sqrt{n}(\bar{X}_n - c)$ for appropriate c . (In 580 quals. (stat))
 the question often asks for the limiting distribution of \bar{X}_n .

So/n] $E(X_i) = \frac{0+1}{2} = \frac{1}{2}$, $V(X_i) = \frac{(1-0)^2}{12} = \frac{1}{12}$ 55.5

so $\sqrt{n} (\bar{X}_n - \frac{1}{2}) \xrightarrow{D} \underbrace{N(0, \frac{1}{12})}_X$ by CLT.

8) Several distributions are such that

$X_n = \sum_{i=1}^n Y_i$ where Y_i are iid.

a) $X_n \sim \text{bin}(nk, p)$, $X_n = \sum_{i=1}^n Y_i$, $Y_i \stackrel{iid}{\sim} \text{bin}(k, p)$

b) $X_n \sim \text{poisson}(na)$, $Y_i \stackrel{iid}{\sim} \text{poisson}(a)$

c) $X_n \sim \chi_{np}^2$, $Y_i \stackrel{iid}{\sim} \chi_p^2$, $E(Y_i) = p$, $V(Y_i) = 2p$

d) $X_n \sim \text{Gamma}(na, B)$, $Y_i \stackrel{iid}{\sim} \text{Gamma}(\alpha, B)$, $E(Y_i) = \frac{\alpha}{B}$, $V(Y_i) = \frac{\alpha}{B^2}$

e) $X_n \sim \text{NB}(nr, p)$, $Y_i \stackrel{iid}{\sim} \text{NB}(r, p)$, $E(Y_i) = \frac{r(1-p)}{p}$, $V(Y_i) = \frac{r(1-p)}{p^2}$

f) $X_n \sim N(n\mu, n\sigma^2)$, $Y_i \stackrel{iid}{\sim} N(\mu, \sigma^2)$

g) $X_n \sim \text{IG}(n\theta, n^2\lambda)$, $Y_i \stackrel{iid}{\sim} \text{IG}(\theta, \lambda)$, $E(Y_i) = \theta$, $V(Y_i) = \frac{\theta^3}{\lambda}$

Show with mgfs or characteristic functions.