

THM: ① \Leftrightarrow ②, WHERE

① $\mathbb{X}_n \Rightarrow \mathbb{X}$

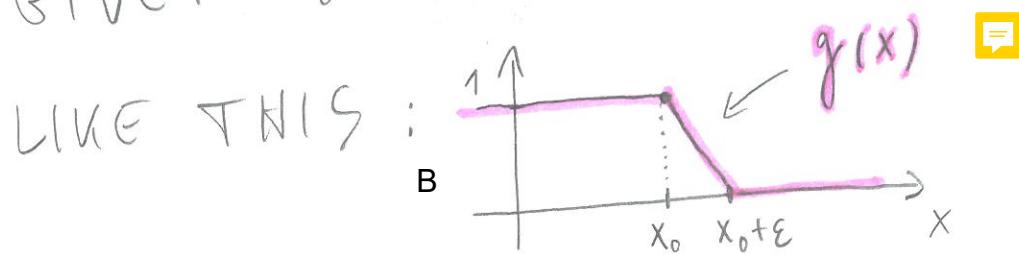
② $E(g(\mathbb{X}_n)) \rightarrow E(g(\mathbb{X}))$ FOR ANY
BOUNDED & CONTINUOUS $g: \mathbb{R} \rightarrow \mathbb{R}$

PROOF OF ② \Rightarrow ①: ENOUGH TO SHOW: $\forall x_0 \in \mathbb{R}, \forall \epsilon > 0:$

A $F(x_0 - \epsilon) \leq \liminf_{n \rightarrow \infty} F_n(x_0) \leq \limsup_{n \rightarrow \infty} F_n(x_0) \leq F(x_0 + \epsilon)$

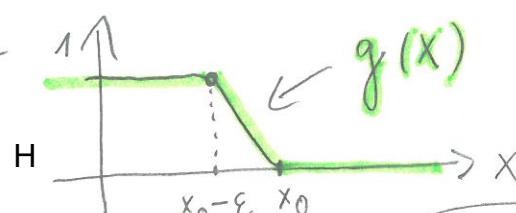
WHERE $F_n(x) = P(\mathbb{X}_n \leq x)$, $F(x) = P(\mathbb{X} \leq x)$

GIVEN x_0 AND ϵ , LET US DEFINE $g: \mathbb{R} \rightarrow \mathbb{R}$



$$\begin{aligned} \limsup_{n \rightarrow \infty} F_n(x_0) &\stackrel{\text{C}}{=} \limsup_{n \rightarrow \infty} E(\mathbb{1}[\mathbb{X}_n \leq x_0]) \stackrel{\text{D}}{\leq} \lim_{n \rightarrow \infty} E(g(\mathbb{X}_n)) \stackrel{\text{E}}{=} \\ &= E(g(\mathbb{X})) \stackrel{\text{F}}{\leq} E(\mathbb{1}[\mathbb{X} \leq x_0 + \epsilon]) = F(x_0 + \epsilon) \checkmark \end{aligned}$$

G FOR THE OTHER BOUND, REPEAT THIS
ARGUMENT USING



PROOF OF $\textcircled{1} \Rightarrow \textcircled{2}$: GIVEN g , ENOUGH TO SHOW:

A $\forall \varepsilon > 0 : \limsup_{n \rightarrow \infty} |\mathbb{E}(g(\hat{x}_n)) - \mathbb{E}(g(\hat{x}))| \leq 4\varepsilon$

LET US FIX $g: \mathbb{R} \rightarrow \mathbb{R}$, $M := \|g\|_\infty = \sup_x |g(x)|$

B USE TIGHTNESS OF $(\hat{x}_n)_{n=1}^\infty$ TO CHOOSE $K \in \mathbb{R}$

SUCH THAT $\mathbb{P}(|\hat{x}_n| \geq K) \leq \frac{\varepsilon}{M}$, $n=1, 2, 3, \dots$

MOREOVER $\mathbb{P}(\hat{x} = K) = \mathbb{P}(\hat{x} = -K) = 0$

E NOTE THAT g IS UNIFORMLY CONTINUOUS

ON $[-K, K]$, SO CHOOSE $\delta > 0$ SUCH THAT

F1 $\forall x, y \in [-K, K], |x-y| \leq \delta \Rightarrow |g(x)-g(y)| \leq \varepsilon$

CHOOSE $-K = x_0 < x_1 < \dots < x_N = K$ SUCH THAT

F2 $|x_{j+1} - x_j| \leq \delta$ AND $\mathbb{P}(\hat{x} = x_j) = 0, j = 0, 1, \dots, N$

H NOTATION: $\mathbb{E}(g(\hat{x}); A) := \mathbb{E}(g(\hat{x}) \cdot \mathbf{1}[A])$

NOTE:

$$|\mathbb{E}(g(\hat{x})) - \mathbb{E}(g(\hat{x}); -K < \hat{x} \leq K)| \leq$$

$$\leq \|g\|_\infty \cdot \mathbb{P}(|\hat{x}| \geq K) \leq M \cdot \frac{\varepsilon}{M} = \varepsilon$$

(SIMILARLY FOR \hat{x}_n)

AN EVENT

$$\text{THUS : } |\mathbb{E}(g(\hat{X}_n)) - \mathbb{E}(g(\hat{X}))| \stackrel{\text{A1}}{\leq} \square$$

$$\text{A2} |\mathbb{E}(g(\hat{X}_n); -K < \hat{X}_n \leq K) - \mathbb{E}(g(\hat{X}); -K < \hat{X} \leq K)| + 2\varepsilon$$

$$\text{B } \mathbb{E}(g(\hat{X}); -K < \hat{X} \leq K) = \sum_{r=1}^N \mathbb{E}(g(\hat{X}); X_{r-1} < \hat{X} \leq X_r)$$

$$\text{D } = \sum_{r=1}^N \mathbb{E}(g(X_r); X_{r-1} < \hat{X} \leq X_r) + \underbrace{\sum_{r=1}^N \mathbb{E}((g(\hat{X}) - g(X_r)); X_{r-1} < \hat{X} \leq X_r)}$$

$$\text{E } = \sum_{r=1}^N g(X_r) \cdot (F(X_r) - F(X_{r-1})) + \star$$

$$|\star| \leq \sum_{r=1}^N \mathbb{E}(|g(\hat{X}) - g(X_r)|; X_{r-1} < \hat{X} \leq X_r) \stackrel{\text{H}}{\leq} \varepsilon$$

$$\leq \sum_{r=1}^N \varepsilon \cdot P(X_{r-1} < \hat{X} \leq X_r) \stackrel{\text{J}}{\leq} \varepsilon, \text{ THUS } \square \square \square \square$$

$$\text{L } |\mathbb{E}(g(\hat{X}_n); -K < \hat{X}_n \leq K) - \mathbb{E}(g(\hat{X}); -K < \hat{X} \leq K)| \stackrel{\text{M}}{\leq}$$

$$\left| \sum_{r=1}^N g(X_r) \cdot (F_n(X_r) - F_n(X_{r-1})) - \sum_{r=1}^N g(X_r) \cdot (F(X_r) - F(X_{r-1})) \right| + 2\varepsilon \stackrel{\text{N}}{\leq}$$

$$\left| \sum_{r=1}^N g(X_r) \cdot (F_n(X_r) - F(X_r)) + \sum_{r=1}^N g(X_r) \cdot (\underbrace{F(X_{r-1}) - F_n(X_{r-1})}_{\rightarrow \infty}) \right| + 2\varepsilon$$

$\text{O} \cancel{\xrightarrow{n \rightarrow \infty}}$

\checkmark

PAGE 83

EX: LET

$$I_m := \int_0^1 \int_0^1 \dots \int_0^1 \frac{x_1^2 + \dots + x_m^2}{x_1 + \dots + x_m} dx_1 \dots dx_m$$

B $\lim_{n \rightarrow \infty} I_m = ?$

SOLUTION: LET X_1, X_2, \dots I.I.D. UNI $[0, 1]$

$$\text{THEN } I_m = \mathbb{E} \left(\frac{X_1^2 + \dots + X_m^2}{X_1 + \dots + X_m} \right)$$

$$\text{LET } Y_m := \frac{X_1^2 + \dots + X_m^2}{m} \quad Z_m := \frac{X_1 + \dots + X_m}{m}$$

WEAK LAW OF LARGE NUMBERS:

$$\begin{array}{l|l} \text{D} \quad Y_m \Rightarrow \mathbb{E}(X_1^2) = \int_0^1 x^2 dx = \frac{1}{3} & \text{F} \quad I_m = \mathbb{E}\left(\frac{Y_m}{Z_m}\right) \\ \text{E} \quad Z_m \Rightarrow \mathbb{E}(X_1) = \int_0^1 x dx = \frac{1}{2} & \end{array}$$

$$\text{SLUTSKY: } \frac{Y_m}{Z_m} \stackrel{\text{G}}{\Rightarrow} \frac{1/3}{1/2} = \frac{2}{3}$$

$$\text{NOTE: } 0 \leq \frac{X_1^2 + \dots + X_m^2}{X_1 + \dots + X_m} \stackrel{\text{H}}{\leq} \frac{X_1 + \dots + X_m}{X_1 + \dots + X_m} = 1$$

$$\text{THUS IF } g(x) := \begin{cases} 0 & \text{IF } x \leq 0 \\ x & \text{IF } 0 \leq x \leq 1 \\ 1 & \text{IF } x \geq 1 \end{cases} \text{ THEN}$$

$$I_m = \mathbb{E}\left(g\left(\frac{Y_m}{Z_m}\right)\right) \xrightarrow{\text{J}} \mathbb{E}\left(g\left(\frac{2}{3}\right)\right) \stackrel{\text{K}}{=} \frac{2}{3}$$

(SINCE g IS BOUNDED, CONTINUOUS)

PAGE 84

DEF: CHARACTERISTIC FUNCTION OF THE

RANDOM VARIABLE X :

$$\varphi: \mathbb{R} \rightarrow \mathbb{C} \quad \varphi(t) := \mathbb{E}(e^{itX})$$

NOTE: $e^{itX} = \cos(tx) + i \cdot \sin(tx)$, THUS

$$\varphi(t) = \mathbb{E}(\cos(tx)) + i \cdot \mathbb{E}(\sin(tx))$$

IF X IS ABS. CONT. WITH P.D.F. f , THEN

$$\varphi(t) = \int_{-\infty}^{\infty} e^{itx} \cdot f(x) dx$$

NOTE: IF THE MOMENT GENERATING FUNCTION

$$z(\lambda) = \mathbb{E}(e^{\lambda X})$$
 IS FINITE FOR $\lambda \in (-R, R)$

THEN BY $e^{a+bi} = e^a \cdot (\cos(b) + i \cdot \sin(b))$

WE HAVE $|e^{a+bi}| = e^a$, THUS

$$|z(a+bi)| \leq z(a) < +\infty \text{ IF } a \in (-R, R)$$

THUS $|\varphi(t)| < +\infty$ IF $\operatorname{Im}(t) \in (-R, R)$

AND φ IS AN ANALYTIC FUNCTION ON

THE STRIP $\{t \in \mathbb{C} : \operatorname{Im}(t) \in (-R, R)\}$

IN PARTICULAR: $\varphi(t) = z(it)$

PAGE 85

IN PARTICULAR:

$$X \sim N(0, 1) \Rightarrow \mathbb{E}(e^{\lambda X}) = e^{\lambda^2/2} \quad (\text{SEE PAGE 14})$$

SO $\Psi(t) = \mathbb{E}(e^{itX}) \stackrel{\text{A}}{=} e^{-t^2/2}$

$$X \sim EXP(1) \Rightarrow \mathbb{E}(e^{\lambda X}) = \frac{1}{1-\lambda} \quad \text{IF } \lambda < 1$$

SO $\Psi(t) \stackrel{\text{B}}{=} \frac{1}{1-i\cdot t}, t \in \mathbb{R}$

SEE
HW 2.1(d)

NOTE: IF THE DISTRIBUTION OF X IS

SYMMETRIC, I.E. IF $X \sim (-X)$, THEN

$$\Psi(t) \stackrel{\text{C}}{=} \mathbb{E}(\cos(t \cdot X))$$

BECAUSE $\sin(-x) = -\sin(x)$, THUS

$$\mathbb{E}(\min(t \cdot X)) \stackrel{\text{D}}{=} \mathbb{E}(\min(-t \cdot X)) = -\mathbb{E}(\max(t \cdot X))$$

$$X \sim (-X)$$

THUS $\mathbb{E}(\min(t \cdot X)) \stackrel{\text{E}}{=} 0$ ✓