

Convergence Concepts

(1) Relations:

① Thm. For $r \geq 1$:

$$X_n \rightarrow X \text{ a.s.}$$

$$\Rightarrow X_n \xrightarrow{p} X \Rightarrow X_n \xrightarrow{\mu} X.$$

$$X_n \rightarrow X \text{ in } L^r$$

$$\text{For } r > s > 0. \quad X_n \rightarrow X \text{ in } L^r \Rightarrow X_n \rightarrow X \text{ in } L^s.$$

Remark: Some counterexamples:

$$\text{i) } X_n \xrightarrow{p} X \not\Rightarrow X_n \rightarrow X \text{ a.s. or in } L^r:$$

$$P(X_n = 1) = 1/n, \quad P(X_n = 0) = 1 - 1/n.$$

$$\text{ii) } X_n \xrightarrow{\mu} X \not\Rightarrow X_n \xrightarrow{p} X:$$

$$X_n = -X \sim N(0, 1)$$

$$\text{iii) } X_n \rightarrow X \text{ a.s.} \not\Rightarrow X_n \rightarrow X \text{ in } L^r.$$

$$\not\Rightarrow: P(X_n = 0) = 1 - \frac{1}{n}, \quad P(X_n = n^2) = \frac{1}{n}.$$

$$\not\Leftarrow: P(X_n = 0) = 1 - \frac{1}{n}, \quad P(X_n = 1) = \frac{1}{n}.$$

② Partial Converse

$$\text{i) } \underline{\text{Thm.}} \quad X_n \xrightarrow{\mu} C \Leftrightarrow X_n \xrightarrow{p} C \text{ for const. } C.$$

Pf: Written in h.f. for $P(|X_n - C| \geq \varepsilon)$.

ii) Monotone r.v's:

Thm. $\{X_n\}$ mono r.v's. $X_n \xrightarrow{p} X \Rightarrow X_n \rightarrow X$ a.s.

Pf: Lemma. $X_n \leq Y_n \leq Z_n$. $X_n, Z_n \rightarrow Y$ a.s.

Then $Y_n \rightarrow Y$ a.s.

Pf: $p(\{\omega \mid Y_n(\omega) \rightarrow Y(\omega)\}) \geq p(\{\omega \mid X_n(\omega) \rightarrow X(\omega)\} \cap \{\omega \mid Z_n(\omega) \rightarrow Z(\omega)\})$
 $\geq 1 - p(\{\omega \mid \dots\}^c) = 1 - p(\{\omega \mid \dots\}^c) = 1.$

\Rightarrow WLOG. $X_n \geq X_{n+1}$. $\forall n \in \mathbb{Z}^+$. Since $\exists \{X_{n_k}\} \subset \{X_n\}$ st.

$X_{n_k} \rightarrow X$ a.s. Then $\forall n$. $\exists k$. $X_{n_k} \geq X_n \geq X_{n_{k+1}}$.

iii) Converge completely:

Def: $X_n \rightarrow X$ completely if $\forall \varepsilon > 0$. $\sum p(|X_n - X| > \varepsilon) < \infty$.

Thm. $X_n \rightarrow X$ completely $\Rightarrow X_n \rightarrow X$ a.s.

Cor. $\sum E(|X_n - X|^r) < \infty \Rightarrow X_n \rightarrow X$ a.s. ($r > 0$).

iv) Converge in another space:

Thm. (Skorokhod's Representation)

If $X_n \xrightarrow{d} X$. Then \exists r.v's $Y, \{Y_n\}$ on space $([0,1], \mathcal{B}_{[0,1]}, m)$. m is Lebesgue measure. st.

$X_n \sim Y_n$. $X \sim Y$. $Y_n \rightarrow Y$ a.s.

Pf: Denote F_n, F are A.f of X_n, X .

Set $Y_n(t) = F_n^{-1}(t)$. $Y(t) = F^{-1}(t)$.

$\therefore Y_n \sim X_n$. $Y \sim X$.

Fix $\varepsilon > 0$, $t' \in (0, 1)$. $\exists X \in \mathcal{C}(F)$. $Y(t') < X < Y(t') + \varepsilon$

$\therefore F(x) \geq t' > t$. $\therefore F_n \rightarrow F$ at X . $\therefore \exists N$, $n > N$, st.

$F_n(x) > t$. $\therefore Y_n(t) < X < Y(t') + \varepsilon$.

$\therefore \overline{\lim}_n Y_n \leq Y(t')$. (Choose $t \in \mathcal{C}(F)$. Let $t' \rightarrow t$.)

$\therefore \overline{\lim}_n Y_n \leq Y$ on $\mathcal{C}(F)$. Similarly, $\underline{\lim}_n Y_n \geq Y$.

$\therefore p(\{t \mid Y_n(t) \rightarrow Y(t)\}) \leq p(\mathcal{C}(0, 1) / \mathcal{C}(F)) = 0$.

(2) Convergence in Moments:

① Pratt's Lemma.

i) $X_n \leq Y_n \leq Z_n$, a.s.

ii) $X_n \rightarrow X$ a.s., $Y_n \rightarrow Y$ a.s., $Z_n \rightarrow Z$ a.s.

iii) $E(X_n) \rightarrow E(X)$, $E(Z_n) \rightarrow E(Z)$, $X, Z \in L^1$.

$\Rightarrow E(Y_n) \rightarrow E(Y)$.

Pf: Apply Fatou's Lemma on $Z_n - Y_n$, $Y_n - X_n$.

② Dominated Convergence:

Lemma. $X_n \xrightarrow{p} X$, $p(|X_n| \leq Y) = 1$. $\Rightarrow p(|X| \leq Y) = 1$.

Pf: $\forall \delta > 0$, $p(|X| > Y + \delta) =$

$p(\emptyset, |X_n| \leq Y) + p(\emptyset, |X_n| > Y)$

$\leq p(|X| > |X_n| + \delta, |X_n| \leq Y)$

$\leq p(|X - X_n| > \delta) \rightarrow 0$.

Lemma. $Y \in L^1$, $P(A_n) \rightarrow 0 \Rightarrow E_{A_n}(Y) \rightarrow 0$.

Pf: By MCT: $E(Y I_{|Y| > N}) \rightarrow 0$.

$$\begin{aligned} \text{Truncation: } E_{A_n}(Y) &= E_{A_n}(Y I_{|Y| \geq N}) + Y I_{|Y| < N} \\ &\leq \varepsilon + N P(A_n) \rightarrow 0 \end{aligned}$$

Thm. If $X_n \xrightarrow{p} X$, $|X_n| \leq Y$, a.s. $\forall n$, $Y \in L^r$, $r > 0$.

Then $X_n \rightarrow X$ in L^r . (s.o.: $E(X_n^r) \rightarrow E(X^r)$).

Pf: By Lemma, $|X_n - X| \leq 2|Y|$, a.s.

$$\begin{aligned} E(|X_n - X|^r) &= E(|X_n - X|^r I_{|X_n - X| \geq \varepsilon} + I_{|X_n - X| < \varepsilon}) \\ &\leq 2^r E(|Y|^r I_{A_n}) + \varepsilon^r. \quad (\text{Apply Lemma.}) \end{aligned}$$

Remark: $X_n \rightarrow X$, a.s., $X \in L^r$ $\nRightarrow X_n \rightarrow X$ in L^r .

e.g., $P(X_n = 2^n) = \frac{1}{2^n}$, $P(X = 0) = 1 - \frac{1}{2^n}$, $X_n \rightarrow 0$, a.s.

Actually, $|X_n| \leq X + \varepsilon$, a.s. for large n is wrong!

Cor. If $|X_n| \leq C$, a.s. Then $\forall r > 0$, we have:

$$X_n \rightarrow X \text{ in } L^r \Leftrightarrow X_n \xrightarrow{p} X.$$

$$\text{Cor. } X_n \xrightarrow{p} 0 \Leftrightarrow E\left(\frac{|X_n|}{1+|X_n|}\right) \rightarrow 0.$$

$$\text{Pf: } P(|X_n| \geq \varepsilon) = P\left(\frac{|X_n|}{1+|X_n|} \geq \frac{\varepsilon}{1+\varepsilon}\right)$$

$$\therefore \frac{|X_n|}{1+|X_n|} \xrightarrow{p} 0 \Leftrightarrow X_n \xrightarrow{p} 0.$$

③ Uniformly Integrable:

i) Definitions:

- Because the dominated condition: $|X_n| \leq Y \in L^1$ is strong. We will introduce a weaker condition: uniform integrability.

Def: $\{Y_i\}_{i \in I}$ r.v.'s on (Ω, \mathcal{A}, P) is u.i. iff

$$\lim_{c \rightarrow \infty} \sup_{i \in I} E(|Y_i| I_{\{|Y_i| \geq c\}}) = 0$$

Remark: (a) It motivates by $X \in L^1 \Leftrightarrow$

$$E(|X| I_{\{|X| \geq k\}}) \rightarrow 0 \text{ (} k \rightarrow \infty \text{)}.$$

(b) It can imply:

$$\sup_{i \in I} E(|X_i|) \leq M < \infty. \text{ But}$$

converse doesn't hold!

Thm. (Criteria)

$\varphi \geq 0$. s.t. $\varphi(x)/x \rightarrow \infty$ ($x \rightarrow \infty$). If $\forall i \in I$.

$E(\varphi(|X_i|)) \leq C < \infty$. Then $\{X_i\}_{i \in I}$ is u.i.

Pf: Denote $\varepsilon_m = \sup\{\varphi(x)/x \mid x \geq m\} \rightarrow 0$ ($m \rightarrow \infty$)

$$\begin{aligned} \text{Then } E(|X_i| I_{\{|X_i| \geq m\}}) &\leq \varepsilon_m E(\varphi(|X_i|) I_{\{|X_i| \geq m\}}) \\ &\leq C \varepsilon_m \rightarrow 0 \text{ (} m \rightarrow \infty \text{)} \end{aligned}$$

Remark: e.g. $\varphi = x^p$, $p > 1$. $(x \log x)^+$.

Lemma. (Absolute Continuity)

If $X \in L^1$. Then $Q(A) = E_A(X)$ is absolutely
conti. i.e. $\forall \varepsilon > 0, \exists \delta > 0, \forall A \in \mathcal{A}, P(A) < \delta \Rightarrow$
 $Q(A) < \varepsilon.$

Pf: Truncate $X = I_{\{|X| \leq m\}} + I_{\{|X| > m\}}.$

Thm. (Equivalence Def for u.i.)

r.v.'s $\{Y_i\}$ on (N.A.P.)
is u.i.

$$\Leftrightarrow \begin{aligned} (a) \quad \sup_i E(|Y_i|) &< \infty \\ (b) \quad \forall \varepsilon > 0, \exists \delta > 0, \text{ s.t. } \forall A \in \mathcal{A}, \\ P(A) < \delta &\Rightarrow \sup_i E_A(|Y_i|) < \varepsilon. \\ P(A_n) \rightarrow 0 &\Rightarrow \sup_i E_{A_n}(|Y_i|) \rightarrow 0 \end{aligned}$$

Pf: (\Rightarrow) Trivial. By truncation.

(\Leftarrow) Denote $M = \sup_{i \in I} E(|Y_i|) < \infty$. For $\forall \varepsilon > 0, \exists \delta > 0$.

$P(|Y_i| \geq c) \leq M/c$. choose $c > M/\delta$.

$$\therefore \sup_{i \in I} E_{\{|Y_i| > c\}}(|Y_i|) < \varepsilon.$$

ii) Thm. (a) $\{X_i\}$ u.i. $\Leftrightarrow \{|X_i|\}$ u.i. $\Leftrightarrow \{X_i^+\}, \{X_i^-\}$ u.i.

(b) $|X_i| \leq |Y_i|, \{Y_i\}$ u.i. $\Rightarrow \{X_i\}$ u.i.

(c) $\{X_i\}, \{Y_i\}$ u.i. $\Rightarrow \{X_i + Y_i\}$ u.i.

(d) If $|X_i| \leq Y \in L^1$ a.s. $\forall i$. Then $\{X_i\}$ u.i.

Pf. (a), (b) are trivial.

$$(c) |X_i + Y_i| I_{\{|X_i + Y_i| \geq c\}} \leq |X_i| I_{\{|X_i| \geq \frac{c}{2}\}} + |Y_i| I_{\{|Y_i| \geq \frac{c}{2}\}}.$$

$$(d) \sup_{i \in I} E(|X_i| I_{\{|X_i| \geq c\}}) \leq E(|Y| I_{\{|Y| \geq c\}}) \rightarrow 0$$

$$\text{Cor. If } \exists Y \in L^1, p(|Y_n| \geq \eta) \leq p(|Y| \geq \eta), \forall \eta > 0$$

Then $\{Y_n\}$ is u.i.

$$\begin{aligned} \text{pf: } E(|Y_n| I_{\{|Y_n| \geq c\}}) &= \int_0^c + \int_c^\infty p(|Y_n| \geq \eta) I_{\eta} \\ &= c p(|Y_n| \geq c) + \int_c^\infty p(|Y_n| \geq \eta) d\eta \\ &\leq c p(|Y| \geq c) + \int_c^\infty p(|Y| \geq \eta) d\eta \\ &= E(|Y| I_{\{|Y| \geq c\}}) \rightarrow 0 \quad (c \rightarrow \infty) \end{aligned}$$

Remark: Denote $Y \geq_{\text{stoch}} Y_n$ if $F_Y(\eta) \leq F_{Y_n}(\eta), \forall \eta$.

iii) Converge in pr. + u.i. \Rightarrow Converge in L^1 :

Thm. (Vitali's)

$X_n \xrightarrow{p} X, X_n \in L^1, \forall n$. Then followings are equivalent.

(a) $\{X_n\}$ u.i.

(b) $X_n \rightarrow X$ in $L^1, X \in L^1$.

(c) $E(|X_n|) \rightarrow E(|X|) < \infty$.

Pf: (a) \Rightarrow (b):

1') $X \in L^r$.

since exist $\{X_{n_k}\} \subseteq \{X_n\}$. $X_{n_k} \rightarrow X$ a.s.

By Fatou's: $E(|X|^r) \leq \liminf E(|X_{n_k}|^r) \leq \sup E(|X_n|) < \infty$.

2') $\{X_n - X\}$ u.i.

Lemma. C_r - Inequality

$$|X + \eta|^r \leq C_r (|X|^r + |\eta|^r), \quad C_r = \begin{cases} 1, & 0 < r < 1 \\ 2^{r-1}, & 1 \leq r. \end{cases}$$

Pf: $0 < r < 1$: $\lambda^r + (1-\lambda)^r \geq \lambda + 1 - \lambda = 1$, $\lambda = |X|/|X| + |\eta|$.

$1 \leq r$: by convexity of $|X|^r$.

$$\Rightarrow |X_n - X|^r \leq C_r (|X_n|^r + |X|^r) \text{ u.i.}$$

3') $X_n \rightarrow X$ in L^r :

$$\begin{aligned} E(|X_n - X|^r) &= E(\square I_{|X_n - X| \geq \varepsilon} + I_{\varepsilon \leq |X_n - X| < \varepsilon} + I_{|X_n - X| \leq \varepsilon}) \\ &\leq C^r P(|X_n - X| \geq \varepsilon) + \varepsilon^r + E(|X_n - X|^r I_{|X_n - X| \geq \varepsilon}) \\ &\rightarrow \varepsilon^r + E(|X_n - X|^r I_{|X_n - X| \geq \varepsilon}) \rightarrow \varepsilon^r. \end{aligned}$$

(Directly. $E(|X_n - X|^r I_{|X_n - X| \geq \varepsilon}) \rightarrow 0$ (n.t.w.), by 2nd def).

(b) \Rightarrow (c): C_r - Inequality. Minkovsky - Inequality

(c) \Rightarrow (a): Set $f_A(x) \in C_B$ s.t.

$$f_A(x) = \begin{cases} |x|^r, & |x| \leq A \\ 0, & |x| > A+1. \end{cases} \quad \therefore \lim E(f_A(X_n)) = E(f_A(X)).$$

$$\liminf E(|X_n|^r I_{|X_n| \leq A+1}) \geq \lim E(f_A(X_n)).$$

$$\geq E(f_A(X)) \geq E(|X|^r I_{|X| \leq A})$$

$$\Rightarrow \lim E(|X_n|^r I_{\{|X_n| > A\}}) \leq E(|X|^r I_{\{|X| > A\}})$$

$$\forall \varepsilon > 0, \exists A(\varepsilon), n_0 \in \mathbb{N}, \sup_{n \geq n_0} E(|X_n|^r I_{\{|X_n| > A\}}) < \varepsilon.$$

Cor. $X_n \rightarrow X$ in L^r , $r > 0$, $X \in L^r$. Then

$$(a) E(|X_n|^r) \rightarrow E(|X|^r), \quad (b) E(X_n^r) \rightarrow E(X^r).$$

Pf. (b) is from (a), since $\{X_n^r\}$ is u.i.

$$X_n^r \rightarrow X^r \text{ in } L^1.$$

(a) By Gr. Minkowski Inequality, analogously!

Remark: $X_n \rightarrow X$ in $L^r \not\Rightarrow X_n$ or $X \in L^r$.

e.g. $X_n \sim X \sim \text{Cauchy}$, $r=1$.

iv) Converge in dist. + u.i. \Rightarrow Converge in moments:

Thm. $X_n \xrightarrow{d} X$, $\{X_n^r\}$ u.i. Then, we have:

$$(a) E(|X|^r) < \infty, \quad (b) \lim E(|X_n|^r) = E(|X|^r), \quad (c) \lim E(X_n^r) = E(X^r).$$

Pf: Apply Skorokhod's Representation.

Rmk: $X_n \rightarrow X \nleftrightarrow Y_n \rightarrow Y$, even if $X_n \sim Y_n$, $X \sim Y$.

$\therefore X_n \not\rightarrow X$ in L^r commonly.

Besides X_n 's may not be in the same prob. space.

3) Closed Operations:

① Algebraic Operations:

Thm. $X_n \rightarrow X, Y_n \rightarrow Y \Rightarrow X_n \pm Y_n \rightarrow X \pm Y.$

It holds for converging n.s./in pr/in L^r .

Pf: By $\mathbb{I}\{X_n + Y_n \geq \varepsilon\} \subseteq \mathbb{I}\{X_n \geq \frac{\varepsilon}{2}\} \cup \mathbb{I}\{Y_n \geq \frac{\varepsilon}{2}\}$. Cr-Inequality.

Rmk: It fails when converge in dist: e.g. $X \sim N(0,1)$.

$X_n = X = -Y = Y_n$. (Actually, $X_1 \sim X_2, Y_1 \sim Y_2$. Then:

$X_1 + Y_1 \not\sim X_2 + Y_2$).

Generally, $X_n \xrightarrow{d} X, Y_n \xrightarrow{d} Y \not\Rightarrow g(X_n, Y_n) \xrightarrow{d} g(X, Y)$ for

$g \in C(\mathbb{R}^2)$, except: $(X_n, Y_n) \xrightarrow{d} (X, Y)$.

Thm. $X_n \xrightarrow{d} X_\infty, Y_n \xrightarrow{d} Y_\infty, X_n, Y_n$ indept. $\forall n \in \mathbb{Z}^+$. Then:

$\exists X \sim X_\infty, Y \sim Y_\infty, X_n \pm Y_n \xrightarrow{d} X \pm Y.$

Pf: $\varphi_{X_n + Y_n} = \varphi_{X_n} \varphi_{Y_n} \rightarrow \varphi_X \varphi_Y$. where we choose X, Y .

are just random sample (size=1) from X_∞, Y_∞ .

$\therefore X, Y$ indept. $\varphi_X \varphi_Y = \varphi_{X+Y}$. identical ch.f.

Rmk: X_∞, Y_∞ may not be indept. e.g. X_∞ isn't degenerated

$Y_n = Y_\infty = X_\infty, \forall n$. even if Y_n indept with $X_k, \forall n, k$.

Also note that we only require X_n, Y_n are in the same probability space.

Thm. $X_n \rightarrow X, Y_n \rightarrow Y \Rightarrow X_n Y_n \rightarrow XY$ holds
for convergence a.s. / in pr.

Pf. i) $\{X_n Y_n \not\rightarrow XY\} \subseteq \{X_n \not\rightarrow X\} \cup \{Y_n \not\rightarrow Y\}.$

ii) $p(\{X_n Y_n - XY \geq \varepsilon\}) \leq p(\{X_n - X \mid Y_n \geq \frac{\varepsilon}{2}\})$
 $+ p(\{Y_n - Y \mid X \geq \frac{\varepsilon}{2}\}).$

For the former, separate: (latter is same)

$$\mathcal{A} = \{|Y_n| = 0\} + \{0 < |Y_n| < M\} + \{M \leq |Y_n|\}.$$

$$\Rightarrow \leq p(|X_n - X| > \frac{\varepsilon}{2M}) + p(|Y_n| \geq M).$$

$$\leq p(|X_n - X| > \frac{\varepsilon}{2M}) + p(|Y_n - Y| \geq \frac{M}{2}) + p(|Y| > \frac{M}{2})$$

Firstly, fix large M . Then $n \rightarrow \infty$.

Remark: i) L^r doesn't hold: choose $X_n = Y_n \in L^1 \cap (L^2)^c$.

$$\therefore \|X_n\|_{L^2}^2 = \infty.$$

dist doesn't hold: $X_n = Y_n = X = -Y \sim N(0,1).$

ii) $X_n \rightarrow X$ in L^p . $Y_n \rightarrow Y$ in L^2 . $1/p + 1/2 = 1$.

$$\Rightarrow X_n Y_n \rightarrow XY \text{ in } L^1.$$

Pf: $X, Y \in L^p, L^2$. It's trivial.

$$X \in L^p, Y \in L^2: \exists N, n > N, X_n, Y_n \in L^p, L^2.$$

$$|X_n Y_n - XY| \leq |X_n| |Y_n - Y| + |Y| |X_n - X|.$$

By Hölder Inequality.

② Transformations:

Thm. X_n, X are k -dimensional random vectors.

$g: \mathbb{R}^k \rightarrow \mathbb{R}^l$ conti. Then we have:

i) $X_n \rightarrow X \text{ a.s.} \Rightarrow g(X_n) \rightarrow g(X) \text{ a.s.}$

ii) $X_n \rightarrow_p X \Rightarrow g(X_n) \rightarrow_p g(X)$.

iii) $X_n \rightarrow_d X \Rightarrow g(X_n) \rightarrow_d g(X)$.

Pf: i) By conti. $\{X_n \rightarrow X\} \subset \{g(X_n) \rightarrow g(X)\}$.

ii) Fix $\varepsilon > 0$. Choose M large enough s.t.

$$P(|X| \geq M) \leq \varepsilon. \text{ Then on } |X| \leq M + \varepsilon:$$

$$\exists \delta < \varepsilon \text{ s.t. } |g(X) - g(Y)| < \varepsilon \text{ if } |X - Y| < \delta, |X| \leq M.$$

$$\text{Then } \{ |g(X_n) - g(X)| \geq \varepsilon, |X_n - X| < \delta \} \subseteq \{ |X| > M \}.$$

$$\text{partition } P(|g(X_n) - g(X)| \geq \varepsilon) = P(\square, |X_n - X| < \delta) + P(\square, |X_n - X| \geq \delta)$$

iii) By Skorokhod's Representation.

Cor. Extend to g is a.s. conti w.r.t P_X , i.e.

$$P(\{\omega \mid g \text{ is conti at } X(\omega)\}) = 0.$$

Pf: i) $\{X_n \rightarrow X\} \subset \{g(X_n) \rightarrow g(X)\} + \{g \text{ is conti at } X(\omega)\}$

ii) Similarly, separate $\{g \text{ conti at } X\} + \{g \text{ is conti}\}$

iii) For $\forall f \in C_b$. By Skorokhod Representation

$$\text{since } P(\{\omega \mid f \circ g \text{ is conti at } X(\omega)\}) = 0.$$

$$\therefore f(g(X_n)) \rightarrow f(g(X)) \text{ a.s.}$$

Then Apply DCT. $\therefore g(X_n) \Rightarrow g(X)$.

③ Slutsky Thm:

If $X_n \rightarrow_d X$, $Y_n \rightarrow_p c$. Then:

i) $X_n \pm Y_n \rightarrow_d X \pm c$

ii) $X_n Y_n \rightarrow_d cX$.

iii) $X_n / Y_n \rightarrow_d X / c$ for $c \neq 0$.

Pf: Consider $F_{X_n+Y_n}(x) = P(X_n+Y_n \leq x) = 1 - P(X_n+Y_n > x)$.

Separate into: $\{ |Y_n - c| \geq \varepsilon \} + \{ |Y_n - c| < \varepsilon \}$.

Written into d.f. Then apply \lim , $\underline{\lim}$.

Remark: The point is: if $X \sim Y$, c is const. Then

$X+c \sim Y+c$. (It's obvious by ch.f's)

④ Cauchy Convergence:

$\{X_n\}$ converge in L^2 , a.s. pr \Leftrightarrow It's Cauchy in L^2 , a.s. pr.

Pf: (\Rightarrow) is trivial.

(\Leftarrow) . Find $\{X_{n_k}\} \subseteq \{X_n\}$. $X_{n_k} \rightarrow X$, a.s.

i) $L^2 = E(|X_n - X|^2) = E(\lim_k |X_n - X_{n_k}|^2)$

$\leq \lim_k E(|X_n - X_{n_k}|^2) \rightarrow 0$. (Fatou's)

ii) a.s.: $I_n \sim / N = X_n(\omega) \rightarrow X(\omega) \in \mathbb{R}$. $\forall \omega$

iii) pr: $\{ |X_n - X| \geq \varepsilon \} \subseteq \{ |X_n - X_{n_k}| \geq \frac{\varepsilon}{2} \} \cup \{ |X_{n_k} - X| \geq \frac{\varepsilon}{2} \}$