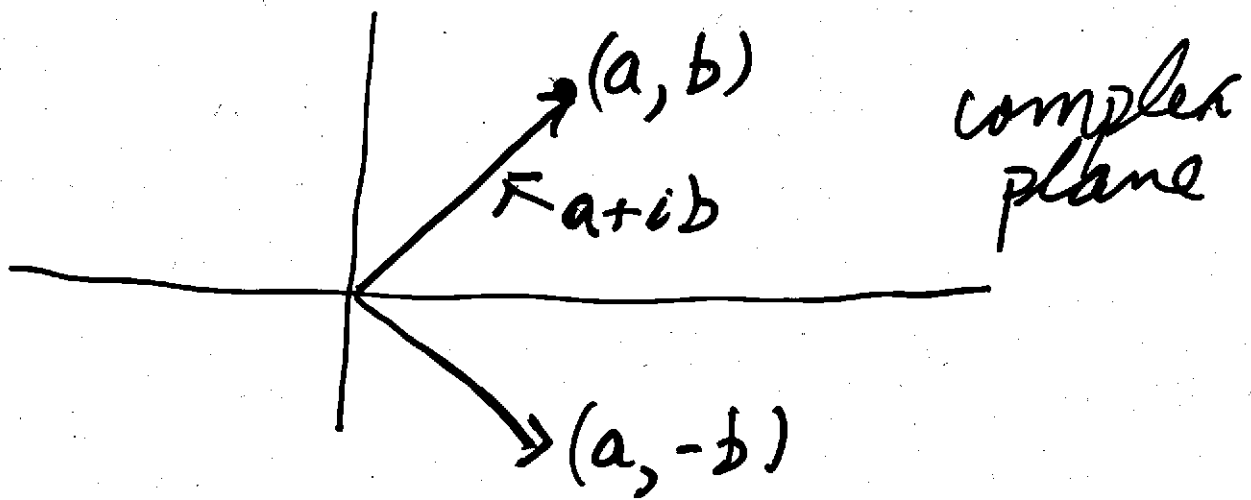


## Some background

$a, b \in \mathbb{R}$  &  $i$  an object st  $i^2 = -1$  then  $a + ib$  is called a complex #.

$\uparrow$  real part       $\uparrow$  imaginary part



complex conjugate  $\overline{a + ib} = a - ib$

If  $z = a + ib$  then  $|z|^2 = a^2 + b^2$

Notice  $z \bar{z} = (a + ib)(a - ib) = a^2 + b^2 = |z|^2$

$e^{it} = \cos t + i \sin t$  (def'n)

Note  $e^z = e^a e^{ib}$

$$\overline{e^{it}} = \cos t - i \sin t$$

$$= \cos(-t) + i \sin(-t) = e^{-it}$$

$$e^{it_1} e^{it_2} = e^{i(t_1+t_2)} \quad \text{— can check}$$

$$|e^{it}|^2 = \cos^2 t + \sin^2 t = 1$$

$$\Rightarrow |e^{it}| = 1$$

$$|e^{it_2} - e^{it_1}| \leq |t_2 - t_1|$$

$$|e^{it_2} - e^{it_1}| \leq 2$$

Note  $|z_1 + z_2| \leq |z_1| + |z_2|$

$$|e^{it_2} - e^{it_1}| = \left| \int_{t_1}^{t_2} \frac{d(e^{it})}{dt} dt \right|$$
$$\leq \int_{t_1}^{t_2} \left| \frac{d(e^{it})}{dt} \right| dt$$

$$= \left| \int_{t_1}^{t_2} \underbrace{|ie^{it}|}_1 dt \right| = |t_2 - t_1|$$

Note,  $|z_1 z_2| = |z_1| |z_2|$

$$\underline{\underline{2}} \quad |\cos t_2 - \cos t_1| \leq |t_2 - t_1|$$

$$|\sin t_2 - \sin t_1| \leq |t_2 - t_1|$$

$X$  rv

$$c(t) = E(e^{itX})$$

characteristic  
— cf fm

$$= E[\cos(tX)] + i E[\sin(tX)]$$

### Proposition

(i)  $c(0) = 1$ ,  $|c(t)| \leq 1$

(ii)  $\overline{c(t)} = c(-t)$

(iii)  $c_{aX+b}(t) = e^{itb} c_X(at)$

# Inversion Theorem

(i) If  $X$  has pdf  $f(x)$  then

$$f(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-itx} c(t) dt$$

$$[ c(t) = \int_{-\infty}^{\infty} e^{itx} f(x) dx ]$$

(ii)  $F(y) - F(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \left( \frac{e^{-ity} - e^{-itx}}{-it} \right) c(t) dt$

$\uparrow$  at  $\uparrow$  at pts of  $F$

Note -  $c(t)$  determines the dist'n

$$\Leftrightarrow c(t) = E(e^{itX})$$

$$\Leftrightarrow c(t) \leftrightarrow F(x) = P(X \leq x)$$

Fact  $F(x) = F(x_1) \cdots F(x_m) \Leftrightarrow$  independent

$$c(t) = c(t_1) \cdots c(t_m) \Rightarrow F(x) = F(x_1) \cdots F(x_m) \Rightarrow \text{ind}$$

Prop  $X \text{ \& } Y \text{ ind} \Rightarrow C_{X+Y}(t) = C_X(t) C_Y(t)$

$$\begin{aligned} \text{Pf } C_{X+Y}(t) &= E[e^{it(X+Y)}] \\ &= E(e^{itX} e^{itY}) \\ &= E(e^{itX}) E(e^{itY}) \end{aligned}$$

Note  $C_{X+Y}(t) = C_X(t) C_Y(t) \not\Rightarrow \text{ind}$

eg  $X$  is Cauchy if

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

$$C(t) = \int_{-\infty}^{\infty} e^{itx} \frac{1}{\pi(1+x^2)} dx = e^{-|t|}$$

Let  $Y = X$ . Then  $Y \text{ \& } X$  are not ind.

$$C_{X+Y}(t) = C_{2X}(t) = C_X(2t) = e^{-2|t|}$$

$$C_X(t) C_Y(t) = e^{-|t|} e^{-|t|} = e^{-2|t|}$$

∴ so  $C_{X+Y}(t) = C_X(t) C_Y(t)$

but  $Y$  &  $X$  are dependent.

### Remarks

—  $m(t) = E(e^{tX})$  — mgf

—  $m(\underline{t}) = E(e^{\underline{t}'X})$  — mgf of  $\underline{X}$

Fact: If  $m(\underline{t})$  exists in a neighborhood of  $\underline{0}$  then  $c(\underline{t}) = m(i\underline{t})$

eg  $X \sim N(0, 1) \Leftrightarrow m(t) = e^{t^2/2}, \forall t$

$$\Rightarrow c(t) = e^{(it)^2/2} = e^{-t^2/2}$$

## 2 A version of Taylor's Theorem

If  $g^{(m)}(0)$  exists then

$$g(x) = \left[ \sum_{j=0}^m \frac{g^{(j)}(0)}{j!} x^j \right] + o(x^m),$$

as  $x \rightarrow 0$ . Here  $o(x^m)$  is the remainder & has the property that

$$\lim_{x \rightarrow 0} \frac{o(x^m)}{x^m} = 0$$

Also,  $g^{(0)} = g$ .

eg  $e^x = 1 + x + o(x)$

$e^{x+o(x)} = 1 + x + o(x)$

not the same expression!

Formally,

$$e^{itX} = \sum_{j=0}^{\infty} \frac{(it)^j}{j!} X^j$$

"& so"

$$E(e^{itX}) = \sum_{j=0}^{\infty} \frac{(it)^j}{j!} E(X^j)$$

"& so"

$$c^{(j)}(0) = i^j E(X^j)$$

Note,  $e^z = \sum_{j=0}^{\infty} \frac{z^j}{j!}$

∴ We "know"  $m^{(j)}(0) = E(X^j)$

Theorem Let  $X$  be finite w.p.1.  
Then  $c(t)$  is uniformly cts.  
If  $E(|X|^k) < \infty$  then  $E(X^k e^{itX})$   
is uniformly cts (in  $t$ ).

Proof Let  $\epsilon > 0$ . Then

$$|c(t_2) - c(t_1)| = |E(e^{it_2 X} - e^{it_1 X})|$$



$$\leq E(|e^{it_2 X} - e^{it_1 X}|)$$

Now choose  $A$  so that

$$2P(|X| \geq A) \leq \epsilon/2$$

Now

$$E(|e^{it_2 X} - e^{it_1 X}|)$$

$$= E(\underbrace{|e^{it_2 X} - e^{it_1 X}|}_{\leq |t_2 - t_1| |X|} I(|X| < A))$$

$$+ E(\underbrace{|e^{it_2 X} - e^{it_1 X}|}_{\leq 2} I(|X| \geq A))$$

$$\leq A |t_2 - t_1| + 2 P(|X| \geq A)$$

$$\leq A |t_2 - t_1| + \epsilon/2$$

$$\circ \circ |t_2 - t_1| \leq \frac{\epsilon}{2A} \Rightarrow |C(t_2) - C(t_1)| \leq \epsilon$$

QED

## Consequence

①  $E(|X|^k) < \infty \Rightarrow C^{(k)}(0)$  exists  
(try to show for  $k=1$  by def'n of a derivative)

$C^{(k)}(0)$  exists  $\Rightarrow E(|X|^k) < \infty$  ?

No unless  $k$  is even.

②  $E(|X|^k) < \infty$

$$\Rightarrow C(t) = \sum_{j=0}^k \frac{(it)^j}{j!} E(X^j) + o(t^k)$$

$$C^{(k)}(0) = i^k E(X^k)$$

# Some more background

real #'s

set of #'s

$\left. \begin{array}{l} \text{bounded above} \\ \text{bounded below} \end{array} \right\} \text{bounded}$

$\exists \text{ lub} = \text{sup}$   
 $\& \text{ glb} = \text{inf}$

sequences

$$a_n \rightarrow a$$

$$\left( \frac{1}{n} \sum_{k=1}^n a_k \rightarrow a \right)$$

$\rightarrow$  has the Cauchy property in

That if  $a_n - a_m \rightarrow 0$  as  $n, m \rightarrow \infty$  ~~mutual conv~~

$n, m \rightarrow \infty$  then  $\exists$  an  $a$  st

$$a_n \rightarrow a$$

# Types of convergence

1  $X_n \xrightarrow{ms} X$  (in mean square) if  
 $E(X_n - X)^2 \rightarrow 0$  as  $n \rightarrow \infty$

Fact  $\xrightarrow{ms}$  has the Cauchy property

2  $X_n \xrightarrow{P} X$  (convergence in probability)

$\forall \epsilon > 0$

$P(|X_n - X| \leq \epsilon) \rightarrow 1$  as  $n \rightarrow \infty$

Fact  $\xrightarrow{P}$  has the Cauchy property

3  $X_n \rightarrow X$  if  $\lim_{n \rightarrow \infty} X_n(\omega) = X(\omega)$ ,

$\forall \omega \in \Omega$ .

almost surely

4  $X_n \xrightarrow{WP} X$  ( $X_n \xrightarrow{as} X$ ) if  
 $P(X_n \rightarrow X) = 1$

Theorem If  $0 \leq X_n$  &  $X_n \uparrow X$  w.p.1  
then  $E(X_n) \rightarrow E(X)$ .

Monotone convergence theorem (MCT)

Note  $0 \leq X_n$  can be replaced by  
assuming  $E(X, 1) < \infty$ .

Theorem (DCT) If  $X_n \xrightarrow{as} X$  &  
 $|X_n| \leq Y$  with  $E(Y) < \infty$  then  
 $E(X_n) \rightarrow E(X)$

Note  $X \stackrel{as}{=} X' \Rightarrow E(X) = E(X')$   
which allows  $X_n \uparrow X$  to be replaced  
by  $X_n \uparrow X$ , w.p.1. (Why?)

## Another type of convergence

Def'n  $X_n$  converges to  $X$  weakly if  $\forall$  bounded cts  $h$  we have

$$E(h(X_n)) \rightarrow E(h(X)) \quad (*)$$

Note — also called convergence in dist'n

$$- X_n \xrightarrow{d} X$$

Separating class of  $h$ 's (subset

of all bd'd cts  $f$ 'ns such that if  $(*)$  holds for them it holds for all bd'd cts  $f$ 'ns)

$$\perp \{ \sin tx, \cos tx \mid \forall t \}$$

$$\{ e^{itx} \mid \forall t \}$$

ie. if  $c_n(t) \rightarrow c(t), \forall t$   
then  $X_n \xrightarrow{d} X$ .

Weak Law of Large Numbers (WLLN)

Let  $X_1, X_2, \dots$  be iid with mean  $\mu$ . Then

$$\bar{X} \xrightarrow{d} \mu$$

Note  $\{ Y_n \xrightarrow{d} c \Rightarrow Y_n \xrightarrow{P} c \}$  problem

Proof: Let  $c(t)$  be the cf of  $X_1$ . We know  $c(t) = 1 + i\mu t + o(t) = e^{i\mu t + o(t)}$

$$\begin{aligned}
\text{So } E(e^{it\bar{X}}) &= E(e^{i\frac{t}{n}(X_1 + \dots + X_n)}) \\
&= E(e^{i\frac{t}{n}X_1}) \dots E(e^{i\frac{t}{n}X_n}) \\
&= \left( c\left(\frac{t}{n}\right) \right)^n \\
&= \left[ e^{in\frac{t}{n} + o\left(\frac{t}{n}\right)} \right]^n \\
&= e^{int + no\left(\frac{t}{n}\right)} \\
&= e^{int + o\left(\frac{t}{n}\right) / \left(\frac{1}{n}\right)} \\
&\rightarrow e^{int}
\end{aligned}$$

which is the cf of the constant  $n$ .

qed



# Central Limit Theorem

Let  $X_1, X_2, \dots$  be iid with mean  $\mu$  & variance  $\sigma^2$ . Then

$$\frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \xrightarrow{d} N(0, 1)$$

Proof The cf of  $X_1$  is

$$c(t) = E(e^{itX_1})$$

2nd moment =  $\sigma^2$   
mean = 0

$$= e^{it\mu} E(e^{it(X_1 - \mu)})$$

$$= e^{it\mu} \left( 1 - \frac{\sigma^2 t^2}{2} + o(t^2) \right)$$

$$= e^{it\mu} e^{-\sigma^2 t^2 / 2} + o(t^2)$$

$$= e^{it\mu - \sigma^2 t^2 / 2} + o(t^2)$$

$$E\left(e^{it \frac{\bar{X} - \mu}{\sigma/\sqrt{n}}}\right)$$

$$= E\left(e^{\frac{it}{\sigma\sqrt{n}}[(X_1 - \mu) + \dots + (X_n - \mu)]}\right)$$

$$= \left(E\left(e^{\frac{it}{\sigma\sqrt{n}}(X_1 - \mu)}\right)\right)^n$$

$$= \left(e^{-\frac{\sigma^2 t^2}{2\sigma^2 n} + o\left(\frac{t^2}{\sigma^2 n}\right)}\right)^n$$

$$= e^{-t^2/2 + n o\left(\frac{t^2}{\sigma^2 n}\right)}$$

$$\rightarrow e^{-t^2/2}$$

qed

Theorem (DCT) Suppose  $X_n \rightarrow X$  &  
 $|X_n| \leq W$  with  $E(W) < \infty$ . Then  
 $E(X_n) \rightarrow E(X)$

Note The conditions  $X_n \rightarrow X$  &  $|X_n| \leq W$  can  
be replaced by  $X_n \xrightarrow{a.s.} X$  &  $|X_n| \leq W$

Proof Let  $Z_n = |X_n - X| \leq |X_n| + |X| \leq 2W$

Then, setting  $Y_n = \sup_{k \geq n} Z_k$ , yields

$$0 \leq Z_n \leq Y_n \leq 2W$$

Now  $E(Y_n) < \infty$  &  $Y_n \downarrow 0 \Rightarrow -Y_n \uparrow 0$   
with  $E(-Y_n)$  existing. Now use the MCT  
to get  $E(Y_n) \rightarrow 0$ . Since  $0 \leq Z_n \leq Y_n$   
we have  $E(Z_n) \rightarrow 0$ . Finally

$$|E(X_n) - E(X)| \leq E|X_n - X| \rightarrow 0$$

so that  $E(X_n) \rightarrow E(X)$

qed

$$X_n \xrightarrow{\text{a.s.}} X \text{ if } P(X_n \rightarrow X) = 1 \quad (*)$$

$\{X_n \rightarrow X\} = \{\omega \mid X_n(\omega) \rightarrow X(\omega)\}$ . Denote this event by  $D$ .

An equivalent def'n of  $\xrightarrow{\text{a.s.}}$  is (\*\*)

$$X_n \xrightarrow{\text{a.s.}} X \text{ if } \forall \epsilon > 0 \quad P(|X_n - X| \leq \epsilon, \forall m \geq n) \rightarrow 1$$

$$(\text{i.e. } \forall \epsilon > 0 \quad P(\sup_{m \geq n} |X_m - X| \leq \epsilon) \rightarrow 1)$$

Remark 1  $X_n \xrightarrow{\text{a.s.}} X \Leftrightarrow |X_n - X| \xrightarrow{\text{a.s.}} 0, \text{ as } n \rightarrow \infty$

$$\Leftrightarrow \sup_{m \geq n} |X_m - X| \xrightarrow{P} 0, \text{ as } n \rightarrow \infty$$

$$\Leftrightarrow X_n \xrightarrow{P} X \Leftrightarrow |X_n - X| \xrightarrow{P} 0$$

From 1 + 2 we see that

$$X_n \xrightarrow{\text{a.s.}} X \Rightarrow X_n \xrightarrow{P} X$$

(\*)  $\Leftrightarrow$  (\*\*)

Assume (\*\*). The events  $\bigcap_{m=1}^{\infty} \{ |X_m - X| \leq \epsilon \}$  form an increasing sequence (in  $m$ ) with limit

$$\bigcup_{m=1}^{\infty} \bigcap_{n=m}^{\infty} \{ |X_m - X| \leq \epsilon \}$$

Now let  $\epsilon_k \downarrow 0$  and set

$$D_k = \bigcup_{m=1}^{\infty} \bigcap_{n=m}^{\infty} \{ |X_m - X| \leq \epsilon_k \}$$

We have  $P(D_k) = 1$ ,  $\forall k$ , and  $D_k$  forms a decreasing sequence in  $k$ . In fact

$$D_k \downarrow D$$

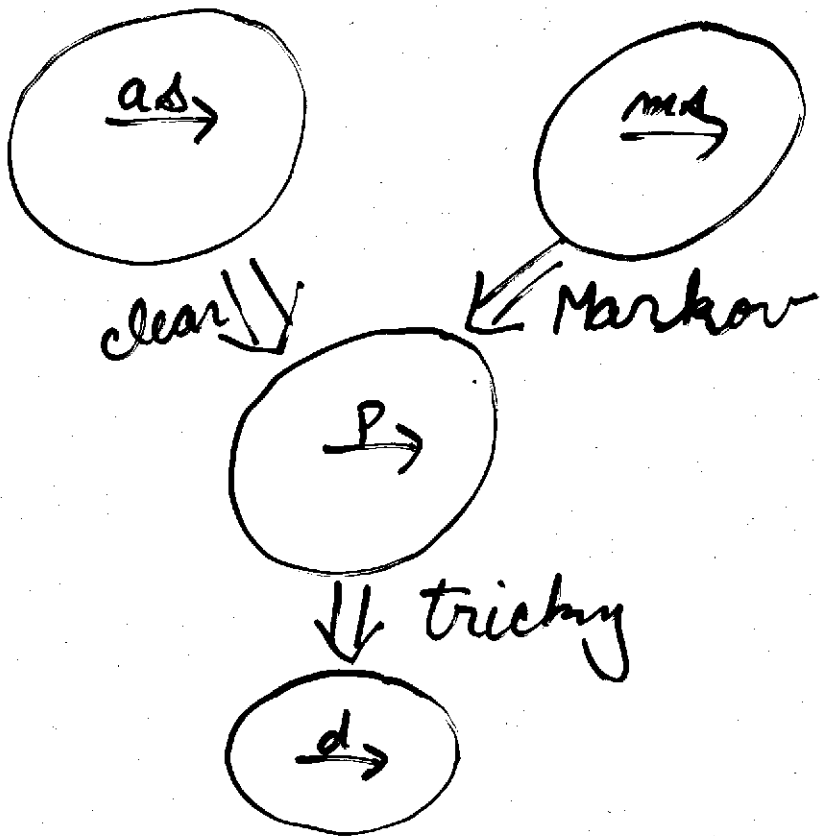
so that  $P(D) = 1$  & hence (\*) holds.

Assume (\*). Then for  $D_k$  as above we have

$$1 \geq P(D_k) \geq P(D) = 1$$

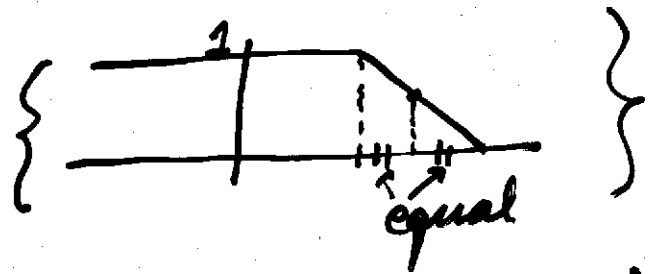
so that  $P(D_k) = 1$ ,  $\forall k$  & consequently  $\forall \epsilon > 0$

$$P\left(\bigcup_{m=1}^{\infty} \bigcap_{n=m}^{\infty} \{ |X_m - X| \leq \epsilon \}\right) = 1 \quad \& \text{ so } (**) \text{ holds.}$$

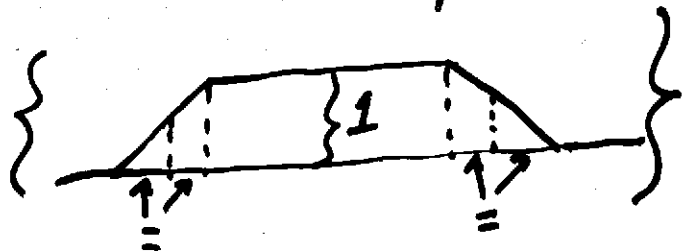


Proposition  $X_n \xrightarrow{P} X \implies X_n \xrightarrow{d} X$

A separating class



Another



Either of these can be used to show  $d$   
 $\Leftrightarrow F_n(x) \rightarrow F(x)$  at continuity points  
of  $F$

Proposition  $X_m \xrightarrow{d} c \Rightarrow X_m \xrightarrow{P} c$

Proof Let  $\epsilon > 0$ . Then

$$\begin{aligned} P(|X_m - c| > \epsilon) &= P(X_m < c - \epsilon) + P(X_m > c + \epsilon) \\ &\leq P(X_m \leq c - \epsilon) + [1 - P(X_m \leq c + \epsilon)] \\ &\rightarrow 0 \end{aligned}$$

qed

Little results

$$X_m \xrightarrow{d} X \Rightarrow \underset{\text{cts}}{g}(X_m) \xrightarrow{d} g(X)$$

- true in the vector case

$$X_m \xrightarrow{P} X \Rightarrow \underset{\text{cts}}{g}(X_m) \xrightarrow{P} g(X)$$

- "

Subsequence

$X_1, X_2, \dots$

$m_1 < m_2 < \dots$  natural #'s

$X_{m_1}, X_{m_2}, \dots$  is a subsequence

Proposition If  $\forall \epsilon > 0 \quad \sum P(|X_m| > \epsilon) < \infty$   
then  $X_m \xrightarrow{a.s.} 0$ .

Proof Let  $\epsilon > 0$ . Then

$$P(|X_m| \leq \epsilon; \forall m \geq N) \\ = 1 - P\left(\bigcup_{m=N}^{\infty} \{|X_m| > \epsilon\}\right)$$

Boole  
 $\geq 1 - \sum_{m=N}^{\infty} P(|X_m| > \epsilon) \rightarrow 1$

Corollary 1 Let  $\epsilon_m \downarrow 0$ . Then  $\sum P(|X_m| > \epsilon_m) < \infty$   
 $\Rightarrow X_m \xrightarrow{a.s.} 0$

Corollary 2  $X_m \rightarrow X \Rightarrow \exists X_{m_k} \xrightarrow{a.s.} X$

Pf of 2 Let  $\epsilon_k \downarrow 0$ . Choose  $k$  st  
 $P(|X_{m_k} - X| > \epsilon_k) \leq c_k$ ,

where  $\sum c_k < \infty$ . The result then follows  
from Corollary 1



Corollary 3  $X_m \rightarrow X \Leftrightarrow$  every subsequence has a further subsequence converging wpl to  $X$

Proof Use contradiction.

Recall  $X_m$  converges mutually wpl if  $\forall \epsilon > 0 \quad P(|X_m - X_n| \leq \epsilon, \forall m, n \geq N) \rightarrow 1$  as  $N \rightarrow \infty$ . This implies  $\exists X$  st  $X_m \xrightarrow{as} X$ .

Note  $X_m \xrightarrow{as} X$  if  $P(|X_m - X| \leq \epsilon, \forall m \geq N) \rightarrow 1$

Proposition Let  $\epsilon_m > 0, \sum \epsilon_m < \infty$  &  
 $\sum P(|X_{m+1} - X_m| > \epsilon_m) < \infty$

Then  $X_m \xrightarrow{as} X$

Proof Let  $c_N = \sum_{m=N}^{\infty} \epsilon_m$ . Then  $c_N \downarrow 0$ .

So, for any  $\epsilon > 0$   $c_N \leq \epsilon$   $\forall N$  large enough. Now for such  $N$

$$P(|X_m - X_n| \leq \epsilon, \forall m, n \geq N)$$

$$\geq P(|X_m - X_n| \leq C_N, \forall m, n \geq N)$$

$$\geq P(|X_{m+1} - X_m| \leq \epsilon_m, \forall m \geq N)$$

$$\geq 1 - \sum_{m=N}^{\infty} P(|X_{m+1} - X_m| > \epsilon_m) \rightarrow 1$$

Hence  $\{X_n\}$  is mutually convergent w.p.1  
 so that  $\exists X$  st  $X_n \xrightarrow{a.s.} X$  qed

PDCT  $X_n \rightarrow X, |X_n| \leq W$   
 with  $E(W) < \infty$ . Then  
 $E(X_n) \rightarrow E(X)$

Pf. Use subsequences

Note In fact  $E|X_n - X| \rightarrow 0$