

## Random Elements

$S$ -valued random variable.

Suppose  $(\Omega, \mathcal{F}, P)$  is probability space, and let

$X : \Omega \rightarrow S$  be measurable

$\mathcal{F}/\mathcal{S}$ , where  $S$  is a metric space.

random variable	$X$ on $\Omega$
random vector	
random sequence	$X$ in $S$
random function	

The distribution of  $X$  is  
the measure  $Q := \underbrace{P \circ X^{-1}}_p$   
on  $(S, \mathcal{S})$  defined as

$$\begin{aligned} Q(A) &:= P(X^{-1}(A)) \\ &= P(\omega : X(\omega) \in A) \\ &=: P(X \in A). \end{aligned}$$

$P$  is defined on an arbitrary  $\mathcal{F}$ .  
 $Q$  is defined on the Borel  $\sigma(S)$ .

## Notice

I. Suppose  $f$  is measurable  $\mathcal{S}/\mathcal{R}$ ,  
then

$$\begin{aligned} E[f(X)] &= \int f(X(\omega)) dP(\omega) \\ &= \int f(z) dQ(z) \\ &= \int f(X^{-1}(z)) d(P \circ X^{-1}) = Qf. \end{aligned}$$

II. Each prob. measure on each  
metric space is the distribution  
of some random variable on some  
probability space.

$$(\Omega, \mathcal{F}, P) = (S, \mathcal{L}, Q)$$

$$\lambda = \perp$$

## Notice

III. Suppose  $X_n$  has distbn.  $P_n$   
and  $X$  has distbn.  $P$ .

We say  $X_n \Rightarrow X$

("conv. in distbn." or "conv. weakly")

if  $P_n \Rightarrow P$ .

The underlying prob. spaces  
almost never enter the fray.

## Portmanteau Theorem

$$(i) \quad X_n \Longrightarrow X$$

$$(ii) \quad E[f(X_n)] \rightarrow E[f(X)]$$

for bounded unif. cont.  $f$ .

(iii) for closed  $F$ ,

$$\limsup_n P(X_n \in F) \leq P(X \in F) \quad (*)$$

(iv) for open  $G$ , ...

$$(v) \quad P(X_n \in A) \rightarrow P(X \in A), \quad A \in \partial X.$$

Just symbols!

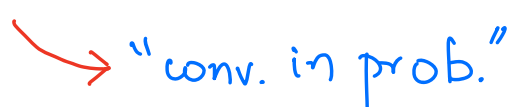
$$\mathbb{P}_n \implies \mathbb{P}$$

$$X_n \implies X$$

$$X_n \implies \mathbb{P}$$

$$\mathbb{P}_n \implies X$$

For an element  $a \in S$ ,

We say  $X_n \xrightarrow{P} a$  if  
 "conv. in prob."

$$P(\rho(X_n, a) > \varepsilon) \rightarrow 0 \quad \forall \varepsilon > 0.$$

- "a" is a const-valued r.v.
- For the above,  $X_n, n \geq 1$  all live in the same space  $\Omega$ .

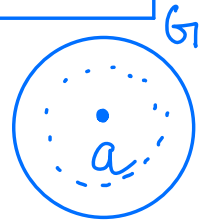
## Theorem 3.0

$$X_n \Rightarrow \delta_a \text{ iff } X_n \xrightarrow{P} a.$$



## Theorem 3.0 Proof Sketch

Suppose  $X_n \xrightarrow{P} a$ .



Take open  $G$  and suppose  $a \in G$ .

$$\liminf_n P_n(G) \geq \liminf_n P(\rho(X_n, a) \leq \varepsilon) = 1.$$

(If  $a \notin G$   $P_n(G) \geq 0 = P(a \in G)$ )

Now suppose  $X_n \xRightarrow{P} a$ . Then.

$$P(\rho(X_n, a) > \varepsilon) =: P_n(B^c(a, \varepsilon)) \rightarrow P(B^c(a, \varepsilon)) = 0.$$

## Theorem 3.1 "Slutsky"

Suppose  $(X_n, Y_n)$  are random elements of  $S \times S$ . If

$$X_n \Rightarrow X \text{ \& } P(X_n, Y_n) \xrightarrow{P} 0$$

then

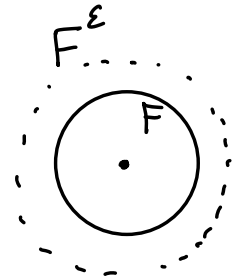
$$Y_n \Rightarrow X$$

$$\left( \begin{array}{l} \text{Slutsky's Thm: } X_n \Rightarrow X \text{ and } A_n \xrightarrow{P} A, \\ B_n \xrightarrow{P} B, \text{ then } A_n X_n + B_n \Rightarrow AX + B \end{array} \right) (*)$$

## Theorem 3.1 Proof Sketch

Take  $F$  closed in  $\mathcal{S}$ .

$$F^\varepsilon := \{x: \rho(x, F) \leq \varepsilon\}$$



Notice:

$$(Y_n \in F) \subseteq (X_n \in F^\varepsilon) \cup (\rho(X_n, Y_n) > \varepsilon)$$

Therefore,

$$P(Y_n \in F) \leq P(X_n \in F^\varepsilon) + P(\rho(X_n, Y_n) > \varepsilon)$$

Take  $\limsup$  and apply Portmanteau,  
and then  $\varepsilon \downarrow 0$ .

## Theorem 3.2

Suppose  $(X_{un}, X_n)$  are random elements of  $S \times S$ . If

$$X_{un} \xRightarrow{n} Z_u \xRightarrow{u} X$$

and

$$\lim_u \limsup_n P(p(X_{un}, X_n) \geq \varepsilon) = 0$$

for  $\varepsilon > 0$ , then  $X_n \xRightarrow{n} X$ .

## Theorem 3.2 Proof Sketch

Following the proof of Thm. 3.1,

$$P(X_n \in F) \leq P(X_{u_n} \in F^\varepsilon) + P(\rho(X_{u_n}, X_n) > \varepsilon)$$

$$\begin{aligned} \limsup_n P(X_n \in F) &\leq P(Z_u \in F^\varepsilon) \\ &\quad + \limsup_n P(\rho(X_{u_n}, X_n) > \varepsilon) \\ &\rightarrow_u P(X \in F^\varepsilon). \end{aligned}$$

Now send  $\varepsilon \downarrow 0$ .

## Integration to the Limit

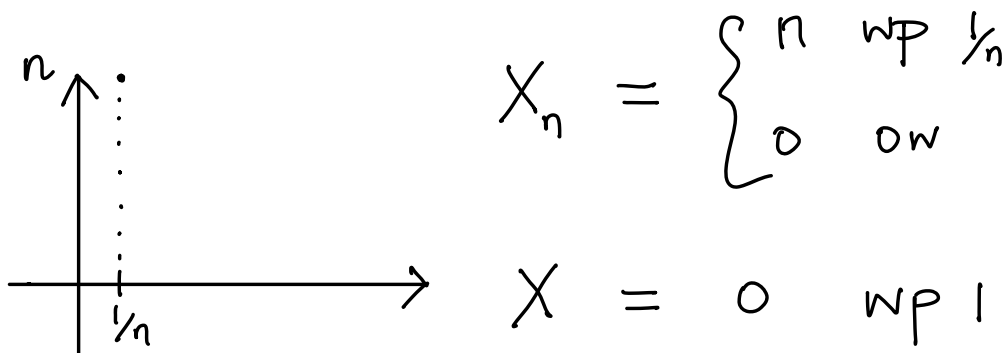
If  $X_n \Rightarrow X$ , what can we say about  $E[X_n]$  in relation  $E[X]$ ?

## Theorem 3.4 (Fatou's Lemma)

For random variables  $X_n, X$   
if  $X_n \Rightarrow X$ , then

$$E[|X|] \leq \liminf_n E[|X_n|].$$

To get the inequality direction:



## Theorem 3.4 Proof Sketch

Due to mapping thm,

$$P(|X_n| > t) \rightarrow P(|X| > t). \quad \text{--- (1)}$$

$$\begin{aligned} E[|X|] &\stackrel{(*)}{=} \int P(|X| > t) \\ &\stackrel{(1)}{=} \int \lim_n P(|X_n| > t) \\ &\stackrel{\text{Fatou}^{(*)}}{\leq} \liminf_n \int P(|X_n| > t). \end{aligned}$$

**Theorem 16.3.** For nonnegative  $f_n$ ,

$$(16.6) \quad \int \liminf_n f_n d\mu \leq \liminf_n \int f_n d\mu.$$

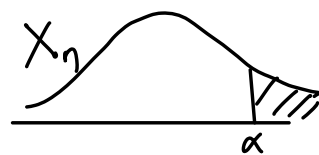


The random variable sequence

$\{X_n, n \geq 1\}$  is uniformly

integrable (U.I) if

$$\lim_{\alpha \rightarrow \infty} \sup_n \int_{|X_n| \geq \alpha} |X_n| dP = 0.$$



If  $\{X_n, n \geq 1\}$  is U.I, then

$$E[|X_n|] \leq 1 + \alpha \quad \forall n.$$

### Theorem 3.5

If  $\{X_n, n \geq 1\}$  are U.I

and  $X_n \Rightarrow X$ , then

$X$  is integrable<sup>(\*)</sup> and

$$E[X_n] \rightarrow E[X].$$

(  $X$  is integrable if  $E[|X|] < \infty$  )  
(  $\Leftrightarrow E[X^+] < \infty$  &  $E[X^-] < \infty$  )

## Theorem 3.5 Proof Sketch

$$E[|X_n|] \leq 1 + \alpha \quad \forall n.$$

Hence,  $E[|X|] < \infty$  and

$X$  is integrable.

Also, it is okay to prove  
 $E[X_n] \rightarrow E[X]$  assuming

$X_n, X$  are non-negative. Why?

## Theorem 3.5 Proof Sketch

$$\mathbb{E}[X_n] = \int X_n \mathbb{I}(X_n < \alpha) dP + \int_{X_n \geq \alpha} X_n dP$$

$$= \int_0^\alpha P(t < X_n < \alpha) dt + \int_{X_n \geq \alpha} X_n dP$$

Similarly,

$$\mathbb{E}[X] = \int_0^\alpha P(t < X < \alpha) dt + \int_{X \geq \alpha} X dP$$

$\leq \varepsilon$  for large  $\alpha$

## Theorem 3.5 Proof Sketch

And, there are at most countable number of  $t, \alpha$  such that

$$P(t < X_n < \alpha) \not\rightarrow P(t < X < \alpha)$$

Use bounded convergence thm (\*)  
to conclude.

**Theorem 16.5.** If  $\mu(\Omega) < \infty$  and the  $f_n$  are uniformly bounded, then  $f_n \rightarrow f$  almost everywhere implies  $\int f_n d\mu \rightarrow \int f d\mu$ .

## Useful U.I Condition.

$$\uparrow \int \sup_n \mathbb{E} [ |X_n|^{1+\varepsilon} ] < \infty, \text{ some } \varepsilon > 0$$

then  $\{X_n, n \geq 1\}$  is U.I.

Notice

$$\mathbb{E} [ |X_n|^{1+\varepsilon} ] = \int_{|X_n| \geq \alpha} |X_n|^{1+\varepsilon} dP + \int_{|X_n| < \alpha} |X_n|^{1+\varepsilon} dP$$

$$\geq \int_{|X_n| \geq \alpha} |X_n| \alpha^\varepsilon dP$$

Therefore,

$$\int_{|X_n| \geq \alpha} |X_n| dP \leq \frac{\mathbb{E} [ |X_n|^{1+\varepsilon} ]}{\alpha^\varepsilon}.$$



