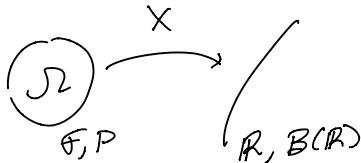


## Lecture 8: Measurable functions, Random variables and distribution functions

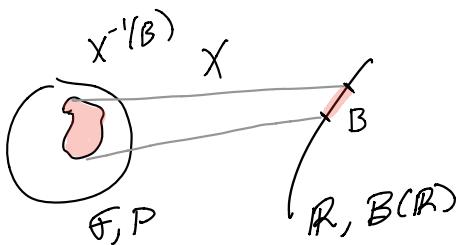
In this lecture we will start by developing the measure theoretic notion of measurable functions & define random variables  $X$  as measurable functions  $X: \Omega \rightarrow \mathbb{R}$  where  $(\Omega, \mathcal{F}, P)$  is a prob. space.



Measurability of  $X$  is required since we want  $P(X \in B)$  to be defined where  $B \in \mathcal{B}(\mathbb{R})$  and  $\{X \in B\} = \{w \in \Omega : X(w) \in B\}$

$$=: X^{-1}(B)$$

$\curvearrowleft$   
pre-image of  $B$  under  $X$ .



## Measurable functions

Let  $(\Omega_1, \mathcal{F}_1)$  &  $(\Omega_2, \mathcal{F}_2)$  be two measurable spaces &  $f: \Omega_1 \rightarrow \Omega_2$ .

Def:  $f$  is measurable between  $\mathcal{F}_1$  &  $\mathcal{F}_2$   
(written  $f @ \mathcal{F}_1/\mathcal{F}_2$  for short) iff

$$f^{-1}(A) \in \mathcal{F}_1, \forall A \in \mathcal{F}_2. \quad (*)$$

Note: It will sometimes be convenient to write  $f @ \mathcal{F}_1/\mathcal{F}_2$  when  $f$  satisfies (\*) even when  $\mathcal{F}_1$  or  $\mathcal{F}_2$  are not  $\sigma$ -fields ... just collections of sets.

A few basic facts about  $f^{-1}(A)$

$$(1) f^{-1}(\Omega_2) = \Omega_1 \text{ since } f \text{ maps into } \Omega_2$$

$$(2) f^{-1}(\emptyset) = \emptyset$$

$$(3) f^{-1}(A^c) = (f^{-1}(A))^c$$

$$\begin{aligned} \text{since } w \in f^{-1}(A^c) &\Leftrightarrow f(w) \in A^c \\ &\Leftrightarrow f(w) \notin A \\ &\Leftrightarrow w \notin f^{-1}(A) \end{aligned}$$

$$(4) f^{-1}\left(\bigcup_p A_p\right) = \bigcup_p f^{-1}(A_p) \text{ even } A_p's \text{ are not disjoint}$$

$$\begin{aligned} \text{since } w \in f^{-1}\left(\bigcup_p A_p\right) &\Leftrightarrow f(w) \in A_p \text{ some } p \\ &\Leftrightarrow w \in f^{-1}(A_p) \text{ some } p \\ &\Leftrightarrow w \in \bigcup_p f^{-1}(A_p) \end{aligned}$$

Thm (Generators are enough)

If  $\Omega_1 \xrightarrow{f} \Omega_2 @ \mathcal{O}$  &  $\mathcal{F}_1$  is a  $\sigma$ -field

then  $f @ \mathcal{F}_1/\mathcal{O} \Leftrightarrow f @ \mathcal{F}_1/\mathcal{O}$ .

Proof:

$\Rightarrow$ : trivial

$\Leftarrow$ : Good sets are

$$\mathcal{Y} = \{A \in \mathcal{R}_2 : f^{-1}(A) \in \mathcal{F}_1\}.$$

$\mathcal{A} \subset \mathcal{Y}$  by assumption &  $\mathcal{Y}$  is a  $\sigma$ -field by facts (1), (3), (4).

QED

e.g.  $\mathbb{R} \xrightarrow{f} (\mathbb{R}, \mathcal{B}(\mathbb{R}))$

$f$  is monotone  $\Rightarrow f^{-1}([-\infty, x])$  is an interval  $\forall x$

show that  $a, b \in f^{-1}(-\infty, x)$   $\Rightarrow f^{-1}(-\infty, x) \in \mathcal{B}(\mathbb{R})$ ,  $\forall x$

$$a \leq b \Rightarrow f(a) \leq f(b) \Rightarrow f^{-1}(-\infty, x) \text{ could be open or closed} \Leftrightarrow f^{-1}(-\infty, x) \in \mathcal{B}(\mathbb{R})$$

$\therefore$  All monotone funcs are measurable.

e.g.  $\mathbb{R}^d \xrightarrow{f} (\mathbb{R}^k, \mathcal{B}(\mathbb{R}^k))$

$f$  is continuous

$$\begin{aligned} &\Leftrightarrow f^{-1}(G) \text{ is open } \forall G \subset \mathbb{R}^k \\ &\Rightarrow f \text{ is } \mathcal{B}(\mathbb{R}^d)/\text{opens in } \mathbb{R}^k \\ &\Leftrightarrow f \text{ is } \mathcal{B}(\mathbb{R}^d)/\underbrace{\{\text{opens in } \mathbb{R}^k\}}_{\mathcal{B}(\mathbb{R}^k)} \end{aligned}$$

This extends to metric spaces  $\mathcal{R}_1, \mathcal{R}_2$  so that

$$f \text{ is } \mathcal{B}(\mathcal{R}_1)/\mathcal{B}(\mathcal{R}_2) \text{ for all metric continuous } f: \mathcal{R}_1 \rightarrow \mathcal{R}_2$$

(3)

Thm (composition of  $\mathcal{M}$  is  $\mathcal{M}$ ).

(4)

If

$$\mathcal{R}_1 \xrightarrow{f} \mathcal{R}_2 \xrightarrow{g} \mathcal{R}_3$$

where  $f \in \mathcal{F}_1/\mathcal{F}_2$  &  $g \in \mathcal{F}_2/\mathcal{F}_3$

then  $g \circ f \in \mathcal{F}_1/\mathcal{F}_3$

Proof:

If  $B \in \mathcal{F}_3$  then

$$(g \circ f)^{-1}(B) = f^{-1} \circ g^{-1}(B)$$

$$= f^{-1} \underbrace{g^{-1}(B)}_{\in \mathcal{F}_2}$$

$$\underbrace{\quad}_{\in \mathcal{F}_1} \quad \text{QED.}$$

e.g. if  $f \in \mathcal{F}/\mathcal{B}(\mathbb{R})$  then

$|f|, f^2, \sin(f) \dots$  are all  $\mathcal{B}(\mathbb{R})/\mathcal{B}(\mathbb{R})$

by continuity.

Corollary (Just check the coordinates)

If  $\mathcal{R} \xrightarrow{f} \mathbb{R}^n \xrightarrow{g} \mathbb{R}^m$  works for  $\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n)$  too

where  $\mathcal{F}$  is a  $\sigma$ -field

and  $f(w) = (f_1(w), \dots, f_n(w))$

then

$$f \in \mathcal{F}/\mathcal{B}(\mathbb{R}^n) \Leftrightarrow \text{each } f_k \text{ is } \mathcal{B}(\mathbb{R})/\mathcal{B}(\mathbb{R})$$

Proof:

$\Rightarrow$ : follows since the coordinate mappings  $\pi_p(x_1, \dots, x_n) = x_p$  are continuous &  $f_p = \pi_p \circ f$

$\Leftarrow$ :

$$f^{-1}((a_1, b_1] \times \dots \times (a_n, b_n]) = \bigcap_{b=1}^n f_p^{-1}([a_p, b_p])$$

rectangle  $\in \mathcal{F}$  since  
 $f_p \in \mathcal{F}/\mathcal{B}(\mathbb{R})$   
 $\in \mathcal{F}$

$\therefore f \in \mathcal{F}/\text{rectangles}$

$\therefore f \in \mathcal{F}/\sigma\langle \text{rectangles} \rangle$

QED.

Thm (Cut & paste over countable &  $\mathbb{N}$  pieces)

If  $\mathcal{R}_1 \xrightarrow{f} \mathcal{R}_2$  where

$\mathcal{F}_1$  &  $\mathcal{F}_2$  are  $\sigma$ -fields and

$$\mathcal{R}_1 = \bigcup_{k=1}^{\infty} A_k \text{ s.t. } A_k \in \mathcal{F}_1, A_k \in \mathcal{F},$$

then

$f \in \mathcal{F}_1/\mathcal{F}_2 \Leftrightarrow f|_{A_k} \in (\mathcal{F}_1 \cap A_k)/\mathcal{F}_2$   
for all  $k = 1, 2, \dots$

(5)

Proof:

$\Rightarrow$ : Suppose  $f \in \mathcal{F}_1/\mathcal{F}_2$ . Let  $B \in \mathcal{F}_2$ .

$$w \in f|_{A_k}^{-1}(B) \Leftrightarrow f(w) \in B \text{ & } w \in A_k$$

$$\Leftrightarrow f(w) \in B \text{ & } w \in A_k$$

$$\therefore f|_{A_k}^{-1}(B) = A_k \cap \underbrace{f^{-1}(B)}_{\in \mathcal{F}_1} \in \mathcal{F}_1 \cap A_k$$

$\Leftarrow$ : Suppose  $f|_{A_k} \in (\mathcal{F}_1 \cap A_k)/\mathcal{F}_2$ . Let  $B \in \mathcal{F}_2$ .

$$f^{-1}(B) = f^{-1}(B) \cap \mathcal{R}_1$$

$$= f^{-1}(B) \cap \bigcup_k A_k$$

$$= \bigcup_k (f^{-1}(B) \cap A_k)$$

$$= \bigcup_k f|_{A_k}^{-1}(B)$$

$\in \mathcal{F}_1 \cap A_k$  by assumption

$\in \mathcal{F}_1$  since  $\mathcal{F}_1 \cap A_k \subset \mathcal{F}_1$   
&  $\mathcal{F}_1$  is a  $\sigma$ -field

QED.

Corollary: Piecewise metric

continuous functions are  $\in \mathcal{F}$   
if the "Pieces" are countable &  
Borel measurable.

Thm: (just check  $\in \mathcal{F}$  on the range)

If  $f: \mathcal{R}_1 \rightarrow \mathcal{R}_2 \subset \mathcal{R}_2$  metric space

and  $\mathcal{F}_1$  is a  $\sigma$ -field on  $\mathcal{R}_1$ , then

$$f \in \mathcal{F}_1/\mathcal{B}(\mathcal{R}_2) \Leftrightarrow f \in \mathcal{F}_1/\mathcal{B}(\mathcal{R}_2^\circ).$$

(6)

Proof: The borel restriction thm says (7)

$$B(\mathcal{J}_2^o) = B(\mathcal{J}_2) \cap \mathcal{J}_2^o.$$

$$\begin{aligned} f @ \mathbb{F}_1 / B(\mathcal{J}_2^o) &\iff f @ \mathbb{F}_1 / B(\mathcal{J}_2) \cap \mathcal{J}_2^o \\ &\iff \underbrace{f^{-1}(B \cap \mathcal{J}_2^o)}_{\text{since } f \text{ maps into } \mathcal{J}_2^o} \in \mathbb{F}_1, \forall B \in B(\mathcal{J}_2) \\ &\iff f^{-1}(B) \in \mathbb{F}_1, \forall B \in B(\mathcal{J}_2) \\ &\quad \text{since } f \text{ maps into } \mathcal{J}_2^o \\ &\iff f @ \mathbb{F}_1 / B(\mathcal{J}_2). \end{aligned}$$

QED.

e.g.  $\sin(x) @ B(\mathbb{R}) / B(\mathbb{R})$

$$\iff \sin(x) @ B(\mathbb{R}) / B([-1, 1])$$

Question:

is  $f(x) = \begin{cases} \frac{1}{x}, & x > 0 \\ -\infty, & x = 0 \\ \sin(x), & x < 0 \end{cases} @ B(\mathbb{R}) / B(\bar{\mathbb{R}})$

Yes since  $f$  is metric continuous on countably many measurable pieces.

Notation:

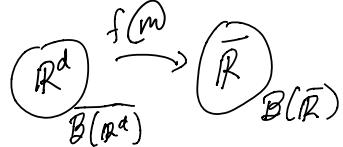
- $f$  is " $\mathbb{F}$ -measurable" if



- $f$  is "Borel measurable" if



- $f$  is "Lebesgue measurable" if

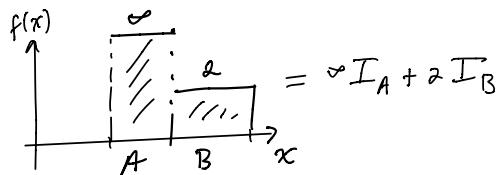


when  $\overline{B(\mathbb{R}^d)}$  is the completion w.r.t. Lebesgue measure-

Convention for  $\infty$

- $\infty + x = \infty$  when  $x \in (-\infty, \infty]$
- $\infty \cdot 0 = 0$
- $\infty \cdot \infty = \infty$
- $\frac{x}{\infty} = 0$  when  $x \neq \infty$
- $\frac{x}{0}$ ,  $\frac{\pm\infty}{\pm\infty}$ ,  $\infty - \infty$  are not defined.

Note: we use the convention  $\infty \cdot 0 = 0$  so



## Thm (Closure thm)

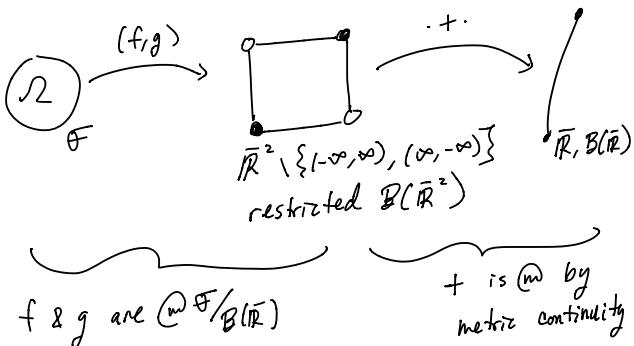
Let  $(\Omega, \mathcal{F})$  be a measurable space.

(i) if  $\begin{array}{c} \Omega \\ \mathcal{F} \end{array} \xrightarrow{\substack{f \in \mathbb{R} \\ g \in \mathbb{R}}} \begin{array}{c} \bar{\mathbb{R}} \\ \mathcal{B}(\bar{\mathbb{R}}) \end{array}$  then  
 $f+g, f \cdot g, f/g, \max(f, g), \min(f, g), f^+, f^-, |f|$  are all  $\cap \mathcal{F}/\mathcal{B}(\bar{\mathbb{R}})$   
provided they are defined  $\forall \omega \in \Omega$   
*i.e. No*  $\infty - \infty, \frac{\infty}{\infty} \dots$

(ii) if  $\begin{array}{c} \Omega \\ \mathcal{F} \end{array} \xrightarrow{\substack{f_n \in \mathbb{R} \\ \vdots \\ f_n \in \mathbb{R}}} \begin{array}{c} \bar{\mathbb{R}} \\ \mathcal{B}(\bar{\mathbb{R}}) \end{array}$   
then  $\sup_n f_n, \inf_n f_n, \limsup_n f_n$   
and  $\liminf_n f_n$  are  $\cap \mathcal{F}/\mathcal{B}(\bar{\mathbb{R}})$ .

Proof:

(i) Just show  $f+g \dots$  the others are similar.  
since  $f(\omega) + g(\omega)$  is defined  $\forall \omega \in \Omega$  we have



$\Leftrightarrow (f, g)$  is  $\cap$  on the  
restricted space, by  
"just check the range"

$\therefore f+g \in \mathcal{F}/\mathcal{B}(\bar{\mathbb{R}})$  by "Composition of  
 $\cap$  is  $\cap$ ".

(9)

To show (ii) just notice

$$\left( \sup_n f_n \right)' \left( [-\infty, c] \right) = \left\{ \omega : \sup_n f_n(\omega) \leq c \right\}$$

$$= \bigcap_{n=1}^{\infty} \left\{ \omega : f_n(\omega) \leq c \right\}$$

Note: this will not be true for  $\in \mathcal{F}$  since  $f_n \in \mathcal{F}/\mathcal{B}(\bar{\mathbb{R}})$   
 $\in \mathcal{F}$  since  $\mathcal{F}$  is a  $\sigma$ -field.

$$\therefore \sup_n f_n \in \mathcal{F}/\cap_{[-\infty, c]} : c \in \bar{\mathbb{R}}$$

$$\therefore \sup_n f_n \in \mathcal{F}/\underbrace{\cap_{[-\infty, c]} : c \in \bar{\mathbb{R}}}_{= \mathcal{B}(\bar{\mathbb{R}})}$$

For the others

$$\inf_n f_n = - \sup_n (-f_n)$$

$$\limsup_n f_n = \lim_{m \rightarrow \infty} \sup_{n \geq m} f_n = \inf_m \sup_{n \geq m} f_n$$

decreases as  $m \rightarrow \infty$

$\liminf_n f_n \dots$  similar.

QED.

e.g. Coin flip model from lecture 1.

$$S_n(\omega) = \sum_{k=1}^n R_k(\omega) \text{ maps } \{0,1\} \rightarrow \bar{\mathbb{R}}$$

Since  $S_n$  is constant over intervals  $(\frac{i-1}{2^n}, \frac{i}{2^n}]$   
we have  $S_n^{-1}([-x, x]) = \text{finite disjoint union of dyadic}$   
union of dyadic

By "generators are enough"



$\therefore \limsup_n \frac{S_n}{\sqrt{2n \log n}}$  is  $\cap \mathcal{B}((0,1))/\mathcal{B}(\bar{\mathbb{R}})$

$\therefore \left\{ \limsup_n \frac{S_n}{\sqrt{2n \log n}} = 1 \right\} \in \mathcal{B}((0,1))$  since  
it is the pre-image of  $\{1\} \in \mathcal{B}(\bar{\mathbb{R}})$ .

(10)

## Random variables, induced measures and c.d.f.s

(11)

Def:  $X$  is a random variable if there exists a probability space  $(\Omega, \mathcal{F}, P)$  where  $X: \Omega \rightarrow \mathbb{R}$  s.t.  $X \in \mathcal{F}/B(\mathbb{R})$ .

$X$  is an extended random variable if  $X: \Omega \rightarrow \bar{\mathbb{R}}$  &  $X \in \mathcal{F}/B(\bar{\mathbb{R}})$ .

Picture:

we write  $X(\omega)$  instead of  $f(\omega)$  to indicate  $(\Omega, \mathcal{F})$  has a probability measure attached. Think of  $P$  as modeling a random draw  $\omega \in \Omega$  &  $X(\omega)$  as a "variable" or "label" associated with each  $\omega \in \Omega$ .

Since  $X \in \mathcal{F}/B(\mathbb{R})$  it makes sense to talk about quantities like:

$$P(X=1) = P\left(X^{-1}\left(\underbrace{\{1\}}_{\in B(\mathbb{R})}\right)\right)$$

$$P(X \leq x) = P\left(X^{-1}\left(\underbrace{(-\infty, x]}_{\in \mathcal{F}}\right)\right)$$

$$P(X \in \mathbb{Q}) = P(X^{-1}(\mathbb{Q})).$$

Def: If  $X$  is a random variable defined on  $(\Omega, \mathcal{F}, P)$ , the distribution of  $X$  (also called the induced probability measure) is a set function  $PX^{-1}: B(\mathbb{R}) \rightarrow [0, 1]$  given by

$$PX^{-1}(B) := P(X^{-1}(B)) = P(X \in B).$$

More generally if



where  $\mathcal{F}_1, \mathcal{F}_2$  are  $\sigma$ -fields &  $(\Omega_1, \mathcal{F}_1, \mu)$  is a measure then

$$\mu f^{-1}(F) := \mu(f^{-1}(F)), \quad \forall F \in \mathcal{F}_2$$

$\nwarrow$  induced measure on  $(\Omega_2, \mathcal{F}_2)$ .

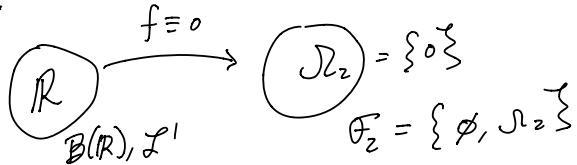
Thm: In the setup above  $\mu f^{-1}$  is a measure on  $(\Omega_2, \mathcal{F}_2)$

Proof: Follows immediately since pre-images & set operations commute.  $\square$

Since  $\mu f^{-1}(\mathcal{R}_2) = \mu(\mathcal{R}_1)$  we have the following facts:

- $\mu(\mathcal{R}_1) = 1 \Rightarrow \mu f^{-1}(\mathcal{R}_2) = 1$   
i.e.  $PX^{-1}$  is a probability measure
- $\mu(\mathcal{R}_1) < \infty \Rightarrow \mu f^{-1}(\mathcal{R}_2) < \infty$
- Warning:**  
 $\mu$  is a  $\sigma$ -finite measure  
 $\not\Rightarrow \mu f^{-1}$  is a  $\sigma$ -finite measure

e.g.



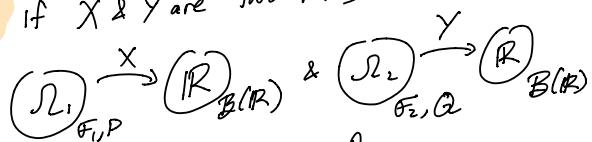
f is  $B(R)/F_2$  measurable and  $\mathbb{Z}'$  is  $\sigma$ -finite but  $\mathbb{Z}'f^{-1}$  is not  $\sigma$ -finite since

$$\mathbb{Z}'f^{-1}(\phi) = 0 \text{ and}$$

$$\mathbb{Z}'f^{-1}(\mathcal{R}_2) = \mathbb{Z}'(R) = \infty.$$

Notice that even if two r.v.s  $X$  &  $Y$  are defined on different probability spaces the induced distributions are both on  $(R, B(R))$ .

**Def:** If  $X$  &  $Y$  are two r.v.s s.t.



we write  $X \sim Y$  or  $X \stackrel{d}{=} Y$  if

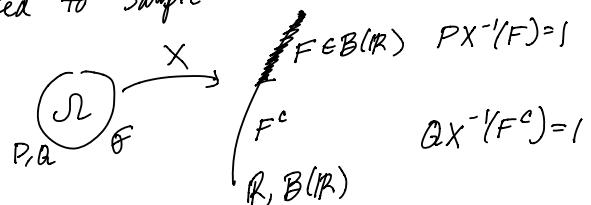
$$PX^{-1} = QY^{-1} \text{ over } B(R).$$

(13)

In the simple statistics set up (14)  
there are two possible measures  $P$  &  $Q$ , both defined on  $(\mathcal{R}, \mathcal{F})$ .  
 $w \in \mathcal{R}$  is picked at random either from  $P$  or  $Q$  & the "observer" only gets to see the value of  $X(w)$  (i.e. the data).

The observer then tries to figure out which  $P$  or  $Q$   $w$  was sampled from.

Note: using the notation from Hawk 3 if  $PX^{-1} \perp QX^{-1}$  then the observer will know exactly which  $P$  or  $Q$  was used to sample  $w$



- If  $X(w) \in F$  then  $w$  was drawn from  $P$ .

- If  $X(w) \in F^c$  then  $w$  was drawn from  $Q$ .

**Def:** The distribution function (sometimes called cumulative distribution function.. cdf) for a random variable  $X$  defined on  $(\mathcal{R}, \mathcal{F}, P)$  is the function  $F: \mathbb{R} \rightarrow [0, 1]$  defined as

$$F(t) = P(X \leq t), \quad \forall t \in \mathbb{R}.$$

$$= PX^{-1}(-\infty, t]$$

Thm: If  $X$  &  $Y$  are two r.v.s with c.d.f.s  $F_X$  &  $F_Y$ , respectively, then

$$X \stackrel{D}{=} Y \Leftrightarrow F_X(t) = F_Y(t) \quad \forall t \in \mathbb{R}.$$

Proof:

$\Rightarrow$ : trivial when taking  $B = (-\infty, t]$

$$\text{since } X^{-1}(-\infty, t] = \{X \leq t\}$$

$\Leftarrow$ :

$$F_X(t) = F_Y(t) \quad \forall t \in \mathbb{R}$$

$$\Rightarrow P(X^{-1}(-\infty, t]) = P(Y^{-1}(-\infty, t]) \quad \forall t \in \mathbb{R}.$$

$$\Rightarrow P(X^{-1}) = P(Y^{-1}) \text{ on } \mathcal{P} = \{(-\infty, t] : t \in \mathbb{R}\}$$

$$\stackrel{\text{II-ning}}{\Rightarrow} P(X^{-1}) = P(Y^{-1}) \text{ on } \sigma(\emptyset) = \mathcal{B}(\mathbb{R}).$$

↑ π-sys

AED

(15)

Proof:

(I): clear since  $A \subset B \Rightarrow P(X^{-1}(A)) \leq P(X^{-1}(B))$

(II):

$$\{X \leq x\} \downarrow \{X \leq y\} \text{ as } x \downarrow y$$

which follows since if  $x_n \downarrow y$  as  $n \rightarrow \infty$   
 then clearly  $\{X \leq y\} \subset \bigcap_{n=1}^{\infty} \{X \leq x_n\}$  &  
 $\omega \in \bigcap_{n=1}^{\infty} \{X \leq x_n\} \Rightarrow X(\omega) \leq x_n \quad \forall n$   
 $\Rightarrow X(\omega) \leq \lim_n x_n = y$

$$\therefore P(X \leq x) \downarrow P(X \leq y)$$

warning: It is not true that  
 $\{X < x\} \downarrow \{X < y\} \text{ as } x \downarrow y$   
 since  $X(n) < x_n \not\Rightarrow X(\omega) < \lim x_n = y$   
 e.g. take  $X(n) = 0, x_n = \frac{1}{n}, y = 0$

(III): Follows since

$$\{X \leq x\} \uparrow \mathbb{R} \quad \& \quad \{X \leq x\} \downarrow \emptyset.$$

AED

Thm (Properties of c.d.f.s)

Let  $F(t) = P(X \leq t)$  for a r.v.  $X$  defined on  $(\Omega, \mathcal{F}, P)$ . Then

$$(I) \quad F(x) \leq F(y), \quad \forall x \leq y$$

$$(II) \quad \lim_{x \downarrow y} F(x) = F(y)$$

$$(III) \quad \lim_{x \rightarrow \infty} F(x) = 1 \quad \& \quad \lim_{x \rightarrow -\infty} F(x) = 0$$

It turns out properties (I), (II), (III) are characterizing properties of c.d.f.s i.e. if  $F: \mathbb{R} \rightarrow [0, 1]$  satisfies

(I), (II), (III) then  $\exists$  a r.v.  $X$

$$\text{s.t. } F(t) = P(X \leq t).$$

In fact, this will be the main tool we use to show the existence of an infinite sequence of indep. r.v.s all with a specified distribution.

(16)

D.of: If  $F: \mathbb{R} \rightarrow [0,1]$  satisfies  
(I), (II) & (III) of the above then

define

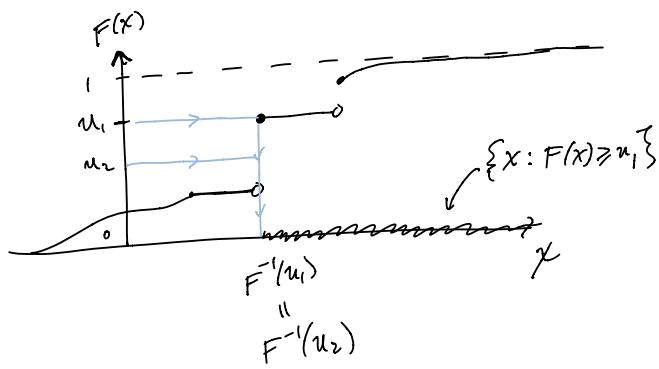
$$F^{-1}(u) := \inf \{x \in \mathbb{R} : F(x) \geq u\}$$

often called the  
inverse c.d.f or the  
quantile function.

for  $u \in (0,1)$ .

Note:  $F^{-1}: (0,1) \rightarrow \mathbb{R}$  is well defined  
since  $\{x : F(x) \geq u\} \neq \emptyset$  when  $0 < u < 1$   
by (III) & (I).

Picture:



There are two important facts about  $F$  &  $F^{-1}$  which are useful to remember.

Suppose  $U$  is a r.v. which is uniform on  $(0,1)$ , i.e.  $\exists (\Omega, \mathcal{F}, P)$  &  $\xrightarrow[\mathcal{F}, P]{} U \sim \text{Uniform}(0,1)$

$$\text{s.t. } P(U \in B) = \mathcal{J}(B).$$

(17)

For example, digit coin flip model:

$$\Omega = \{0,1\}, \mathcal{F} = \mathcal{B}(\{0,1\}), P = \mathcal{J}|_{\{0,1\}}$$

where

$$U(w) = \begin{cases} w & \text{if } w \leq 1 \\ 1 & \text{if } w = 1 \end{cases}$$

is (17) by cut & paste &

$$P(U \in B) = P(B) = \mathcal{J}(B)$$

↑ since  $P(U=1) = 0$

Now the two important facts are ...

if  $F$  is the c.d.f of a r.v.  $X$  then

$$(a) F^{-1}(u) \stackrel{D}{=} X$$

$$(b) F(X) \stackrel{D}{=} U \text{ when } F \text{ is continuous.}$$

Lemma (switching formula):

If  $F: \mathbb{R} \rightarrow [0,1]$  satisfies (I), (II) & (III)  
above then

$$(*) F(x) \geq u \iff x \geq F^{-1}(u)$$

$\forall u \in (0,1) \text{ & } x \in \mathbb{R}$ .

Proof:

A useful restatement of (\*) is simply that

$$(**) \{x : F(x) \geq u\} = [F^{-1}(u), \infty).$$

To show (\*\*), notice that

$\{x : F(x) \geq u\}$  must be an interval of the form  $[F^{-1}(u), \infty)$  or  $(F^{-1}(u), \infty)$  since

$$F^{-1}(u) \stackrel{\text{def}}{=} \inf \{x : F(x) \geq u\}$$

and

$$x' \in \{x : F(x) \geq u\} \xrightarrow{\text{by (I)}} \left\{ \begin{array}{l} x'' \in \{x : F(x) \geq u\} \\ x'' > x' \end{array} \right\}$$

$\therefore$  to show  $\{x : F(x) \geq u\} = [F^{-1}(u), \infty)$  just (1a)  
prove the inf of the lts is attained.

Let  $x_p \in \{x : F(x) \geq u\}$  s.t.  $x_p \downarrow \underbrace{F^{-1}(u)}_{=\inf \text{ of lts}}$

$\therefore F(\lim_p x_p) \stackrel{(II)}{=} \lim_p F(x_p) \geq u$   
 $\uparrow$  since  $x_p$  is  
in  $\{x : F(x) \geq u\}$

$\therefore F^{-1}(u) = \lim_p x_p \in \{x : F(x) \geq u\}$   
 $\uparrow$   
so the inf is attained.

(Note we used III implicitly to show  
 $F^{-1}(u)$  is well defined and a r.v.)

QED

### Lemma (c.d.f sandwich)

If  $F: \mathbb{R} \rightarrow [0,1]$  satisfies (I), (II) & (III)  
then  $\forall u \in (0,1)$   
 $F(F^{-1}(u)-) \leq u \leq F(F^{-1}(u)).$

Proof:  
 $u \leq F(F^{-1}(u))$  holds since  $F^{-1}(u) \in \{x : F(x) \geq u\}$ .

The contrapositive of the switching formula is

$$F(x) < u \Leftrightarrow x < F^{-1}(u)$$

$\therefore F(F^{-1}(u)-) = \lim_{x \uparrow F^{-1}(u)} \underbrace{F(x)}_{< u} \leq u$   
 $\text{these satisfy } x < F^{-1}(u)$   
so  $F(x) < u$  by switch

QED

Maybe the best way to remember these

$$\{x : F(x) \geq u\} = [F^{-1}(u), \infty)$$

$$F(F^{-1}(u)-) \leq u \leq \underbrace{F(F^{-1}(u)+)}_{= F(F^{-1}(u)) \text{ by (II)}}$$

### Thm (c.d.f representation)

(20)  
If  $F: \mathbb{R} \rightarrow [0,1]$  satisfies (I), (II) & (III)  
then  $\exists$  a r.v.  $X$  on a prob space  $(\Omega, \mathcal{F}, P)$   
s.t.

$$(1) \quad P(X \leq x) = F(x) \quad \forall x \in \mathbb{R}$$

and, moreover, any rv  $U$  which is  
uniformly distributed on  $(0,1)$  satisfies

$$(2) \quad F^{-1}(U) \stackrel{D}{=} X$$

$$(3) \quad P(F(X) \leq u) \leq u \quad \text{for all } u \in (0,1).$$

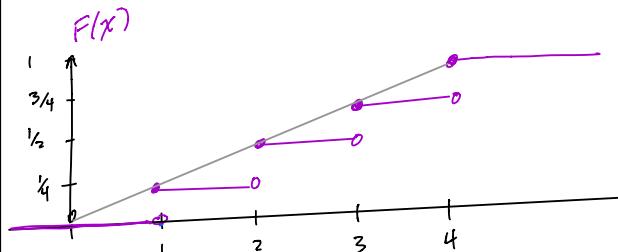
If, in addition,  $F$  is continuous then

$$(4) \quad F(X) \stackrel{D}{=} U.$$

Here is the picture of why we don't  
always get (4).

Suppose  $P(X=i) = \frac{1}{4}$  for  $i=1, 2, 3, 4$   
So  $X$  is uniformly distributed on  $\{1, 2, 3, 4\}$

The c.d.f of  $X$  is



$\therefore F(X)$  assigns  $\frac{1}{4}$  prob  
to  $\{\frac{1}{4}, \frac{1}{2}, \frac{3}{4}, 1\}$ .

$$\text{i.e. } P(F(X) \leq 0.25) = 0.25$$

$$\text{but } P(F(X) \leq 0.3) = 0.25 \leq 0.3$$

Proof:

(21)

First note that  $F^{-1}$  is CDF by monotonicity.

Let  $U$  be a r.v uniform on  $(0,1)$ .

$$\begin{aligned}\therefore P(F^{-1}U \leq x) &= P(U \leq F(x)) \\ &\quad \text{by switching lemma} \\ &= F'(F(x)) \quad U \text{ is uniform} \\ &= F(x)\end{aligned}$$

Now set  $X := F^{-1}U$  to get (1) & (2) using  
1-1 uniqueness.

For (3),

$$\begin{aligned}P(F(X) \leq u) &= P(F(F^{-1}U) \leq u) \\ &\quad \text{since } X = F^{-1}U \\ &\leq P(U \leq u) \\ &\quad \text{since } U \leq F(F^{-1}U) \leq u \\ &\quad \uparrow \\ &\quad \text{by c-d-f sandwich} \\ &= u, \quad \forall u \in (0,1)\end{aligned}$$

If  $F$  is continuous c-d-f sandwich

gives  $F(F^{-1}U^-) = U = F(F^{-1}U)$  so

$$P(F(X) \leq u) = P(U \leq u)$$

which shows (4)

QED