Some overdue derivations

Define the left-continuous (how do we know?) inverse of F

$$F^{\longleftarrow}(y) = \inf\{s : F(s) \ge y\}, \ 0 < y \le 1$$

 $A(y) = \{s : F(s) \ge y\}$

Properties of A(y): (a) A(y) is closed Indeed

$$A(y)
i s_n \downarrow s \implies F(s_n) \downarrow F(s) \ge y \implies s \in A(y)$$

$$A(y)
i s_n \uparrow s \implies y \le F(s_n) \uparrow F(s-0) \le F(s) \implies s \in A(y)$$

 \square

(b) Therefore,

$$\inf A(y) \in A(y) \Longleftrightarrow F(F^{\leftarrow}(y)) \ge y$$

The part on the left follows by closedness.

But this means that the defining property $F(s) \geq y$ holds for the $s = \inf A(y) = F^{\longleftarrow}(y)$

(c) Furthermore

$$[F^{\longleftarrow}(y) > t \iff y > F(t)] \iff [F^{\longleftarrow}(y) \le t \iff y \le F(t)]$$
 Indeed,

$$t < F^{\longleftarrow}(y) \iff t \notin A(y) \iff F(t) < y$$
$$t \ge F^{\longleftarrow}(y) \iff t \in A(y) \iff F(t) \ge y$$

Now we can prove that $F^{\leftarrow}(y)$ is left continuous

$$y_n \uparrow y \implies F^{\longleftarrow}(y_n) \le F^{\longleftarrow}(y_{n+1}) \le F^{\longleftarrow}(y)$$

Therefore

$$\phi = \lim_{n \uparrow \infty} F^{\longleftarrow}(y_n) \le F^{\longleftarrow}(y)$$

How to exclude $\phi < F^{\leftarrow}(y)$?

$$\forall n : F^{\leftarrow}(y_n) \le \phi \stackrel{(c)}{\Longrightarrow} y_n \le F(\phi)$$
$$\implies y \le F(\phi) \stackrel{(c)}{\Longrightarrow} F^{\leftarrow}(y) \le \phi$$

RANDOM VARIABLES, ELEMENTS AND MEASURABLE MAPS

Intuitive introduction (blackboard)

- Reproducible repeat experiments
- How to measure outcomes of experiments?
- Intuitive notion of random variable

- Random variable as a function
- Mathematical conditions to be satisfied by random variables: measurability
- Induced probability measures = distributions of rv's
- Generalizations: Stochastic processes

Inverse Maps

$$X: \Omega \longrightarrow \Omega'$$
$$X^{-1}: 2^{\Omega'} \longrightarrow 2^{\Omega}: X^{-1}(A') = \{\omega \in \Omega: X(\omega) \in A'\}$$

 $(drawing \dots)$

Properties (prove ...): (i) $X^{-1}(\emptyset) = \emptyset; \ X^{-1}(\Omega') = \Omega$ (ii) $X^{-1}(A'^{c}) = X^{-1}(A')^{c}$ (iii) $X^{-1}(\bigcup_{t} A'_{t}) = \bigcup_{t} X^{-1}(A'_{t})$

Indeed

$$\begin{aligned} X^{-1}(\bigcup_{t} A'_{t}) &= \{\omega \in \Omega : X(\omega) \in \bigcup_{t} A'_{t}\} \\ &= \{\omega \in \Omega : [\exists t : X(\omega) \in A'_{t}]\} \\ &= \{\omega \in \Omega : [\exists t : \omega \in X^{-1}(A'_{t})]\} \\ &= \bigcup_{t} X^{-1}(A'_{t}) \end{aligned}$$

(iv)

$$X^{-1}(\bigcap_{t} A'_{t}) = \bigcap_{t} X^{-1}(A'_{t})$$

Notation If $\mathcal{C}' \subset 2^{\Omega'}$ $X^{-1}(\mathcal{C}') := \{X^{-1}(C') : C' \in \mathcal{C}'\}$

Proposition 3.1.1. If \mathcal{B}' is a σ -field of subsets of Ω' , then $X^{-1}(\mathcal{B}')$ is a σ -field of subsets of Ω .

Proof
(i)

$$\Omega' \in \mathcal{B}' \implies \Omega = X^{-1}(\Omega') \in X^{-1}(\mathcal{B}')$$
(ii)

$$A' \in \mathcal{B}' \implies X^{-1}(A') \in X^{-1}(\mathcal{B}')$$

$$A' \in \mathcal{B}' \implies A'^c \in \mathcal{B}' \implies X^{-1}(A'^c) = X^{-1}(A')^c \in X^{-1}(\mathcal{B}')$$

therefore

$$X^{-1}(A') \in X^{-1}(\mathcal{B}') \implies X^{-1}(A')^c \in X^{-1}(\mathcal{B}')$$

(iii)

$$X^{-1}(A'_n) \in X^{-1}(\mathcal{B}') \implies \bigcup_n X^{-1}(A'_n) = X^{-1}(\bigcup_n A'_n) \in X^{-1}(\mathcal{B}')$$

Proposition 3.1.2. If $\mathcal{C}' \in 2^{\Omega'}$, then

$$X^{-1}(\sigma(\mathcal{C}')) = \sigma(X^{-1}(\mathcal{C}'))$$

Proof self-study

Measurable maps, random elements, induced probability measures

Measurable space: (Ω, \mathcal{B})

Measurable map: Given (Ω, \mathcal{B}) and (Ω', \mathcal{B}') , a map

 $X:\Omega\longrightarrow\Omega'$

is called measurable if

 $X^{-1}(\mathcal{B}') = \mathcal{B}$

X is also called a random element of Ω' :

$$X \in \mathcal{B}/\mathcal{B}' \iff X : (\Omega, \mathcal{B}) \longrightarrow (\Omega', \mathcal{B}')$$

This latter is by abuse of notation (why?)

Random variable is a special case of random element when $(\Omega', \mathcal{B}') = (\mathbb{R}, \mathcal{B}(\mathbb{R}))$

Random vector is a special case of random element when $(\Omega', \mathcal{B}') = (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n))$ Induced probabilities

Let (Ω, \mathcal{B}, P) be a probability space and $X : (\Omega, \mathcal{B}) \longrightarrow (\Omega', \mathcal{B}')$ a measurable map

Define

$$[X \in A'] := X^{-1}(A') = \{\omega : X(\omega) \in A'\}$$

and a set function $P \circ X^{-1}$ on \mathcal{B}' by

$$P \circ X^{-1}(A') = "(P \circ X^{-1})(A')" = P(X^{-1}(A'))$$

Is $P \circ X^{-1}$ a (*induced*) probability measure on \mathcal{B}' ? Yes!

Verify $\dots(!)$

Notation and jargon

$$P \circ X^{-1}(A') = P[X \in A']$$

If X is a rv (random variable), then $P \circ X^{-1}$ is a measure induced on \mathbb{R} by the distribution function

$$P \circ X^{-1}(-\infty, x] = P[X \le x] = F_X(x)$$

Induced distribution is a condensed summary of the probability measure on (Ω, \mathcal{B}) , which is sufficient to describe all events (measurable sets) induced by rv X.

Example: Counts of nucleotides in a random DNA sequence

Consider a DNA sequence of length N.

Each site can be independently occupied by A, C, G, or T equally likely.

$$\Omega = \{ \omega = (\omega_1, \omega_2, \dots, \omega_N), \ \omega_i \in \{A, C, G, T\} \}$$
$$\mathcal{A} = \sigma(\{\omega\}) = 2^{\Omega}$$

The set function

$$P(A) = \#\{A\} \cdot 4^{-N}, \ A \in \mathcal{A}$$

satisfies axioms of probability measure (this is an example of combinatorial probability).

Define

$$X: \Omega \to \Omega' := \{0, 1, 2, \dots, N\}$$
$$X(\omega) = \#\{G \text{ in } \omega\}$$
$$X^{-1}(\{0\}) = \{(\ G, \ G, \dots, \ G)\}$$

etc.

$$P'(\{i\}) = P \circ X^{-1}(\{i\})$$

= $P(X^{-1}(\{i\}))$
= $\begin{cases} (3/4)^N & i = 0\\ (1/4)(3/4)^{N-1} & i = 1\\ \dots & \\ \binom{N}{i}(1/4)^i(3/4)^{N-i} & i\\ \dots & \\ (1/4)^N & i = N \end{cases}$

Test for measurability

Proposition 3.2.1. Suppose

$$X:\Omega\to\Omega'$$

where (Ω, \mathcal{B}) and (Ω', \mathcal{B}') are measurable spaces. Suppose

$$\mathcal{B}'=\sigma(\mathcal{C}')$$

Then X is measurable iff

$$X^{-1}(\mathcal{C}') \subset \mathcal{B}$$

i.e. we do not have to check

$$X^{-1}(\sigma(\mathcal{C}')) \subset \mathcal{B}$$

Proof by counterimage properties (self-study)

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Corollary 3.2.1. The real-valued function

$$X:\Omega\to\mathbb{R}$$

is a random variable iff

$$X^{-1}((-\infty,\lambda]) = [X \le \lambda] \in \mathcal{B}, \ \forall \lambda \in \mathbb{R}$$

Topics for whiteboard explanation

- Composition of measurable maps
- Random elements of metric spaces
- Measurability and continuity
- Maesurability and limits
- σ -fields generated by maps

σ -fields generated by maps

Consider rv $X : (\Omega, \mathcal{B}) \longrightarrow (\mathbb{R}, \mathcal{B}(\mathbb{R}))$

The σ -field generated by X

$$\sigma(X) = X^{-1}[\mathcal{B}(\mathbb{R})]$$

or

 $\sigma(X) = \{ [X \in A], A \in \mathcal{B}(\mathbb{R}) \} = (!) \sigma\{ [X \in A], A \in \mathcal{B}(\mathbb{R}) \}$

More general construction, if $X : (\Omega, \mathcal{B}) \longrightarrow (\Omega', \mathcal{B}')$ $\sigma(X) = X^{-1}(\mathcal{B}')$

If $\mathcal{F} \subset \mathcal{B}$, is a sub- σ -field of \mathcal{B} and

 $\sigma(X) \subset \mathcal{F} \subset \mathcal{B}$

then we say X is measurable wrt \mathcal{F} .

$\sigma\mbox{-fields}$ generated by an indexed set of maps (ie. by almost a process)

Take $t \in T$. Then

$$X_t: (\Omega, \mathcal{B}) \longrightarrow (\Omega', \mathcal{B}')$$

is a family of maps (almost random elements), and

$$\sigma(X_t, t \in T) = \bigvee_{t \in T} \sigma(X_t) = \sigma\left[\bigcup_{t \in T} \sigma(X_t)\right]$$

is the smallest σ -field containing all $\sigma(X_t)$.

Examples:

 $X = c, X = 1_A$ (whiteboard)

Further examples:

Simple functions, σ -fields generated by sequences of rv's

Proposition 3.3.1. X is a rv and $\mathcal{C} \in 2^{\mathbb{R}}$ st

$$\sigma(\mathcal{C}) = \mathcal{B}(\mathbb{R})$$

Then

$$\sigma(X) = \sigma\{[X \in B], \ B \in \mathcal{C}\}$$

Proof

$$\sigma\{[X \in B], B \in \mathcal{C}\} = \sigma\{X^{-1}(B), B \in \mathcal{C}\}$$
$$= \sigma\{X^{-1}(\mathcal{C})\}$$
$$= X^{-1}\{\sigma(\mathcal{C})\}$$
$$= X^{-1}\{\mathcal{B}(\mathbb{R})\}$$
$$= \sigma(X)$$

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