

Zero-One Law

• Borel-Cantelli $\Rightarrow \forall$ independent events E_1, E_2, \dots satisfy
 $P\{E_n \text{ occur i.o.}\} \in \{0, 1\}$.

• General framework:

Let X_1, X_2, \dots be independent random variables.

Recall (p. 37-39):

(a) $\sigma(X_1, \dots, X_n)$ is generated by events $\{X_1 \in B_1, \dots, X_n \in B_n\}$ where $B_i \in \mathcal{B}(\mathbb{R})$

(b) $\sigma(X_1, X_2, \dots)$ is generated by events $\{X_1 \in B_1, X_2 \in B_2, \dots\}$ where $B_i \in \mathcal{B}(\mathbb{R})$
and $B_i = \mathbb{R}$ for all but finitely many i

(i.e. events $X^{-1}(B)$ where $B = \text{cylinder set in } \mathbb{R}^{\mathbb{N}}$)

Def The tail σ -algebra is

$$\tau(X_1, \dots, X_n) := \bigcap_{n=1}^{\infty} \sigma(X_n, X_{n+1}, \dots) \quad (\sigma\text{-algebra indeed})$$

i.e. $\tau = \tau(X_1, \dots, X_n)$ consists of events that are unaffected by finitely many X_i 's

Examples of tail σ -algebras:

(a) $\{X_n \in B_n \text{ i.o.}\} \quad \forall B_n \in \mathcal{B}(\mathbb{R})$

$$\left(\bigcap_{n=1}^{\infty} \bigcup_{m \geq n} \{X_m \in B_m\} \right) = \sigma(X_n, X_{n+1}, \dots)$$

(b) $\{\lim X_n = c\} \quad \forall c \in \mathbb{R} \cup \{\pm\infty\}$

$$\left(\bigcap_{n=1}^{\infty} \{\lim (X_n, X_{n+1}, X_{n+2}, \dots) = c\} \right)$$

(c) $\{X_n \text{ converges}\}$ (d) $\{X_n \text{ is bounded}\}$ (e) $\{\sum X_n \text{ converges}\}$

(f) $\{\frac{X_1 + \dots + X_n}{n} \text{ converges}\}$

But: $\{\sup X_n \leq 1\}$ and $\{\limsup X_n \leq 1\}$ are NOT tail σ -algebras.

Thm (Kolmogorov's zero-one law)

Let X_1, X_2, \dots be independent random variables.

Then every event $E \in \mathcal{T}(X_1, X_2, \dots)$ satisfies $P(E) \in \{0, 1\}$.

Proof. It suffices to show that $\boxed{T \perp T} \quad \forall T \in \mathcal{T}(X_1, X_2, \dots)$

($P(T \cap T) = P(T) \cdot P(T) \Rightarrow P(T) \in \{0, 1\}$ as claimed.)

• X_i are independent and $T \in \sigma(X_{n+1}, X_{n+2}, \dots) \Rightarrow$

$T \perp \sigma(X_1, \dots, X_n) \quad \forall n$ (see p. 39: these two σ -algebras are independent)

$\Rightarrow T \perp \bigcup_{n=1}^{\infty} \sigma(X_1, \dots, X_n)$

!!
 \mathcal{A} is a π -system

$\boxed{E, F \in \mathcal{A} \Rightarrow E, F \in \sigma(X_1, \dots, X_n) \text{ for some } n}$
 $\Rightarrow E \cap F \in \sigma(X_1, \dots, X_n) \subset \mathcal{A} \quad \boxed{}$

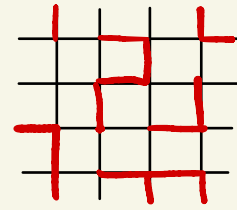
$\Rightarrow T \perp \sigma(\mathcal{A})$ (Stability of independence: Thm p. 36)

$\stackrel{\text{def}}{\parallel} \sigma(X_1, X_2, \dots)$. But $T \in \sigma(X_1, X_2, \dots) \Rightarrow T \perp T. \quad \square$

Examples: (a) All examples on p. 73

(b) Bond percolation on \mathbb{Z}^2 :

keep each edge with prop. p (remove with prob. $1-p$) independently.



$E_p = \{ \exists \text{ an infinite connected component} \}$

$E_p \in \mathcal{T}(X_{ij} : i, j \in \mathbb{Z})$ where $X_{ij} = \begin{cases} 1 & \text{if } (ij) \text{ is an edge} \\ 0 & \text{if not} \end{cases}$
 ↑
 arranged in \forall order.

Hence $P(E_p) \in \{0, 1\}$.

$P(E_p)$ is monotone in $p \Rightarrow \exists$ critical $p_c := \inf \{ p : P(E) = 1 \}$

$p > p_c : \exists \infty$ connected component a.s.

$p < p_c : \nexists \infty$ connected component a.s.

Thm [Kesten'1982]: $p_c = \frac{1}{2}$

(And for $p = p_c$, $\nexists \infty$ connected component a.s.)
 ← I think

Open \uparrow for \mathbb{Z}^3 .

(c) Simple random walk ($S_n = X_1 + \dots + X_n$ with iid Rademacher X_i) is called "recurrent" if $S_n = k$ i.o. $\forall k \in \mathbb{Z}$. otherwise called "transient".

0-1 law $\Rightarrow P\{\text{recurrent}\} \in \{0, 1\}$.

Actually, this probability = 1.

Thm $P\{S_n = 0 \text{ i.o.}\} = 1$

(KW): deduce $P(\text{recurrent}) = 1$.
 For that, prove $P(\limsup S_n = \infty) = 1$
 \downarrow
 $P(\exists n : S_n = k) = 1$

Proof $N := \#(\text{returns to } 0) = \#\{n : S_{2n} = 0\} = \sum_{n=1}^{\infty} \mathbb{1}_{\{S_{2n} = 0\}}$

$$\Rightarrow \mathbb{E}N = \sum_{n=1}^{\infty} P\{S_{2n} = 0\} = \sum_{n=1}^{\infty} \frac{1}{2^{2n}} \binom{2n}{n} = \infty$$

\parallel integrated tail (KW1)

$$\sum_{n=1}^{\infty} P\{N \geq n\} = \sum_{n=1}^{\infty} p^n < \infty \quad \text{if } p < 1$$

\parallel $P\{\underbrace{\text{returns; returns; } \dots \text{ returns}}_{n \text{ times}}\} = (P\{\text{returns}\})^n = p^n$

$\Rightarrow p = P\{\text{ever returns}\} = 1$.
 \downarrow repeat
 $P\{\text{returns i.o.}\} = 1$.
 \square

← slightly informal proof (uses conditioning).

More systematic & general treatment \rightarrow Markov chains.

SUMS OF INDEPENDENT RANDOM VARIABLES

THM Let X, Y be indep. r.v.'s with densities f_X, f_Y
 Then $X+Y$ has density $f_{X+Y} = f_X * f_Y$

↑
convolution: $(f_X * f_Y)(z) := \int_{-\infty}^{\infty} f_X(x) f_Y(z-x) dx = \int_{-\infty}^{\infty} f_X(z-y) f_Y(y) dy$

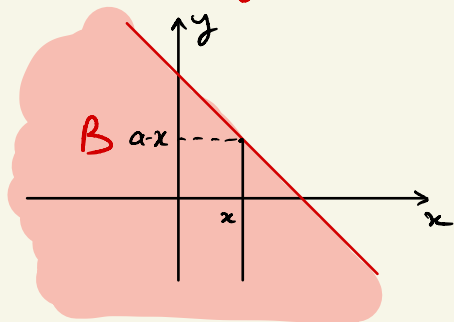
Proof Thm (p.42) $\Rightarrow (X, Y)$ are jointly abs. continuous, and joint density is
 $f_{X,Y}(x,y) = f_X(x) \cdot f_Y(y)$. (*)

CDF of $X+Y$:

$F_{X+Y}(a) = P\{X+Y \leq a\} = P\{(X,Y) \in B\}$ where $B = \{(x,y) \in \mathbb{R}^2 : x+y \leq a\}$

$= \iint_B f_{X,Y}(x,y) dx dy \stackrel{(*)}{=} \int_{-\infty}^{\infty} \int_{-\infty}^{a-x} f_X(x) f_Y(y) dy dx$

Change of var: $y = z - x$ ||
 $\int_{-\infty}^a f_X(x) f_Y(z-x) dz dx$



$= \int_{-\infty}^a \left(\int_{-\infty}^{\infty} f_X(x) f_Y(z-x) dx \right) dz$

\Rightarrow this must be the density of $X+Y$. □

Examples

