

CHAPTER 3:  
KOLMOGOROV 0-1 LAW

- Goals:
- Independence reminders
  - Importance of abstract theory:  
The elements of  $\sigma(X_1, X_2, \dots)$  can be approximated by elements of  $\sigma(X_1, \dots, X_n)$  for  $n$  large.
  - Notion of tail-events.

Setup:

- $(\Omega, \mathcal{F}, P)$  probability space
- $(E, \mathcal{E})$  measured space.

We will consider general n.v.s  $X: \Omega \rightarrow E$

I TAIL  $\sigma$ -ALGEBRA.

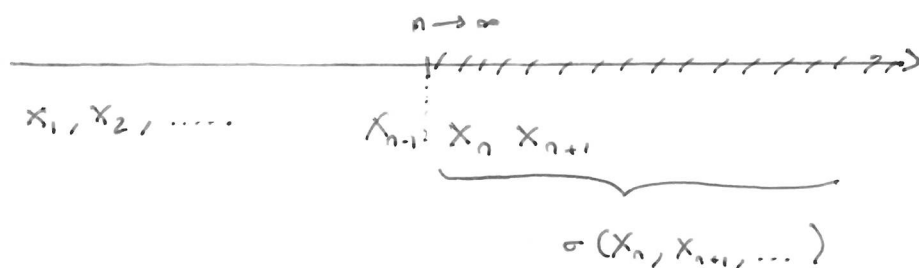
Def. Let  $X_1, X_2, \dots$  n.v.s with values in  $E$ .

The tail  $\sigma$ -algebra associated to  $X_1, X_2, \dots$  is defined by

$$\mathcal{G} = \bigcap_{n \geq 1} \sigma(X_n, X_{n+1}, \dots).$$

Rk:  $\mathcal{G}$  is indeed a  $\sigma$ -algebra (as an intersection of  $\sigma$ -algebras)

Why "tail"?



The tail  $\sigma$ -algebra contains the events  $A$  whose occurrence does not depend on any fixed finite subfamily of the  $X_i$ 's.

To put it differently, for any  $n$ , we can change the values of  $X_1, \dots, X_n$  arbitrarily without changing whether  $A$  occurs or not.

Examples:  $\emptyset, \Omega \in \mathcal{G}$

Are there interesting events in  $\mathcal{G}$ ?

$\rightarrow E =$  metric space  $(E, d)$ ,  $d$  distance.

Examples:

- $\{(X_n)_{n \geq 1} \text{ is a Cauchy sequence}\} \in \mathcal{G}$ .
- $\{(X_n)_{n \geq 1} \text{ converges}\} \in \mathcal{G}$ .
- $\{(X_n)_{n \geq 1} \text{ is bounded}\} \in \mathcal{G}$ .  
 $\hookrightarrow = \{ \exists C < \infty \quad \forall n \geq 1 \quad d(x, X_n) \leq C \}$  where  $x$  fixed in  $E$ .
- $\{ \forall n \geq 1 \quad d(x, X_n) \leq 10 \} \notin \mathcal{G}$ .

$E = \mathbb{R}$

- $\{ \limsup X_n \leq 10 \} \in \mathcal{G}$ ,
- $\{ \sup X_n \leq 10 \} \notin \mathcal{G}$ ,
- $\{ \sup X_n < \infty \} \in \mathcal{G}$ ,
- $\{ \sum X_n < \infty \} \in \mathcal{G}$ .

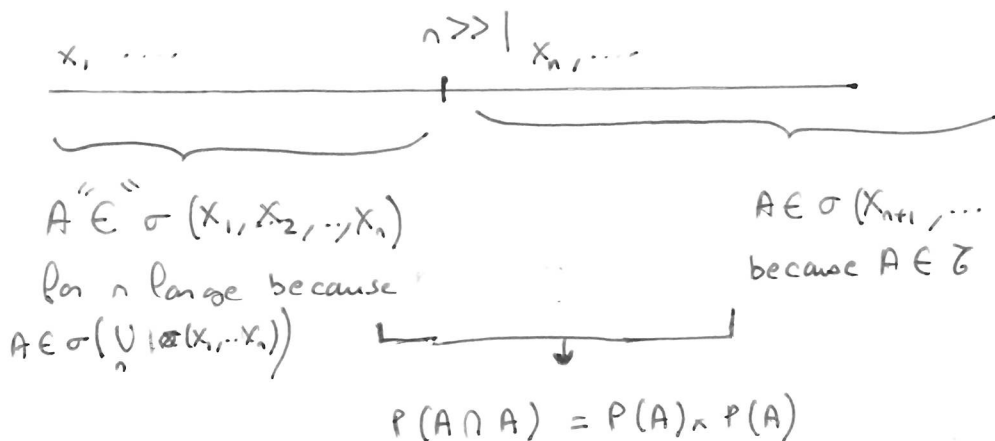
2 KOLMOGOROV THEOREM.

Thm: Let  $X_1, X_2, \dots$  be independent random variables with values in  $E$ . Let  $\mathcal{G}$  be the associated tail  $\sigma$ -algebra.

$$\forall A \in \mathcal{G} \quad P(A) \in \{0, 1\}.$$

" $X_1, X_2, \dots$  is tail-trivial"

Proof idea  $A \in \mathcal{G}$



Rk: Let  $\mathcal{E} \subset \mathcal{F}$  be a class of sets.

$$\mathcal{E} \text{ independent of } \mathcal{E} \Rightarrow \forall A \in \mathcal{E} \quad \overbrace{P(A \cap A)}^{= P(A)} = P(A)^2$$

$$\Rightarrow \forall A \in \mathcal{E} \quad P(A) = P(A)^2$$

$$\left( \text{if } x \in \mathbb{R} \quad x^2 = x \Leftrightarrow x(1-x) = 0 \Leftrightarrow x \in \{0, 1\} \right)$$

Proof. Let  $\mathcal{D}_n = \sigma(X_1, \dots, X_n) \quad , n \geq 1$

By grouping, we have

$$\forall n \geq 1 \quad \mathcal{D}_n \text{ indep. of } \sigma(X_{n+1}, \dots)$$

Since  $\mathcal{G} \subset \sigma(X_{n+1}, \dots)$ , we deduce

$$\forall n \geq 1 \quad \mathcal{D}_n \text{ indep. of } \mathcal{G},$$

and therefore

$$\bigcup_{n \geq 1} \mathcal{D}_n \text{ indep. of } \mathcal{G}.$$

Both  $\bigcup_{n \geq 1} \mathcal{D}_n$  and  $\mathcal{G}$  are  $\Pi$ -systems containing  $\emptyset, \Omega$ ,

hence

$$\sigma\left(\bigcup_{n \geq 1} \mathcal{D}_n\right) \text{ indep. of } \mathcal{G}.$$

Since  $\mathcal{G} \subset \sigma\left(\bigcup_{n \geq 1} \mathcal{D}_n\right)$ , we deduce that

$$\mathcal{G} \text{ indep. of } \mathcal{G} \quad \blacksquare$$

Rk: independence is important. Consider

$$X \sim \text{Ber}(\frac{1}{2}) \quad X_n = X \text{ for every } n.$$

$$A = \{ \exists n_0 : \forall n \geq n_0, X_n = 1 \} \in \mathcal{G}$$

$$\text{but } P(A) = P(X=1) = \frac{1}{2}.$$

### 3 ALTERNATIVE APPROACH: CYLINDRICAL APPROXIMATION.

Prop. Let  $X_1, X_2, \dots$  be arbitrary r.v. with values in  $E$ .  
For every  $A \in \sigma(X_1, X_2, \dots)$ , for every  $\epsilon > 0$ ,  
$$\exists n \geq 1 \exists A_\epsilon \in \sigma(X_1, \dots, X_n) \quad P(A \Delta A_\epsilon) \leq \epsilon$$

( $A \Delta B := A \setminus B \cup B \setminus A$  symmetric difference)

Rk: It is a particular case of a more general result in measure theory. It is important here that  $P$  is  $\sigma$ -finite (because it is finite).

• It can be difficult to represent the elements of  $\sigma(X_1, X_2, \dots)$   
The proposition above helps to understand such events: they are the events that can be well approximated by events of  $\sigma(X_1, \dots, X_n)$  for  $n$  large.

Proof. Let  $\mathcal{C} = \bigcup_{n=1}^{\infty} \sigma(X_1, \dots, X_n)$ .

NS:  $\mathcal{C}$  is stable under complement and finite union.

$$\sigma(\mathcal{C}) = \sigma(X_1, X_2, \dots)$$

Define  $\mathcal{G} = \{A \in \mathcal{F} : \forall \varepsilon > 0 \exists A_\varepsilon \in \mathcal{C} P(A \Delta A_\varepsilon) \leq \varepsilon\}$

We have  $\mathcal{C} \subset \mathcal{G}$  and  $\mathcal{G}$  is a  $\sigma$ -algebra. Indeed.

•  $A \in \mathcal{G} \Rightarrow A^c \in \mathcal{G}$  (because  $A \Delta D^c = A \Delta D \ \forall A, D \in \mathcal{F}$ )

• Let  $A_1, A_2, \dots \in \mathcal{G}$ . We claim that  $A = \bigcup_{i=1}^{\infty} A_i$  is also in  $\mathcal{G}$ . Let  $\varepsilon > 0$ .

By continuity of  $P$ , we can fix  $N$  s.t.

$$P\left(A \setminus \bigcup_{i=1}^N A_i\right) \leq \frac{\varepsilon}{2}.$$

For  $i=1, \dots, N$ , let  $A_{\varepsilon,i} \in \mathcal{C}$  s.t.  $P(A_i \Delta A_{\varepsilon,i}) \leq \frac{\varepsilon}{2^i}$ .

For  $i > N$  let  $A_{\varepsilon,i} = \emptyset$ . Since  $A_\varepsilon = \bigcup_{i \geq 1} A_{\varepsilon,i}$  is

a finite union of elements of  $\mathcal{C}$ , we have  $A_\varepsilon \in \mathcal{C}$  and

$$\begin{aligned} P(A \Delta A_\varepsilon) &\stackrel{\textcircled{*}}{=} P\left(\left(\bigcup_{i=1}^N A_i \Delta A_{\varepsilon,i}\right) \cup \bigcup_{i > N} A_i\right) \\ &\leq \underbrace{\sum_{i=1}^N \frac{\varepsilon}{2^i}}_{\leq \frac{\varepsilon}{2}} + \underbrace{P\left(\bigcup_{i > N} A_i\right)}_{\leq \frac{\varepsilon}{2}} \leq \varepsilon \end{aligned}$$

Therefore,  $\sigma(\mathcal{B}) \subset \mathcal{G}$ , which concludes the proof. ■

⊕ Here we use  $\forall I \quad \forall (A_i)_{i \in I} \quad \forall (B_i)_{i \in I}$

$$\left( \bigcup_{i \in I} A_i \right) \Delta \left( \bigcup_{i \in I} B_i \right) \subset \bigcup_{i \in I} (A_i \Delta B_i). \quad (\text{exercise})$$

(if  $x \in (\bigcup_i A_i) \setminus (\bigcup_i B_i)$ , then  $\exists i$  s.t.  $x \in A_i \setminus B_i$ )

We can now use this approximation result to deduce the Thm:

second proof of the Kolmogorov 0-1 law:

Let  $A \in \mathcal{G}$ . Let  $\varepsilon > 0$ .

Since  $A \in \sigma(X_1, X_2, \dots)$ , there exists  $n \geq 1$  and  $A_\varepsilon \in \sigma(X_1, \dots, X_n)$  such that

$$P(A \Delta A_\varepsilon) \leq \varepsilon.$$

We have  $A \in \sigma(X_{n+1}, \dots)$ . Hence, by grouping,

$$P(A_\varepsilon \cap A) = P(A_\varepsilon) P(A)$$

Since  $A_\varepsilon \subset A \cup (A_\varepsilon \setminus A) \subset A \cup (A \Delta A_\varepsilon)$  and

we have  $P(A_\varepsilon) \leq P(A) + \varepsilon$ . Similarly,

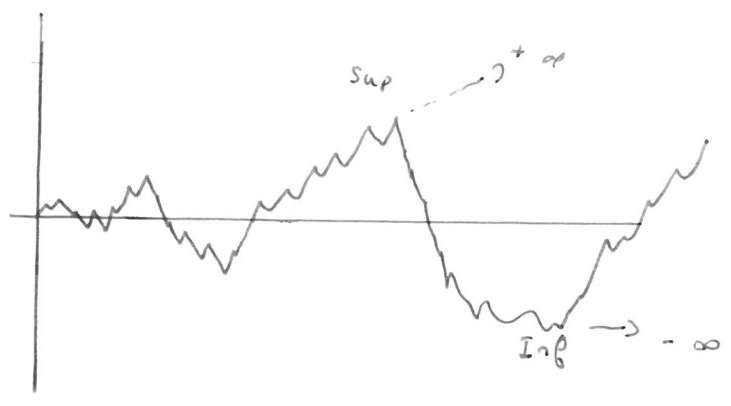
$$P(A \cap A) \leq P(A_\varepsilon \cap A) + \varepsilon.$$

Hence  $P(A) - \epsilon \leq P(A)^2 + \epsilon$ .

Since  $\epsilon$  is arbitrary, we conclude  $P(A) \in \{0, 1\}$

4 APPLICATION : RECURRENCE OF 1D RANDOM WALK.

Prop. Let  $(X_n)_{n \geq 1}$  be a sequence of iid n.v.s in  $\{-1, 1\}$  a.t.  
 $\forall n \quad P(X_n = -1) = P(X_n = +1) = \frac{1}{2}$ .  
 Setting  $S_n = X_1 + \dots + X_n$  for every  $n \geq 1$ , we have  
 $\sup S_n = +\infty$  a.s and  $\inf S_n = -\infty$  a.s.



Condition: Almost surely, we have  $S_n = 0$  for infinitely many  $n$ .

(The condition is left as an exercise)



Proof of the proposition:

Step 1: We first prove

$$\forall a \in \mathbb{N} \quad P(-a \leq \text{Inf } S_n \leq \text{Sup } S_n \leq a) = 0. \quad (*)$$

Let  $a \in \mathbb{N}$ . By Borel Cantelli II, for  $b > 2a + 1$

$$P\left(\bigcup_{j \in \mathbb{N}} \{X_{j b + 1} = +1, \dots, X_{j b + b} = 1\}\right) = 1$$

Indeed, the events in the union are independent and

$$\sum_{j \in \mathbb{N}} P(X_{j b + 1} = 1, \dots, X_{j b + b} = 1) = +\infty \\ = \frac{1}{2^b} > 0 \text{ indep. of } j.$$

Since  $\bigcup_{j \in \mathbb{N}} \{X_{j b + 1} = 1, \dots, X_{j b + b} = 1\} \subset \{-a \leq \text{Inf } S_n \leq \text{Sup } S_n \leq a\}^c$ ,

we obtain (\*).

Step 2 Letting  $a$  tend to infinity in (\*) and taking the complement, we get

$$1 = P(\{\text{Inf } S_n = -\infty\} \cup \{\text{Sup } S_n = +\infty\}) \\ \leq P(\text{Inf } S_n = -\infty) + P(\text{Sup } S_n = +\infty) \\ = 2 \times P(\text{Sup } S_n = +\infty) \\ \text{Sym.}$$

Therefore  $P(\sup S_n = +\infty) > 0$ .

Since  $\{\sup S_n = +\infty\} \in \mathcal{I}$

(indeed  $\forall k \in \mathbb{N} \quad \{\sup S_n = +\infty\} = \left\{ \sup_{n \geq k} X_k + \dots + X_n = +\infty \right\} \in \sigma(X_k, \dots)$ )

by the 0-1 Law, we have

$$P(\sup S_n = +\infty) = 1,$$

and by symmetry  $P(\inf S_n = -\infty) = 1$ . ■