

### 3 Kolmogorov's 0-1 law

With the Borel–Cantelli lemma, we have seen a first 0–1 law for independent events. We now come to another 0–1 law for independent events and for independent  $\sigma$ -algebras. To this end, we first introduce the notion of the tail  $\sigma$ -algebra.

**Definition 3.1** (Tail  $\sigma$ -algebra).

- (i) Let  $\mathcal{A}_1, \mathcal{A}_2, \dots$  be  $\sigma$ -algebras on  $\Omega$ . Then

$$\mathcal{T}(\mathcal{A}_1, \mathcal{A}_2, \dots) := \bigcap_{J \subset \mathbb{N} \text{ finite}} \sigma(\mathcal{A}_i ; i \in \mathbb{N} \setminus J) \quad (3.1)$$

is called the **tail  $\sigma$ -algebra of the  $\sigma$ -algebras  $\mathcal{A}_1, \mathcal{A}_2, \dots$**

- (ii) If  $A_1, A_2, \dots$  are events of  $\mathcal{A}$ , then we define the **tail  $\sigma$ -algebra of the events  $A_1, A_2, \dots$**  as the tail  $\sigma$ -algebra of the  $\sigma$ -algebras  $\sigma(A_1), \sigma(A_2), \dots$  and denote it by

$$\mathcal{T}(A_1, A_2, \dots).$$

- (iii) Finally, if  $X_1, X_2, \dots$  are random variables all defined on  $\Omega$ , then we define the **tail  $\sigma$ -algebra of the random variables  $X_1, X_2, \dots$**  as the tail  $\sigma$ -algebra of the  $\sigma$ -algebras  $\sigma(X_1), \sigma(X_2), \dots$  and denote it by

$$\mathcal{T}(X_1, X_2, \dots).$$

**Theorem 3.2.** Let  $\mathcal{A}_1, \mathcal{A}_2, \dots$  be  $\sigma$ -algebras on  $\Omega$ . Then

$$\mathcal{T}(\mathcal{A}_1, \mathcal{A}_2, \dots) = \bigcap_{k \in \mathbb{N}} \sigma(\mathcal{A}_i ; i = k, k+1, \dots). \quad (3.2)$$

**Remark 3.3.** The name “tail  $\sigma$ -algebra” is due to the interpretation of  $\mathbb{N}$  as a set of times. As is made clear in the theorem, any event in  $\mathcal{T}$  does not depend on the first finitely many time points.

**Remark 3.4.** Maybe at first glance it is not evident that there are any interesting events in the tail  $\sigma$ -algebra at all. It might not even be clear that we do not have  $\mathcal{T} = \{\emptyset, \Omega\}$ . Hence we now present simple examples of tail events and tail  $\sigma$ -algebra measurable random variables.

**Example 3.5.**

- (i) Let  $A_1, A_2, \dots$  be events. Then the events  $A_*$  and  $A^*$  are in  $\mathcal{T}(A_1, A_2, \dots)$ . Indeed, let  $k \in \mathbb{N}$ . Since

$$\bigcap_{i \geq k} A_i \in \sigma(\sigma(A_i) ; i = k, k+1, \dots)$$

for any  $m = k, k + 1, \dots$  and

$$A_* = \bigcup_{m \geq k} \bigcap_{i \geq m} A_i,$$

$A_* \in \sigma(\sigma(A_i) ; i = k, k + 1, \dots)$ . Then,  $A_* \in \mathcal{T}(A_1, A_2, \dots)$ . Analogously, it holds that  $A^* \in \mathcal{T}(A_1, A_2, \dots)$ .

- (ii) Let  $X_1, X_2, \dots$  be  $\overline{\mathbb{R}}$ -valued random variables. Then the maps  $X_*$  and  $X^*$  are  $\mathcal{T}(X_1, X_2, \dots)$ -measurable. Indeed, let  $k \in \mathbb{N}$ . Since

$$\inf_{i \geq m} X_i \text{ is } \sigma(\sigma(X_i) ; i = k, k + 1, \dots)\text{-measurable}$$

for any  $m = k, k + 1, \dots$  and

$$X_* = \sup_{m \geq k} \inf_{i \geq m} X_i,$$

$X_*$  is  $\sigma(\sigma(X_i) ; i = k, k + 1, \dots)$ -measurable. Then,  $X_*$  is  $\mathcal{T}(X_1, X_2, \dots)$ -measurable. Analogously, it holds that  $X^*$  is  $\mathcal{T}(X_1, X_2, \dots)$ -measurable.

- (iii) Let  $X_1, X_2, \dots$  be real random variables. Then the *Cesàro limits*:

$$\liminf_{n \rightarrow +\infty} \frac{1}{n} \sum_{i=1}^n X_i \quad \text{and} \quad \limsup_{n \rightarrow +\infty} \frac{1}{n} \sum_{i=1}^n X_i$$

are  $\mathcal{T}(X_1, X_2, \dots)$ -measurable. Indeed, let  $k \in \mathbb{N}$ . Since,

$$\liminf_{n \rightarrow +\infty} \frac{1}{n} \sum_{i=1}^n X_i = \liminf_{n \rightarrow +\infty} \frac{1}{n} \sum_{i=k}^n X_i,$$

by the example above,

$$\liminf_{n \rightarrow +\infty} \frac{1}{n} \sum_{i=1}^n X_i \text{ is } \sigma(\sigma(X_i) ; i = k, k + 1, \dots)\text{-measurable.}$$

Then,

$$\liminf_{n \rightarrow +\infty} \frac{1}{n} \sum_{i=1}^n X_i$$

is  $\mathcal{T}(X_1, X_2, \dots)$ -measurable. Analogously, it holds that

$$\limsup_{n \rightarrow +\infty} \frac{1}{n} \sum_{i=1}^n X_i$$

is  $\mathcal{T}(X_1, X_2, \dots)$ .

*Proof of theorem 3.2.* Let  $\mathcal{B}$  and  $\mathcal{C}$  be the  $\sigma$ -algebras defined by equations 3.1 and 3.2, respectively. Let  $k \in \mathbb{N}$ ; then

$$\mathcal{B} \subset \sigma(\mathcal{A}_i ; i = k, k+1, \dots).$$

Thus,  $\mathcal{B} \subset \mathcal{C}$ . Let  $J \subset \mathbb{N}$  be finite; then

$$\mathcal{C} \subset \sigma(\mathcal{A}_i ; i = k, k+1, \dots) \subset \sigma(\mathcal{A}_i ; i \in \mathbb{N} \setminus J)$$

for some  $k \in \mathbb{N}$ . Hence,  $\mathcal{C} \subset \mathcal{B}$ . □

**Theorem 3.6** (Kolmogorov's 0–1 law). Let  $\mathcal{A}_1, \mathcal{A}_2, \dots$  be independent  $\sigma$ -algebras on  $\Omega$ . Then the tail  $\sigma$ -algebra of  $\mathcal{A}_1, \mathcal{A}_2, \dots$  is  $\mathbf{P}$ -trivial, that is,

$$A \in \mathcal{T}(\mathcal{A}_1, \mathcal{A}_2, \dots) \Rightarrow \mathbf{P}(A) \in \{0, 1\}.$$

**Example 3.7.**

- (i) Let  $A_1, A_2, \dots$  be independent events. Then  $\mathbf{P}(A_*), \mathbf{P}(A^*) \in \{0, 1\}$ . Indeed,  $A_*, A^* \in \mathcal{T}(A_1, A_2, \dots)$ .
- (ii) Let  $X_1, X_2, \dots$  be independent  $\overline{\mathbb{R}}$ -valued random variables. Then the maps  $X_*$  and  $X^*$  are deterministic. Indeed, since  $X^*$  is  $\mathcal{T}(X_1, X_2, \dots)$ -measurable,

$$\mathbf{P}(X^* \leq x) \in \{0, 1\} \quad \text{for every } x \in \mathbb{R}.$$

Let

$$x^* := \inf\{x \in \mathbb{R} ; \mathbf{P}(X^* \leq x) = 1\}.$$

Then we have three cases:

If  $x^* = -\infty$ , then

$$\mathbf{P}(X^* = -\infty) = \lim_{n \rightarrow +\infty} \mathbf{P}(X^* \leq -n) = 1.$$

If  $x^* \in \mathbb{R}$ , then

$$\mathbf{P}(X^* \leq x^*) = \lim_{n \rightarrow +\infty} \mathbf{P}\left(X^* \leq x^* + \frac{1}{n}\right) = 1,$$

and

$$\mathbf{P}(X^* < x^*) = \lim_{n \rightarrow +\infty} \mathbf{P}\left(X^* \leq x^* - \frac{1}{n}\right) = 0.$$

If  $x^* = +\infty$ , then

$$\mathbf{P}(X^* < +\infty) = \lim_{n \rightarrow +\infty} \mathbf{P}(X^* \leq n) = 0.$$

Since the conclusion that  $X_*$  is deterministic only depends from the fact that  $X_*$  is  $\mathcal{T}(X_1, X_2, \dots)$ -measurable, we can also conclude that  $X^*$  is deterministic.

(iii) Let  $X_1, X_2, \dots$  be independent real random variables. Then the *Cesàro limits*

$$\liminf_{n \rightarrow +\infty} \frac{1}{n} \sum_{i=1}^n X_i \quad \text{and} \quad \limsup_{n \rightarrow +\infty} \frac{1}{n} \sum_{i=1}^n X_i$$

are deterministic. Indeed, the Cesàro limits are  $\mathcal{T}(X_1, X_2, \dots)$ -measurable.

*Proof of theorem 3.6.* It is enough to prove that  $\mathcal{T}$  and  $\mathcal{T}$  are independent. Indeed, let  $n \in \mathbb{N}$ . Then, by virtue of theorem ??,

$$\mathcal{C}_n := \sigma(\mathcal{A}_i; i = 1, \dots, n) \quad \text{and} \quad \sigma(\mathcal{A}_i; i = n+1, n+2, \dots)$$

are independent. Since

$$\mathcal{T} := \mathcal{T}(\mathcal{A}_1, \mathcal{A}_2, \dots) \subset \sigma(\mathcal{A}_i; i = n+1, n+2, \dots),$$

$\mathcal{C}_n$  and  $\mathcal{T}$  are independent. Hence  $\mathcal{C}$  and  $\mathcal{T}$  are independent, where  $\mathcal{C}_n \uparrow \mathcal{C}$ . Then, because  $\mathcal{C}$  is a  $\pi$ -system and generates

$$\mathcal{C}_\infty := \sigma(\mathcal{A}_i; i = 1, 2, \dots),$$

$\mathcal{C}_\infty$  and  $\mathcal{T}$  are independent. Finally, the assertion follows since  $\mathcal{T} \subset \mathcal{C}_\infty$ .  $\square$

**Exercise 3.1.** Let  $X_1, X_2, \dots$  be independent real random variables. Then  $\mathbf{P}(\lim_{n \rightarrow +\infty} X_n \text{ exists})$  must be 0 or 1.

**Exercise 3.2.** Let  $X_1, X_2, \dots$  be independent real random variables with  $X_n \sim \mathcal{U}_{\{1, \dots, n\}}$ . Compute  $\mathbf{P}(X_n = 5 \text{ i.o.})$ .

**Exercise 3.3.** Let  $A_1, A_2, \dots$  be events such that

1.  $A_{i_1}, A_{i_2}, \dots, A_{i_k}$  are independents whenever  $i_{j+1} \geq i_j + 2$  for  $j = 1, \dots, k-1$ , and
2.  $\sum_{n=1}^{+\infty} \mathbf{P}(A_n) = +\infty$ .

Prove that  $\mathbf{P}(A^*) = 1$ .

**Exercise 3.4.** Consider infinite, independent, fair coin tossing, and let  $H_n$  be the event that the  $n$ th coin is heads. Determine the following probabilities.

1.  $\mathbf{P}[H_{n+1} \cap H_{n+2} \cap \dots \cap H_{n+9} \text{ i.o.}]$ .
2.  $\mathbf{P}[H_{n+1} \cap H_{n+2} \cap \dots \cap H_{2n} \text{ i.o.}]$ .

**Exercise 3.5.** Let  $A_1, A_2, \dots$  be independent events, let  $x \in \mathbb{R}$ , and let

$$S_x = \left( \lim_{n \rightarrow +\infty} \frac{1}{n} \sum_{i=1}^n \mathbf{1}_{A_i} \leq x \right).$$

Prove that  $\mathbf{P}(S_x)$  must equal 0 or 1.

**Exercise 3.6.** Let  $A_1, A_2, \dots$  be independent events. Let  $X$  be a real random variable which is  $\sigma(A_n, A_{n+1}, \dots)$ -measurable for each  $n \in \mathbb{N}$ . Prove that  $X$  is deterministic.