

### 3 Convergence in Distribution

Given: a metric space  $(M, \rho)$ . Put

$$C^b(M) = \{f : M \rightarrow \mathbb{R} : f \text{ bounded, continuous}\},$$

and consider the Borel- $\sigma$ -algebra  $\mathfrak{B}(M)$  in  $M$ . Moreover, let  $\mathfrak{M}(M)$  denote the set of all probability measures on  $\mathfrak{B}(M)$ .

**Definition 1.**

(i) A sequence  $(Q_n)_{n \in \mathbb{N}}$  in  $\mathfrak{M}(M)$  *converges weakly* to  $Q \in \mathfrak{M}(M)$  if

$$\forall f \in C^b(M) : \lim_{n \rightarrow \infty} \int f dQ_n = \int f dQ.$$

Notation:  $Q_n \xrightarrow{w} Q$ .

(ii) A sequence  $(X_n)_{n \in \mathbb{N}}$  of random elements with values in  $M$  *converges in distribution* to a random element  $X$  with values in  $M$  iff  $\mathfrak{L}(X_n) \xrightarrow{w} \mathfrak{L}(X)$  (recall  $\mathfrak{L}(X)$  is the distribution of  $X$ ).

**Notation:**  $X_n \xrightarrow{d} X$ .

**Remark 1.** For convergence in distribution the random elements need not be defined on a common probability space.

In the sequel:  $Q_n, Q \in \mathfrak{M}(M)$  for  $n \in \mathbb{N}$ .

**Example 1.**

(i) For  $x_n, x \in M$

$$\delta_{x_n} \xrightarrow{w} \delta_x \Leftrightarrow \lim_{n \rightarrow \infty} \rho(x_n, x) = 0.$$

For the proof of ' $\Leftarrow$ ', note that

$$\int f d\varepsilon_{x_n} = f(x_n), \quad \int f d\varepsilon_x = f(x).$$

For the proof of ' $\Rightarrow$ ', suppose that  $\limsup_{n \rightarrow \infty} \rho(x_n, x) > 0$ . Take

$$f(y) = \min(\rho(y, x), 1), \quad y \in M,$$

and observe that  $f \in C^b(M)$  and

$$\limsup_{n \rightarrow \infty} \int f d\delta_{x_n} = \limsup_{n \rightarrow \infty} (\min(\rho(x_n, x), 1)) > 0$$

while  $\int f d\delta_x = 0$ .

(ii) Let  $(M, \rho) = (\mathfrak{R}, |\cdot|)$ ,  $Q_n = N(\mu_n, \sigma_n^2)$  with  $\mu_n \in \mathfrak{R}$ ,  $\sigma_n > 0$ .

**Claim:**

$$Q_n \xrightarrow{w} Q \Leftrightarrow Q = N(\mu, \sigma), \mu_n \rightarrow \mu, \sigma_n \rightarrow \sigma.$$

(Here,  $N(\mu, 0) = \delta_\mu$ ).

**Proof:** ' $\Leftarrow$ ': For  $f \in C_b(\mathfrak{R})$ ,

$$\int f dQ_n = 1/\sqrt{2\pi} \cdot \int_{\mathbb{R}} \underbrace{f(\sigma_n \cdot x + \mu_n)}_{\rightarrow f(\sigma x + \mu), \leq b} \cdot \exp(-1/2 \cdot x^2) \lambda_1(dx) \rightarrow \int_{\mathbb{R}} f(\sigma \cdot x + \mu) \cdot \exp(-1/2 \cdot x^2) \lambda_1(dx)$$

' $\Rightarrow$ ': Übung 8.4.

**Remark 2.** Note that  $Q_n \xrightarrow{w} Q$  does not imply

$$\forall A \in \mathfrak{B}(M) : \lim_{n \rightarrow \infty} Q_n(A) = Q(A).$$

For instance, assume  $\lim_{n \rightarrow \infty} \rho(x_n, x) = 0$  with  $x_n \neq x$  for every  $n \in \mathbb{N}$ . Then

$$\delta_{x_n}(\{x\}) = 0, \quad \delta_x(\{x\}) = 1.$$

**Theorem 1 (Portmanteau Theorem).** The following properties are equivalent:

- (i)  $Q_n \xrightarrow{w} Q$ ,
- (ii)  $\forall f \in C^b(M)$  uniformly continuous :  $\lim_{n \rightarrow \infty} \int f dQ_n = \int f dQ$ ,
- (iii)  $\forall A \subset M$  closed :  $\limsup_{n \rightarrow \infty} Q_n(A) \leq Q(A)$ ,
- (iv)  $\forall A \subset M$  open :  $\liminf_{n \rightarrow \infty} Q_n(A) \geq Q(A)$ ,
- (v)  $\forall A \in \mathfrak{B}(M) : Q(\partial A) = 0 \Rightarrow \lim_{n \rightarrow \infty} Q_n(A) = Q(A)$ .

*Proof.* See Gänsler, Stute (1977, Satz 8.4.9). □

In the sequel, we study the particular case  $(M, \mathfrak{B}(M)) = (\mathbb{R}, \mathfrak{B})$ , i.e., convergence in distribution for random variables. The Central Limit Theorem deals with this notion of convergence, see the introductory Example I.1 and Section IV.5.

Notation: for any  $Q \in \mathfrak{M}(\mathbb{R})$

$$F_Q(x) = Q(]-\infty, x]), \quad x \in \mathbb{R},$$

and for any function  $F : \mathbb{R} \rightarrow \mathbb{R}$

$$\text{Cont}(F) = \{x \in \mathbb{R} : F \text{ continuous at } x\}.$$

**Theorem 2.**

$$Q_n \xrightarrow{w} Q \iff \forall x \in \text{Cont}(F_Q) : \lim_{n \rightarrow \infty} F_{Q_n}(x) = F_Q(x).$$

Moreover, if  $Q_n \xrightarrow{w} Q$  and  $\text{Cont}(F_Q) = \mathbb{R}$  then

$$\lim_{n \rightarrow \infty} \sup_{x \in \mathbb{R}} |F_{Q_n}(x) - F_Q(x)| = 0.$$

*Proof.* ‘ $\Rightarrow$ ’: If  $x \in \text{Cont}(F_Q)$  and  $A = ]-\infty, x]$  then  $Q(\partial A) = Q(\{x\}) = 0$ , see Theorem 1.2. Hence Theorem 1 implies

$$\lim_{n \rightarrow \infty} F_{Q_n}(x) = \lim_{n \rightarrow \infty} Q_n(A) = Q(A) = F_Q(x).$$

‘ $\Leftarrow$ ’: Consider a non-empty open set  $A \subset \mathbb{R}$ . Take pairwise disjoint open intervals  $A_1, A_2, \dots$  such that  $A = \bigcup_{i=1}^{\infty} A_i$ . Fatous’s Lemma implies

$$\liminf_{n \rightarrow \infty} Q_n(A) = \liminf_{n \rightarrow \infty} \sum_{i=1}^{\infty} Q_n(A_i) \geq \sum_{i=1}^{\infty} \liminf_{n \rightarrow \infty} Q_n(A_i).$$

Note that  $\mathbb{R} \setminus \text{Cont}(F_Q)$  is countable. Fix  $\varepsilon > 0$ , and take

$$A'_i = ]a'_i, b'_i] \subset A_i$$

for  $i \in \mathbb{N}$  such that

$$a'_i, b'_i \in \text{Cont}(F_Q) \wedge Q(A_i) \leq Q(A'_i) + \varepsilon \cdot 2^{-i}.$$

Then

$$\liminf_{n \rightarrow \infty} Q_n(A_i) \geq \liminf_{n \rightarrow \infty} Q_n(A'_i) = Q(A'_i) \geq Q(A_i) - \varepsilon \cdot 2^{-i}.$$

We conclude that

$$\liminf_{n \rightarrow \infty} Q_n(A) \geq Q(A) - \varepsilon,$$

and therefore  $Q_n \xrightarrow{w} Q$  by Theorem 1.

Uniform convergence, Übung 9.1. □

**Corollary 1.**

$$Q_n \xrightarrow{w} Q \wedge Q_n \xrightarrow{w} \tilde{Q} \Rightarrow Q = \tilde{Q}.$$

*Proof.* By Theorem 2  $F_Q(x) = F_{\tilde{Q}}(x)$  if  $x \in D = \text{Cont}(F_Q) \cap \text{Cont}(F_{\tilde{Q}})$ . Since  $D$  is dense in  $\mathbb{R}$  and  $F_Q$  as well as  $F_{\tilde{Q}}$  are right-continuous, we get  $F_Q = F_{\tilde{Q}}$ . Apply Theorem 1.3. □

Given: random variables  $X_n, X$  on  $(\Omega, \mathfrak{A}, P)$  for  $n \in \mathbb{N}$ .

**Theorem 3.**

$$X_n \xrightarrow{P} X \Rightarrow X_n \xrightarrow{d} X$$

and

$$X_n \xrightarrow{d} X \wedge X \text{ constant a.s.} \Rightarrow X_n \xrightarrow{P} X.$$

*Proof.* Assume  $X_n \xrightarrow{P} X$ . For  $\varepsilon > 0$  and  $x \in \mathbb{R}$

$$\begin{aligned} & P(\{X \leq x - \varepsilon\}) - P(\{|X - X_n| > \varepsilon\}) \\ & \leq P(\{X \leq x - \varepsilon\} \cap \{|X - X_n| \leq \varepsilon\}) \\ & \leq P(\{X_n \leq x\}) \\ & \leq P(\{X_n \leq x\} \cap \{X \leq x + \varepsilon\}) + P(\{X_n \leq x\} \cap \{X > x + \varepsilon\}) \\ & \leq P(\{X \leq x + \varepsilon\}) + P(\{|X - X_n| > \varepsilon\}). \end{aligned}$$

Thus

$$F_X(x - \varepsilon) \leq \liminf_{n \rightarrow \infty} F_{X_n}(x) \leq \limsup_{n \rightarrow \infty} F_{X_n}(x) \leq F_X(x + \varepsilon).$$

For  $x \in \text{Cont}(F_X)$  we get  $\lim_{n \rightarrow \infty} F_{X_n}(x) = F_X(x)$ . Apply Theorem 2.

Now, assume that  $X_n \xrightarrow{d} X$  and  $P_X = \varepsilon_x$ . Let  $\varepsilon > 0$  and take  $f \in C^b(\mathbb{R})$  such that  $f \geq 0$ ,  $f(x) = 0$ , and  $f(y) = 1$  if  $|x - y| \geq \varepsilon$ . Then

$$P(\{|X - X_n| > \varepsilon\}) = P(\{|x - X_n| > \varepsilon\}) \leq \int 1_{\mathbb{R} \setminus [x - \varepsilon, x + \varepsilon]} dP_{X_n} \leq \int f dP_{X_n}$$

and

$$\lim_{n \rightarrow \infty} \int f dP_{X_n} = \int f dP_X = 0.$$

□

**Example 2.** Consider the uniform distribution  $P$  on  $\Omega = \{0, 1\}$ . Put

$$X_n(\omega) = \omega, \quad X(\omega) = 1 - \omega.$$

Then  $P_{X_n} = P_X$  and therefore

$$X_n \xrightarrow{d} X.$$

However,  $\{|X_n - X| < 1/2\} = \emptyset$  and therefore

$$X_n \xrightarrow{P} X \text{ does not hold.}$$

We quote without proof an interesting converse to Theorem 3:

**Theorem 4 (Skorohod).** Let  $\Omega = ]0, 1[$  with the uniform distribution, and  $Q_n, Q \in \mathfrak{M}(\mathfrak{R})$ . If  $Q_n \xrightarrow{w} Q$ , then for the random variables

$$X_{Q_n}(\omega) := F_{Q_n}^-(z) := \inf\{z \in \mathfrak{R} : \omega \leq F_{Q_n}(z)\}, \quad \omega \in ]0, 1[$$

we have  $P_{X_{Q_n}} = Q_n$  and  $X_{Q_n} \rightarrow X_Q$   $P$ -a.s..

**Remark 3.** Skorohod's Theorem is based on a general method to transform uniformly distributed 'random numbers' from  $]0, 1[$  into 'random numbers' with distribution  $Q$ . Namely, if  $U$  is uniformly distributed on  $]0, 1[$  and  $F$  is some distribution function,  $X := F^{-1}(U)$  has  $F$  as distribution function.

**Remark 4.**

Let  $\mu, \mu_n \in \mathfrak{M}(M)$ . Then

$$\mu_n \xrightarrow{w} \mu \iff \forall n_k \exists n_{k_l} \mu_{n_{k_l}} \xrightarrow{w} \mu,$$

see Übung8.3.

Finally, we present a compactness criterion, which is very useful for construction of probability measures on  $\mathfrak{B}(M)$ . We need a generalized Bolzano–Weierstrass Theorem:

**Lemma 1.** Let  $x_{n,\ell} \in \mathbb{R}$  for  $n, \ell \in \mathbb{N}$  with

$$\forall \ell \in \mathbb{N} : \sup_{n \in \mathbb{N}} |x_{n,\ell}| < \infty.$$

Then there exists an increasing sequence  $(n_i)_{i \in \mathbb{N}}$  in  $\mathbb{N}$  such that

$$\forall \ell \in \mathbb{N} : (x_{n_i,\ell})_{i \in \mathbb{N}} \text{ converges.}$$

*Proof.* (Sketch, see Billingsley (1979, Thm. 25.13 for details): For fixed  $l$ ,  $x_{n,l}$  is a bounded sequence. Hence, by the original Bolzano–Weierstrass Theorem, there is a subsequence  $\pi_1(n)$  (where  $\pi_1 : \mathfrak{N} \rightarrow \mathfrak{N}$  is a monotone mapping) such that  $(x_{\pi_1(n),1})$  converges. Next,  $(x_{\pi_1(n),2})$  is bounded, hence there is  $\pi_2 : \mathfrak{N} \rightarrow \mathfrak{N}$  monotone such that  $x_{\pi_2(\pi_1(n),2)}$  converges. Set  $\pi^2 := \pi_2 \circ \pi_1$ . Iterating, we can find for each  $m$  some  $\pi^m = \pi_m \circ \pi^{m-1}$  such that for  $l \leq m$ ,  $x_{\pi^m(n),k}$  converges. Define  $n_i = \pi^i(i)$ .  $\square$

**Definition 2.**

(i)  $\mathfrak{P} \subset \mathfrak{M}(M)$  *tight* if

$$\forall \varepsilon > 0 \exists K \subset M \text{ compact } \forall P \in \mathfrak{P} : P(K) \geq 1 - \varepsilon.$$

(ii)  $\mathfrak{P} \subset \mathfrak{M}(M)$  *relatively compact* if every sequence in  $\mathfrak{P}$  contains a subsequence that converges weakly.

**Theorem 5 (Prohorov).** Assume that  $M$  is a complete separable metric space and  $\mathfrak{P} \subset \mathfrak{M}(M)$ . Then

$$\mathfrak{P} \text{ relatively compact} \iff \mathfrak{P} \text{ tight.}$$

*Proof.* We only treat the case  $M = \mathfrak{R}$ ; see Parthasarathy (1967, Thm. II.6.7) for the general case.

‘ $\Rightarrow$ ’: Suppose that  $\mathfrak{P}$  is not tight. Then, for some  $\varepsilon > 0$ , there exists a sequence  $(P_n)_{n \in \mathbb{N}}$  in  $\mathfrak{P}$  such that

$$P_n([-n, n]) < 1 - \varepsilon.$$

For a suitable subsequence,  $P_{n_k} \xrightarrow{w} P \in \mathfrak{M}(\mathbb{R})$ . Take  $m > 0$  such that

$$P(]-m, m[) > 1 - \varepsilon.$$

Theorem 1 implies

$$P(]-m, m[) \leq \liminf_{k \rightarrow \infty} P_{n_k}(]-m, m[) \leq \liminf_{k \rightarrow \infty} P_{n_k}([-n_k, n_k]) < 1 - \varepsilon,$$

which is a contradiction.

‘ $\Leftarrow$ ’: Consider any sequence  $(P_n)_{n \in \mathbb{N}}$  in  $\mathfrak{P}$  and the corresponding sequence  $(F_n)_{n \in \mathbb{N}}$  of distribution functions. Use Lemma 1 to obtain a subsequence  $(F_{n_i})_{i \in \mathbb{N}}$  and a non-decreasing function  $G : \mathbb{Q} \rightarrow [0, 1]$  with

$$\forall q \in \mathbb{Q} : \lim_{i \rightarrow \infty} F_{n_i}(q) = G(q).$$

Put

$$F(x) = \inf\{G(q) : q \in \mathbb{Q} \wedge x < q\}, \quad x \in \mathbb{R}.$$

Claim (*Helly's Theorem*):

(i)  $F$  is non-decreasing and right-continuous,

(ii)  $\forall x \in \text{Cont}(F) : \lim_{i \rightarrow \infty} F_{n_i}(x) = F(x)$ .

*Proof:* Ad (i): Obviously  $F$  is non-decreasing. For  $x \in \mathbb{R}$  and  $\varepsilon > 0$  take  $\delta_2 > 0$  such that

$$\forall q \in \mathbb{Q} \cap ]x, x + \delta_2[ : G(q) \leq F(x) + \varepsilon.$$

Thus, for  $z \in ]x, x + \delta_2[$ ,

$$F(x) \leq F(z) \leq F(x) + \varepsilon.$$

Ad (ii): If  $x \in \text{Cont}(F)$  take  $\delta_1 > 0$  such that

$$F(x) - \varepsilon \leq F(x - \delta_1).$$

Thus, for  $q_1, q_2 \in \mathbb{Q}$  with

$$x - \delta_1 < q_1 < x < q_2 < x + \delta_2,$$

we get

$$\begin{aligned} F(x) - \varepsilon &\leq F(x - \delta_1) \leq G(q_1) \leq \liminf_{i \rightarrow \infty} F_{n_i}(x) \leq \limsup_{i \rightarrow \infty} F_{n_i}(x) \\ &\leq G(q_2) \leq F(x) + \varepsilon. \end{aligned}$$

Claim:

$$\lim_{x \rightarrow -\infty} F(x) = 0 \wedge \lim_{x \rightarrow \infty} F(x) = 1.$$

Proof: For  $\varepsilon > 0$  take  $m \in \mathbb{Q}$  such that

$$\forall n \in \mathbb{N} : P_n([-m, m]) \geq 1 - \varepsilon.$$

Thus

$$G(m) - G(-m) = \lim_{i \rightarrow \infty} (F_{n_i}(m) - F_{n_i}(-m)) = \lim_{i \rightarrow \infty} P_{n_i}([-m, m]) \geq 1 - \varepsilon.$$

Since  $F(m) \geq G(m)$  and  $F(-m - 1) \leq G(-m)$ , we obtain

$$F(m) - F(-m - 1) \geq 1 - \varepsilon.$$

It remains to apply Theorems 1.3 and 2. □