## 7. Convergence in distribution

7.1. Compactness of distribution functions. Recall that a distribution is a non-decreasing right-continuous function  $F: \mathbb{R} \to [0,1]$  satisfying

$$F(-\infty) = \lim_{x \to -\infty} F(x) = 0, \quad F(\infty) = \lim_{x \to \infty} F(x) = 1.$$

Throughout this section, we use  $\mathcal{M}$  to denote the set of all non-decreasing right-continuous functions F satisfying  $F(-\infty) \ge 0$  and  $F(\infty) \le 1$ .

**Definition 7.1.** Let  $f, f_n$  be functions defined on  $\mathbb{R}$ .  $f_n$  is said to converge in distribution (or  $\mathcal{D}$ -converge) to f if

$$f_n(x) \to f(x) \quad \forall x \in \mathcal{C}(f)$$

where  $\mathcal{C}(f)$  is the set of continuous points of f.

Remark 7.1. If  $f_n \to f$  and  $f_n \to g$  in distribution and f, g are monotonic and right-continuous, then f = g.

**Lemma 7.1.** Let  $f_n$ , f be functions defined on  $\mathbb{R}$ . Then, the following are equivalent.

- (1)  $f_n$  converges to f in distribution.
- (2) For any subsequence (f<sub>nk</sub>)<sup>∞</sup><sub>k=1</sub> of (f<sub>n</sub>)<sup>∞</sup><sub>n=1</sub>, f<sub>nk</sub> converges to f in distribution.
  (3) For any subsequence (f<sub>nk</sub>)<sup>∞</sup><sub>k=1</sub> of (f<sub>n</sub>)<sup>∞</sup><sub>n=1</sub>, there is a further subsequence (f<sub>mk</sub>)<sup>∞</sup><sub>k=1</sub> such that f<sub>mk</sub> converges to f in distribution.

**Lemma 7.2.** The set  $\mathcal{M}$  is sequentially compact under  $\mathcal{D}$ -convergence.

*Proof.* Let  $(f_n)_{n=1}^{\infty}$  be a sequence in  $\mathcal{M}$  and  $\mathbb{Q}$  be the set of all rational numbers. For  $x \in \mathbb{Q}$ , since  $\{f_n(x) : n = 1, 2, ...\}$  is bounded, one may choose a convergent subsequence. By Cantor's diagonalization method, we may choose a subsequence  $(f_{n_k})_{k=1}^{\infty}$  of  $(f_n)_{n=1}^{\infty}$  such that  $f_{n_k}$ converges on  $\mathbb{Q}$ . Set, for  $x \in \mathbb{Q}$ ,

$$f(x) = \lim_{k \to \infty} f_{n_k}(x)$$

and define, by the monotonicity of f on  $\mathbb{Q}$ ,

$$f(x) = \lim_{y \downarrow x} f(y) \quad \forall y \in \mathbb{R} \setminus \mathbb{Q}.$$

Clearly, f is non-decreasing in  $\mathbb{R}$ . Let  $\mathcal{C}(f)$  be the set of continuous points of f and  $x \in \mathcal{C}(f)$ . Since f is continuous at x, we may choose, for each  $\epsilon > 0$ ,  $y, z \in \mathbb{Q}$  such that y < x < z and

$$f(x) - \epsilon/2 < f(y) \le f(z) < f(x) + \epsilon/2.$$

Since  $f_{n_k}(y) \to f(y)$  and  $f_{n_k}(z) \to f(z)$ , we may choose an integer  $K = K(\epsilon)$  such that

$$f_{n_k}(y) > f(y) - \epsilon/2, \quad f_{n_k}(z) < f(z) + \epsilon/2 \quad \forall k \ge K.$$

This implies

$$f(x) - \epsilon < f_{n_k}(y) \le f_{n_k}(x) \le f_{n_k}(z) < f(x) + \epsilon, \quad \forall k \ge K,$$

which proves that  $f_n$  converges to f in distribution. By setting g as a right-continuous version of f, one has  $g \in \mathcal{M}$ , f = g on  $\mathcal{C}(f)$  and  $\mathcal{C}(g) = \mathcal{C}(f)$ . Thus,  $f_n$  converges to g in distribution. 

*Remark* 7.2. The set of all distribution functions is not sequentially compact. Consider the following sequence of functions

$$F_n(x) = \begin{cases} 0 & \text{for } x < n \\ 1 & \text{for } x \ge n \end{cases}$$

 $F_n \to 0$  in distribution.

For any  $F \in \mathcal{M}$ , let F(B),  $B \in \mathcal{B}_1$ , be the measure induced by F.

**Definition 7.2.** Let  $\mathcal{N}$  be the set of all distribution functions. A set  $\mathcal{A} \subset \mathcal{N}$  is

- (1) **mass-preserving** if, for all  $\epsilon > 0$ , there exists a finite interval  $I_{\epsilon}$  such that  $F(I_{\epsilon}^{c}) < \epsilon$  for all  $F \in \mathcal{A}$ .
- (2) conditionally compact in  $\mathcal{N}$  under  $\mathcal{D}$ -convergence if any sequence  $(f_n)_{n=1}^{\infty}$  in  $\mathcal{A}$  contains a subsequence that  $\mathcal{D}$ -converges to a distribution function in  $\mathcal{N}$ .

**Lemma 7.3.** A set  $\mathcal{A} \subset \mathcal{N}$  is conditionally compact under  $\mathcal{D}$ -convergence if and only if  $\mathcal{A}$  is mass-preserving.

*Proof.* Assume first that  $\mathcal{A}$  is mass-preserving and let  $(F_n)_{n=1}^{\infty}$  be a sequence of distribution functions in  $\mathcal{A}$ . By Lemma 7.2, there exists a subsequence  $(F_{n_k})_{k=1}^{\infty}$  and  $F \in \mathcal{M}$  such that  $F_{n_k}$  converges to F in distribution. We will show in the following that  $F \in \mathcal{N}$  and it remains to prove  $F(-\infty) = 0$  and  $F(\infty) = 1$ .

Let  $\epsilon > 0$  and choose an interval I such that  $f(I^c) < \epsilon$  for all  $f \in \mathcal{A}$ . Let  $a, b \in \mathcal{C}(F)$  be such that  $I \subset (a, b]$ . This implies

$$F_{n_k}(b) - F_{n_k}(a) = F_{n_k}((a, b]) > 1 - \epsilon \quad \forall k \ge 1,$$

and, then,

$$F((a,b]) = F(b) - F(a) = \lim_{k \to \infty} [F_{n_k}(b) - F_{n_k}(a)] \ge 1 - \epsilon.$$

Letting  $a \to -\infty$  and  $b \to \infty$  yields  $1 \ge F(\infty) - F(-\infty) \ge 1 - \epsilon$  for all  $\epsilon > 0$ . As it is known that  $F(-\infty) \ge 0$  and  $F(\infty) \le 1$ , we have  $F \in \mathcal{N}$ .

For the converse, assume that  $\mathcal{A}$  is conditionally compact and assume the inverse that  $\mathcal{A}$  is not mass-preserving. Equivalently, the latter means that there is  $\epsilon > 0$  such that, for any finite interval I, one may select  $F \in \mathcal{A}$  such that  $F(I) \leq 1 - \epsilon$ . For  $n \geq 1$ , let  $F_n \in \mathcal{A}$  be such that  $F_n([-n,n]) \leq 1 - \epsilon$ . Since  $\mathcal{A}$  is conditionally compact, we may choose a subsequence  $(F_{n_k})_{k=1}^{\infty}$  and  $F \in \mathcal{N}$  such that

$$F_{n_k} \to F$$
 in distribution.

This implies that, for  $a, b \in \mathcal{C}(F)$ ,

$$F((a,b]) = \lim_{k \to \infty} F_{n_k}((a,b]) \le \lim_{k \to \infty} F_{n_k}([-n_k,n_k]) \le 1 - \epsilon.$$

Letting  $a \to -\infty$  and  $b \to \infty$  yields  $F(\infty) - F(-\infty) \le 1 - \epsilon < 1$ . This contradicts the fact  $F \in \mathcal{N}$ .

**Proposition 7.4.** Suppose  $F_n \in \mathcal{N}$  converges to F in distribution. Then,  $F \in \mathcal{N}$  if and only if  $\{F_n | n = 1, 2, ...\}$  is mass-preserving.

Proof. First, assume that  $F \in \mathcal{N}$  and set  $\mathcal{A} = \{F_n | n \geq 1\} \subset \mathcal{N}$ . Let  $(f_n)_{n=1}^{\infty}$  be a sequence in  $\mathcal{A}$  and write  $f_n = F_{\ell_n}$ . Set  $\mathcal{I} = \{\ell_n | n \geq 1\}$ . If  $|\mathcal{I}| < \infty$ , then there is  $M \in \mathcal{I}$  and a subsequence  $(m_k)_{k=1}^{\infty}$  of  $\mathbb{N}$  such that  $f_{m_k} = F_M$  for all  $k \geq 1$ . Clearly,  $f_{m_k}$  converges to  $F_M \in \mathcal{N}$  in distribution. When  $|\mathcal{I}| = \infty$ , we may choose a subsequence  $(m'_k)_{k=1}^{\infty}$  of  $\mathbb{N}$  such that  $(\ell_{m'_k})_{k=1}^{\infty}$  is increasing. This implies that  $f_{m'_k}$  converges to F in distribution. Consequently,  $\mathcal{A}$  is conditionally compact and, by Lemma 7.3,  $\mathcal{A}$  is mass-preserving.

Next, assume that  $(F_n)_{n=1}^{\infty}$  is mass-preserving. By Lemma 7.3, there is a subsequence  $(F_{n_k})_{k=1}^{\infty}$  and  $G \in \mathcal{N}$  such that  $F_{n_k} \to G$  in distribution. Since  $F_{n_k} \to F$ , by Remark 7.1, F = G.

**Exercise 7.1.** For a < b, show that the set of all distribution functions F satisfying F(b) = 1 and  $F(a - \epsilon) = 0$  for all  $\epsilon > 0$  is sequentially compact.

**Exercise 7.2.** Let  $F_n, F \in \mathcal{N}$  and assume that  $F_n \to F$  in distribution. Prove that, if  $B \subset \mathbb{R}$  is closed and  $F_n(B) = 1$  for all  $n \ge 1$ , then F(B) = 1. Find an example that  $F_n(B) = 1$  for  $n \ge 1$  but F(B) < 1.

**Exercise 7.3.** Let  $g : \mathbb{R} \to \mathbb{R}$  be a random variable satisfying  $|g(x)| \to \infty$  as  $|x| \to \infty$  and  $A \subset \mathcal{N}$  be a set satisfying  $\sup_{F \in A} \int |g(x)| dF(x) < \infty$ . Show that A is mass-preserving.

Remark 7.3. For  $n \ge 1$ , let  $X_n$  be a random variable with distribution function  $F_n$ . By the above exercise, if  $\limsup_n \mathbb{E}|X_n|^r < \infty$  for some r > 0, then  $\{F_n | n \in \mathbb{N}\}$  is mass-preserving.

7.2. Preserving the integrations. In this subsection, we consider the following question. Suppose  $F_n \in \mathcal{N}$  converges to  $F \in \mathcal{N}$  in distribution. Under what assumption does the limit  $\int f dF_n \to \int f dF$  hold? For  $n \ge 1$ , let  $F_n$  be defined by

$$F_n(x) = \begin{cases} 0 & \text{for } x < -\frac{1}{n} \\ 1 & \text{for } x \ge -\frac{1}{n} \end{cases}$$

Then,  $F_n \to F$  in distribution, where

$$F(x) = \begin{cases} 0 & \text{for } x < 0\\ 1 & \text{for } x \ge 0 \end{cases}$$

Let f be a function defined by f(x) = 1 for x < 0 and f(x) = 0 for  $x \ge 0$ . Clearly, one may compute that  $\int f dF_n = 1$  and  $\int f dF = 0$ .

**Proposition 7.5.** Let X and  $X_n$  with  $n, \ge 1$  be random variables with distribution functions F and  $F_n$ . Then, the following are equivalent.

- (1)  $F_n \to F$  in distribution.
- (2)  $\mathbb{E}\phi(X_n) \to \mathbb{E}\phi(X)$  for any bounded continuous function  $\phi$ .
- (3)  $\mathbb{E}\phi(X_n) \to \mathbb{E}\phi(X)$  for any bounded  $\mathcal{B}(\mathbb{R})$ -measurable function  $\phi$  satisfying  $F(\mathcal{C}(\phi)) = 1$ .
- (4) There are  $Y_n$  and Y such that  $X_n \stackrel{d}{=} Y_n$ ,  $X \stackrel{d}{=} Y$  and  $Y_n \to Y$ .

Remark 7.4. For any  $\mathcal{B}(\mathbb{R})$ -measurable function  $\phi, \mathcal{C}(\phi) \in \mathcal{B}(\mathbb{R})$ .

Proof of Proposition 7.5. For (4) $\Rightarrow$ (3), let  $Y_n \stackrel{d}{=} X_n$  and  $Y \stackrel{d}{=} X$  be random variables on  $(\Omega, \mathcal{F}, P)$  such that  $Y_n \to Y$ . By Theorem 4.15, we have

$$\mathbb{E}\phi(X_n) = \mathbb{E}\phi(Y_n), \quad \mathbb{E}\phi(X) = \mathbb{E}\phi(Y)$$

Set  $A = \{Y \in \mathcal{C}(\phi)\}$ . Note that  $\mathbb{P}(A) = F(\mathcal{C}(\phi)) = 1$  and  $\phi(Y_n)$  converges to  $\phi(Y)$  on A. By the Lebesgue dominant convergence theorem,  $\mathbb{E}\phi(Y_n) \to \mathbb{E}\phi(Y)$ , as desired.

 $(3) \Rightarrow (2)$  is clear. For  $(2) \Rightarrow (1)$ , let  $x \in \mathcal{C}(F)$  and, for  $\epsilon > 0$ , set

$$\phi(t) = \begin{cases} 1 & \text{for } t < x \\ \frac{x + \epsilon - t}{\epsilon} & \text{for } x \le t \le x + \epsilon \\ 0 & \text{for } t > x + \epsilon \end{cases}$$

It is easy to see that  $\mathbb{E}\phi(X_n) \ge F_n(x)$  and this implies

$$\limsup_{n \to \infty} F_n(x) \le \mathbb{E}\phi(X) \le F(x+\epsilon).$$

Similarly, one may use  $\varphi(t) = \phi(t + \epsilon)$  to derive

$$\liminf_{n \to \infty} F_n(x) \ge F(x - \epsilon).$$

By the continuity of F at x, letting  $\epsilon \to 0$  implies  $F_n(x) \to F(x)$ .

For  $(1) \Rightarrow (4)$ , we define, for  $n \ge 1$ ,

$$f_n(t) = \inf\{x | F_n(x) \ge t\} = \min\{x | F_n(x) \ge t\}, \quad \forall t \in (0, 1).$$

In the above setting,  $f_n$  is non-decreasing and satisfies

(7.1) 
$$f_n(t) \le x \quad \Leftrightarrow \quad F_n(x) \ge t.$$

To see the left-continuity of  $f_n$ , observe that

$$F_n(f_n(t)) \ge t, \quad \forall t \in (0,1).$$

Fix  $t \in (0, 1)$ . By the monotonicity of  $f_n$  and  $F_n$ , the above inequality implies

$$F_n\left(\lim_{s \to t^-} f_n(s)\right) \ge \lim_{s \to t^-} F_n\left(f_n(s)\right) \ge t,$$

which yields

$$f_n(t) \le \lim_{n \to t^-} f_n(s) \le f_n(t).$$

Let f be the function defined by  $f(t) = \inf\{x | F(x) \ge t\}$ .

**Step 1:**  $f_n \to f$  on  $\mathcal{C}(f)$ . Let  $t \in \mathcal{C}(f)$  and x = f(t). Since f is right-continuous at  $t, F(x + \epsilon) > t$  for all  $\epsilon > 0$ , otherwise  $f(s) \ge x + \epsilon$  for all s > t. This implies that, for  $\epsilon > 0$  and  $x + \epsilon \in \mathcal{C}(F)$ , one may choose  $\delta > 0$  such that  $F(x + \epsilon) \ge t + 2\delta$ . Since  $F_n(x + \epsilon) \to F(x + \epsilon)$ , we may choose  $N \in \mathbb{N}$  such that  $F_n(x + \epsilon) \ge t + \delta$  for  $n \ge N$ . This implies  $f_n(t) \le f_n(t + \delta) \le x + \epsilon = f(t) + \epsilon$  for  $n \ge N$ . Letting  $n \to \infty$  and  $\epsilon \to 0$  yields  $\limsup_n f_n(t) \le x$ .

Next, observe that  $F(x-\epsilon) < t$  for  $\epsilon > 0$ . This implies, for  $\epsilon > 0$  and  $x-\epsilon \in \mathcal{C}(F)$ , one may select  $\delta > 0$  such that  $F(x-\epsilon) < t-2\delta$ . Since  $F_n(x-\epsilon) \to F(x-\epsilon)$ , we may choose  $N' \in \mathbb{N}$ such that  $F_n(x-\epsilon) < t-\delta$  for  $n \ge N'$ . This implies  $f_n(t) \ge f_n(t-\delta) > x-\epsilon = f(t)-\epsilon$  for  $n \ge N'$ . Letting  $n \to \infty$  and  $\epsilon \to 0$  yields  $\liminf_n f_n(t) \ge x$ .

**Step 2:** Let  $\mu$  be the Lebesgue measure on (0, 1). By (7.1), one has

$$\mu(\{t \in (0,1) : f_n(t) \le x\}) = \mu(\{t \in (0,1) : F_n(x) \ge t\}) = F_n(x).$$

This implies  $X_n \stackrel{d}{=} f_n$  and similarly  $X \stackrel{d}{=} f$ . To achieve the pointwise convergence, one only needs to set  $Y_n = f_n \mathbf{1}_A$  and  $Y = f \mathbf{1}_A$ , where  $A = (0, 1) \cap \mathcal{C}(f)$ .

**Definition 7.3.** A sequence of random variables  $(X_n)_{n=1}^{\infty}$  is said to converge to X in distribution if the distribution of  $X_n \mathcal{D}$ -converges to that of X.

**Corollary 7.6.** Let  $X_1, X_2, ...$  be random variables converging in distribution to X. If  $\phi$  is a nonnegative function satisfying  $\mathbb{P}(X \in \mathcal{C}(\phi)) = 1$ , then  $\mathbb{E}[\phi(X)] \leq \liminf_n \mathbb{E}[\phi(X_n)]$ .

**Exercise 7.4.** Let  $F_n, F \in \mathcal{N}$  and suppose  $F_n \to F$  in distribution. Show that if  $B \in \mathcal{B}(\mathbb{R})$  satisfies  $F(\partial B) = 0$ , where  $\partial B = \overline{B} \cap \overline{B^c}$  is the boundary of B, then  $F_n(B) \to F(B)$ .

**Exercise 7.5.** Let  $X_1, X_2, ...$  be a sequence of random variables converging to X in distribution. Let g, h be continuous functions satisfying

$$\lim_{|x|\to\infty} |g(x)| = \infty, \quad \lim_{|x|\to\infty} \frac{|h(x)|}{|g(x)|} = 0.$$

Prove that if  $\limsup_{n} \mathbb{E}|g(X_n)| < \infty$ , then  $\mathbb{E}[h(X_n)] \to \mathbb{E}[h(X)]$ .

## 7.3. Classes of functions that separate.

**Definition 7.4.** Let  $\mathcal{N}$  be the set of all distribution functions. A class  $\mathcal{C}$  of bounded continuous functions defined on  $\mathbb{R}$  (not necessarily real-valued) is called  $\mathcal{N}$ -separating if, for any  $F, G \in \mathcal{N}$ ,

$$\int_{\mathbb{R}} f(x) dF(x) = \int_{\mathbb{R}} f(x) dG(x), \quad \forall f \in \mathcal{C},$$

implies F = G.

**Proposition 7.7.** Let C be an  $\mathcal{N}$ -separating class. Then,  $F_n \in \mathcal{N}$  converges in distribution to some distribution  $F \in \mathcal{N}$  if and only if  $(F_n)_{n=1}^{\infty}$  is mass-preserving and

$$\lim_{n \to \infty} \int_{\mathbb{R}} f(x) dF_n(x) \quad exists, \quad \forall f \in \mathcal{C}.$$

Proof. By Proposition 7.5, if  $F_n \mathcal{D}$ -converges to F, then  $\int f dF_n \to \int f dF$  for all  $f \in \mathcal{C}$ . Conversely, assume that  $(F_n)_{n=1}^{\infty}$  is mass-preserving and  $\int f dF_n$  converges for all  $f \in \mathcal{C}$ . By Lemma 7.2 and Proposition 7.4, one may choose a  $\mathcal{D}$ -convergent subsequence  $(F_{n_k})_{k=1}^{\infty}$  with limit  $F \in \mathcal{N}$  and, then,  $\int f dF_n \to \int f dF$ . Let  $(F_{m_k})_{k=1}^{\infty}$  be any subsequence of  $(F_n)_{n=1}^{\infty}$  and  $(F_{m'_k})_{k=1}^{\infty}$  be a further subsequence that is  $\mathcal{D}$ -convergent with limit  $G \in \mathcal{N}$ . As a result, this implies

$$\int f dF = \lim_{n \to \infty} \int f dF_n = \lim_{k \to \infty} \int f dF_{m'_k} = \int f dG, \quad \forall f \in \mathcal{C}.$$
  
Since  $\mathcal{C}$  separates  $\mathcal{N}, F = G$ . By Lemma 7.1,  $F_n \to F$  in distribution.

**Proposition 7.8.** Let  $C_0$  be the set of functions of the following form.

(7.2) 
$$f(x) = \begin{cases} 0 & x < a - \epsilon \\ 1 + \frac{x-a}{\epsilon} & a - \epsilon \le x < a \\ 1 & a \le x < b \\ 1 - \frac{x-b}{\epsilon} & b \le x < b + \epsilon \\ 0 & x \ge b + \epsilon \end{cases}$$

where a < b and  $\epsilon > 0$ . Then  $C_0$  is  $\mathcal{N}$ -separating.

*Proof.* Let  $F, G \in \mathcal{N}$  and a < b with  $a \in \mathcal{C}(F) \cap \mathcal{C}(G)$ . Let f be the function defined by (7.2). Then,

$$\int_{\mathbb{R}} f(x)dF(x) = \int_{\mathbb{R}} f(x)dG(x).$$

This implies

$$F((a,b]) \le \int_{\mathbb{R}} f(x) dF(x) = \int_{\mathbb{R}} f(x) dG(x) \le G((a-\epsilon,b+\epsilon]).$$

Similarly, one has  $G((a, b]) \leq F((a - \epsilon, b + \epsilon])$ . Letting  $\epsilon \to 0$  yields F((a, b]) = G((a, b]) for all  $a \in \mathcal{C}(F) \cap \mathcal{C}(G)$  and, then, for all  $a \in \mathbb{R}$ . This proves F = G.

**Corollary 7.9.** Let  $C_0$  be the class in Proposition 7.8 and C be a class of bounded continuous functions on  $\mathbb{R}$  satisfying the property that, for any  $f \in C_0$ , there exists a sequence  $f_n \in C$  such that  $f_n$  converges to f boundedly. Then C is  $\mathcal{N}$ -separating.

Proof. Let  $F, G \in \mathcal{N}$  and assume that  $\int_{\mathbb{R}} f(x) dF(x) = \int_{\mathbb{R}} f(x) dG(x)$  for all  $f \in \mathcal{C}$ . For  $g \in \mathcal{C}_0$ , let  $g_n \in \mathcal{C}$  converge to g boundedly. By the Lebesgue dominant convergence theorem, we have

$$\int_{\mathbb{R}} g(x)dF(x) = \lim_{n \to \infty} \int_{\mathbb{R}} g_n(x)dF(x) = \lim_{n \to \infty} \int_{\mathbb{R}} g_n(x)dG(x) = \int_{\mathbb{R}} g(x)dG(x).$$
position 7.8  $F = G$ 

By Proposition 7.8, F = G.

**Exercise 7.6.** Let  $C_0^{\infty}(\mathbb{R})$  be the set of all infinitely differentiable functions on  $\mathbb{R}$  with compact support. Prove that  $C_0^{\infty}(\mathbb{R})$  is  $\mathcal{N}$ -separating. (*Hint*: Let  $\phi$  be a function defined by

$$\phi(x) = \begin{cases} c \exp\left\{-\frac{1}{1-x^2}\right\} & |x| < 1\\ 0 & |x| \ge 1 \end{cases}$$

where c is a constant such that  $\int_{-\infty}^{\infty} \phi(x) dx = 1$ . For  $\epsilon > 0$ , set  $\phi_{\epsilon}(x) = \epsilon^{-1} \phi(x \epsilon^{-1})$ . Consider the convolution  $f * \phi_{\epsilon}$  and show that, for any compactly supported function  $f, f * \phi_{\epsilon} \in C_0^{\infty}(\mathbb{R})$ and converges to f uniformly as  $\epsilon \to 0$ .)

7.4. Characteristic functions. Let  $C = \{e^{inx} : n \in \mathbb{Z}\}$  and  $C(S^1)$  be the set of all continuous functions f defined on  $[0, 2\pi]$  satisfying  $f(0) = f(2\pi)$ . A trigonometric polynomial is a finite linear combination of functions in C. The following theorem is a well-known result about the denseness of trigonometric polynomials in  $C(S^1)$  under the sup-norm.

**Theorem 7.10.** For  $f \in C(S^1)$  and  $\epsilon > 0$ , there exists a trigonometric polynomial g such that

$$\sup_{0 \le x \le 2\pi} |f(x) - g(x)| < \epsilon.$$

*Proof.* See Rudin's "Real and complex analysis" (Chapter 4).

**Theorem 7.11.** The class  $\{e^{iux} : u \in \mathbb{R}\}$  is  $\mathcal{N}$ -separating.

*Proof.* Let  $C_0$  be the function in Proposition 7.8. We prove this theorem by the following two steps.

Step 1: Let  $f \in C_0$  and choose  $N \in \mathbb{N}$  such that f = 0 on  $[-N\pi, N\pi]^c$ . By Theorem 7.10, we may select, for each  $n \geq N$ , a trigonometric polynomial  $f_n(x) = \sum_k c_{n,k} e^{ikx}$  such that  $\sup_{|x|\leq \pi} |f_n(x) - f(nx)| < 1/n$ . By setting  $g_n(x) = f_n(x/n) = \sum_k c_{n,k} e^{ikx/n}$ , this is equivalent to  $\sup_{|x|\leq n\pi} |g_n(x) - f(x)| < 1/n$ , which implies  $g_n \to f$  on  $\mathbb{R}$ . Furthermore, since f is bounded and  $g_n$  is periodic,  $g_n$  is uniformly bounded. This implies that any function in  $C_0$  is the limit of some boundedly convergent sequence of trigonometric polynomials.

**Step 2:** Let F, G be two distribution functions satisfying

(7.3) 
$$\int_{\mathbb{R}} e^{iux} dF(x) = \int_{\mathbb{R}} e^{iux} dG(x), \quad \forall u \in \mathbb{R}.$$

Let  $f \in C_0$  and, by Step 1, let  $(g_n)_{n=1}^{\infty}$  be a sequence of trigonometric polynomials converging to f boundedly. By (7.3), it is clear that  $\int_{\mathbb{R}} g_n(x) dF(x) = \int_{\mathbb{R}} g_n(x) dG(x)$  for all  $n \ge 1$ . The LDCT then implies  $\int_{\mathbb{R}} f(x) dF(x) = \int_{\mathbb{R}} f(x) dG(x)$ .

**Definition 7.5.** Let F be a distribution function and X be a random variable.

(1) The characteristic function of F is defined to be the following complex-valued function

$$f(u) = \int_{\mathbb{R}} e^{iux} dF(x).$$

(2) The characteristic functions of X is defined by

$$f(u) = \mathbb{E}[e^{iuX}].$$

**Proposition 7.12.** Let F be a distribution function with characteristic function f. Then,

(1) f(0) = 1,

(2) 
$$|f(u)| \le 1$$
,

(3) 
$$f(-u) = f(u)$$
,

(4) f is uniformly continuous on  $\mathbb{R}$ ,

(5) There exists a constant  $A \in (0, \infty)$  independent of F such that

$$F([-1/u, 1/u]^c) \le \frac{A}{u} \int_0^u (1 - \Re f(v)) dv \quad \forall u > 0$$

where  $\Re a$  is the real part of a.

*Proof.* The proofs of (1)-(3) are clear. For (4), observe that, when  $h \in \mathbb{R}$ ,

$$|f(u+h) - f(u)| = \left| \int_{\mathbb{R}} e^{iux} (e^{ihx} - 1) dF(x) \right| \le \int_{\mathbb{R}} |e^{ihx} - 1| dF(x).$$

This implies that, for  $\delta > 0$ ,

$$\Delta(\delta) := \sup_{u, |h| < \delta} |f(u+h) - f(u)| \le \sup_{|h| \le \delta} \int_{\mathbb{R}} |e^{ihx} - 1| dF(x)| dF(x) \le |h| \le \delta \int_{\mathbb{R}} |e^{ihx} - 1| dF(x)| dF(x$$

Since  $\int_{\mathbb{R}} |e^{ihx} - 1| dF(x) \to 0$  when  $|h| \to 0$ , we have  $\Delta(\delta) \to 0$  when  $\delta \to 0$ . For (5), note that

$$\frac{1}{u} \int_0^u [1 - \Re f(v)] dv = \frac{1}{u} \int_0^u \left( \int_{\mathbb{R}} [1 - \cos(vx)] dF(x) \right) dv$$
$$= \int_{\mathbb{R}} \left( \frac{1}{u} \int_0^u [1 - \cos(vx)] dv \right) dF(x) = \int_{\mathbb{R}} \left( 1 - \frac{\sin(ux)}{ux} \right) dF(x).$$

This yields

$$\frac{1}{u} \int_0^u [1 - \Re f(v)] dv \ge \int_{|ux| > 1} \left( 1 - \frac{\sin(ux)}{ux} \right) dF(x) \ge A^{-1} F\left( [-1/u, 1/u]^c \right),$$
  
$$I^{-1} := \inf \left\{ 1 - \frac{\sin t}{t} : |t| \ge 1 \right\} > 0.$$

where  $A^{-1} := \inf \left\{ 1 - \frac{\sin t}{t} : |t| \ge 1 \right\} > 0.$ 

The next theorem simplifies the assumption of mass-preservation in Proposition 7.7.

**Theorem 7.13** (The continuity theorem). For  $n \ge 1$ , let  $F_n$  be a distribution function and  $f_n$  be the corresponding characteristic function  $f_n$ . Then, there exists  $F \in \mathcal{N}$  such that  $F_n$ converges in distribution to F if and only if  $f_n$  converges pointwise to some function f, which is continuous at 0. Furthermore, f is the characteristic function of F.

*Proof.* By Proposition 7.7, since  $\{e^{iux} : u \in \mathbb{R}\}$  is  $\mathcal{N}$ -separating, it remains to prove that if  $f_n \to f$  and f is continuous at 0, then  $(F_n)_{n=1}^{\infty}$  is mass-preserving. By Proposition 7.12(5), one may choose a universal constant A > 0 such that

$$F_n([-1/u, 1/u]^c) \le \frac{A}{u} \int_0^u [1 - \Re f_n(v)] dv$$

By the Lebesgue dominant convergence theorem, we have

$$\limsup_{n \to \infty} F_n([-1/u, 1/u]^c) \le \frac{A}{u} \int_0^u [1 - \Re f(v)] dv$$

Since f is continuous at 0, the above inequality implies

$$\lim_{u \downarrow 0} \limsup_{n \to \infty} F_n([-1/u, 1/u]^c) = 0.$$

This is equivalent to say that, for any  $\epsilon > 0$ , there is u > 0 and  $N \in \mathbb{N}$  such that

$$F_n([-1/u, 1/u]^c) \le \epsilon, \quad \forall n \ge N.$$

This implies  $(F_n)_{n=1}^{\infty}$  is mass-preserving.

**Lemma 7.14.** Let  $(F_n)_{n=1}^{\infty}$  be a sequence in  $\mathcal{N}$  with characteristic functions  $(f_n)_{n=1}^{\infty}$ . If  $F_n$   $\mathcal{D}$ -converges to  $F \in \mathcal{N}$ , then, for any  $\epsilon > 0$ , there is  $\delta > 0$  such that

$$|f_n(u+h) - f_n(u)| < \epsilon, \quad \forall |h| < \delta, u \in \mathbb{R}, n \ge 1.$$

*Proof.* Since  $F_n$  converges to F in distribution,  $(F_n)_{n=1}^{\infty}$  is mass-preserving. For  $\epsilon > 0$ , we choose a finite interval I such that  $F_n(I^c) < \epsilon/3$  for all  $n \ge 1$ . This implies that, for  $u, h \in \mathbb{R}$ ,

$$|f_n(u+h) - f_n(u)| \le \int_{\mathbb{R}} |e^{ihx} - 1| dF_n(x) \le \int_I |e^{ihx} - 1| dF_n(x) + 2F_n(I^c) \le \Delta(h) + \frac{2\epsilon}{3},$$

where  $\Delta(h) = \max_{x \in I} |e^{ihx} - 1|$ . It is clear that  $\Delta(h) \to 0$  as  $|h| \to 0$ . This implies that one may choose  $\delta > 0$  such that  $\Delta(h) < \epsilon/3$  when  $|h| < \delta$ .

**Theorem 7.15.** For  $n \ge 1$ , let  $F_n \in \mathcal{N}$  and  $f_n(u) = \int_{\mathbb{R}} e^{iux} dF_n(x)$ . Then, there exists  $F \in \mathcal{N}$  such that  $F_n$   $\mathcal{D}$ -converges to F if and only if  $f_n$  converges uniformly to some function f in every finite interval.

*Proof.* By Theorem 7.13, it remains to show that if  $F_n \mathcal{D}$ -converges to F, then  $f_n$  converges uniformly on any finite interval. Let I be a finite interval. By Lemma 7.14, we may choose, for each  $\epsilon > 0$ , a positive constant  $\delta > 0$  such that

$$|f_n(u+h) - f_n(u)| < \epsilon/3, \quad \forall |h| < \delta, u \in \mathbb{R}, n \ge 1$$

and

$$|f(u+h) - f(u)| < \epsilon/3, \quad \forall |h| < \delta, u \in \mathbb{R}.$$

Let  $u_1 < u_2 < \cdots < u_k$  be points in I such that  $I \subset \bigcup_{i=1}^k (u_i - \delta, u_i + \delta)$ . Since  $f_n(u_i) \to f(u_i)$  for  $1 \le i \le k$ , we may choose  $N \in \mathbb{N}$  such that

$$\max_{1 \le i \le k} |f_n(u_i) - f(u_i)| < \epsilon/3, \quad \forall n \ge N.$$

This implies that, for  $u \in I$  with  $|u - u_i| < \delta$ ,

$$|f_n(u) - f(u)| \le |f_n(u) - f_n(u_i)| + |f_n(u_i) - f(u_i)| + |f(u_i) - f(u)| < \epsilon.$$

when  $n \geq N$ .

**Exercise 7.7.** A random variable is said to have a symmetric distribution if  $\mathbb{P}(X \in B) = \mathbb{P}(-X \in B)$  for all  $B \in \mathcal{B}(\mathbb{R})$ . Prove that the characteristic function of X is real if and only if X has a symmetric distribution.

**Exercise 7.8.** Show that any characteristic function f is non-negative definite, that is, for any  $\lambda_1, ..., \lambda_n \in \mathbb{C}$  and  $u_1, ..., u_n \in \mathbb{R}$ ,

$$\sum_{k,\ell=1}^{n} f(u_k - u_\ell) \lambda_k \overline{\lambda}_\ell \ge 0.$$

**Exercise 7.9.** For  $n \ge 1$ , let  $b_n \in \mathbb{R}$  and  $X_n = b_n$  a.s. with characteristic function  $f_n$ . Prove that the following are equivalent.

- (1)  $X_n \xrightarrow{\mathcal{D}} X$  for some  $X \in \mathcal{N}$ .
- (2)  $b_n$  is Cauchy.
- (3) There is  $\epsilon > 0$  such that the sequence  $(f_n(u))_{n=1}^{\infty}$  is Cauchy for all  $|u| < \epsilon$ .

**Corollary 7.16.** Let  $X_n$  be random variables with characteristic function  $f_n$ . Then,  $X_n \xrightarrow{\mathcal{D}} 0$  if and only if  $f_n \to 1$  on (-a, a) for some  $a < \infty$ .

*Proof.* One direction is obvious from Theorem 7.13. For the other direction, assume that  $f_n$ converges to 1 on (-a, a). Let  $F_n$  be the distribution of  $X_n$ . By Proposition 7.12 (5), there is  $A < \infty$  such that

$$F_n([-1/a, 1/a]^c) \le \frac{A}{a} \int_0^a (1 - \Re f_n(v)) dv, \quad \forall n \ge 1.$$

Since  $f_n \to 1$  on (-a, a), the right hand side converges to 0 and, thus,  $F_n$  is mass-preserving. Let  $(F_{k_n})_{n=1}^{\infty}$  be a subsequence of  $(F_n)_{n=1}^{\infty}$  and  $(F_{n'_k})_{k=1}^{\infty}$  be a  $\mathcal{D}$ -convergent further subsequence with limit  $F \in \mathcal{N}$ . Let f be the characteristic function of F. Then, f is identically 1 on (-a, a). This implies

$$F(\{2\ell\pi/u : \ell \in \mathbb{Z}\}) = 1, \quad \forall 0 < |u| < a.$$

Hence,  $F(\{0\}) = 1$ , as desired.