

(1) **Continuous mapping theorem**

Consider an arbitrary subsequence  $n_1 < n_2 < \dots$ . Since  $f_{n_k} \xrightarrow{\mathbb{P}} f$ , by Theorem 9.2.4 (4) there is a further subsequence with  $f_{n_{k_l}} \xrightarrow{a.s.} f$ . Since  $\varphi$  is continuous, if  $f_{n_{k_l}}(\omega) \rightarrow f(\omega)$  then  $\varphi(f_{n_{k_l}}(\omega)) \rightarrow \varphi(f(\omega))$ , so  $\varphi(f_{n_{k_l}}) \xrightarrow{a.s.} \varphi(f)$ . Therefore by Theorem 9.2.4 (4) again,  $\varphi(f_{n_k}) \xrightarrow{\mathbb{P}} \varphi(f)$ .

(2) **Sequence of coin tosses**

The probability is 0. Indeed, consider the probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  where  $\Omega = \{H, T\}$ ,  $\mathcal{F} = 2^{\{H, T\}}$ , and  $\mathbb{P}(\{H\}) = \mathbb{P}(\{T\}) = 1/2$  (this represents one toss of a fair coin). A sequence of independent tosses of a fair coin is represented by the infinite product probability space  $(\times_{n=1}^{\infty} \Omega, \otimes_{n=1}^{\infty} \mathcal{F}, \otimes_{n=1}^{\infty} \mathbb{P})$ . For  $n \in \mathbb{N}$  define  $X_n: \times_{n=1}^{\infty} \Omega \rightarrow \mathbb{R}$  by

$$X_n((\omega_1, \omega_2, \dots)) = \begin{cases} 1 & \text{if } \omega_n = H, \\ -1 & \text{if } \omega_n = T. \end{cases}$$

Then  $X_1, X_2, \dots$  are i.i.d. random variables with  $\mathbb{E}(X_n) = 0$  and  $\mathbb{E}(X_n^4) = 1 < \infty$ . Therefore by the Strong Law of Large Numbers,

$$\otimes_{n=1}^{\infty} \mathbb{P} \left( \frac{X_1 + \dots + X_n}{n} \xrightarrow{n \rightarrow \infty} 0 \right) = 1.$$

But if  $\frac{X_1 + \dots + X_n}{n} \xrightarrow{n \rightarrow \infty} 0$  then there exists  $N \in \mathbb{N}$  such that for all  $n \geq N$  it holds that  $\frac{X_1 + \dots + X_n}{n} < 0.001$ , so there are only finitely many  $n_k$  such that at least 51% of the first  $n_k$  coin tosses are Heads.

(3) **convergence?**

(a)  $f_n(x) \rightarrow \mathbf{1}_{\{1\}}(x)$  for all  $x \in [0, 1]$ . So  $f_n \xrightarrow{\lambda-a.s.} 0$  and  $f_n \xrightarrow{\delta_1-a.s.} 1$ . Since  $|f_n| \leq 1$  for all  $n$ , by the dominated convergence theorem,  $\mathbb{E}_{\mu} \lim_n f_n = \lim_n \mathbb{E}_{\mu} f_n$  for both  $\lambda$  and  $\delta_1$ .

(b)  $f_n(x) \rightarrow \infty \mathbf{1}_{\{1\}}(x)$  for all  $x \in [0, 1]$ . Indeed, for  $x \in [0, 1)$ ,  $nx^n \rightarrow 0$  because “exponential beats polynomial.” So  $f_n \xrightarrow{\lambda-a.s.} 0$  and  $f_n \xrightarrow{\delta_1-a.s.} \infty$ . Now

$$\mathbb{E}_{\delta_1} \lim_n f_n = \infty = \lim_n n = \lim_n \mathbb{E}_{\delta_1} f_n.$$

But

$$\mathbb{E}_{\lambda}(f_n) = \int_0^1 nx^n dx = \frac{n}{n+1} [x^{n+1}]_0^1 = \frac{n}{n+1} \rightarrow 1,$$

while  $\mathbb{E}_{\lambda}(\lim_n f_n) = 0$ .

(c)  $f_n(x) \rightarrow \infty \mathbf{1}_{\{0\}}(x)$  for all  $x \in [0, 1]$ . So  $f_n \rightarrow 0$  a.s. for both  $\lambda$  and  $\delta_0$ . So for all  $n$ ,  $0 = \mathbb{E}_{\delta_1} f_n = \mathbb{E}_{\delta_1} \lim_n f_n$ . Now,  $\mathbb{E}_{\lambda} f_n = \frac{4^n}{2^n} = 2^n \rightarrow \infty$ , but  $\mathbb{E}_{\lambda}(\lim_n f_n) = 0$ .

(4) **convergence in probability: equivalences**

For the forward implication, since  $x \mapsto \min\{1, x\}$  is continuous for all  $p \in (0, \infty)$ , by the continuous mapping theorem  $f_n \xrightarrow{\mathbb{P}} f$  implies  $\min\{1, |f_n - f|^p\} \xrightarrow{\mathbb{P}} 0$ . Moreover  $\sup_n (\min\{1, |f_n - f|^p\}) \leq 1$  so  $\mathbb{E}(\sup_n (\min\{1, |f_n - f|^p\})) \leq 1 < \infty$ . Now Proposition 9.3.3 (4) implies  $\mathbb{E}(\min\{1, |f_n - f|^p\}) \rightarrow 0$ .

For the backward implication, let  $p \in (0, \infty)$  be such that  $\mathbb{E}(\min\{1, |f_n - f|^p\}) \rightarrow 0$ , and let  $0 < \varepsilon \leq 1$ . Then by Markov’s inequality,

$$\mathbb{P}(|f_n - f| \geq \varepsilon) = \mathbb{P}(|f_n - f|^p \geq \varepsilon^p) = \mathbb{P}(\min\{1, |f_n - f|^p\} \geq \varepsilon^p) \leq \varepsilon^{-p} \mathbb{E}(\min\{1, |f_n - f|^p\}) \rightarrow 0$$

by assumption. The  $\varepsilon > 1$  case follows trivially since  $\mathbb{P}(|f_n - f| > \varepsilon) \leq \mathbb{P}(|f_n - f|^p \geq \min\{\varepsilon, 1\})$ .

(5)  $\mathcal{L}_0(\Omega, \mathcal{F}, \mathbb{P})$  a metric space?

(a) is true:  $D(f, g) = 0 \Leftrightarrow \min\{1, |f - g|\} = 0 \text{ a.s.} \Leftrightarrow |f - g| = 0 \text{ a.s.} \Leftrightarrow f = g \text{ a.s.}$

(b) is clearly true.

(c) is also true. Indeed, for  $a, b \geq 0$ ,  $\min\{1, a\} + \min\{1, b\} = a + b = \min\{1, a + b\}$  if  $a + b \leq 1$ , while  $\min\{1, a\} + \min\{1, b\} \geq 1 = \min\{1, a + b\}$  if  $a + b > 1$ . Therefore by the definition of  $D$ ,

$$D(f, h) \leq \mathbb{E}(\min\{1, |f - g| + |g - h|\}) \leq D(f, g) + D(g, h).$$

Remark:  $\mathcal{L}_0(\Omega, \mathcal{F}, \mathbb{P})$  turns into a metric space if we identify functions which are a.s. equal, i.e. the space of equivalence classes for the equivalence relation  $f \sim g$  if  $f = g$  a.e. becomes a metric space.

(6) **uniform integrability**

(a) First note that if  $a, b \in \mathbb{R}$  and  $|a + b| \geq c$  then either  $|a| \geq c/2$  or  $|b| \geq c/2$  (or both), and  $|a + b| \leq 2 \max\{|a|, |b|\}$ . Therefore

$$\mathbf{1}_{|f+g| \geq c/2} |f + g| \leq 2[\mathbf{1}_{|f| \geq c/2} |f| + \mathbf{1}_{|g| \geq c/2} |g|].$$

Integrating gives the desired inequality.

(b) Using part (a), for all  $c > 0$ ,

$$\begin{aligned} \sup_i \mathbb{E}(\mathbf{1}_{|\alpha f_i + \beta g_i| \geq c} |f + g|) &\leq 2 \sup_i (\mathbb{E}(\mathbf{1}_{|\alpha f_i| \geq c/2} |\alpha f_i|) + \mathbb{E}(\mathbf{1}_{|\beta g_i| \geq c/2} |\beta g_i|)) \\ &\leq 2(\alpha \sup_i \mathbb{E}(\mathbf{1}_{|f_i| \geq c/(2|\alpha|)} |f_i|) + \beta \sup_j \mathbb{E}(\mathbf{1}_{|g_j| \geq c/(2|\beta|)} |g_j|)) \xrightarrow{c \rightarrow \infty} 0, \end{aligned}$$

where we used the obvious convention that if  $\alpha$  or  $\beta$  is 0 then we ignore the corresponding terms.

(c) We observed in lectures that by the dominated convergence theorem, for each  $i$ ,  $\{f_i\}$  is u.i. Therefore for all  $\varepsilon > 0$  we can find  $c_1, \dots, c_N > 0$  such that for each  $i$ ,  $\mathbb{E}(\mathbf{1}_{|f_i| \geq c_i} |f_i|) \leq \varepsilon$ . Therefore if  $c := \max\{c_1, \dots, c_N\}$  then

$$\max_{1 \leq i \leq N} \mathbb{E}(\mathbf{1}_{|f_i| \geq c} |f_i|) \leq \max_{1 \leq i \leq N} \mathbb{E}(\mathbf{1}_{|f_i| \geq c_i} |f_i|) \leq \varepsilon,$$

so  $\{f_1, \dots, f_N\}$  is u.i.