

# 1 强大数定律的证明

这份笔记整理了 Etemadi 在 1981 年给出的强大数定律的证明。Etemadi 版本的大数定律只求随机变量序列两两独立、同分布、可积，这比 Kolmogorov 大数定律的假设要弱（相互独立、同分布、可积）。

**Theorem 1.1** (Strong law of large number, Etemadi). *Suppose  $\{X_n : n \geq 1\}$  is a sequence of pairwise independent identically distributed r.v. with  $\mathbb{E}(|X_i|) < \infty$ . Let  $\mu = \mathbb{E}(X_i)$ ,*

$$S_n = X_1 + X_2 + \cdots + X_n,$$

then

$$\frac{S_n}{n} \rightarrow \mu \quad a.s. \quad (1)$$

as  $n \rightarrow \infty$ .

**Lemma 1.2.** *Suppose  $X$  is a r.v. with  $X \geq 0$ , then we have*

1.

$$\sum_{n=1}^{\infty} \mathbb{P}(X \geq n) \leq \mathbb{E}(X) \leq 1 + \sum_{n=1}^{\infty} \mathbb{P}(X \geq n)$$

2. For any  $p > 0$ ,

$$\mathbb{E}(X^p) = \int_0^{\infty} pt^{p-1} \mathbb{P}(X > t) dt.$$

**Lemma 1.3.** *Let  $Y_k = X_k \mathbb{1}_{\{|X_k| \leq k\}}$  and  $T_n = Y_1 + Y_2 + \cdots + Y_n$ , then*

$$\frac{T_n}{n} \rightarrow \mu \quad a.s.$$

implies (1).

*Proof.* 1. By Lemma 1.2,

$$\sum_{k=1}^{\infty} \mathbb{P}(|X_k| > k) = \sum_{n=1}^{\infty} \mathbb{P}(|X_1| > n) \leq \mathbb{E}(|X_1|) < \infty,$$

then by Borel-Cantelli Lemma,

$$\mathbb{P}(X_k \neq Y_k \text{ i.o.}) = \mathbb{P}(|X_k| > k \text{ i.o.}) = 0. \quad (2)$$

2. Let  $A = \{X_k \neq Y_k \text{ i.o.}\}$ , for any  $\omega \in A^c$  (w.p.1.),  $X_k \neq Y_k$  only for finitely many  $k$ , i.e. there exists  $N(\omega)$ , s.t.  $X_k(\omega) = Y_k(\omega)$  for all  $k \geq N(\omega)$ . Therefore for almost sure  $\omega$ ,

$$\lim_{n \rightarrow \infty} \frac{S_n(\omega)}{n} - \lim_{n \rightarrow \infty} \frac{T_n(\omega)}{n} = \lim_{n \rightarrow \infty} \frac{S_n - T_n}{n} = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^{N(\omega)} (X_k(\omega) - Y_k(\omega)) = 0. \quad \square$$

**Lemma 1.4.** For all  $x \geq 0$ ,

$$x \sum_{k>x}^{\infty} \frac{1}{k^2} \leq 2.$$

*Proof.* By comparison with the integral<sup>1</sup>, we have

$$x \sum_{k>x}^{\infty} \frac{1}{k^2} = x \sum_{k=[x]+1}^{\infty} \frac{1}{k^2} \leq \frac{x}{([x]+1)^2} + x \int_{[x]+1}^{\infty} \frac{1}{t^2} dt = \frac{x}{([x]+1)^2} + \frac{x}{[x]+1} \leq 2. \quad \square$$

**Lemma 1.5.**

$$\sum_{k=1}^{\infty} \frac{\mathbb{E}(Y_k^2)}{k^2} \leq 4\mathbb{E}(|X_1|).$$

*Proof.* By Lemma 1.2,

$$\mathbb{E}(Y_k^2) = \int_0^{\infty} \mathbb{P}(Y_k^2 > t) dt = \int_0^{\infty} 2x\mathbb{P}(|Y_k| > x) dx = \int_0^k 2x\mathbb{P}(|Y_k| > x) dx \leq \int_0^k 2x\mathbb{P}(|X_1| > x) dx,$$

therefore

$$\begin{aligned} \sum_{k=1}^{\infty} \frac{\mathbb{E}(Y_k^2)}{k^2} &\leq \sum_{k=1}^{\infty} \frac{1}{k^2} \int_0^{\infty} \mathbb{1}_{\{x < k\}} 2x\mathbb{P}(|X_1| > x) dx \\ &= \int_0^{\infty} 2x\mathbb{P}(|X_1| > x) \sum_{k=1}^{\infty} \frac{1}{k^2} \mathbb{1}_{\{x < k\}} dx \\ &= \int_0^{\infty} 2\mathbb{P}(|X_1| > x) \left( x \sum_{k>x}^{\infty} \frac{1}{k^2} \right) dx \\ &\leq 4 \int_0^{\infty} \mathbb{P}(|X_1| > x) dx \\ &= 4\mathbb{E}(|X_1|). \end{aligned}$$

□

*Proof of Theorem 1.1.* 1. Since  $X_k = X_k^+ - X_k^-$ , (1) holds if

$$\frac{\sum_{k=1}^n X_k^+}{n} \rightarrow \mathbb{E}(X_1^+) \quad \text{and} \quad \frac{\sum_{k=1}^n X_k^-}{n} \rightarrow \mathbb{E}(X_1^-) \quad a.s.$$

hence it is sufficient to prove the case when  $X_k \geq 0$ .

2. Fix  $\alpha > 1$ , let  $k_n = [\alpha^n]$ . Then we will consider the subsequence  $\{T_{k_n} : n \geq 1\}$ . For any  $\varepsilon > 0$ , by Chebyshev's inequality, we have

$$\mathbb{P}(|T_{k_n} - \mathbb{E}(T_{k_n})| > \varepsilon k_n) \leq \frac{1}{\varepsilon^2 k_n^2} \mathbb{E}(|T_{k_n} - \mathbb{E}(T_{k_n})|^2) = \frac{\text{Var}(T_{k_n})}{\varepsilon^2 k_n^2},$$

<sup>1</sup>See [https://en.wikipedia.org/wiki/Integral\\_test\\_for\\_convergence](https://en.wikipedia.org/wiki/Integral_test_for_convergence)

thus

$$\begin{aligned}
\sum_{n=1}^{\infty} \mathbb{P}(|T_{k_n} - \mathbb{E}(T_{k_n})| > \varepsilon k_n) &\leq \frac{1}{\varepsilon^2} \sum_{n=1}^{\infty} \frac{\text{Var}(T_{k_n})}{k_n^2} \\
&= \frac{1}{\varepsilon^2} \sum_{n=1}^{\infty} \frac{1}{k_n^2} \sum_{m=1}^{k_n} \text{Var}(Y_m) \quad (\text{by pairwise independence}) \\
&= \frac{1}{\varepsilon^2} \sum_{n=1}^{\infty} \sum_{m=1}^{\infty} \frac{1}{k_n^2} \mathbb{1}_{\{m \leq k_n\}} \text{Var}(Y_m) \\
&= \frac{1}{\varepsilon^2} \sum_{m=1}^{\infty} \text{Var}(Y_m) \sum_{n=1}^{\infty} \frac{1}{k_n^2} \mathbb{1}_{\{m \leq k_n\}} \\
&= \frac{1}{\varepsilon^2} \sum_{m=1}^{\infty} \text{Var}(Y_m) \sum_{n: k_n \geq m}^{\infty} \frac{1}{k_n^2} \\
&\leq \frac{1}{\varepsilon^2} \sum_{m=1}^{\infty} \mathbb{E}(Y_m^2) \sum_{n: k_n \geq m}^{\infty} \frac{1}{k_n^2}.
\end{aligned}$$

Since  $k_n = \lceil \alpha^n \rceil \geq \alpha^n/2$ , let  $n_0 = \inf\{n \in \mathbb{Z}_+ : \lceil \alpha^n \rceil \geq m\}$ , then  $\alpha^{n_0} \geq \lceil \alpha^{n_0} \rceil \geq m$ , we have

$$\sum_{n: k_n \geq m}^{\infty} \frac{1}{k_n^2} \leq \sum_{n=n_0}^{\infty} \frac{4}{\alpha^{2n}} = \frac{4}{\alpha^{2n_0}} \sum_{n=0}^{\infty} \frac{1}{\alpha^{2n}} \leq \frac{4}{m^2} \cdot \frac{1}{1 - \alpha^{-2}},$$

therefore

$$\sum_{n=1}^{\infty} \mathbb{P}(|T_{k_n} - \mathbb{E}(T_{k_n})| > \varepsilon k_n) \leq \frac{4}{\varepsilon^2(1 - \alpha^{-2})} \sum_{m=1}^{\infty} \frac{\mathbb{E}(Y_m^2)}{m^2} \leq \frac{16}{\varepsilon^2(1 - \alpha^{-2})} \cdot \mathbb{E}(|X_1|) < \infty,$$

then by Borel-Cantelli Lemma,

$$\mathbb{P}\left(\frac{|T_{k_n} - \mathbb{E}(T_{k_n})|}{k_n} > \varepsilon \text{ i.o.}\right) = 0,$$

i.e. almost surely,  $\frac{|T_{k_n} - \mathbb{E}(T_{k_n})|}{k_n} \leq \varepsilon$  for all large enough  $n$ , in other word,

$$\frac{T_{k_n} - \mathbb{E}(T_{k_n})}{k_n} \rightarrow 0, \quad a.s.$$

Since  $X_1 \mathbb{1}_{\{|X_1| \leq k\}} \uparrow X_1$ , by monotone convergence theorem,

$$\mathbb{E}(Y_k) = \mathbb{E}(X_k \mathbb{1}_{\{|X_k| \leq k\}}) = \mathbb{E}(X_1 \mathbb{1}_{\{|X_1| \leq k\}}) \rightarrow \mathbb{E}(X_1) = \mu,$$

thus by Stolz–Cesàro theorem,

$$\lim_{n \rightarrow \infty} \frac{\mathbb{E}(T_{k_n})}{k_n} = \lim_{n \rightarrow \infty} \frac{1}{k_n} \sum_{m=1}^{k_n} \mathbb{E}(Y_m) = \lim_{m \rightarrow \infty} \mathbb{E}(Y_m) = \mu,$$

and hence

$$\frac{T_{k_n}}{k_n} \rightarrow \mu \quad a.s. \quad (3)$$

3. The last step is to consider intermediate terms. For  $k_n \leq m \leq k_{n+1}$ , by Step 1,  $X_k \geq 0$ , so  $Y_k \geq 0$  and hence  $T_{k_n} \leq T_m \leq T_{k_{n+1}}$ , then

$$\frac{T_{k_n}}{k_{n+1}} \leq \frac{T_m}{m} \leq \frac{T_{k_{n+1}}}{k_n},$$

i.e.

$$\frac{T_{k_n}}{k_n} \cdot \frac{k_n}{k_{n+1}} \leq \frac{T_m}{m} \leq \frac{T_{k_{n+1}}}{k_{n+1}} \cdot \frac{k_{n+1}}{k_n}. \quad (4)$$

Since

$$\frac{\alpha^{n+1} - 1}{\alpha^n} \leq \frac{k_{n+1}}{k_n} = \frac{[\alpha^{n+1}]}{[\alpha^n]} \leq \frac{\alpha^{n+1}}{\alpha^n - 1},$$

we have  $k_{n+1}/k_n \rightarrow \alpha$  as  $n \rightarrow \infty$ , then (3) and (4) implies

$$\frac{\mu}{\alpha} \leq \liminf_{m \rightarrow \infty} \frac{T_m}{m} \leq \limsup_{m \rightarrow \infty} \frac{T_m}{m} \leq \alpha\mu, \quad a.s.$$

let  $\alpha \rightarrow 1^+$ , we have

$$\lim_{m \rightarrow \infty} \frac{T_m}{m} = \mu \quad a.s. \quad \square$$