

Martingale Central Limit Theorem. Let $\{l_t^N = \sum_{n=1}^t X_n^N : 1 \leq t \leq t_N\}_{N \geq 1}$ be an array of zero-mean discrete martingales. Then $l_{t_N}^N \rightarrow N(0, \sigma^2)$ in law if

- (i) $[l^N]_{t_N} = \sum_{n=1}^{t_N} (X_n^N)^2 \rightarrow \sigma^2$ in probability, for some constant σ^2 , and
- (ii) $\mathbf{E} \max_{1 \leq n \leq t_N} |X_n^N| \rightarrow 0$.

Proof. We first modify the martingale while preserving (i) and (ii). We then define a product martingale and work on it until we can apply Levy's convergence theorem.

Step one. Without loss of generality suppose that $\sup_N \{\max_{1 \leq n \leq t_N} |X_n^N|\} \leq C$ for some constant $C > 0$. Define stopping time

$$\tau_N = t_N \wedge \min\{1 \leq t \leq t_N : [l^N]_t > \sigma^2 + 1\},$$

then $\mathbf{P}(\tau_N \neq t_N) \rightarrow 0$ because of (i), and hence $[l_{t \wedge \tau_N}]_{t_N} \rightarrow \sigma^2$.

Embed discrete martingale $(l_t^N)_{1 \leq t \leq t_N}$ into a continuous martingale $(\bar{l}_t^N)_{1 \leq t \leq t_N}$ with continuous sample paths [Heath (1977)] and define similarly stopping time

$$\sigma_N = t_N \wedge \min\{1 \leq t \leq t_N : |\bar{l}_t^N - \bar{l}_{[t]}^N| \geq C\}.$$

By Jensen's inequality and note that $\bar{l}_{[\sigma_N]}^N$ is predictable (measurable) at $\bar{l}_{\sigma_N}^N$,

$$\left| \bar{l}_{\sigma_N}^N - \bar{l}_{[\sigma_N]}^N \right| = \left| \mathbf{E} \left(\bar{l}_{\sigma_N}^N - \bar{l}_{[\sigma_N]}^N \mid \bar{l}_{\sigma_N}^N \right) \right| \leq \mathbf{E} \left(\left| \bar{l}_{\sigma_N}^N - \bar{l}_{[\sigma_N]}^N \right| \mid \bar{l}_{\sigma_N}^N \right) \leq \mathbf{E} \left(\left| \bar{l}_{[\sigma_N]}^N - \bar{l}_{[\sigma_N]}^N \right| \mid \bar{l}_{\sigma_N}^N \right).$$

Hence by (ii)

$$\begin{aligned} \mathbf{P}(\sigma_N \neq t_N) &= \mathbf{P} \left(\left| \bar{l}_{\sigma_N}^N - \bar{l}_{[\sigma_N]}^N \right| \geq C \right) \\ &\leq \frac{1}{C} \mathbf{E} \left| \bar{l}_{\sigma_N}^N - \bar{l}_{[\sigma_N]}^N \right| \\ &\leq \frac{1}{C} \mathbf{E} \left| \bar{l}_{[\sigma_N]}^N - \bar{l}_{[\sigma_N]}^N \right| \leq \frac{1}{C} \max_{1 \leq n \leq t_N} |X_n^N| \rightarrow 0, \end{aligned}$$

which implies that $\mathbf{E} \max_{1 \leq n \leq t_N} \left| l_{t \wedge \sigma_N}^N - l_{(t \wedge \sigma_N)-1}^N \right| \rightarrow 0$. Now replace $(l_t^N)_{1 \leq t \leq t_N}$ by the modified martingale $(l_t^{N,C})_{1 \leq t \leq t_N} = (\bar{l}_{t \wedge \sigma_N \wedge \tau_N}^N)_{1 \leq t \leq t_N}$ which satisfies in addition to (i) and (ii),

(iii) $[l^{N,C}]_{t_N} \leq \sigma^2 + 1 + C^2$ and

(iv) $|X_n^{N,C}| \leq C$

for any N and n , where $X_n^{N,C} = l_n^{N,C} - l_{n-1}^{N,C}$.

Step two. Construct product martingale $m_t^{N,C}(\theta) = \prod_{n=1}^t (1 + \theta X_n^{N,C})$, then for $|\theta| < C^{-1}$ expand complex logarithm to get $\log m_{t_N}^{N,C}(\theta) = \sum_{p \geq 1} \frac{1}{p} \theta^p [l^{N,C}]_{t_N}^p$ where $[l^{N,C}]_t^p = \sum_{n=1}^t (X_n^{N,C})^p$ is the p -variation. Since $\max_{1 \leq n \leq t_N} |X_n^{N,C}| \rightarrow 0$ as $N \rightarrow \infty$,

$$\begin{aligned} \left| \log m_{t_N}^{N,C}(\theta) - \left(\theta l_{t_N}^{N,C} - \frac{1}{2} \theta^2 [l^{N,C}]_{t_N} \right) \right| &\leq \sum_{p \geq 3} \frac{1}{p} \theta^p [l^{N,C}]_{t_N}^p \\ &\leq [l^{N,C}]_{t_N} \sum_{p \geq 3} \frac{1}{p} \theta^p \left(\max_{1 \leq n \leq t_N} |X_n^{N,C}| \right)^{p-2} \rightarrow 0 \end{aligned}$$

in probability. Observe that $l_{t_N}^{N,C}$ is bounded in L^2 as $\mathbf{E}(l_{t_N}^{N,C})^2 = \mathbf{E}[l^{N,C}]_{t_N} \leq \sigma^2 + 1 + C^2$, hence $l_{t_N}^{N,C}$ is uniformly integrable and thus tight. But $[t^{N,C}]_{t_N}$ is also tight, hence $\theta l_{t_N}^{N,C} - \theta^2[l^{N,C}]_{t_N}/2$ is tight, and

$$m_{t_N}^{N,C}(\theta) = (1 + o_p(1)) \exp\left\{\theta l_{t_N}^{N,C} - \frac{1}{2}\theta^2\sigma^2\right\}. \quad (*)$$

Let $\theta = ih$ where $h \in \mathbf{R}$ and $|\theta| = |h| < C^{-1}$. Similarly $m_{t_N}^{N,C}$ is bounded in L^2 as

$$\mathbf{E}\left|m_{t_N}^{N,C}\right|^2 = \mathbf{E} \prod_{n=1}^{t_N} \left(1 + h^2(X_{t_N}^{N,C})^2\right) \leq \mathbf{E} \exp\{h^2[l^{N,C}]_{t_N}\} \leq \mathbf{E} \exp\{h^2(\sigma^2 + 1 + C^2)\}$$

hence $m_{t_N}^{N,C}$ is uniformly integrable, and from (*),

$$\mathbf{E} \exp\left\{\theta l_{t_N}^{N,C} - \frac{1}{2}\theta^2\sigma^2\right\} \rightarrow \mathbf{E} m_{t_N}^{N,C} = 1.$$

Hence $\mathbf{E} \exp\{ihl_{t_N}^{N,C}\} \rightarrow \exp\{-h^2\sigma^2/2\}$. But $l_{t_N}^{N,C} \neq l_{t_N}^N$ only when $t_N \neq \tau_N$ or $t_N \neq \sigma_N$, hence

$$\left|\mathbf{E} \exp\{ihl_{t_N}^{N,C}\} - \mathbf{E} \exp\{ihl_{t_N}^N\}\right| = \left|\mathbf{E} \exp\{ihl_{t_N}^{N,C}\} - \mathbf{E} \exp\{ihl_{t_N}^N\}\right| \mathbf{1}_{\{t_N \neq \tau_N \text{ or } t_N \neq \sigma_N\}} \rightarrow 0.$$

Hence $\mathbf{E} \exp\{ihl_{t_N}^N\} \rightarrow \exp\{-h^2\sigma^2/2\}$, and by Levy's convergence theorem $l_{t_N}^N \rightarrow N(0, \sigma^2)$ in law as desired. □

References

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