

# Necessary and Sufficient Conditions for the Self-normalized Central Limit Theorem

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# 1. Introduction

## ► 1.1 Classical Central Limit Theorem

Let  $\{X_i, 1 \leq i \leq n\}$  be a sequence of independent random variables.

Put

$$S_n = \sum_{i=1}^n X_i.$$

What is the necessary and sufficient condition for the central limit theorem (CLT)?

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What is the necessary and sufficient condition for the central limit theorem (CLT)?

Under what conditions, are there  $a_n$  and  $c_n$  such that

$$\frac{S_n - c_n}{a_n} \xrightarrow{d.} N(0, 1) ?$$

- ▶  $\{X_i\}$  are independent and identically distributed (i.i.d.) random variables with  $E(X_1) = 0$

The following statements are equivalent

- CLT holds
- $EX_1^2 I\{|X_1| \leq x\}$  is a slowly varying function, i.e.,  $X_1$  is in the domain of attraction of the normal distribution, denoted by  $X_1 \in DAN$ .

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$$\frac{f(tx)}{f(x)} \rightarrow 1 \text{ as } x \rightarrow \infty.$$

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- $\max_{1 \leq i \leq n} |X_i|/V_n \rightarrow 0$  in probability, where  $V_n^2 = \sum_{i=1}^n X_i^2$

## ► Independent Random Variables

Assume that

$$\forall \varepsilon > 0, \max_{1 \leq i \leq n} P(|X_i| \geq \varepsilon a_n) \rightarrow 0.$$

Then

$$S_n/a_n \xrightarrow{d.} N(0, 1)$$

if and only if

- (i)  $\sum_{i=1}^n P(|X_i| \geq \varepsilon a_n) \rightarrow 0$  for any  $\varepsilon > 0$ ;
- (ii)  $\frac{1}{a_n} \sum_{i=1}^n EX_i I_{\{|X_i| \leq a_n\}} \rightarrow 0$ ;
- (iii)  $\frac{1}{a_n^2} \sum_{i=1}^n (EX_i^2 I_{\{|X_i| \leq a_n\}} - (EX_i I_{\{|X_i| \leq a_n\}})^2) \rightarrow 1$

## ► 1.2. Self-normalized central limit theorem for i.i.d. random variables

Let  $\{X_i, 1 \leq i \leq n\}$  be independent and identically distributed random variables. Put

$$S_n = \sum_{i=1}^n X_i \quad \text{and} \quad V_n^2 = \sum_{i=1}^n X_i^2.$$

- Self-normalized sum:  $S_n/V_n$

Under what conditions, does the CLT hold for the self-normalized sum?



Assume  $E(X_1) = 0$ .

Logan, Mallows, Rice and Shepp (1973)'s conjecture:

$S_n/V_n \xrightarrow{d.} N(0, 1)$  if and only if  $X_1 \in DAN$ .

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- Maller (1981):

If  $X_1 \in DAN$ , then  $S_n/V_n \xrightarrow{d.} N(0, 1)$

- Griffin and Mason (1991):

Assume that  $X_1$  is **symmetric**. Then CLT holds for  $S_n/V_n$  if and only if  $X_1 \in DAN$

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- Gine, Götze and Mason (1997):

CLT holds for  $S_n/V_n$  if and only if  $X_1 \in DAN$

## 2. Toward to a General Self-normalized Central Limit Theorem

Let  $\{X_i, 1 \leq i \leq n\}$  be independent random variables with  $E(X_i) = 0$ .  
Set

$$S_n = \sum_{i=1}^n X_i \quad \text{and} \quad V_n^2 = \sum_{i=1}^n X_i^2.$$

- **Aim:** Find the necessary and sufficient condition for the self-normalized CLT, i.e.,

$$S_n/V_n \xrightarrow{d.} N(0, 1)$$

## Recall the classical Lindeberg CLT:

If the Lindeberg condition is satisfied

$$\forall \varepsilon > 0, \sum_{i=1}^n E(X_i^2/B_n^2)I_{\{|X_i|>\varepsilon B_n\}} \rightarrow 0,$$

where  $B_n^2 = \sum_{i=1}^n EX_i^2$ , then

$$S_n/B_n \xrightarrow{d.} N(0, 1)$$

Also recall the self-normalized CLT for i.i.d. random variables:

$$S_n/V_n \xrightarrow{d.} N(0, 1)$$

if and only if  $\max_{1 \leq i \leq n} |X_i|/V_n \rightarrow 0$  in probability, which is equivalent to

$$\forall \varepsilon > 0, \sum_{i=1}^n E(X_i^2/V_n^2)I_{\{|X_i|>\varepsilon V_n\}} \rightarrow 0,$$

which looks like a Lindeberg condition.

- Egorov (1996):

Let  $X_i, i \geq 1$  be independent and **symmetric** random variables.  
Then

$$S_n/V_n \xrightarrow{d.} N(0, 1)$$

if and only if  $\max_{1 \leq i \leq n} |X_i|/V_n \rightarrow 0$  in probability .

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- **Question:** Is  $\max_{1 \leq i \leq n} |X_i|/V_n \rightarrow 0$  in probability a sufficient condition in general?

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- **Question:** Is  $\max_{1 \leq i \leq n} |X_i|/V_n \rightarrow 0$  in probability a sufficient condition in general?

**Answer:** No.



### Example 1:

Let  $X_i, i \geq 1$  be independent random variables satisfying

$$\begin{aligned}P(X_{2i-1} = -i^2) &= \frac{1}{1+i^2}, & P(X_{2i-1} = 1) &= \frac{i^2}{1+i^2} \\P(X_{2i} = i^2) &= \frac{1}{1+i^2}, & P(X_{2i} = -1) &= \frac{i^2}{1+i^2}\end{aligned}$$

for  $i = 1, 2, \dots$ . Then  $E(X_i) = 0$  and  $\max_{1 \leq i \leq n} |X_i|/V_n \rightarrow 0$  in probability, but  $S_n/V_n \rightarrow 0$  in probability.

## Theorem (1)

Let  $\{X_n, n \geq 1\}$  be independent non-degenerate random variables.  
Then

$$S_n/V_n \xrightarrow{d.} N(0, 1)$$

if the following conditions are satisfied.

- (i)  $\max_{1 \leq i \leq n} |X_i|/V_n \rightarrow 0$  in probability;
- (ii)  $\sum_{i=1}^n \{E(X_i/V_n)\}^2 \rightarrow 0$ ;
- (iii)  $E\left(\frac{S_n}{\max(V_n, a_n)}\right) \rightarrow 0$ , where  $a_n$  satisfies

$$\sum_{i=1}^n E\left(\frac{X_i^2}{a_n^2 + X_i^2}\right) = 1. \quad (1)$$

- **Remark:** Theorem remains valid if (iii) is replaced by

$$(iii)^* \quad E(S_n/V_n) \rightarrow 0.$$

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However,  $(iii)^*$  is not a necessary condition.

- **Example 2.**

Let  $Y_i := Y_{n,i}$ ,  $1 \leq i \leq n$  be i.i.d. with pdf

$$P(Y_i = 1/n) = 1 - \frac{\ln n}{4n}, \quad P(Y_i = 1) = P(Y_i = -1) = \frac{\ln n}{8n}.$$

Define  $X_i = Y_i - E(Y_i)$ . Then we have

$$S_n/V_n \xrightarrow{d.} N(0, 1)$$

but

$$E(S_n/V_n) \rightarrow \infty$$

## Theorem (2)

*If (i) is satisfied, then (ii) and (iii) are necessary for the self-normalized CLT.*

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*If (i) is satisfied, then (ii) and (iii) are necessary for the self-normalized CLT.*

**Conjecture:** Conditions (i), (ii) and (iii) are necessary and sufficient for the self-normalized CLT.

A key step of the proof is the following results:

### Proposition (1)

*Let  $\{X_n, n \geq 1\}$  be independent non-degenerate random variables satisfying*

$$\max_{i \leq n} |X_i|/V_n \rightarrow 0 \text{ in probability}$$

*Let  $a_n > 0$  satisfy*

$$\sum_{i=1}^n E \left( \frac{X_i^2}{a_n^2 + X_i^2} \right) = 1$$

*Then*

$$V_n/a_n \rightarrow 1 \text{ in probability}$$

## Proposition (2)

Assume that

$$\max_{1 \leq i \leq n} |X_i|/V_n \rightarrow 0 \text{ in probability}$$

and let  $a_n > 0$  satisfy  $\sum_{i=1}^n E\left(\frac{X_i^2}{a_n^2 + X_i^2}\right) = 1$ . Then

$\sum_{i=1}^n \{E(X_i/V_n)\}^2 \rightarrow 0$  is equivalent to

$$\frac{1}{a_n^2} \sum_{i=1}^n (EX_i I_{\{|X_i| \leq a_n\}})^2 \rightarrow 0$$

and  $E\left(\frac{S_n}{\max(V_n, a_n)}\right) \rightarrow 0$  is equivalent to

$$\frac{1}{a_n} \sum_{i=1}^n EX_i I_{\{|X_i| \leq a_n\}} \rightarrow 0.$$



