

On the Another proof of Liapounov's theorem in the Central Limit Theorem

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中心極限定理에 있어서 Liapounov 定理의 別證에 대하여

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Summary

In this paper, we shall study the another proof that the normalized sum converges in distribution to a random variable that is normal with mean and variance 1 by comparing the expectation of test function,

of proofs of the theorem under such restrictions.

1. Introduction

Let X_1, X_2, \dots be independent random variables (r.v.'s) with each X_j having finite mean μ_j and finite variance δ_j^2 . Let $S_n = \sum_{j=1}^n X_j$, $n=1, 2, \dots$; then $E(S_n) = \sum_{j=1}^n \mu_j$, $\text{Var } S_n = s_n^2 = \sum_{j=1}^n \delta_j^2$.

We consider normalized sum $S_n^* = s_n^{-1}(S_n - E(S_n))$ which has mean 0 and variance 1 assuming that $s_n > 0$ for sufficiently large n . If X^* is a r.v. having the normal distribution with mean 0 and variance 1, $N(0, 1)$, so that the distribution function (df) of X^* is

$$\Phi(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{1}{2}t^2} dt. \quad \dots\dots\dots(1)$$

Let F_j be the df of X_j , f_j be the Characteristic function (Ch.f) of X_j , F_n^* be the df of S_n^* and $g_n(t)$ be the Ch.f of r.v. S_n^* . We shall investigate the conditions, especially Liapounov's, under which S_n^* converges in distribution to X^* and the method

2. Converges to a Normal Distribution

I. Let F be a df of r.v. X and $E(X) = \mu$, if $\int_{-\infty}^{\infty} |X|^n dF(x)$ exists and is finite, then the Ch.f $f(t)$ of F may be written as

$$f(t) = 1 + \sum_{j=1}^{n-1} \mu_j' \frac{(it)^j}{j!} + R_n(t), \quad \dots\dots\dots(2)$$

where

$$R_n(t) = t^n \int_0^1 \frac{(1-u)^{n-1}}{(n-1)!} f^{(n)}(tu) du \quad \dots\dots\dots(3)$$

and

$$f^{(j)}(t) = \int_{-\infty}^{\infty} (ix)^j e^{itx} dF(x),$$

$$f^{(j)}(0) = i^j \int_{-\infty}^{\infty} x^j dF(x) = i^j \mu_j$$

Also

$$R_n(t) = \frac{(it)^n}{n!} \int_{-\infty}^{\infty} x^n dF(x) + o(t^n), \dots\dots\dots (4)$$

where $o(t^n)$ denotes a function of t such that

$$\frac{o(t^n)}{t^n} \rightarrow 0 \text{ as } t \rightarrow \infty.$$

Furthermore

$$|R_n(t)| = \theta \frac{|t|^n}{n!} \int_{-\infty}^{\infty} |x|^n dF(x), \quad |\theta| \leq 1 \dots (5)$$

for every real t .

Pf. The exponential function satisfies

$$e^v = 1 + \sum_{j=1}^{n-1} \frac{v^j}{j!} + r_n(v),$$

where

$$r_n(v) = \frac{v^n}{(n-1)!} \int_0^1 (1-u)^{n-1} \frac{d^n}{du^n} (e^{uv}) du,$$

or

$$r_n(v) = \frac{v^n}{n!} + o(v^n).$$

If we place v by it x and take expected values, we obtain (3) and (4). Also (5) is obtained by applying) the Maclaurin expansion theorem of calculus to $f(t)$.

II. Liapounov's inequality

Let X and Y be $r, v, s, 0 < p < \infty$ and $p^{-1} + q^{-1} = 1$, Hölder's inequality

$$E(|XY|) \leq \{E(|X|^p)\}^{1/p} \cdot \{E(|Y|^q)\}^{1/q} \dots\dots\dots (6)$$

If $Y \equiv 1$ in (6) we obtain

$$E(|X|) \leq \{E(|X|^p)\}^{1/p}$$

Replacing $|X|$ by $|X|^r$, where $1 < r < p$ and writing $s = pr$ we obtain the Liapounov's inequality;

$$E(|X|^r)^{1/r} \leq E(|X|^p)^{1/p}, \quad 1 < r < s < \infty. \dots\dots\dots (7)$$

Theorem. Let $S_n = \sum_{j=1}^n X_j, n=1, 2, \dots$, where the X_j are independent r, v, s with μ_j and σ_j^2 , they are finite. Let $S_n^* = s_n^{-1}(S_n - E(S_n))$, where $s_n^2 = \sum_{j=1}^n \sigma_j^2$ and let F_j be the df of X_j . If for every $\epsilon > 0$,

$$\frac{1}{s_n^2} \sum_{j=1}^n \int_{\{x : |x - \mu_j| \geq \epsilon s_n\}} (x - \mu_j)^2 dF_j(x) \rightarrow 0$$

as $n \rightarrow \infty. \dots\dots\dots (8)$

then S_n^* converges in distribution to X^* . This theorem implies that $S^* \xrightarrow{d} X^*$ under any one of the following conditions.

1) The uniformly bounded case.

Assume $|X_j| \leq M$ for all j , and $s_n \rightarrow \infty$.

Then

$$\begin{aligned} & \int_{\{x : |x - \mu_j| \geq \epsilon s_n\}} (x - \mu_j)^2 dF_j(x) \\ &= E[(X_j - \mu_j)^2 I\{|X_j - \mu_j| \geq \epsilon s_n\}] \\ &\leq \frac{(2M)^2 \sigma_j^2}{\epsilon^2 s_n^2} \dots\dots\dots (6) \end{aligned}$$

by Chebyshev's inequality.

Thus

$$\frac{1}{s_n^2} \sum_{j=1}^n \int_{\{x : |x - \mu_j| \geq \epsilon s_n\}} (x - \mu_j)^2 dF_j(x) \leq \frac{(2M)^2}{\epsilon^2 s_n^2} \rightarrow 0$$

2) The identically distributed case.

Assume that the X_j are independent identically distributed r, v, s with finite μ and finite $\sigma^2 > 0$, then

$$(8) = \frac{1}{n\sigma^2} \sum_{j=1}^n \int_{\{x : |x - \mu| \geq \epsilon \sqrt{n}\}} (x - \mu)^2 dF(x)$$

$$= \frac{1}{\epsilon^2} \int_{\{x: |x-\mu| \geq \epsilon \sqrt{n}\}} (x-\mu)^2 dF(x) \rightarrow 0, \dots\dots\dots(10) \quad \geq \epsilon^2 s_n^2 \int_{\{x: |x-\mu_j| \geq \epsilon s_n\}} (x-\mu_j)^2 dF_j(x)$$

since σ^2 is finite and $\{x: |x-\mu| \geq \epsilon \delta \sqrt{n}\} \downarrow \phi$ as $n \rightarrow \infty$. Thus

3) The Bernoulli case.

Let S_n be the number of success in n Bernoulli trials, with probability of success p on a given trial. We may write

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

where the X_j are independent and $P(X_j=1)=p$, $P(X_j=0)=q=1-p$. We may take X_j as the indicator of a success on trial j , thus case 2 applies with $\mu = E(X_j) = p$, $\sigma^2 = E(X_j^2) - [E(X_j)]^2 = p(1-p)$, $E(S_n) = n\mu = np$, $s_n^2 = n\sigma^2 = np(1-p)$. Thus

$$S_n^* = \frac{S_n - np}{(npq)^{1/2}} \dots\dots\dots(11)$$

and

$$S_n^* \xrightarrow{d} X^*, \text{ that is, } P(S_n^* \leq x) \rightarrow \Phi(x) \text{ for all } x.$$

4) Liapounov's condition.

Assume that

$$\frac{1}{s_n^{2+\delta}} \sum_{j=1}^n E[|X_j - \mu_j|^{2+\delta}] \rightarrow 0 \quad \dots\dots\dots(12)$$

for some $\delta > 0$.

where $E[|X_j - \mu_j|^{2+\delta}]$ exist. Then

$$E[|X_j - \mu_j|^{2+\delta}] = \int_{-\infty}^{\infty} |x - \mu_j|^{2+\delta} dF_j(x) \geq \int_{\{x: |x - \mu_j| \geq \epsilon s_n\}} |x - \mu_j|^{2+\delta} dF_j(x)$$

$$(8) \leq \frac{1}{s_n^2} \sum_{j=1}^n \frac{E[|X_j - \mu_j|^{2+\delta}]}{\epsilon^\delta s_n^\delta} = \frac{\sum_{j=1}^n E[|X_j - \mu_j|^{2+\delta}]}{\epsilon^\delta s_n^{2+\delta}} \rightarrow 0.$$

Now we shall show that Theorem holds under case

4). Let $E(|X_j|^3) = \tau_j$ and $\Gamma_n = \sum_{j=1}^n \tau_j = \sum_{j=1}^n E(|X_j|^3)$.

To prove Theorem in (12) we need only show that

$$g_n(t) \rightarrow e^{-\frac{t^2}{2}} \text{ or } \log g_n(t) \rightarrow -\frac{t^2}{2} \text{ as } n \rightarrow \infty.$$

Furthermore we may assume without loss of generality that all $\mu_j = 0$ and that the $r.v$'s X_j all have finite third moments, that is, $\delta = 1$.

Proof 1.

The condition (12) in case 4) is written as

$$\Gamma_n / s_n^3 \rightarrow 0. \quad \dots\dots\dots(13)$$

By I, the Ch.f $f_j(t)$ of X_j has the expansion :

$$f_j(t) = 1 - \frac{1}{2} \sigma_j^2 t^2 + \frac{1}{6} E(X_j^3) (it)^3 + o(E(X_j^3)t^3) \dots\dots\dots(14)$$

The Ch.f $g(t)$ of the $r.v.S$ is given by

$$g_n(t) = \prod_{j=1}^n f_j\left(\frac{t}{s_n}\right) = \prod_{j=1}^n \left\{ 1 - \frac{\sigma_j^2}{2s_n^2} t^2 + \frac{E(X_j^3)}{6s_n^3} (it)^3 + o\left[\frac{E(X_j^3)t^3}{s_n^3}\right] \right\}. \quad \dots\dots\dots(15)$$

Writing

$$\eta_j = -\frac{\sigma_j^2}{2s_n^2} t^2 + \frac{E(X_j^3)}{6s_n^3} (it)^3 + o\left[\frac{E(X_j^3)t^3}{s_n^3}\right]$$

we note that :

$$\sum_{j=1}^n \eta_j = -\frac{t^2}{2} + \frac{1}{6s_n^3} (it)^3 \sum_{j=1}^n E(X_j^3) + \sum_{j=1}^n 0 \left[\frac{E(X_j^3)t^3}{s_n^3} \right]$$

$$\begin{aligned} \log(1+\eta_j) &= \sum_{r=1}^{\infty} (-1)^{r-1} \cdot \frac{1}{r} \eta_j^r \\ &= \eta_j + \eta_j^2 \sum_{r=2}^{\infty} (-1)^{r+1} \frac{1}{r} \eta_j^{r-2} = \eta_j + \eta_j^2 \phi(\eta_j), \end{aligned} \dots\dots\dots(16)$$

step 1 : $\eta_j \rightarrow 0$ for all j , uniformly in j as $n \rightarrow \infty$.
Clearly the 0-term $\rightarrow 0$ as $n \rightarrow \infty$.
Recalling(7)

$$\{E(|X_j|^3)\}^{1/3} = \sigma_j \leq \{E(|X_j|^3)\}^{1/3},$$

we obtain

$$0 \leq \frac{\sigma_j^2}{2s_n^2} t^2 \leq \frac{1}{2} [\gamma_j]^{2/3} \frac{t^2}{s_n^2} \leq \frac{1}{2} [\Gamma_n]^{2/3} \cdot \frac{t^2}{s_n^2}$$

The first and third inequalities are obvious and, by (13), the upper bound on the right hand side tends to zero as $n \rightarrow \infty$, i, e ;

$$\frac{1}{2} \cdot \frac{\sigma_j^2}{s_n^2} \cdot t^2 \rightarrow 0, \text{ uniformly in } j \text{ as } n \rightarrow \infty,$$

Finally

$$\begin{aligned} 0 \leq \left| \frac{1}{6} \cdot \frac{E(X_j^3)}{s_n^3} (it)^3 \right| &\leq \frac{1}{6} \cdot \frac{\gamma_j |t|^3}{s_n^3} \\ &\leq \frac{1}{6} \frac{|t|^3}{s_n^3} \Gamma_n. \end{aligned}$$

By (13) the quantity on the right tends to zero as $n \rightarrow \infty$, i, e ;

$$\frac{1}{6} \cdot \frac{E(X_j^3)}{s_n^3} (it)^3 \rightarrow 0,$$

uniformly in j as $n \rightarrow \infty$.

step 2 : Given $\epsilon > 0$, $0 < \epsilon < \frac{1}{2}$, we can find $N(t, \epsilon)$

such that for $n > N(t, \epsilon)$, $|\eta_j| < \epsilon$ for all $j \leq n$.

Using the logarithmic expansion we obtain

where

$$\phi(\eta_j) = \sum_{r=2}^{\infty} (-1)^{r+1} \frac{1}{r} \eta_j^{r-2}$$

However

$$\begin{aligned} |\phi(\eta_j)| &\leq \sum_{r=2}^{\infty} \frac{1}{2+r} |\eta_j|^{r-2} \leq \frac{1}{2} [1-|\eta_j|]^{-1} \\ &< \frac{1}{2} \cdot \frac{1}{1-\epsilon} < 1, \end{aligned} \dots\dots\dots(17)$$

since $(2+r)^{-1} < \frac{1}{2}$ for all $r \geq 1$ and since η_j tends to zero uniformly in j . The logarithm of the Ch.f $g_n(t)$ of S_n^* may be written as follows;

$$\begin{aligned} \log g_n(t) &= \sum_{j=1}^n \log(1+\eta_j) = \sum_{j=1}^n \eta_j + \sum_{j=1}^n \eta_j^2 \phi(\eta_j) \\ &= -\frac{t^2}{2} + \frac{1}{6s_n^3} (it)^3 \sum_{j=1}^n E(X_j^3) + \sum_{j=1}^n 0 \left[\frac{E(X_j^3)t^3}{s_n^3} \right] \\ &\quad + \sum_{j=1}^n \eta_j^2 \phi(\eta_j) \end{aligned} \dots\dots\dots(18)$$

step 3 : $\sum_{j=1}^n \eta_j^2 \phi(\eta_j) \rightarrow 0$ as $n \rightarrow \infty$

By (17), $\left| \sum_{j=1}^n \eta_j^2 \phi(\eta_j) \right| \leq \sum_{j=1}^n \eta_j^2 \left| \phi(\eta_j) \right| \leq \sum_{j=1}^n \eta_j^2,$

Now, let show $\sum_{j=1}^n \eta_j^2 \rightarrow 0$ as $n \rightarrow \infty$.

$$\left| \sum_{j=1}^n \eta_j^2 \right| \leq \textcircled{1} \frac{|t|^4}{4s_n^4} \sum_{j=1}^n \sigma_j^4 + \textcircled{2} \frac{|t|^6}{36s_n^6} \sum_{j=1}^n [\gamma_j]^2$$

$$\begin{aligned}
 & + \textcircled{3} \sum_{j=1}^n \left[0 \left(\frac{t^3}{s_n^3} E(X_j^3) \right) \right]^2 + \textcircled{4} \frac{t^5}{6s_n^3} \sum_{j=1}^n \sigma_j^2 \gamma_j \\
 & + \textcircled{5} \frac{t^2}{s_n^2} \sum_{j=1}^n \sigma_j^2 \left| 0 \left(\frac{E(X_j^3)t^3}{s_n^3} \right) \right| \\
 & + \textcircled{6} \frac{t^3}{3s_n^3} \sum_{j=1}^n \gamma_j \left| 0 \left(\frac{E(X_j^3)t^3}{s_n^3} \right) \right|
 \end{aligned}$$

Appealing twice to (7) we have

$$\frac{\sum_{j=1}^n \sigma_j^4}{s_n^4} \leq \frac{\sum_{j=1}^n [\gamma_j]^{4/3}}{s_n^4} \leq \frac{\Gamma_n}{s_n^3}$$

and the upper bound tends to zero by (13), this takes care of ①.

Next, by expanding the square on the right that

$$\frac{\sum_{j=1}^n [\gamma_j]^2}{s_n^6} \leq \left(\frac{\Gamma_n}{s_n^3} \right)^2$$

add upper bound of ② tends to zero.

Also ③ tends to zero because of the $o(s)$, so do ⑤ and ⑥. Finally ④ satisfies

$$|t|^5 \frac{\sum_{j=1}^n \sigma_j^2 \gamma_j}{s_n^5} \leq \frac{\Gamma_n}{s_n} \cdot |t|^5$$

since $\sigma_j^2 \leq s_n^2$, $j=1, 2, \dots, n$.

The upper bounds on the right go to zero as $n \rightarrow \infty$ by (13), hence completes the proof of step 3.

step 4: $\sum_{j=1}^n \gamma_j^2 + \frac{t^2}{2} \rightarrow 0$ as $n \rightarrow \infty$

We have;

$$\left| \sum_{j=1}^n \gamma_j^2 + \frac{t^2}{2} \right| \leq \frac{|t|^3}{6s_n^3} \Gamma_n + \sum_{j=1}^n 0 \left[\frac{E(X_j^3)t^3}{s_n^3} \right]$$

and both on the right tend to zero as $n \rightarrow \infty$. The first by (13) and the second because of the

small o .

We conclude that

$$\log g_n(t) \rightarrow -\frac{t^2}{2}$$

for every t and so complete the proof of Theorem.

Let proceed to another proof of Theorem by the idea that is to approximate the sum S_n successively by replacing one X at a time with a normal r, v, Y .

Proof 2.

Let $\{Y_j; j \geq 1\}$ be r, v 's having $N(0, \sigma_j^2)$, thus Y_j has the same mean and variance as the corresponding X_j ; let all the X 's and Y 's be totally independent.

Now put

$$Z_j = Y_1 + Y_2 + \dots + Y_{j-1} + X_j + \dots + X_n, \quad 1 \leq j \leq n,$$

with the convention that

$$Z_1 = X_2 + \dots + X_n, \quad Z_n = Y_1 + \dots + Y_{n-1}.$$

We now write

$$\begin{aligned}
 & f(X_1 + X_2 + \dots + X_n) - f(Y_1 + \dots + Y_n) \\
 & = \sum_{j=1}^n \left\{ f(X_j + Z_j) - f(Y_j + Z_j) \right\}.
 \end{aligned}$$

Let estimate the difference for a suitable class of function f to compare the distribution of $(X_j + Z_j)/s_n$ with that of $(Y_j + Z_j)/s_n$. i. e ;

$$\begin{aligned}
 & E \left\{ f \left(\frac{X_1 + \dots + X_n}{s_n} \right) \right\} - E \left\{ f \left(\frac{Y_1 + \dots + Y_n}{s_n} \right) \right\} \\
 & = \sum_{j=1}^n \left[E \left\{ f \left(\frac{X_j + Z_j}{s_n} \right) \right\} - E \left\{ f \left(\frac{Y_j + Z_j}{s_n} \right) \right\} \right]. \quad (19)
 \end{aligned}$$

On the other hand, if we take f in C^3 , the class of bounded continuous function with three bounded continuous derivatives, it suffices to show that

$$E\{f(S_n^*)\} \rightarrow E\{f(X^*)\}.$$

Now by Taylor's *Th*, we have for every x and y ;

$$\left| f(x+y) - \left[f(x) + f'(x)y + \frac{f''(x)}{2}y^2 \right] \right| \leq \frac{M|y|^3}{6}$$

where $M = \sup_{x \in R^1} |f^{(3)}(x)|$.

Hence if ξ and η are independent $r.v$'s such that $E\{|\eta|^3\} < \infty$, by substitution and integration,

$$\begin{aligned} |E\{f(\xi+\eta)\} - E\{f(\xi)\} - E\{f'(\xi)\}E\{\eta\} \\ - \frac{1}{2}E\{f''(\xi)E\{\eta^2\}\}| \leq \frac{M}{6}E\{|\eta|^3\} \end{aligned} \quad \dots\dots\dots(20)$$

since the $r.v$'s $f(\xi)$, $f'(\xi)$ and $f''(\xi)$ are bounded hence integrable.

If ζ is another $r.v$. independent of ξ and having the same mean and variance as η , and $E\{|\zeta|^3\} < \infty$, we obtain by replacing η with ζ and taking the difference;

$$|E\{f(\xi+\eta)\} - E\{f(\xi+\zeta)\}| \leq \frac{M}{6}E\{|\eta|^3 + |\zeta|^3\}. \quad \dots\dots\dots(21)$$

Applying to the right side of (19) with

$\zeta = \frac{Z_j}{s_n}$, $\eta = \frac{X_j}{s_n}$, $\zeta = \frac{X_j}{s_n}$ and the bounds on the right-hand side of (21) then add up to

$$\frac{M}{6} \sum_{j=1}^n \left\{ \frac{\gamma_j}{s_n^3} + \frac{c\sigma_j^3}{s_n^3} \right\} \quad \dots\dots\dots(22)$$

where $c = \sqrt{8/\pi}$ since the absolute third moment of $N(0, \sigma^2)$ is equal to $c\sigma^3$. By (7), $\sigma_j^3 \leq \gamma_j$, so that the quantity in (22) is $o(\Gamma_n/s_n^3)$. We have thus obtained the following estimate;

$$\forall f \in C^3; |E\{f(S_n^*)\} - E\{f(X^*)\}| \leq o\left(\frac{\Gamma_n}{s_n^3}\right)$$

and under (13) this converges to zero as $n \rightarrow \infty$.

Hence

$$E\{f(S_n^*)\} \rightarrow E\{f(X^*)\} \text{ as } n \rightarrow \infty.$$

Example :

Let $\epsilon > 0$ and $f_\epsilon \in C^3$

$$f_\epsilon(x) = \begin{cases} 1 & \text{if } x \leq 0 \\ [1 - (x\epsilon^{-1})^4]^4 & \text{if } 0 \leq x \leq \epsilon \\ 0 & \text{if } x \geq \epsilon \end{cases}$$

Then

$$\Phi(-x+\epsilon) \geq \int_{-\infty}^{-x+\epsilon} f_\epsilon(x+y) d\Phi(y) \geq \Phi(-x), \quad x \in R.$$

Similarly

$$F_{n^*}(-x+\epsilon) \geq \int_{-\infty}^{-x+\epsilon} f_\epsilon(x+y) dF_{n^*}(y) \geq F_{n^*}(-x), \quad x \in R,$$

$$\text{since } \lim_{n \rightarrow \infty} \int_{-\infty}^{-x+\epsilon} f(x+y) dF_{n^*}(y) = \int_{-\infty}^{-x+\epsilon} f(x+y) d\Phi(y)$$

for any x (uniformly in x), $f \in C^3$ (Ref [6])

We conclude that

$$\begin{aligned} \overline{\lim}_{n \rightarrow \infty} F_{n^*}(-x) &\geq \lim_{n \rightarrow \infty} \int_{-\infty}^{-x+\epsilon} f_\epsilon(x+y) dF_{n^*}(y) \\ &= \int_{-\infty}^{-x+\epsilon} f(x+y) d\Phi(y) \leq \Phi(-x+\epsilon) \end{aligned}$$

and

$$\begin{aligned} \underline{\lim}_{n \rightarrow \infty} F_{n^*}(-x+\epsilon) &\geq \lim_{n \rightarrow \infty} \int_{-\infty}^{-x+\epsilon} f_\epsilon(x+y) dF_{n^*}(y) \\ &= \int_{-\infty}^{-x+\epsilon} f_\epsilon(x+y) d\Phi(y) \geq \Phi(-x) \end{aligned} \quad \dots\dots\dots(23)$$

for all real x and $\epsilon > 0$.

By (23),

$$\lim_{n \rightarrow \infty} F_{n^*}(-x) \geq \Phi(-x-\epsilon),$$

and we get

$$\Phi(-x-\epsilon) \leq \liminf_{n \rightarrow \infty} F_n^*(-x) \leq \overline{\lim}_{n \rightarrow \infty} F_n^*(-x) \\ \leq \Phi(-x+\epsilon)$$

for all $x \in R$ and $\epsilon > 0$. Since ϵ is arbitrary,

$$\lim_{n \rightarrow \infty} F_n^*(-x) = \Phi(-x), \quad x \in R.$$

Literature Cited

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〈國文抄錄〉

中心極限定理에 있어서 Liapounov 定理의 別證에 對하여

本論文에서는 Liapounov의 條件下에서 中心極限定理가 成立한다는 사실을 C^3 의 범위안에서 test function을 비교함에 의하여 고찰하고 實例로서 이를 確認하였다.