

NEW EXPONENTIAL PROBABILITY INEQUALITY AND COMPLETE CONVERGENCE FOR $\mathcal{F} - LNQD$ RANDOM VARIABLES SEQUENCE WITH APPLICATION TO $AR(1)$ MODEL GENERATED BY $\mathcal{F} - LNQD$ ERRORS

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Abstract. *The exponential probability inequalities have been important tools in probability and statistics. In this paper, we prove a new tail probability inequality for the distributions of sums of conditionally linearly negative quadrant dependent ($\mathcal{F} - LNQD$, in short) random variables, and obtain a result dealing with conditionally complete convergence of first-order autoregressive processes with identically distributed ($\mathcal{F} - LNQD$) innovations.*

Keywords: *autoregressive processes; random variables; $\mathcal{F} - LNQD$ sequence; conditionally complete convergence; exponential inequalities.*

1. INTRODUCTION

The exponential inequality plays an important role in various proofs of limit theorems. In particular, it provides a measure of the complete convergence for partial sums.

Firstly, we will recall the definitions of conditionally negative quadrant dependent, conditionally negatively associated, and conditionally linearly negative quadrant dependent sequence. Let $(\Omega, \mathcal{A}, \mathbb{P})$ be a probability space, and all random variables in this paper are defined on it unless otherwise mentioned. Let \mathcal{F} be a sub-algebra of \mathcal{A} , two random variables ζ_1 and ζ_2 are said to be conditionally negative quadrant dependent given \mathcal{F} ($\mathcal{F} - NQD$), in short if, for all $\epsilon_1, \epsilon_2 \in \mathbb{R}$

$$\mathbb{P}^{\mathcal{F}}(\zeta_1 \leq \epsilon_1, \zeta_2 \leq \epsilon_2) \leq \mathbb{P}^{\mathcal{F}}(\zeta_1 \leq \epsilon_1) \mathbb{P}^{\mathcal{F}}(\zeta_2 \leq \epsilon_2) \quad (1)$$

One of the many possible multivariate generalizations of conditionally negative quadrant dependence is conditionally negatively association introduced by Yuan et al [1]. A finite collection of random variables $\zeta_1, \zeta_2, \dots, \zeta_n$ is said to be conditionally negatively associated ($\mathcal{F} - NA$, in short) if for every pair of disjoint subsets A, B of $\{1, 2, \dots, n\}$

$$\text{cov}^{\mathcal{F}}(f(\zeta_i; i \in A), g(\zeta_j; j \in B)) \leq 0,$$

whenever f and g are coordinatewise nondecreasing such that this covariance exists. An infinite sequence $\{\zeta_n, n \geq 1\}$ is $\mathcal{F} - NA$ if every finite subcollection is $\mathcal{F} - NA$.

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We now propose another multivariate generalization of conditionally negative quadrant dependence called conditionally linearly negative quadrant dependence, which is weaker than $\mathcal{F} - NA$ property.

1.1. DEFINITION

A finite sequence of random variables $\{\zeta_n, n \geq 1\}$ is said to be conditionally linearly negative quadrant dependent given $(\mathcal{F} - LNQD)$, in short) if for any disjoint subsets $A, B \subset \mathbb{Z}$ and positive r_j 's,

$$\sum_{k \in A} r_k \zeta_k \text{ and } \sum_{j \in B} r_j \zeta_j \text{ are } \mathcal{F} - NQD$$

As mentioned earlier, it can be shown that the concepts of linearly negative quadrant dependent and conditional linearly negative quadrant dependent are not equivalent. See, for example, Yuan and Xie [2], where various of counterexamples are given.

A concrete example where conditional limit theorems are useful is the study of statistical inference for non-ergodic models as discussed in Bassawa and Prakasa Rao [3] and Basawa and Scott [4]. For instance, if one wants to estimate the mean off-spring for a Galton-Watson branching process, the asymptotic properties of the maximum likelihood estimator depend on the set of non-extinction.

As it was pointed out earlier, the conditional $LNQD$ property does not imply the $LNQD$ property and the opposite implication is also not true. Hence one does have to derive limit theorems under conditioning if there is a need for such results even through the results and proofs of such results may be analogous to those under the non-conditioning setup. This one of the reasons for developing results for sequences of $\mathcal{F} - LNQD$ random variables in this paper.

As mentioned earlier, large numbers of results for $LNQD$ random variables have been achieved. However, nothing is variable for conditional $LNQD$ random variables. Yuan and Wu [5] extended many results from negative association to asymptotically negative association, Yuan and Yang [6] extended many results from association to conditional association, Yuan et al [1] extended many results from negative association to conditional negative association, and these motivate our original interest in conditional $LNQD$.

On the other hand, the concept of complete convergence of a sequence of random variables was introduced by [7]. Note that complete convergence implies almost sure convergence in view of the Borel-Cantelli lemma. Now we extend this concept a conditionally converge completely given \mathcal{F} to a constant a if $\sum_{i=1}^{\infty} P(|X_i - a| > \varepsilon/\mathcal{F}) < \infty$ for every $\varepsilon > 0$, and we write $X_n \rightarrow a$ conditionally completely given \mathcal{F} .

The main purpose of this paper is to establish a new probability inequality and conditional complete convergence for the $\mathcal{F} - LNQD$ random variables and to extend and improve the results of Wang et al [8].

Throughout the paper, let $S_n = \sum_{i=1}^n X_i$ for a sequence $\{X_n, n \geq 1\}$ of random variables defined on a probability space (Ω, \mathcal{A}, P) . Let \mathcal{F} is a sub- σ -algebra of \mathcal{A} , $\{X_n, n \geq 1\}$ will be called \mathcal{F} -centered if $\mathbb{E}^{\mathcal{F}} X_n = 0$ for every $n \geq 0$. Denote $B_n = \sum_{i=1}^n \mathbb{E}^{\mathcal{F}} |X_i|^2$ for each $1 \leq i \leq n$.

2. SOME LEMMAS

Lemma 2.1. [2] Let random variables X and Y be $\mathcal{F} - NQD$. Then

- i. $\mathbb{E}^{\mathcal{F}}(XY) \leq \mathbb{E}^{\mathcal{F}}(X)\mathbb{E}^{\mathcal{F}}(Y)$;
- ii. $\mathbb{P}^{\mathcal{F}}(X > x, Y > y) \leq \mathbb{P}^{\mathcal{F}}(X > x)\mathbb{P}^{\mathcal{F}}(Y > y)$;
- iii. If c and g are both nondecreasing (or both nonincreasing) functions, then $f(X)$ and $g(Y)$ are $\mathcal{F} - NQD$.

Corollary 2.1. [2] Let $\{X_n, n \geq 1\}$ be a sequence of $\mathcal{F} - LNQD$ random variables and $t > 0$, then for each $n \geq 1$,

$$\mathbb{E}^{\mathcal{F}} \left[\sum_{i=1}^n \exp(tX_i) \right] \leq \prod_{i=1}^n \mathbb{E}^{\mathcal{F}}(\exp(tX_i)) \quad (2)$$

Lemma 2.2. [9] For any $x \in \mathbb{R}$, we have

$$\exp(x) \leq 1 + x + \frac{|x|}{2} \ln(1 + |x|) \exp(2|x|).$$

Lemma 2.3. Let $\{X_n, n \geq 1\}$ be a sequence of $\mathcal{F} - LNQD$ random variables with $\mathbb{E}^{\mathcal{F}}(X_n) = 0$ for each $n \geq 1$. If there exists a sequence of positive numbers $\{c_n, n \geq 1\}$ such that $|X_i| \leq c_i$ for each $i \geq 1$, then for any $t > 0$,

$$\mathbb{E}^{\mathcal{F}} \exp\{t \sum_{i=1}^n X_i\} \leq \exp\left\{\frac{t^2}{2} \sum_{i=1}^n e^{2tc_i} \mathbb{E}^{\mathcal{F}} |X_i|^2\right\} \quad (3)$$

Proof: By Lemma 2.2, for all $x \in \mathbb{R}$, $\exp(x) \leq 1 + x + \frac{|x|}{2} \ln(1 + |x|) \exp(2|x|)$. Thus, by $\mathbb{E}^{\mathcal{F}}(X_i) = 0$ and $|X_i| \leq c_i$ for each $i \geq 1$, we have

$$\begin{aligned} \mathbb{E}^{\mathcal{F}} \exp(tX_i) &\leq \mathbb{E}^{\mathcal{F}} \left\{ 1 + tX_i + \frac{t}{2} |X_i| \ln(1 + |tX_i|) \exp(2|tX_i|) \right\} \\ &= 1 + t\mathbb{E}^{\mathcal{F}} X_i + \frac{t}{2} \mathbb{E}^{\mathcal{F}} \{ |X_i| \ln(1 + |tX_i|) \exp(2|tX_i|) \} \\ &= 1 + \frac{t}{2} \mathbb{E}^{\mathcal{F}} \{ |X_i| \ln(1 + |tX_i|) \exp(2|tX_i|) \} \\ &= 1 + \frac{t}{2} \mathbb{E}^{\mathcal{F}} \{ |X_i| \ln(1 + |tX_i|) \exp(2tc_i) \} \\ &= 1 + \frac{t}{2} \exp(2tc_i) \mathbb{E}^{\mathcal{F}} \{ t|X_i|^2 \} \\ &= 1 + \frac{t^2}{2} \exp(2tc_i) \mathbb{E}^{\mathcal{F}} \{ |X_i|^2 \} \\ &\leq \exp\left\{ \frac{t^2}{2} \exp(2tc_i) \mathbb{E}^{\mathcal{F}} \{ |X_i|^2 \} \right\} \quad (\text{using } 1 + y \leq \exp(y) \text{ for all } y \in \mathbb{R}) \end{aligned} \quad (4)$$

for any $t > 0$. By Lemma 2.1 and (4) we have can see that

$$\mathbb{E}^{\mathcal{F}} \exp \left\{ t \sum_{i=1}^n X_i \right\} \leq \prod_{i=1}^n \mathbb{E}^{\mathcal{F}} \exp \{ t X_i \} \quad (5)$$

$$\leq \exp \left\{ \frac{t^2}{2} \sum_{i=1}^n e^{2tc_i} \mathbb{E}^{\mathcal{F}} |X_i|^2 \right\}. \quad (6)$$

The lemma is thus proved.

Lemma 2.4. Let $\{X_n, n \geq 1\}$ be a sequence of \mathcal{F} – LNQD random variables with $\mathbb{E}^{\mathcal{F}}(X_n) = 0$ for each $n \geq 1$. If there exists a sequence of positive numbers $\{c_n, n \geq 1\}$ such that $|X_i| \leq c_i$ for each $i \geq 1$, then for any $t > 0$ and $\varepsilon > 0$

$$\mathbb{P}^{\mathcal{F}} \left\{ \left| \sum_{i=1}^n X_i \right| \geq \varepsilon \right\} \leq \exp \left\{ -t\varepsilon + \frac{t^2}{2} \sum_{i=1}^n e^{2tc_i} \mathbb{E}^{\mathcal{F}} |X_i|^2 \right\}. \quad (7)$$

Proof: By Markov's inequality and Lemma 2.3, we can see that

$$\begin{aligned} \mathbb{P}^{\mathcal{F}} (\sum_{i=1}^n X_i \geq \varepsilon) &\leq \exp(-t\varepsilon) \mathbb{E}^{\mathcal{F}} \exp \{ t \sum_{i=1}^n X_i \} \\ &\leq \exp(-t\varepsilon) \prod_{i=1}^n \mathbb{E}^{\mathcal{F}} \exp \{ t X_i \} \\ &\leq \exp \left\{ -t\varepsilon + \frac{t^2}{2} \sum_{i=1}^n e^{2tc_i} \mathbb{E}^{\mathcal{F}} |X_i|^2 \right\}. \end{aligned} \quad (8)$$

The desired result follows by replacing X_i by $-X_i$ in (8). This completes the proof of the lemma.

3. MAIN RESULTS AND PROOFS

Theorem 3.1. Let $\{X_n, n \geq 1\}$ be a sequence of \mathcal{F} – LNQD random variables with $\mathbb{E}^{\mathcal{F}}(X_i) = 0$. If there exists a positive number c such that $|X_i| \leq c_i, i \geq 1$ where $B_n = \sum_{i=1}^n \mathbb{E}^{\mathcal{F}} X_i^2$, then for any $p > 1, \varepsilon > 0$ and $n > 0$, then

$$\mathbb{P}^{\mathcal{F}} (S_n/B_n \geq \varepsilon) \leq \exp \left\{ \frac{1}{q} b^{q/p} e \right\} \exp \left\{ - \left(\frac{\varepsilon 2^{p-1} b p}{B_n^{p-1}} \right)^{\frac{1}{2p-1}} \varepsilon B_n \left(1 - \frac{1}{p-1} \right) \right\} \quad (9)$$

Proof: By Markov's inequality, we have that for any $t > 0$;

$$\begin{aligned} \mathbb{P}^{\mathcal{F}} (S_n/B_n \geq \varepsilon) &= \mathbb{P}^{\mathcal{F}} (e^{tS_n} \geq e^{t\varepsilon B_n}), \\ &\leq e^{t\varepsilon B_n} \mathbb{E}^{\mathcal{F}} \left(\prod_{i=1}^n e^{tX_i} \right), \\ &\leq \exp \left\{ -t\varepsilon B_n + \frac{t^2}{2} e^{2t \max_{1 \leq i \leq n} c_i} B_n \right\}. \end{aligned} \quad (10)$$

Let $p > 1$. It is well known that

$$uv = \inf_{b>0} \left\{ \frac{1}{pb} u^p + \frac{1}{q} b^{q/p} v^q \right\} \text{ for } u > 0, v > 0 \text{ and } \frac{1}{p} + \frac{1}{q} = 1.$$

This yields the inequality

$$\frac{t^2}{2} e^{2t \max_{1 \leq i \leq n} c_i} B_n \leq \frac{1}{pb} \frac{t^{2p}}{2^p} B_n^p + \frac{1}{q} b^{q/p} e^{2tq \max_{1 \leq i \leq n} c_i} \tag{11}$$

We can thus conclude that for every $p > 1$, there for all $t > 0$, such that

$$\begin{aligned} \mathbb{P}^{\mathcal{F}}(S_n/B_n \geq \varepsilon) &\leq \exp \left\{ -t\varepsilon B_n + \frac{1}{pb} \frac{t^{2p}}{2^p} B_n^p \right\} \\ &\quad \times \exp \left\{ \frac{1}{q} b^{q/p} e^{2tq \max_{1 \leq i \leq n} c_i} \right\} \\ &= \exp \left\{ \frac{1}{q} b^{q/p} e^{2tq \max_{1 \leq i \leq n} c_i} \right\} \exp(\Phi(t, n)) \end{aligned} \tag{12}$$

The equation $\frac{\partial \Phi(t, n)}{\partial t} = 0$ has the unique solution .

$$t = \left(\frac{\varepsilon 2^{p-1} bp}{B_n^{p-1}} \right)^{\frac{1}{2p-1}} \tag{13}$$

which minimizes $\Phi(t, n)$. Then from (12), (13)) and taking $2tq \max_{1 \leq i \leq n} c_i \leq 1$ we obtain (9).

Theorem 3.2. Let $\{X_n, n \geq 1\}$ be a sequence of \mathcal{F} – LNQD random variables $\mathbb{E}^{\mathcal{F}}(X_i) = 0$. If there exists a positive number c such that $|X_i| \leq c_i, i \geq 1$, then for any $p > 1, \varepsilon > 0$ and $n \geq 1$,

$$\mathbb{P}^{\mathcal{F}}(|S_n| \geq \varepsilon) \leq 2 \exp \left\{ \frac{1}{q} b^{q/p} e \right\} \left\{ - \left(\frac{\varepsilon 2^{p-1} bp}{B_n^p} \right)^{\frac{1}{2p-1}} \varepsilon \left(1 - \frac{1}{p-1} \right) \right\} \tag{14}$$

Proof: From conditions $\mathbb{E}^{\mathcal{F}}(X_i) = 0$ and $|X_i| \leq c_i$ for each $i \geq 1$. By Markov’s inequality and Lemma 2.4, Corollary 2.1 with the fact that $1 + x \leq e^x$, then

$$\begin{aligned} \mathbb{P}^{\mathcal{F}}(S_n \geq \varepsilon) &\leq e^{-t\varepsilon} \mathbb{E}^{\mathcal{F}}(e^{tS_n}), \\ &\leq e^{-t\varepsilon} \prod_{i=1}^n \exp \left(\frac{t^2}{2} e^{2tc_i} \mathbb{E}^{\mathcal{F}}|X_i|^2 \right), \\ &\leq \exp \left\{ -t\varepsilon + \frac{t^2}{2} e^{2t \max_{1 \leq i \leq n} c_i} B_n \right\} \end{aligned} \tag{15}$$

Let $p > 1$. It is well known that

$$uv = \inf_{b>0} \left\{ \frac{1}{pb} u^p + \frac{1}{q} b^{q/p} v^q \right\} \text{ for } u > 0, v > 0 \text{ and } 1/p + 1/q = 1.$$

This yields the inequality

$$\frac{t^2}{2} e^{2t \max_{1 \leq i \leq n} c_i} B_n \leq \frac{1}{pb} \frac{t^{2p}}{2^p} B_n^p + \frac{1}{q} b^{q/p} e^{2tq \max_{1 \leq i \leq n} c_i} \quad (16)$$

We can thus conclude that for every $p > 1$, there for all $t > 0$, such that

$$\begin{aligned} \mathbb{P}^{\mathcal{F}}(|S_n| \geq \varepsilon) &\leq 2 \exp \left\{ -t\varepsilon + \frac{1}{pb} \frac{t^{2p}}{2^p} B_n^p \right\} \\ &\quad \times \exp \left\{ \frac{1}{q} b^{q/p} e^{2tq \max_{1 \leq i \leq n} c_i} \right\} \\ &= 2 \exp \left\{ \frac{1}{q} b^{q/p} e^{2tq \max_{1 \leq i \leq n} c_i} \right\} \exp(\Phi(t, n)) \end{aligned} \quad (17)$$

The equation $\frac{\partial \Phi(t, n)}{\partial t} = 0$ has the unique solution

$$t = \left(\frac{\varepsilon 2^{p-1} bp}{B_n^p} \right)^{\frac{1}{2p-1}} \quad (18)$$

which minimizes $\Phi(t, n)$. Then from (17),(18)) and taking $2tq \max_{1 \leq i \leq n} c_i \leq 1$ we obtain upper bound for the tail probability as

$$\mathbb{P}^{\mathcal{F}}(|S_n| \geq \varepsilon) \leq 2 \exp \left\{ \frac{1}{q} b^{q/p} e \right\} \exp \left\{ - \left(\frac{\varepsilon 2^{p-1} bp}{B_n^p} \right)^{\frac{1}{2p-1}} \varepsilon \left(1 - \frac{1}{p-1} \right) \right\} \quad (19)$$

Since $\{-X_n, n \geq 1\}$ is also a sequence of \mathcal{F} -LNQD random variables it follows from (19) that

$$\begin{aligned} \mathbb{P}^{\mathcal{F}}(S_n \leq -\varepsilon) &= \mathbb{P}^{\mathcal{F}}(-S_n \geq \varepsilon) \leq \exp \left\{ \frac{1}{q} b^{q/p} e \right\} \\ &\quad \times \exp \left\{ - \left(\frac{\varepsilon 2^{p-1} bp}{B_n^p} \right)^{\frac{1}{2p-1}} \varepsilon \left(1 - \frac{1}{p-1} \right) \right\} \end{aligned} \quad (20)$$

From (19) and (20) we obtain

$$\begin{aligned} \mathbb{P}^{\mathcal{F}}(|S_n| \geq \varepsilon) &= \mathbb{P}^{\mathcal{F}}(S_n \geq -\varepsilon) + \mathbb{P}^{\mathcal{F}}(S_n \leq \varepsilon) \leq 2 \exp \left\{ \frac{1}{q} b^{q/p} e \right\} \\ &\quad \times \exp \left\{ - \left(\frac{\varepsilon 2^{p-1} bp}{B_n^p} \right)^{\frac{1}{2p-1}} \varepsilon \left(1 - \frac{1}{p-1} \right) \right\} \end{aligned} \quad (21)$$

Theorem 3.3. Let $\{X_n, n \geq 1\}$ be a sequence of \mathcal{F} -LNQD random variables with mean zero and finite variances. If there exists a positive number c such that $|X_i| \leq c_i, i \geq 1$, where $\{c_n, n \geq 1\}$ is a sequence of positive numbers. Then for any $p > 1, \varepsilon > 0$ and $n \geq 1$,

$$p^F(|S_n - E^F S_n| \geq \varepsilon) \leq 2 \exp\left\{\frac{1}{q} b^{q/p} e\right\} \exp\left\{-\left(\frac{\varepsilon 2^{p-1} b p}{B_n^p}\right)^{\frac{1}{2p-1}} \varepsilon \left(1 - \frac{1}{p-1}\right)\right\} \tag{22}$$

Proof: By Markov's inequality and Lemma (2.2), we have that for any $t > 0$,

$$\begin{aligned} p^F(S_n - E^F S_n \geq \varepsilon) &\leq e^{-t\varepsilon} E^F[\exp(t \sum_{i=1}^n (X_i - E^F X_i))], \\ &\leq e^{-t\varepsilon} E^F \prod_{i=1}^n [e^{-t(X_i - E^F X_i)}], \\ &\leq \exp\left\{-t\varepsilon + \frac{t^2}{2} e^{2t \max_{1 \leq i \leq n} c_i} B_n\right\}. \end{aligned} \tag{23}$$

Let $p > 1$. It is well known that

$$uv = \inf_{b>0} \left\{ \frac{1}{pb} u^p + \frac{1}{q} b^{q/p} v^q \right\} \text{ for } u > 0, v > 0 \text{ and } 1/p + 1/q = 1.$$

This yields the inequality

$$\frac{t^2}{2} e^{2t \max_{1 \leq i \leq n} c_i} B_n \leq \frac{1}{pb} \frac{t^{2p}}{2^p} B_n^p + \frac{1}{q} b^{q/p} e^{2tq \max_{1 \leq i \leq n} c_i} \tag{24}$$

We can thus conclude that for every $p > 1$, there for all $t > 0$, such that

$$\begin{aligned} p^F(|S_n - E^F S_n| \geq \varepsilon) &\leq 2 \exp\left\{-t\varepsilon + \frac{1}{pb} \frac{t^{2p}}{2^p} B_n^p\right\} \\ &\times \exp\left\{\frac{1}{q} b^{q/p} e^{2tq \max_{1 \leq i \leq n} c_i}\right\} \\ &= 2 \exp\left\{\frac{1}{q} b^{q/p} e^{2tq \max_{1 \leq i \leq n} c_i}\right\} \exp(\Phi(t, n)) \end{aligned} \tag{25}$$

The equation $\frac{\partial \Phi(t, n)}{\partial t} = 0$ has the unique solution.

Taking $t = \left(\frac{\varepsilon 2^{p-1} b p}{B_n^p}\right)^{\frac{1}{2p-1}}$. Hence it follows from (23) that

$$p^F(S_n - E^F S_n \geq \varepsilon) \leq \exp\left\{\frac{1}{q} b^{q/p} e\right\} \exp\left\{-\left(\frac{\varepsilon 2^{p-1} b p}{B_n^p}\right)^{\frac{1}{2p-1}} \varepsilon \left(1 - \frac{1}{p-1}\right)\right\} \tag{26}$$

Let $-S_n = T_n = \sum_{i=1}^n (-X_n)$. Since $\{-X_n, n \geq 1\}$ is also a sequence of \mathcal{F} -LNQD random variables we also have

$$\begin{aligned} p^F(S_n - E^F S_n \leq -\varepsilon) &= p^F(T_n - E^F T_n \geq \varepsilon) \leq \exp\left\{\frac{1}{q} b^{q/p} e\right\} \\ &\times \exp\left\{-\left(\frac{\varepsilon 2^{p-1} b p}{B_n^p}\right)^{\frac{1}{2p-1}} \varepsilon \left(1 - \frac{1}{p-1}\right)\right\} \end{aligned} \tag{27}$$

by Combining (26) and(27) we get (22)

Corollary 3.1. Let $\{X_n, n \geq 1\}$ be a sequence of \mathcal{F} -LNQD random variables. Assume that there exists a positive integer n_0 such that $|X_i| \leq c_n$, for each $1 \leq i \leq n, n \geq n_0$, where $\{c_n, n \geq 1\}$ is a sequence of positive numbers. Then for any $\varepsilon > 0$ and $p > 1$

$$P^F(|S_n - E^F S_n| \geq n\varepsilon) \leq 2 \exp\left\{\frac{1}{q} b^{q/p} e\right\} \exp\left\{-\left(\frac{n\varepsilon 2^{p-1} bp}{B_n^p}\right)^{\frac{1}{2p-1}} n\varepsilon \left(1 - \frac{1}{p-1}\right)\right\} \quad (28)$$

Theorem 3.4. Let $\{X_n, n \geq 1\}$ be a sequence of \mathcal{F} -LNQD random variables with $E^F(X_i) = 0$. If there exists a positive numbers c such that $|X_i| \leq c_i, i \geq 1$. Then for any $r > 0$

$$n^{-r} S_n \rightarrow 0 \text{ completely, } n \rightarrow \infty. \quad (29)$$

Proof: Let $B = \sum_{n=1}^{\infty} E^F(X_n)^2 \leq \infty$. For any $\varepsilon > 0$, it follows from Theorem 3.2 we have

$$\begin{aligned} \sum_{n=1}^{\infty} P^F(|S_n| \geq n^r \varepsilon) &\leq 2 \sum_{n=1}^{\infty} \exp\left\{\frac{1}{q} b^{q/p} e\right\} \exp\left\{-\left(\frac{n^r \varepsilon 2^{p-1} bp}{B_n^p}\right)^{\frac{1}{2p-1}} \varepsilon n^r \left(1 - \frac{1}{p-1}\right)\right\} \\ &\leq 2 \sum_{n=1}^{\infty} \exp\left\{\frac{1}{q} b^{q/p} e\right\} \exp\left\{-\left(\frac{\varepsilon 2^{p-1} bp}{B_n^p}\right)^{\frac{1}{2p-1}} \varepsilon \left(1 - \frac{1}{p-1}\right)\right\}^{\frac{2rp}{n^{2p-1}}} \\ &\leq 2 \exp\left\{\frac{1}{q} b^{q/p} e\right\} \sum_{n=1}^{\infty} [\exp(-c)]^{\frac{2rp}{n^{2p-1}}} \end{aligned} \quad (30)$$

where C is positive number not depending on n . (by the inequality $e^{-y} \leq \left(\frac{a}{ey}\right)^a$), choosing $a = \frac{2p-1}{rp}$, since $a > 0, y > 0$. Then the right-hand side of (30) become,

$$\begin{aligned} \sum_{n=1}^{\infty} P^F(|S_n| \geq n^r \varepsilon) &\leq 2 \exp\left\{\frac{1}{q} b^{q/p} e\right\} \sum_{n=1}^{\infty} \left(\frac{a}{ec}\right)^a \left(\frac{1}{n}\right)^{\left(\frac{2rp}{n^{2p-1}}\right)^a} \\ &\leq 2 \exp\left\{\frac{1}{q} b^{q/p} e\right\} \frac{a^a}{(ec)^a} \sum_{n=1}^{\infty} \frac{1}{\frac{2rpa}{n^{2p-1}}} \\ &\leq 2 \exp\left\{\frac{1}{q} b^{q/p} e\right\} \frac{a^a}{(ec)^a} \sum_{n=1}^{\infty} \frac{1}{n^2}, \\ &\leq 2 \exp\left\{\frac{1}{q} b^{q/p} e\right\} \frac{a^a}{(ec)^a} \frac{\pi^2}{6}, \\ &< \infty \end{aligned} \quad (31)$$

Theorem 3.5. Let $\{X_n, n \geq 1\}$ be a sequence of \mathcal{F} -LNQD random variables. Assume that there exists a positive integer n_0 such that $|X_i| \leq c_n$, for each $1 \leq i \leq n, n \geq n_0$, where $\{c_n, n \geq 1\}$ is a sequence of positive numbers. Then,

$$\sum_{n=1}^{\infty} P^F\left(\frac{1}{n} |S_n - E^F S_n| \geq \varepsilon_n\right) < \infty \quad (32)$$

Theorem 3.6. Let $\{X_n, n \geq 1\}$ be a sequence of \mathcal{F} -LNQD random variables with $E^F(X_i) = 0$. If there exists a positive number c such that $|X_i| \leq c_n, i \geq 1$. Then for any $r > 0$

$$n^{-r} (S_n - E^F S_n) \rightarrow 0 \text{ completely, } n \rightarrow \infty. \tag{33}$$

Proof: For any $\varepsilon > 0$, it follows from Corollary 3.1 that

$$\begin{aligned} \sum_{n=1}^{\infty} \mathbb{P}^F (|S_n - E^F S_n| \geq n^r \varepsilon) &\leq \\ 2 \sum_{n=1}^{\infty} \exp \left\{ \frac{1}{q} b^{q/p} e \right\} \exp \left\{ - \left(\frac{n^r \varepsilon 2^{p-1} b p}{B_n^p} \right)^{\frac{1}{2p-1}} \varepsilon n^r \left(1 - \frac{1}{p-1} \right) \right\} & \\ \leq 2 \sum_{n=1}^{\infty} \left[\exp \left\{ \frac{1}{q} b^{q/p} e \right\} \right] & \\ \times \left[\exp \left\{ - \left(\frac{\varepsilon 2^{p-1} b p}{B_n^p} \right)^{\frac{1}{2p-1}} \varepsilon \left(1 - \frac{1}{p-1} \right) \right\} \right]^{\frac{2rp}{n^{2p-1}}} & \end{aligned} \tag{34}$$

By this result we get (33).

4. APPLICATIONS TO THE RESULTS TO AR(1) MODEL

The basic object of this section is applying the results to first-order autoregressive processes (AR(1)).

We consider an autoregressive time series of first order AR(1) defined by this

$$X_{n+1} = \theta X_n + \zeta_{n+1}, \quad n = 1, 2, \dots, \tag{35}$$

where $\{\zeta_n, n \geq 0\}$ is a sequence of identically distributed \mathcal{F} -LNQD random variables with $\zeta_0 = X_0 = 0, 0 < E^F \zeta_k^4 < \infty, k = 1, 2, \dots$, and where θ is a parameter with $|\theta| < 1$. Here, we can rewrite X_{n+1} in (35) as follows:

$$X_{n+1} = \theta^{n+1} X_0 + \theta^n \zeta_1 + \theta^{n-1} \zeta_2 + \dots + \zeta_{n+1}. \tag{36}$$

The coefficient θ is fitted least squares, giving the estimator

$$\widehat{\theta}_n = \frac{\sum_{j=1}^n X_j X_{j-1}}{\sum_{j=1}^n X_{j-1}^2} \tag{37}$$

It immediately follows from (35) and (37) that

$$\widehat{\theta}_n - \theta = \frac{\sum_{j=1}^n \zeta_j X_{j-1}}{\sum_{j=1}^n X_{j-1}^2} \tag{38}$$

Theorem 4.1. Let the conditions of Theorem 3.3 be satisfied then for any $\frac{(E^F \zeta_1^2)^{1/2}}{\rho^2} < \xi$ positive, we have

$$\mathbb{P}^F (\sqrt{n} |\widehat{\theta}_n - \theta| > \rho) \leq 2 \exp \left\{ - \left(\frac{(\rho^2 \xi^2 - E^F \zeta_1^2) n 2^{p-1} b p}{B_n^p} \right)^{\frac{1}{2p-1}} (\rho^2 \xi^2 - E^F \zeta_1^2) n \left(1 - \right. \right. \tag{39}$$

$$\left. \frac{1}{p-1} \right\} \times \exp \left\{ \frac{1}{q} b^{q/p} e \right\} + \left\{ -\frac{1}{2} n \frac{(K_1 - n\xi^2)^2}{K_2} \right\}$$

where $K_1 = \mathbb{E}^{\mathcal{F}}(X_i^2) < \infty$, $K_2 = \mathbb{E}^{\mathcal{F}}(X_i^4) < \infty$.

Proof: Firstly, we notice that:

$$\widehat{\theta}_n - \theta = \frac{\sum_{j=1}^n \zeta_j X_{j-1}}{\sum_{j=1}^n X_{j-1}^2}.$$

It follows that

$$\mathbb{P}^{\mathcal{F}}(\sqrt{n}|\widehat{\theta}_n - \theta| > \rho) \leq \mathbb{P}^{\mathcal{F}}\left(\left| \frac{1/\sqrt{n} \sum_{j=1}^n \zeta_j X_{j-1}}{1/n \sum_{j=1}^n X_{j-1}^2} \right| > \rho\right)$$

By virtue of the probability properties and Hölder’s inequality, we have for any ξ positive

$$\begin{aligned} \mathbb{P}^{\mathcal{F}}(\sqrt{n}|\widehat{\theta}_n - \theta| > \rho) &\leq \mathbb{P}^{\mathcal{F}}\left(1/n \sum_{j=1}^n \zeta_j^2 \geq \rho^2 \xi^2\right) + \mathbb{P}^{\mathcal{F}}\left(1/n^2 \sum_{j=1}^n X_{j-1}^2 \leq \xi^2\right) \\ &= \mathbb{P}^{\mathcal{F}}(\sum_{j=1}^n \zeta_j^2 \geq (\rho^2 \xi^2)n) + \mathbb{P}^{\mathcal{F}}(\sum_{j=1}^n X_{j-1}^2 \leq n^2 \xi^2) \\ &= I_{1n} + I_{2n}. \end{aligned}$$

Next we estimate I_{1n} and I_{2n} .

$$\begin{aligned} I_{1n} &= \mathbb{P}^{\mathcal{F}}(\sum_{j=1}^n \zeta_j^2 \geq (\rho^2 \xi^2)n) \\ &= \mathbb{P}^{\mathcal{F}}(\sum_{j=1}^n (\zeta_j^2 - \mathbb{E}^{\mathcal{F}} \zeta_j^2 + \mathbb{E}^{\mathcal{F}} \zeta_j^2) \geq (\rho^2 \xi^2)n) \\ &= \mathbb{P}^{\mathcal{F}}(\sum_{j=1}^n (\zeta_j^2 - \mathbb{E}^{\mathcal{F}} \zeta_j^2) \geq (\rho^2 \xi^2 - \mathbb{E}^{\mathcal{F}} \zeta_j^2)n) \\ &\leq \mathbb{P}^{\mathcal{F}}(|\sum_{j=1}^n (\zeta_j^2 - \mathbb{E}^{\mathcal{F}} \zeta_j^2)| \geq (\rho^2 \xi^2 - \mathbb{E}^{\mathcal{F}} \zeta_j^2)n) \end{aligned} \tag{40}$$

By using the Theorem 3.3 the right hand side of (40) become

$$\begin{aligned} I_{1n} &= \mathbb{P}^{\mathcal{F}}(\sum_{j=1}^n \zeta_j^2 \geq (\rho^2 \xi^2)n) \\ &\leq 2 \exp \left\{ -\left(\frac{(\rho^2 \xi^2 - \mathbb{E}^{\mathcal{F}} \zeta_1^2) n 2^{p-1} b p}{B_n^p} \right)^{\frac{1}{2p-1}} (\rho^2 \xi^2 - \mathbb{E}^{\mathcal{F}} \zeta_1^2) n \left(1 - \frac{1}{p-1}\right) \right\} \\ &\times \exp \left\{ \frac{1}{q} b^{q/p} e \right\} \end{aligned} \tag{41}$$

We will bound now, the second probability of the right-hand side of the expression I_{2n} . According to the Markov’s inequality, it follows for any positive t .

$$\begin{aligned} I_{2n} &= \mathbb{P}^{\mathcal{F}}\left(\frac{1}{n^2} \sum_{i=1}^n X_{i-1}^2 \leq \xi^2\right) \\ &= \mathbb{P}^{\mathcal{F}}(\xi^2 n^2 - \sum_{i=1}^n X_{i-1}^2 \geq 0) \end{aligned}$$

$$\begin{aligned}
 &= \mathbb{E}^{\mathcal{F}} \left(\mathbb{1}_{\{n^2 - \sum_{i=1}^n X_{i-1}^2 \geq 0\}} \right) \\
 &\leq \mathbb{E}^{\mathcal{F}} \left(\exp(t(n^2 \xi^2 - \sum_{i=1}^n X_{i-1}^2)) \right) \quad (t > 0) \\
 &\leq e^{tn^2 \xi^2} \mathbb{E}^{\mathcal{F}} \left(\exp - t \sum_{i=1}^n X_{i-1}^2 \right) \\
 &\leq e^{tn^2 \xi^2} \prod_{i=1}^n \mathbb{E}^{\mathcal{F}} \left(\exp - t X_{i-1}^2 \right).
 \end{aligned}$$

Since

$$I_{2n} \leq e^{tn^2 \xi^2} \prod_{i=1}^n \mathbb{E}^{\mathcal{F}} \left(\exp - t X_{i-1}^2 \right).$$

we first claim that for $x \geq 0$

$$e^{-x} \leq 1 - x + \frac{1}{2}x^2. \tag{42}$$

To see this let $\psi(x) = e^{-x}$ and $\phi(x) = 1 - x + \frac{1}{2}x^2$, $(\psi(x))' = -e^{-x}$ and recall that for every x

$$e^x \geq 1 + x \quad \forall x, \tag{43}$$

so that $\psi(x)' = -e^{-x} \leq -1 + x = \phi(x)'$. Since $\psi(0) = 1 = \phi(0)$ this implies $\psi(x) \leq \phi(x)$ for all $x \geq 0$ and (42) is claimed.

From (42) and (43) it follows that for $t > 0$

$$\begin{aligned}
 e^{tn^2 \xi^2} \prod_{i=1}^n \mathbb{E}^{\mathcal{F}} \left(\exp(-t X_{i-1}^2) \right) &\leq e^{tn^2 \xi^2} \left(1 - tK_1 + \frac{t^2}{2}K_2 \right)^n \\
 &\leq e^{tn^2 \xi^2} \left(\exp \left(-tK_1 + \frac{t^2}{2}K_2 \right) \right)^n \\
 &\leq e^{tn^2 \xi^2} \exp \left(-ntK_1 + \frac{t^2}{2}nK_2 \right)
 \end{aligned} \tag{44}$$

where $K_1 = \mathbb{E}^{\mathcal{F}}(X_i^2) < \infty$, $K_2 = \mathbb{E}^{\mathcal{F}}(X_i^4) < \infty$.

Hence

$$I_{2n} = \mathbb{P}^{\mathcal{F}} \left(\sum_{i=1}^n X_{i-1}^2 \leq n^2 \xi^2 \right) \leq \left[t(n^2 \xi^2 - nK_1) + \frac{nt^2 K_2}{2} \right]$$

with $h(t) = n^2 \xi^2 - nK_1 + \frac{nt^2 K_2}{2}$ and $t > 0$, the equation $h'(t) = 0$ has the unique solution $t = \frac{K_1 - n\xi^2}{K_2}$ which minimize $h(t)$. Hence

$$\mathbb{P}^{\mathcal{F}} \left(\sum_{i=1}^n X_{i-1}^2 \leq n^2 \xi^2 \right) \leq \exp \left\{ -\frac{1}{2}n \frac{(K_1 - n\xi^2)^2}{K_2} \right\} \tag{45}$$

Then for every $\rho > 0$, $K_1 < \infty$, $K_2 < \infty$, and by the assumption

$$\mathbb{P}^{\mathcal{F}}(\sqrt{n}|\widehat{\theta}_n - \theta| > \rho) \leq 2 \exp \left\{ - \left(\frac{(\rho^2 \xi^2 - \mathbb{E}^{\mathcal{F}} \zeta_1^2) n 2^{p-1} \text{bp}}{B_n^p} \right)^{\frac{1}{2p-1}} (\rho^2 \xi^2 - \mathbb{E}^{\mathcal{F}} \zeta_1^2) n \left(1 - \frac{1}{p-1} \right) \right\} \tag{46}$$

$$\times \exp \left\{ \frac{1}{q} b^{q/p} e \right\} + \left\{ -\frac{1}{2} n \frac{(K_1 - n\xi^2)^2}{K_2} \right\}$$

This completes the proof.

Corollary 4.1. The sequence $(\widehat{\theta}_n)_{n \in \mathbb{N}}$ completely converges to the parameter θ of autoregressive process $AR(1)$ model. Then we have

$$\sum_{n=1}^{\infty} \mathbb{P}^F(\sqrt{n}|\widehat{\theta}_n - \theta| > \rho) < \infty. \quad (47)$$

Proof: By using Theorem 3.6 and $\mathbb{E}^F(X_i^2) < \infty$, $\mathbb{E}^F(X_i^4) < \infty$ we get the result of (47) immediately.

5. CONCLUSION

Our work consists in establishing some new exponential inequalities for the distribution of sums of \mathcal{F} – LNQD random variables. Using these inequalities, we proved the conditionally complete convergence of first-order autoregressive processes $AR(1)$ with identically distributed (\mathcal{F} – LNQD) errors.

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