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Author(s): Tze Leung Lai

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SUMMABILITY METHODS FOR INDEPENDENT, IDENTICALLY DISTRIBUTED RANDOM VARIABLES¹

TZE LEUNG LAI

ABSTRACT. In this paper, we present certain theorems concerning the Cesaro (C, α) , Abel (A), Euler (E, q) and Borel (B) summability of ΣY_i , where $Y_i = X_i - X_{i-1}$, $X_0 = 0$ and X_1 , X_2 , \cdots are i.i.d. random variables. While the Kolmogorov strong law of large numbers and the Hartman-Wintner law of the iterated logarithm are related to (C, 1) summability and involve the finiteness of, respectively, the first and second moments of X_1 , their analogues for Euler and Borel summability involve different moment conditions, and the analogues for (C, α) and Abel summability remain essentially the same.

Let X_1, X_2, \cdots be independent, identically distributed (i.i.d.) random variables. Let $Y_n = X_n - X_{n-1}$ ($X_0 = Y_0 = 0$). Kolmogorov's strong law of large numbers asserts that $EX_1 = \mu$ iff $\sum Y_i$ is a.e. (C, 1) summable to μ , i.e., the (C, 1) limit of X_n is μ a.e. By the well-known inclusion theorems involving Cesaro and Abel summability (cf. [5, Theorems 43 and 55]), this implies that $\sum Y_i$ is a.e. (C, α) summable to μ for any $\alpha \ge 1$ and that $\sum Y_i$ is a.e. (A) summable to μ . In fact, the converse also holds in the present case, and we have the following theorem.

Theorem 1. If X_1, X_2, \cdots is a sequence of i.i.d. random variables and $\alpha \ge 1$ and μ are given real numbers, then the following statements are equivalent:

(1)
$$EX_{1} = \mu_{0}$$

(2)
$$X_n \to \mu(C, 1)$$
 a.e., i.e., $\lim_{n \to \infty} (1/n)(X_1 + \cdots + X_n) = \mu$ a.e.

(3)
$$X_n \to \mu(C, \alpha)$$
 a.e., i.e., $\lim_{n \to \infty} \sum_{i=1}^{n-1} {n \choose i} X_{n-i} / {n+\alpha \choose n} = \mu$ a.e., where ${i+\beta \choose j} = (\beta+1) \cdots (\beta+j)/(j!)$.

(4)
$$X_n \to \mu(A)$$
 a.e., i.e., $\lim_{\lambda \to 1^-} (1 - \lambda) \sum_{i=1}^{\infty} \lambda^i X_i = \mu$ a.e.

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Proof. The implications (2) \Rightarrow (3) \Rightarrow (4) are well known (cf. [5, pp. 96, 100, 108]). We shall now prove that $(4) \Rightarrow (1)$. By (4),

$$\lim_{m \to \infty} \frac{1}{m} \sum_{n=1}^{\infty} e^{-n/m} X_n^s = 0 \quad \text{a.e.,}$$

where $X_n^s = X_n - X_n'$ with X_n' , $n \ge 1$, and X_n , $n \ge 1$, being i.i.d. Let

$$Y_m = \frac{1}{m} \sum_{n=1}^m e^{-n/m} X_n^s, \qquad Z_m = \frac{1}{m} \sum_{n=m+1}^\infty e^{-n/m} X_n^s.$$

Then $Y_m + Z_m \xrightarrow{P} 0$, Y_m and Z_m are independent and symmetric. Therefore it follows easily from the Lévy inequality [8, p. 247] that $Z_m \xrightarrow{P} 0$. Since Z_m and (Y_1, \dots, Y_m) are independent and $Y_m + Z_m \to 0$ a.e., $Z_m \xrightarrow{P} 0$, we obtain by Lemma 3 of [2] that $Y_m \to 0$ a.e. Letting $Y_m^{(1)} = Y_m - (em)^{-1} X_m^s$, since $(em)^{-1} X_m^s \xrightarrow{P} 0$, we again obtain by Lemma 3 of [2] that $X_m^s/m \to 0$ a.e. By the Borel-Cantelli lemma, this implies that $E[X_1]$ $< \infty$. As established before, we then have $X_n \to EX_1(A)$ and so by (4), $\mu = EX_1$

The classical law of the iterated logarithm gives us the rate at which the convergence in (2) takes place. Gaposhkin [4] has established the law of the iterated logarithm for the (C, α) and Abel methods in the case where X, is bounded. Actually by using either the Hartman-Wintner truncation scheme [6] or Strassen's strong invariance principle [9], we can extend Gaposhkin's results to the case where $0 \le EX_1^2 \le \infty$. The extension is the sharpest possible in the sense that its converse also holds, and this converse can be established by a modification of Feller's argument in [3]. The details of the proof are omitted here, and we simply state the results in the following theorem.

Theorem 2. With the same notation as in Theorem 1, for any given positive number σ , the following statements are equivalent:

- (5) $EX_1 = 0$, $EX_1^2 = \sigma^2$.
- (6) $\limsup_{n\to\infty} \frac{|\Sigma_1^n X_i|}{|\Sigma_1^n X_i|} = \sigma a.e.$ (7) $\limsup_{n\to\infty} \frac{|\Sigma_1^n X_i|}{|\Sigma_{i=1}^n (i+\alpha-1)X_{n-i}|}{|\Sigma_{n-i}^n X_{n-i}|} = \sigma a.e.$ a.e.

(8)
$$\limsup_{\lambda \to 1^{-}} (1 - \lambda)^{1/2} |\sum_{i=1}^{\infty} \lambda^{i} X_{i}| / \{\log |\log (1 - \lambda)|\}^{1/2} = \sigma a.e.$$

Our primary interest in this paper lies in the analogue of the law of the iterated logarithm for the Euler and Borel methods of summation (cf. [5,

Chapters 8 and 9]). Chow [1] has shown that unlike the Cesaro and Abel methods which require $E|X_1| < \infty$ for summability, the Euler and Borel methods require $EX_1^2 < \infty$ for summability. Specifically, if X_1, X_2, \cdots are i.i.d., then the following statements are equivalent:

(9)
$$EX_1 = \mu, EX_1^2 < \infty$$
.

(10) $X_n \to \mu(E, q)$ for some (or equivalently for every) q > 0, i.e.,

$$\lim_{n \to \infty} (q+1)^{-n} \sum_{i=1}^{n} \binom{n}{i} q^{n-i} X_i = \mu \quad \text{a.e.}$$

(11)
$$X_n \rightarrow \mu(B)$$
, i.e.,

$$\lim_{\lambda \to \infty} e^{-\lambda} \sum_{i=1}^{\infty} \frac{\lambda^i}{i!} X_i = \mu \quad \text{a.e.}$$

The following theorem gives the rate at which the convergence in (10) or (11) takes place. It is interesting to compare the result with Theorem 2.

Theorem 3. Suppose q>0, $\sigma>0$ and X_1,X_2,\cdots are i.i.d. random variables. Then the following statements are equivalent:

(12)
$$EX_1 = 0$$
, $EX_1^2 = \sigma^2$, $EX_1^4(\log^+|X_1| + 1)^{-2} < \infty$.

(13)
$$\limsup_{\lambda \to \infty} (\pi \lambda)^{\frac{1}{4}} (\log \lambda)^{-\frac{1}{2}} \left| \sum_{i=1}^{\infty} e^{-\lambda} (\lambda^{i}/i!) X_{i} \right| = \sigma/\sqrt{2} \ a.e.$$

(14)
$$\limsup_{n\to\infty} (\pi q n)^{\frac{1}{4}} (\log n)^{-\frac{1}{2}} \left| \sum_{i=1}^{n} (q+1)^{-n} {n \choose i} q^{n-i} X_i \right|$$

$$= \sigma(\frac{1}{2}(q+1))^{\frac{1}{2}} \text{ a.e.}$$

Lemma. Let X_1, X_2, \cdots be independent random variables with $EX_n = 0$, $EX_n^2 = \sigma_n^2$ and $\lim_{n \to \infty} \sigma_n^2 = \sigma^2 > 0$. Suppose for $j \ge j_0$, there exists $r_j \ge 0$ such that $r_j = o(j^{\frac{1}{4}}(\log j)^{-\frac{1}{2}})$ and

(15)
$$\exp\{t^2\sigma_j^2(1-|t|r_j)/2\} \le E \exp(tX_j) \le \exp\{t^2\sigma_j^2(1+|t|r_j/2)/2\}$$

whenever $|t|r_j \leq 1$. Then for any c > 0 and $\alpha > \frac{1}{2}$,

(16)
$$\lim_{M \to \infty} \sup \left(\frac{\pi M}{2c} \right)^{-\frac{1}{4}} (\log M)^{-\frac{1}{2}} \sum_{|m-M| \le M} X_m \exp \left(\frac{-c(m-M)^2}{M} \right) = \sigma \quad a.e.$$

Proof. To simplify the notation, we shall set b = m - M. Let $0 < \delta < \delta' < 1$ such that $(1 - \delta')^2 (1 + \delta) < 1$. Choose $\zeta > 1$ such that

$$\delta > 3 \left(2 \int_{\zeta}^{\infty} e^{-2ct^2} dt \right)^{\frac{1}{2}} \text{ and } \int_{-\zeta}^{\zeta} e^{-2ct^2} dt > (1 - \delta)^2 (\pi/2c)^{\frac{1}{2}}.$$

Let

$$s_M^2 = \sum_{\zeta \sqrt{M} < |b| \le M^{\alpha}} (\exp(-2ch^2/M)) \sigma_m^2 \sim 2\sigma^2 \sqrt{M} \int_{\zeta}^{\infty} \exp(-2ct^2) dt,$$

$$Y_{M} = \left(\frac{1}{s_{M}}\right) \sum_{\zeta \sqrt{M} < |h| \le M^{\alpha}} (\exp(-ch^{2}/M)) X_{m}.$$

Using (15) and Kolmogorov's (upper) exponential bounds (cf. [10, Lemma 1]), we obtain that for all large M,

$$P\left[\sum_{\zeta\sqrt{M}<|h|\leq M^{\alpha}} (\exp(-ch^{2}/M))X_{m} \geq \delta\sigma M^{\frac{1}{4}} (\log M)^{\frac{1}{2}}\right]$$

$$\leq P[Y_{M} \geq 3(\log M)^{\frac{1}{2}}] \leq \exp\{-(9/4)\log M\}.$$

Replacing X_m by $-X_m$ in (17) gives a similar inequality, and an application of the Borel-Cantelli lemma gives

(18)
$$\lim_{M \to \infty} \sup_{M \to \infty} M^{-\frac{1}{4}} (\log M)^{-\frac{1}{2}} \left| \sum_{\zeta \setminus M < |h| \le M} (\exp(-ch^2/M)) X_m \right| \le \delta \sigma \quad \text{a.e.}$$

Choose $\eta > \zeta^2$ and set $m_k = [\eta k^2]$. Using (15) and Kolmogorov's (lower) exponential bounds (cf. [10, Lemma 1]), it can be shown that for all large k,

$$P\left[\left(\frac{\pi m_k}{2c}\right)^{-\frac{1}{4}} \sum_{|h| \leq \zeta \sqrt{m_k}} X_m \exp\left(\frac{-ch^2}{m_k}\right) \geq (1-\delta)\sigma(2 \log k)^{\frac{1}{4}}\right]$$

$$\geq \exp\left\{-(1+\delta)(1-\delta')^2 \log k\right\}_c$$

Now the σ -fields $\mathcal{F}_k = \Re(X_m: |m-m_k| \leq \zeta \sqrt{m_k})$ are independent for all large k. Hence it follows from (19) and the Borel-Cantelli lemma that

$$\limsup_{k \to \infty} \left(\frac{\pi m_k}{2c} \right)^{-\frac{1}{4}} (2 \log k)^{-\frac{1}{2}} \sum_{|h| \le \zeta \sqrt{m_k}} X_m \exp \left(\frac{-ch^2}{m_k} \right) \ge (1 - \delta)\sigma \quad a.e.$$

Since 2 log $k \sim \log m_k$ and δ is arbitrary, we obtain from (18) and (20) that the \limsup in (16) is $\geq \sigma$ a.e.

To prove that the \limsup in (16) is $\le \sigma$ a.e., let $\eta_k = (\log k)^{-2}$, $M_k = [\eta_k k^2]$. An application of Kolmogorov's (upper) exponential bounds gives

$$(21) \sum P \left[\left(\frac{\pi M_k}{2c} \right)^{-\frac{1}{4}} \sum_{|h| \leq \zeta \sqrt{M_k}} X_m \exp \left(\frac{-ch^2}{M_k} \right) \geq (1 + \delta) \sigma (2 \log k)^{\frac{1}{4}} \right] < \infty.$$

Obviously (21) still holds with X_m replaced by $-X_m$, and noting that $\log M_b \sim 2 \log k$, we obtain that

$$\lim_{k \to \infty} \sup \left(\frac{\pi M_k}{2c} \right)^{-\frac{1}{4}} (\log M_k)^{-\frac{1}{2}} \left| \sum_{|h| \le \zeta \sqrt{M_k}} X_m \exp \left(\frac{-ch^2}{M_k} \right) \right| \le (1+\delta)\sigma \quad \text{a.e.}$$

We now assert that

(23)
$$\limsup_{k \to \infty} M_k^{-1/4} (\log M_k)^{-1/2} \max_{M_k < M \le M_{k+1}} |U_{k,M}| \le 6\delta \quad \text{a.e.,}$$

where we define, for $M_k \le M \le M_{k+1}$,

$$U_{k,M} = \sum_{|m-M| \leq \zeta \sqrt{M}} X_m \exp\left(\frac{-c(m-M)^2}{M}\right) - \sum_{|m-M_k| \leq \zeta \sqrt{M_k}} X_m \exp\left(\frac{-c(m-M_k)^2}{M_k}\right);$$

$$U_{k,M}^{(1)} = \sum_{M_k - \zeta \sqrt{M_k} \le m \le M - \zeta \sqrt{M}} X_m \exp\left(\frac{-c(m - M_k)^2}{M_k}\right);$$

$$U_{k,M}^{(2)} = \sum_{M - \zeta \sqrt{M} \le m \le M} X_m \left\{ \exp\left(\frac{-c(m - M_k)^2}{M_k}\right) - \exp\left(\frac{-c(m - M)^2}{M}\right) \right\};$$

$$U_{k,M}^{(3)} = \sum_{M_k \le m \le M} X_m \exp \left(\frac{-c(m - M_k)^2}{M_k} \right);$$

$$U_{k,M}^{(4)} = \sum_{M, \leq m < M} X_m \exp \left(\frac{-c(m-M)^2}{M} \right);$$

$$U_{k,M}^{(5)} = \sum_{M \le m < M_k + \zeta \sqrt{M_k}} X_m \left\{ \exp\left(\frac{-c(m-M)^2}{M}\right) - \exp\left(\frac{-c(m-M_k)^2}{M_k}\right) \right\};$$

$$U_{k,M}^{(6)} = \sum_{M, +\zeta\sqrt{M}, \leq m \leq M + \zeta\sqrt{M}} X_m \exp\left(\frac{-c(m-M)^2}{M}\right).$$

We note that if k is sufficiently large and $M_k < M \le M_{k+1}$, then $M_k - \zeta \sqrt{M_k} < M - \zeta \sqrt{M} < M \le M_k + \zeta \sqrt{M_k} < M + \zeta \sqrt{M}$, and so $|U_{k,M}| \le \sum_{i=1}^6 |U_{k,M}^{(i)}|$.

First consider $U_{k,M}^{(1)}$. Setting $t = (8\delta^{-2} \log k)^{\frac{1}{2}}$, we have from (15) that $P[U_{k,M}^{(1)} \ge \delta M_k^{\frac{1}{2}} (\log M_k)^{\frac{1}{2}}]$

$$\leq \{ \exp(-t\delta(\log M_k)^{\frac{1}{2}}) \} E \exp\{tM_k^{-\frac{1}{4}}U_{k,M}^{(1)} \}$$

(24)
$$\leq k^{-4} \exp \left\{ t^2 M_k^{-\frac{1}{2}} \sum_{\substack{M_k - \zeta \sqrt{M_k} \leq m < M - \zeta \sqrt{M}}} \sigma_m^2 \exp \left(\frac{-2c(m - M_k)^2}{M_k} \right) \right\}$$

$$= O(k^{-4}).$$

Replacing X_m by $-X_m$ in the above argument, we then obtain from (24) that

$$P\left[\max_{\substack{M_k < M \le M_{k+1}}} |U_{k,M}^{(1)}| \ge \delta M_k^{\frac{1}{4}} (\log M_k)^{\frac{1}{2}}\right] = O(k^{-4}(M_{k+1} - M_k)) = O(k^{-3}).$$

Therefore by the Borel-Cantelli lemma, we have for i = 1,

(25)
$$\limsup_{k \to \infty} M_k^{-\frac{1}{4}} (\log M_k)^{-\frac{1}{2}} \max_{M_k < M \le M_{k+1}} |U_{k,M}^{(i)}| \le \delta \quad \text{a.e.}$$

By a similar argument, we can show that (25) also holds for i = 2, 3, 4, 5, 6. Hence (23) holds, and the desired conclusion follows from (18), (22) and (23).

Proof of Theorem 3. We first assume (12) and show that (13) and (14) both hold. Set $b_m(\lambda) = e^{-\lambda} \lambda^m/(m!)$, let $1/2 < \alpha < 2/3$ and let k be an integer such that $k(1-\alpha) > 1$. Given $\delta > 0$, we choose $\epsilon > 0$ such that $\epsilon k < \delta$. From (12), it follows that

(26)
$$\lim_{n \to \infty} n^{-\frac{1}{4}} (\log n)^{-\frac{1}{2}} X_n = 0 \quad \text{a.e.}$$

To prove (13), we set $M = [\lambda]$, h = m - M. Since there exists $\nu > 0$ such that $\Sigma_{|h|>M}{}^{\alpha}b_{m}(\lambda) = O(\exp(-\lambda^{\nu}))$ (cf. [5, p. 200]), it follows from (26) that

(27)
$$\lim_{\lambda \to \infty} \lambda^{\frac{1}{4}} (\log \lambda)^{-\frac{1}{2}} \sum_{|h| > M} b_m(\lambda) |X_m| = 0 \quad a.e.$$

Furthermore we note that (cf. [5, p. 200])

(28)
$$\sum_{|h| \le M^{\alpha}} \left| (2\pi M)^{-\frac{1}{2}} \exp\left(\frac{-h^2}{2M}\right) - b_m(\lambda) \right|$$

$$= \sum_{|h| \le M^{\alpha}} \left\{ (2\pi M)^{-\frac{1}{2}} \exp\left(\frac{-h^2}{2M}\right) \right\} \left\{ O\left(\frac{(1+|h|)}{\lambda}\right) + O\left(\frac{|h|^3}{\lambda^2}\right) \right\} = O(\lambda^{-\frac{1}{2}})$$

Hence it follows from (26) that

(29)
$$\lim_{\lambda \to \infty} \lambda^{\frac{1}{4}} (\log \lambda)^{-\frac{1}{2}} \left\{ \sum_{|h| \le M^{\alpha}} \left| (2\pi M)^{-\frac{1}{2}} \exp\left(\frac{-h^2}{2M}\right) - b_m(\lambda) \right| |X_m| \right\} = 0 \text{ a.e.}$$

Define

$$X'_{n} = X_{n}^{I} [|X_{n}| \ge \epsilon_{n}^{1/4} (\log_{n})^{1/2}], \qquad X''_{n} = X_{n}^{I} [|X_{n}| \le n^{1/4} / (\log_{n})]$$

and $X_n''' = X_n - X_n' - X_n''$. Let $Y_n = X_n'' - EX_n''$. Then $EY_n = 0$, $EY_n^2 = \sigma_n^2 \rightarrow \sigma^2$, $|Y_n| \leq 2n^{\frac{1}{4}}/(\log n) = o(n^{\frac{1}{4}}(\log n)^{-\frac{1}{2}})$ and therefore satisfies the hypothesis of the lemma (cf. [8, p. 255]). Hence

(30)
$$\limsup_{\lambda \to \infty} (\pi \lambda)^{-\frac{1}{4}} (\log \lambda)^{-\frac{1}{2}} \sum_{|h| \le M^{\alpha}} Y_m \exp\left(\frac{-h^2}{2M}\right) = \sigma \quad \text{a.e.}$$

Noting that $EX_1 = 0$, we have for $|h| \leq M^{\alpha}$,

(31)
$$|EX_m''| = |EX_1 I_{[|X_1| > m^{\frac{1}{4}}/(\log m)]}| \le m^{-\frac{1}{4}}(\log m)^2 E|X_1|^3.$$

Hence

$$\lim_{\lambda \to \infty} \lambda^{\frac{1}{4}} (\log \lambda)^{-\frac{1}{2}} \sum_{|h| \le M^{\alpha}} \left(M^{-\frac{1}{2}} \exp \left(\frac{-h^2}{2M} \right) \right) |EX_m''| = 0.$$

Also by (26), $P[X'_n \neq 0 \text{ i.o.}] = 0$. Since $m^{\frac{1}{4}}/(\log m) < |X'''_m| < \epsilon m^{\frac{1}{4}}(\log m)^{\frac{1}{2}}$, we have for all large M,

$$P\left[\sum_{|h| \leq M^{\alpha}} |X_{m}^{"'}| \geq \delta M^{\frac{1}{4}} (\log M)^{\frac{1}{2}}\right] \leq P\left[\sum_{|h| \leq M^{\alpha}} \frac{|X_{m}^{"'}|}{(\epsilon m^{\frac{1}{4}} (\log M)^{\frac{1}{2}})} \geq k\right]$$

 $\leq P[X_{m}^{"''} \neq 0 \text{ for at least } k \text{ of the indices } M - M^{\alpha} \leq m \leq M + M^{\alpha}]$

$$\leq {1 + 2[M^{\alpha}] \choose b} P^{k}[|X_{1}|^{4} > (M - M^{\alpha})\{\log (M - M^{\alpha})\}^{-4}]$$

= $O(M^{\alpha k - k}(\log M)^{6k})$, by the Markov inequality.

Since $k - \alpha k > 1$, an application of the Borel-Cantelli lemma then gives

(32)
$$\limsup_{\lambda \to \infty} \lambda^{\frac{1}{4}} (\log \lambda)^{-\frac{1}{2}} \sum_{|h| \leq M^{\alpha}} M^{-\frac{1}{2}} |X_m'''| \leq \delta \quad \text{a.e.}$$

Hence we have proved (13). To prove (14), setting M = [(n+1)/(q+1)], c = (q+1)/2q, we can use a similar argument as before (cf. [5, p. 201]).

We now prove that (13) implies (12). From (13), it is clear that $X_n \to 0$ (B) a.e. By the equivalence of (9) and (11), we have $EX_1 = 0$ and $EX_1^2 < \infty$. For $n = 3, 4, \cdots$, let $d_m(n) = (\pi n)^{1/4} (\log n)^{-\frac{1/2}{2}} b_m(n), Y_n = \sum_{m=1}^n d_m(n) X_m$, $Z_n = \sum_{m=n+1}^\infty d_m(n) X_m$. Since $EX_1 = 0$ and $EX_1^2 < \infty$, it follows from Tchebychev's inequality that $Z_n \xrightarrow{P} 0$. Now Z_n is independent of (Y_1, \cdots, Y_n) and $\lim \sup_{n \to \infty} |Y_n| \leq \sigma/\sqrt{2}$ a.e. Hence by Lemma 1 of [7], $\lim \sup_{n \to \infty} |Y_n| \leq \sigma/\sqrt{2}$ a.e. But $Y_n = \sum_{m=1}^{n-1} d_m(n) X_m + d_n(n) X_n$ and $d_n(n) X_n \xrightarrow{P} 0$. Therefore applying Lemma 1 of [7] again, we have

$$\lim_{n \to \infty} \sup_{m=1}^{n-1} d_m(n) X_m \le \frac{\sigma}{\sqrt{2}} \quad \text{a.e.}$$

and

$$\limsup_{n \to \infty} |d_n(n)X_n| \le \sqrt{2}\sigma \quad \text{a.e.}$$

Hence

$$\limsup_{n \to \infty} n^{-\frac{1}{4}} (\log n)^{-\frac{1}{2}} |X_n| < \infty \quad \text{a.e.,}$$

and since this lim sup can only be ∞ a.e. of 0 a.e., it follows that $n^{-\frac{1}{4}}(\log n)^{-\frac{1}{2}}X_n \to 0$ a.e. Therefore $EX_1^4(\log^+|X_1|+1)^{-2} < \infty$. It is now obvious from (13) that $EX_1^2 = \sigma^2$. In a similar way, we can prove that (14) implies (12).

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DEPARTMENT OF MATHEMATICAL STATISTICS, COLUMBIA UNIVERSITY, NEW YORK, NEW YORK 10027