

A note on the length of maximal arithmetic progressions in random subsets

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Abstract Let $U^{(n)}$ denote the maximal length arithmetic progression in a non-uniform random subset of $\{0, 1\}^n$, where 1 appears with probability p_n . By using dependency graph and Stein-Chen method, we show that $U^{(n)} - c_n \ln n$ converges in law to an extreme type distribution with $\ln p_n = -2/c_n$. Similar result holds for $W^{(n)}$, the maximal length aperiodic arithmetic progression (mod n).

Keywords Arithmetic progression, random subset, Stein-Chen method.

§1. Introduction

An arithmetic progression is a sequence of numbers such that the difference of any two successive members of the sequence is a constant. A celebrated result of Szemerédi [5] says that any subset of integers of positive upper density contains arbitrarily long arithmetic progressions. The recent work [6] reviews some extremal problems closely related with arithmetic progressions and prime sequences, under the name of the Erdős-Turán conjectures, which are known to be notoriously difficult to solve.

Let $\xi_1, \xi_2, \dots, \xi_n$ be a uniformly chosen random word in $\{0, 1\}^n$ and Ξ_n be the random set consisting elements i such that $\xi_i = 1$. Benjamini et al. [3] studies the length of maximal arithmetic progressions in Ξ_n . Denote by $U^{(n)}$ the maximal length arithmetic progression in Ξ_n and $W^{(n)}$ the maximal length aperiodic arithmetic progression (mod n) in Ξ_n . They show, among others, that the expectation of $U^{(n)}$ and $W^{(n)}$ is roughly $2 \ln n / \ln 2$.

In view of the random graph theory [4], a natural extension of [3] is to consider non-uniform random subset of $\{0, 1\}^n$, which is the main interest of this note. Let $\xi_i = 1$ with probability p_n and $\xi_i = 0$ with probability $1 - p_n$, where $p_n \in [0, 1]$ is a function of n . Following [3], the key to our work is to construct proper dependency graph and apply the Stein-Chen method of Poisson approximation (see e.g. [1,4]). Our result implies that, in the non-uniform scenarios, the expectation of $U^{(n)}$ and $W^{(n)}$ is roughly $c_n \ln n$, with $\ln p_n = -2/c_n$. Obviously, taking $p_n \equiv 1/2$ and $c_n \equiv 2/\ln 2$, we then recover the main result of Benjamini et al.

The rest of the note is organized as follows. We present the main results in Section 2. Section 3 is devoted to the proofs.

§2. Results

Let ξ_1, ξ_2, \dots be i.i.d. random variables with $P(\xi_i = 1) = p_n$ and $P(\xi_i = 0) = 1 - p_n$. For integers $1 \leq s, t \leq n$, define

$$W_{s,t}^{(n)} := \max \left\{ 1 \leq k \leq n : \xi_s = 0, \prod_{i=1}^k \xi_{s+it} \pmod{n} = 1 \right\}. \quad (1)$$

Therefore, $W_{s,t}^{(n)}$ is the length of the longest arithmetic progression (\pmod{n}) in $\{1, 2, \dots, n\}$ starting at s with difference t . Moreover, set $W^{(n)} = \max_{1 \leq s, t \leq n} W_{s,t}^{(n)}$. Similarly, define

$$U_{s,t}^{(n)} := \max \left\{ 1 \leq k \leq \left\lfloor \frac{n-s}{t} \right\rfloor : \xi_s = 0, \prod_{i=1}^k \xi_{s+it} = 1 \right\}, \quad (2)$$

and $U^{(n)} = \max_{1 \leq s, t \leq n} U_{s,t}^{(n)}$, where $[a]$ is the integer part of a .

Theorem 2.1. Suppose that $\ln p_n = -2/c_n$ and $\alpha < c_n = o(\ln n)$ for some $\alpha > 0$. Let $\{x_n\}$ be a sequence such that $c_n \ln n + x_n \in \mathbb{Z}$ for all n , and $\inf_n x_n \geq \beta$ for some $\beta \in \mathbb{R}$. We have

$$\lim_{n \rightarrow \infty} e^{\lambda(x_n)} P(W^{(n)} \leq c_n \ln n + x_n) = 1, \quad (3)$$

where $\lambda(x) = p_n^{x+2}$. In particular, $W^{(n)}/c_n \ln n$ converges to 1 in probability, as $n \rightarrow \infty$.

Theorem 2.2. Suppose that $\ln p_n = -2/c_n$ and $\alpha < c_n = o(\ln n)$ for some $\alpha > 0$. Let $\{y_n\}$ be a sequence such that $c_n \ln n - \ln(2c_n \ln n) + y_n \in \mathbb{Z}$ for all n , and $\inf_n y_n \geq \beta$ for some $\beta \in \mathbb{R}$. We have

$$\lim_{n \rightarrow \infty} e^{\lambda(y_n)} P(U^{(n)} \leq c_n \ln n - \ln(2c_n \ln n) + y_n) = 1, \quad (4)$$

where $\lambda(x) = p_n^{x+2}$. In particular, $U^{(n)}/c_n \ln n$ converges to 1 in probability, as $n \rightarrow \infty$.

The relationship between p_n and c_n is depicted in Fig. 1. We observe that the probability p_n , by our assumptions, should within the regime $e^{-2/\alpha} < p_n = e^{-2/o(\ln n)}$ for $\alpha > 0$. For the case $p_n = o(1)$ (i.e., $c_n = o(1)$), by letting $\alpha \rightarrow 0$, we can infer that $W^{(n)} \ll \ln n$ and $U^{(n)} \ll \ln n$ whp.

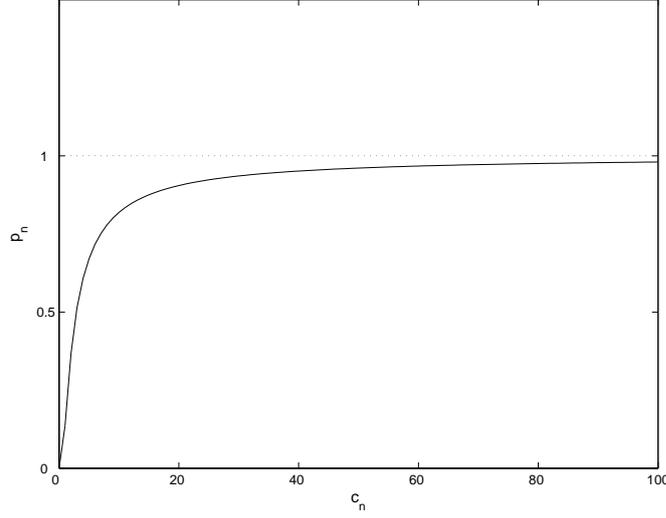
§3. Proofs

In this section, we will only consider Theorem 2.1 since the proofs are very similar. Theorem 2.1 will be proved through a series of lemmas by similar reasoning to [3] with some modifications.

For a collection of random variables $\{X_i\}_{i=1}^n$, a graph G of order n is called a dependency graph [4] of $\{X_i\}_{i=1}^n$ if for any vertex i , X_i is independent of the set $\{X_j : \text{vertices } i \text{ and } j \text{ are not adjacent}\}$. The following is a result of Arratia et al. [2], which is an instrumental version of the Stein-Chen method in numerous probabilistic combinatorial problems [1].

Lemma 3.1.([2]) Let $\{X_i\}_{i=1}^n$ be n Bernoulli random variables with $EX_i = p_i > 0$. Let G be a dependency graph of $\{X_i\}_{i=1}^n$. Set $S_n = \sum_{i=1}^n X_i$ and $\lambda = ES_n = \sum_{i=1}^n p_i$. Define

$$B_1(G) = \sum_{i=1}^n \sum_{j:j \sim i} EX_i EX_j \quad (5)$$

Figure 1: The probability p_n versus c_n .

and

$$B_2(G) = \sum_{i=1}^n \sum_{j \neq i: j \sim i} E(X_i X_j). \quad (6)$$

Let Z be a Poisson random variable with $EZ = \lambda$. For any $A \subset \mathbb{N}$, we have

$$|P(S_n \in A) - P(Z \in A)| \leq B_1(G) + B_2(G). \quad (7)$$

Fix $\varepsilon > 0$ and set $m = \lfloor (c_n + \varepsilon) \ln n \rfloor$. Define the truncated version

$$W'_{s,t}{}^{(n)} := \max \left\{ 1 \leq k \leq m : \xi_s = 0, \prod_{i=1}^k \xi_{s+it \pmod n} = 1 \right\} \quad (8)$$

and $W^{(n)} = \max_{1 \leq s, t \leq n} W'_{s,t}{}^{(n)}$. For $x \in \mathbb{R}$ define the indicator variable

$$I_{s,t}(x) = 1_{\{W'_{s,t}{}^{(n)} > c_n \ln n + x\}} \quad \text{and} \quad S(x) = \sum_{1 \leq s, t \leq n} I_{s,t}(x). \quad (9)$$

By definition, it is clear that $W^{(n)} > c_n \ln n + x$ if and only if $S(x) > 0$. Set $A(s, t) = \{s + it\}_{i=0}^m$. Fix $x \in \mathbb{R}$ such that $x < \varepsilon \ln n$. Hence, as in [3], we can construct a dependency graph G of random variables $\{I_{s,t}(x)\}_{s,t=1}^n$ by setting the vertex set $\{(s, t)\}_{s,t=1}^n$ and edges $(s_1, t_1) \sim (s_2, t_2)$ if and only if $A(s_1, t_1) \cap A(s_2, t_2) \neq \emptyset$.

The following combinatorial lemma is useful.

Lemma 3.2. ([3]) Let $D_{s,t}(k)$ be the number of pairs (s_1, t_1) such that $t \neq t_1$ and $|A(s, t) \cap A(s_1, t_1)| = k$. Then we have

$$D_{s,t}(k) \leq \begin{cases} (m+1)^2 n, & k = 1 \\ (m+1)^2 m^2, & 2 \leq k \leq \frac{m}{2} + 1 \\ 0, & k > \frac{m}{2} + 1 \end{cases} \quad (10)$$

Recall the definitions (5) and (6). Let

$$B_1(x, G) = \sum_{s_1, t_1} \sum_{\substack{s_2, t_2 \\ (s_1, t_1) \sim (s_2, t_2)}} EI_{s_1, t_1}(x) EI_{s_2, t_2}(x) \quad (11)$$

and

$$B_1(x, G) = \sum_{s_1, t_1} \sum_{\substack{(s_1, t_1) \neq (s_2, t_2) \\ (s_1, t_1) \sim (s_2, t_2)}} E[I_{s_1, t_1}(x) I_{s_2, t_2}(x)]. \quad (12)$$

Lemma 3.3. For all $x < \varepsilon \ln n$ and $\delta > 0$, we have

$$B_1(x, G) + B_2(x, G) = O(p_n^{2(x+1)} n^{\delta-1}). \quad (13)$$

Proof. From (9), we have $EI_{s,t}(x) = P(W_{s,t}^{(n)} > c_n \ln n + x) \leq p_n^{c_n \ln n + x + 1}$. Since for fixed s and t , the number of pairs (s_1, t_1) such that $|A(s, t) \cap A(s_1, t_1)| = k$ is at most $D_{s,t}(k) + 1$, we obtain by Lemma 3.2

$$\begin{aligned} B_1(x, G) &\leq \sum_{s,t} \sum_{k=1}^{m+1} (D_{s,t}(k) + 1) p_n^{2(c_n \ln n + x + 1)} \\ &\leq p_n^{2(x+1)} \cdot \frac{1}{n^4} \sum_{s,t} \left((m+1)^2 n + 1 + \sum_{k=2}^{m/2+1} ((m+1)^2 m^2 + 1) \right) \\ &= p_n^{2(x+1)} \cdot O\left(\frac{m^2 n + m^6}{n^2}\right) \\ &= O(p_n^{2(x+1)} n^{\delta-1}), \end{aligned} \quad (14)$$

for all $\delta > 0$, where the last equality holds using the assumption $c_n = o(\ln n)$.

Next, we have $E(I_{s,t}(x) I_{s_1, t_1}(x)) \leq p_n^{2(c_n \ln n + x + 1) - k}$ when $|A(s, t) \cap A(s_1, t_1)| = k$. Therefore, by Lemma 3.2

$$\begin{aligned} B_2(x, G) &\leq \sum_{s,t} \sum_{k=1}^m D_{s,t}(k) p_n^{2(c_n \ln n + x + 1) - k} \\ &\leq p_n^{2(x+1)} \cdot \frac{1}{n^4} \sum_{s,t} \left(2(m+1)^2 n + (m+1)^2 m^2 \sum_{k=2}^{m/2+1} p_n^{-k} \right). \end{aligned} \quad (15)$$

Since $c_n > \alpha > 0$, we obtain

$$\sum_{k=2}^{m/2+1} p_n^{-k} = O(p_n^{-\frac{m}{2}}) = O(n^{\frac{c_n + \varepsilon}{c_n}}). \quad (16)$$

Combining (15), (16) and the assumption $c_n = o(\ln n)$, we derive

$$\begin{aligned} B_2(x, G) &= p_n^{2(x+1)} \cdot O\left(\frac{m^2 n + m^4 n^{\frac{c_n + \varepsilon}{c_n}}}{n^2}\right) \\ &= O(p_n^{2(x+1)} n^{\delta-1}) \end{aligned} \quad (17)$$

for all $\delta > 0$. \square

The following lemma is a simplified version of Theorem 2.1.

Lemma 3.4. $W^{(n)}/c_n \ln n$ converges to 1 in probability, as $n \rightarrow \infty$; i.e., for any $\delta > 0$,

$$\lim_{n \rightarrow \infty} P \left(\left| \frac{W^{(n)}}{c_n \ln n} - 1 \right| > \delta \right) = 0. \quad (18)$$

Proof. Fix $\varepsilon > 0$, we have

$$P(W_{s,t}^{(n)} > (c_n + \varepsilon) \ln n) \leq p_n^{(c_n + \varepsilon) \ln n + 1}. \quad (19)$$

Since $c_n = o(\ln n)$, it follows that

$$P(W^{(n)} > (c_n + \varepsilon) \ln n) \leq n^2 p_n^{(c_n + \varepsilon) \ln n + 1} \leq e^{-\frac{2\varepsilon \ln n}{c_n}} \rightarrow 0 \quad (20)$$

as $n \rightarrow \infty$.

Next, let $x = -\varepsilon \ln n$ and $Z(x)$ be a Poisson random variable with

$$EZ(x) = \lambda(x) = ES(x) = n^2 p_n^{\lfloor c_n \ln n + x + 2 \rfloor} \geq e^{\frac{2\varepsilon \ln n - 4}{c_n}}. \quad (21)$$

Note that $\{W^{(n)} \leq (c_n - \varepsilon) \ln n\}$ implies that $\{W'^{(n)} \leq (c_n - \varepsilon) \ln n\}$. By Lemma 3.1 and Lemma 3.3,

$$\begin{aligned} P(W^{(n)} \leq (c_n - \varepsilon) \ln n) &\leq P(S(x) = 0) \\ &\leq B_1(x, G) + B_2(x, G) + P(Z(x) = 0) \\ &= O(p_n^{2(x+1)} n^{\delta-1} + e^{-e^{\frac{2\varepsilon \ln n - 4}{c_n}}}) \rightarrow 0, \end{aligned} \quad (22)$$

as $n \rightarrow \infty$, for $\delta > 0$ and $\varepsilon < \alpha/5$. Thus, by (20) and (22), it follows that

$$\lim_{n \rightarrow \infty} P \left(\left| \frac{W^{(n)}}{c_n \ln n} - 1 \right| > \delta \right) = 0. \quad (23)$$

for any $0 < \delta < 1/5$. \square

To prove of Theorem 2.1, we need to further refine the proof of Lemma 3.4.

Proof of Theorem 2.1. As in the proof of Lemma 3.4, let $Z(x)$ be a Poisson random variable with

$$EZ(x) = \lambda(x) = ES(x) = n^2 p_n^{\lfloor c_n \ln n + x + 2 \rfloor}. \quad (24)$$

If $c_n \ln n + x \in \mathbb{Z}$, then $\lambda(x) = p_n^{x+2}$. Recall that $W'^{(n)} > c_n \ln n + x$ if and only if $S(x) > 0$. Thus, by Lemma 3.1 and Lemma 3.3

$$\begin{aligned} |P(W'^{(n)} > c_n \ln n + x) - P(Z(x) \neq 0)| &= |P(S(x) > 0) - P(Z(x) > 0)| \\ &\leq B_1(x, G) + B_2(x, G) \\ &= O(p_n^{2(x+1)} n^{\delta-1}). \end{aligned} \quad (25)$$

Note that $x < \varepsilon \ln n$, and then we have

$$\{W^{(n)} > c_n \ln n + x\} = \{W^{(n)} > (c + \varepsilon) \ln n\} \cup \{W'^{(n)} > c_n \ln n + x\}. \quad (26)$$

Hence, by (20), (25) and (26), we obtain

$$\begin{aligned}
|P(W^{(n)} \leq c_n \ln n + x) - e^{-\lambda(x)}| &= |P(W^{(n)} > c_n \ln n + x) - P(Z(x) \neq 0)| \\
&\leq P(W^{(n)} > (c_n + \varepsilon) \ln n) \\
&\quad + |P(W^{(n)} > c_n \ln n + x) - P(Z(x) \neq 0)| \\
&\leq e^{-\frac{2\varepsilon \ln n}{c_n}} + O(p_n^{2(x+1)} n^{\delta-1}), \tag{27}
\end{aligned}$$

for $0 < \delta < 1$, where the first item on the right-hand side of (27) tends to 0 as $n \rightarrow \infty$.

Let $\{x_n\}$ be a sequence such that $c_n \ln n + x_n \in \mathbb{Z}$ for all n . If $\inf_n x_n \geq \beta \in \mathbb{R}$, then $p_n^{2(x_n+1)} n^{\delta-1} \rightarrow 0$ and $e^{\lambda(x_n)}$ is a bounded sequence. Thus, from (27) it follows that

$$|e^{\lambda(x_n)} P(W^{(n)} \leq c_n \ln n + x_n) - 1| = O\left(e^{-\frac{2\varepsilon \ln n}{c_n}} + p_n^{2(x_n+1)} n^{\delta-1}\right) \rightarrow 0, \tag{28}$$

as $n \rightarrow \infty$. \square

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