# ON INDEPENDENT SETS IN PURELY ATOMIC PROBABILITY SPACES WITH GEOMETRIC DISTRIBUTION 

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#### Abstract

We are interested in constructing concrete independent events in purely atomic probability spaces with geometric distribution. Among other facts we prove that there are uncountable many sequences of independent events.


## 1. Introduction

Let us assume a fixed ratio $r$ is given, $r \in(0,1)$. In the paper we will work with the discrete probability space $\mathbb{N}_{0}=\{0,1,2,3, \ldots\}$ and the usual geometric probability on $\mathcal{A}$ (all subsets of $\mathbb{N}_{0}$ ) defined by

$$
P_{r}(E):= \begin{cases}\frac{1-r}{r} \sum_{k \in E \backslash\{0\}} r^{k} & \text { for every set } E \in \mathcal{A}, E \backslash\{0\} \neq \emptyset, \\ 0 & \text { if } E=\{0\} \text { or } E=\emptyset .\end{cases}
$$

We are interested in studying the class of independent sets in this probability space. We are going to follow [2] and define:
$A, B$ subsets of $\mathbb{N}_{0}$ are called independent if $P(A \cap B)=P(A) P(B)$.
With this definition for every $E \subset \mathbb{N}_{0}, \mathbb{N}_{0}$ and $E$ are independent. Sets $\emptyset$ and $E$ are also independent. These are clear trivial examples. Three or more subsets of $\mathbb{N}_{0}, A_{1}, \ldots, A_{n}$ are called mutually independent or simply independent if for every choice of $k(n \geq k \geq 2)$ such sets, say $A_{i_{1}}, \ldots, A_{i_{k}}$, we have

$$
\begin{equation*}
P\left(\bigcap_{j=1}^{k} A_{i_{j}}\right)=\prod_{j=1}^{k} P\left(A_{i_{j}}\right) . \tag{1}
\end{equation*}
$$

So, for $n(n \geq 2)$ independent sets one needs to have $2^{n}-n-1$ relations as in (1) to be satisfied. An infinite family of subsets is called independent if each finite collection of these subsets is independent. Events are called trivial if their probability is 0 or 1 .

[^0]If $n \in \mathbb{N}$, then $\Omega(n)$ usually denotes the number of primes dividing $n$ counting their multiplicities (see [8]). In [1] and [6], independent families of events have been studied for finite probability spaces with uniform distribution. Eisenberg and Ghosh [6] show that the number of nontrivial independent events in such spaces cannot be more than $\Omega(m)$ where $m$ is the cardinality of the space. This result should be seen in view of the known fact (see [7, Problem 50, Section 4.1]) that if $A_{1}, A_{2}, \ldots, A_{n}$ are independent non-trivial events of a sample space $X$, then $|X| \geq 2^{n}$. One can observe that in general $\Omega(m)$ is considerably smaller than $\log _{2} m$. It is worth mentioning that according to [5] the first paper dealing with this problem in uniform finite probability spaces is [9]. In their paper, Shiflett and Shultz [9] raise the question of the existence of spaces with no non-trivial independent pairs, called dependent probability spaces. A space containing non-trivial independent events is called independent. For a uniform distributed probability space $X$, as a result of the work in [6] and [1], $X$ is dependent if $|X|$ is a prime number and independent if $|X|$ is composite. For denumerable sets $X$ one can see the construction given in [5] or look at [10, Example 1.1]. For our spaces, the Example 1.1 does not apply to and in fact, we will construct explicitly lots of independent sets.

For every $n \in \mathbb{N}$, one can consider the following space of geometric probability distribution, denoted here by $\mathcal{G}_{n}:=([n], \mathcal{P}([n]), P)$ where $P(k)=q^{k}$ with $k \in[n]:=\{1,2,3, \ldots, n\}$ and of course, $q$ is the positive solution of the equation

$$
\sum_{k=1}^{n} q^{k}=1
$$

This space is independent for every $n \geq 4$ with $n$ composites. Indeed, if $n=s t$ with $s, t \in \mathbb{N} s, t \geq 2$, one can check that the sets $A:=\{1,2,3, \ldots, s\}, B:=$ $\{1, s+1,2 s+1 \ldots,(t-1) s+1\}$ represent non-trivial independent events. To match the uniform distribution situation it would be interesting if $\mathcal{G}_{n}$ was a dependent space for every prime $n$.

The class of independent sets is important in probability theory for various reasons. Philosophically speaking, the concept of independence is at the heart of the axiomatic system of modern probability theory introduced by A. N. Kolmogorov in 1933. More recently, it was shown in [3] that two probability measures on the same space which have the same independent (pairs of) events must be equal if at least one of them is atomless. This was in fact a result of A. P. Yurachkivsky from 1989 as the same authors of [3] pointed out in the addendum to their paper that appeared in [4].

On the other hand, Szekely and Mori [10] showed that if the probability space is atomic then there may be no independent sets or one may have a sequence of such sets. The following result that appeared in [10] is a sufficient condition for the existence of a sequence of independent events in the probability space.

Theorem 1. If the range of a purely atomic probability measure contains an interval of the form $[0, \epsilon)$ for some $\epsilon>0$, then there are infinitely many independent sets in the underlying probability space.

Let us observe that if $r=1 / 2$, the probability space ( $\mathbb{N}_{0}, \mathcal{A}, P_{1 / 2}$ ) does satisfy the hypothesis of the above theorem with $\epsilon=1$ because every number in $[0,1]$ has a representation in base 2 . On the other hand, if let us say $r=1 / 3$, then the range of $P_{1 / 3}$ is the usual Cantor set which has Lebesgue measure zero, so Theorem 1 does not apply to $\left(\mathbb{N}_{0}, \mathcal{A}, P_{1 / 3}\right)$. However, we will show that there are uncountably many pairs of sets that are independent in $\left(\mathbb{N}_{0}, \mathcal{A}, P_{r}\right)$ for every $0<r<1$ (these sets do not depend on $r$ ).

## 2. Independent pairs of events for denumerable spaces

The first result we would like to include is in fact a characterization, under some restrictions of $r$, of all pairs of independent events $(A, B)$, in which one of them, say $B$, is fixed and of a certain form. This will show in particular that there are uncountably many such pairs. In order to state this theorem we need to start with a preliminary ingredient.

Lemma 1. For $m \geq 1$, consider the function given by

$$
f(x)=(2 x-1)\left(1+x^{m}\right)-x^{m} \text { for all } x \in[0,1] .
$$

The function $f$ is strictly increasing and it has unique zero in $[0,1]$ denoted by $t_{m}$. Moreover, for all $m$ we have $t_{m}>1 / 2$, the sequence $\left\{t_{m}\right\}$ is decreasing and

$$
\lim _{m \rightarrow \infty} t_{m}=\frac{1}{2}
$$

Having $t_{m}$ defined as above we can state our first theorem.
Theorem 2. For every natural number $n \geq 2$, we define the events $E:=$ $\{0, n-1\}$ and

$$
B:=\{\underbrace{1,2, \ldots, n-1}_{n-1}, \underbrace{2 n-1,2 n, \ldots, 3 n-3}_{n-1},
$$

$$
\begin{equation*}
\underbrace{4 n-3,4 n-2, \ldots, 5 n-5}_{n-1}, \ldots\} . \tag{2}
\end{equation*}
$$

Also, for an arbitrary nonempty subset $T \subset B$ we set $A:=E+T$ with the usual definition of addition of two sets in a semigroup. Then $A$ and $B$ are independent events in $\left(\mathbb{N}_{0}, \mathcal{A}, P_{r}\right)$.

Conversely, if $r<t_{m}$ (where $m=n-1$ and $t_{m}$ as in Lemma 1), $B$ is given as in (2) and $A$ forms an independent pair with $B$, then $A$ must be of the above form, i.e. $A=E+T$ for some $T \subset B$.

Proof of Lemma 1. The function $f$ has derivative $f^{\prime}(x)=2\left(1+x^{m}\right)-$ $2 m(1-x) x^{m-1}, x \in(0,1]$. For $m \geq 2$, using the Geometric-Arithmetic Mean inequality we have

$$
(m-1)(1-x) x^{m-1} \leq[\frac{(m-1)(1-x)+\underbrace{x+x+\ldots+x}_{m-1}}{m}]^{m}=\left(\frac{m-1}{m}\right)^{m}
$$

and so $m(1-x) x^{m-1} \leq\left(\frac{m-1}{m}\right)^{m-1} \leq 1$ which implies $m(1-x) x^{m-1} \leq 1$. This last inequality is true for $m=1$, too. This implies that

$$
f^{\prime}(x)=2\left(1+x^{m}\right)-2 m(1-x) x^{m-1} \geq 2 x^{m}>0
$$

for all $x \in(0,1]$. Therefore the function $f$ is strictly increasing and because $f(1 / 2)=-\frac{1}{2^{m}}<0$ and $f(1)=1>0$, by the Intermediate Values Theorem there must be a unique solution $x=t_{m}$, of the equation $f(x)=0$ in the interval $(1 / 2,1)$. Because $f\left(t_{m-1}\right)=\left(\frac{1-t_{m-1}}{1+t_{m-1}^{m-1}}\right) t_{m-1}^{m-1}>0$ we see that $t_{m}<t_{m-1}$ for all $m \geq 2$. Since $\left(2 t_{m}-1\right)\left(1+t_{m}^{m}\right)=t_{m}^{m}$ we can let $m$ go to infinity in this equality and obtain $t_{m} \rightarrow 1 / 2$.

Using Maple, we got some numerical values for the sequence $t_{m}: t_{1}=\frac{1}{\sqrt{2}} \approx$ $0.707, t_{2} \approx 0.648, t_{3} \approx 0.583, t_{4} \approx 0.539$ and, for instance, $t_{10} \approx 0.5005$.

Proof of Theorem 2. First let us check that $E_{1}=E+1=\{1, n\}$ and $B$ are independent. Since $E_{1} \cap B=\{1\}, P(\{1\})=\frac{1-r}{r} r=1-r$ and $P_{r}\left(E_{1}\right)=$ $\frac{1-r}{r}\left(r+r^{n}\right)=(1-r)\left(1+r^{n-1}\right)$, we have to show that $P_{r}(B)=\frac{1}{1+r^{n-1}}$. We have

$$
P_{r}(B)=\frac{1-r}{r}\left(\sum_{j=1}^{m} r^{j}\right)\left(\sum_{i=0}^{\infty} r^{2 m i}\right)=\frac{r-r^{m+1}}{r} \frac{1}{1-r^{2 m}}=\frac{1}{1+r^{m}}
$$

which is what we needed. Now, suppose $b \in B$ and consider $E_{b}=E+b=$ $\{b, b+n-1\}$. We notice that by the definition of $B$, the intersection $B \cap E_{b}$ is $\{b\}$. Hence, $P_{r}\left(B \cap E_{b}\right)=\frac{1-r}{r} r^{b}=(1-r) r^{c}($ with $c=b-1)$ and

$$
P_{r}(B) P_{r}\left(E_{b}\right)=\frac{1}{1+r^{m}} \frac{1-r}{r}\left(r^{b}+r^{b+m}\right)=(1-r) r^{c}
$$

Hence, $B$ and $E_{b}$ are independent for every $b \in B$.
Next we would like to observe that if $\left(F_{1}, B\right)$ and $\left(F_{2}, B\right)$ are independent pairs of events and $F_{1} \cap F_{2}=\emptyset$, then $F_{1} \cup F_{2}$ and $B$ are independent events as well.
Indeed, by the given assumption we can write

$$
\begin{aligned}
P_{r}\left(B \cap\left(F_{1} \cup F_{2}\right)\right) & =P_{r}\left(\left(B \cap F_{1}\right) \cup\left(B \cap F_{2}\right)\right)=P_{r}\left(B \cap F_{1}\right)+P_{r}\left(B \cap F_{2}\right) \\
& =P_{r}(B) P_{r}\left(F_{1}\right)+P_{r}(B) P_{r}\left(F_{2}\right)=P_{r}(B)\left(P_{r}\left(F_{1}\right)+P_{r}\left(F_{2}\right)\right) \\
& =P_{r}(B) P_{r}\left(F_{1} \cup F_{2}\right) .
\end{aligned}
$$

In fact, the above statement can be generalized to a sequence of sets $F_{k}$ which are pairwise disjoint, due to the fact that $P_{r}$ is a genuine finite measure and so it is continuous (from below and above). Then if $T \subset B$ is nonempty, $A=E+T=$ $\bigcup_{b \in B} E_{b}$ is a countable union and since $E_{b} \cap E_{b^{\prime}}=\emptyset$ for all $b, b^{\prime} \in B\left(b \neq b^{\prime}\right)$ the above observation can be applied to $\left\{E_{b}\right\}_{b \in T}$. So, we get that $B$ and $A$ are independent.

For the converse, we need the following lemma.

Lemma 2. If $L \subset \mathbb{N}_{0} \backslash B$ and the smallest element of $L$ is $s=(2 i-1) m+j$, where $i, j \in \mathbb{N}, j \leq m$, then

$$
P_{r}(L) \leq r^{s-1}-\frac{r^{2 i m}}{1+r^{m}}
$$

Proof of Lemma 2. Indeed, we have

$$
\begin{aligned}
P_{r}(L) & \leq \frac{1-r}{r}\left[\left(r^{s}+r^{s+1}+\ldots+r^{2 i m}\right)+\left(r^{(2 i+1) m+1}+\ldots\right)\right] \\
& =r^{s-1}-r^{2 i m}+r^{2 i m} P_{r}(\Omega \backslash B)=r^{s-1}-r^{2 i m}+r^{2 i m}\left(1-\frac{1}{1+r^{m}}\right) \\
& =r^{s-1}-\frac{r^{2 i m}}{1+r^{m}}
\end{aligned}
$$

So, let us assume that $r<t_{m}, B$ is as in (2) and $A$ is independent of $B$. We let $T$ be the intersection of $A$ and $B$ and put $\alpha:=P_{r}(T) / P_{r}(B)$. Also, we define $A^{\prime}:=T+\{0, n-1\}, L=A \backslash A^{\prime}$ and $L^{\prime}=A^{\prime} \backslash A$. We have clearly $L, L^{\prime} \subset \Omega \backslash B$. By the first part of our theorem $P_{r}\left(A^{\prime}\right)=\alpha$. Because $A$ and $B$ are independent $P_{r}(A)$ must be equal to $\alpha$ as well. Hence $P_{r}(A)=P_{r}\left(A^{\prime}\right)$ which attracts

$$
\begin{equation*}
\sum_{k \in L^{\prime}} r^{k}=\sum_{k \in L} r^{k} \Longleftrightarrow \sum_{k \in L \cup L^{\prime}} r^{k}=2 \sum_{k \in L^{\prime}} r^{k} \tag{3}
\end{equation*}
$$

From (3), it is clear that $L^{\prime}=\emptyset$ if an only if $L=\emptyset$ and so if $L^{\prime}$ is empty then $A=A^{\prime}$ which is what we need in order to conclude our proof. By way of contradiction, suppose $L^{\prime} \neq \emptyset$ (or equivalently $L \neq \emptyset$ ). We can assume without loss of generality that $L^{\prime}$ contains the smallest number of $L^{\prime} \cup L$, say $s$, which is written as in Lemma 2. Thus from equality (3) we have $P_{r}\left(L \cup L^{\prime}\right) \geq 2 P_{r}\left(L^{\prime}\right)$ and then by Lemma 2 we get

$$
\begin{aligned}
r^{s-1}-\frac{r^{2 i m}}{1+r^{m}} \geq 2(1-r) r^{s-1} & \Longleftrightarrow 2 r \geq 1+\frac{r^{2 i m+1-s}}{1+r^{m}} \\
& \Longleftrightarrow 2 r \geq 1+\frac{r^{n-j}}{1+r^{m}}
\end{aligned}
$$

Therefore for every $n$ and $1 \leq j \leq m$,

$$
\Longrightarrow \quad \begin{aligned}
2 r & \geq 1+\frac{r^{n-j}}{1+r^{m}} \geq 1+\frac{r^{m}}{1+r^{m}} \\
f(r) & =(2 r-1)\left(1+r^{m}\right)-r^{m} \geq 0
\end{aligned}
$$

By Lemma 1 we see that $r \geq t_{m}$ which is a contradiction. It remains that $L$ and $L^{\prime}$ must be empty and so $A=A^{\prime}$.

In the previous theorem, since $T$ is an arbitrary subset of an infinite set we obtain an uncountable family of pairs of independent sets.

Remark 1. If $r=\sqrt{\frac{1}{\phi}}$ where $\phi$ stands for the classical notation of the golden ratio (i.e. $\phi=\frac{\sqrt{5}+1}{2}$ ) $n=2, B=\{1,3,5,7, \ldots\}$ as in (2), and $A=\{1,4,6\}$, then one can check that $P_{r}(B)=\frac{1}{1+r}, P_{r}(A \cap B)=1-r, P_{r}(A)=(1-r)\left(1+r^{3}+r^{5}\right)$. So the equality $P_{r}(A \cap B)=P_{r}(A) P_{r}(B)$ is equivalent to $1+r=1+r^{3}+r^{5}$ which is the same as $r^{4}+r^{2}-1=0$. One can easily see that this last equation is satisfied by $r=\sqrt{\frac{1}{\phi}}$. Hence $A$ and $B$ are independent so clearly $A$ is not a translation of $\{0,1\}$ with a subset of $B$. Therefore the converse part in Theorem 1 cannot be extended to numbers $r \geq t_{m}$ such as $r=\sqrt{\frac{1}{\phi}}$. In fact, we believe that the constants $t_{m}$ are sharp in the sense that for all $r>t_{m}$ the converse part is false, but an argument for showing this is beyond the scope of this paper.

Remark 2. Another family of independent events which seems to have no connection with those constructed so far is given by $A=\{1,2,3,4, \ldots, n-1, n\}$ and $B=\{n, 2 n, 3 n, \ldots\}$, with $n \in \mathbb{N}$. A natural question arises as a result of this wealth of independent events: can one characterize all pairs $(A, B)$ which are independent regardless the value of the parameter $r$ ?

## 3. Three independent events

The next theorem deals with the situation in which two sets the same as in the construction of Theorem 2 and $B$ given by (2), form a triple of independent sets.

Let us observe that if $A_{1}, A_{2}$, and $B$ are mutually independent, then by Theorem 2 (at least if $\left.r \in\left(0, t_{m}\right)\right), A_{1}$ and $A_{2}$ must be given by $A_{i}=T_{i}+E$ with $T_{i} \subset B, i=1,2$. Therefore $A_{1} \cap A_{2}=\left(T_{1} \cap T_{2}\right)+E$.

Also, we note that $P_{r}\left(A_{i}\right)=P_{r}\left(T_{i}\right)\left(1+r^{n-1}\right), i=1,2$, and $P_{r}\left(A_{1} \cap A_{2}\right)=$ $P_{r}\left(T_{1} \cap T_{2}\right)\left(1+r^{n-1}\right)$. This means that the equality $P_{r}\left(A_{1} \cap A_{2}\right)=P_{r}\left(A_{1}\right) P\left(A_{2}\right)$ is equivalent to

$$
\begin{equation*}
P_{r}\left(T_{1} \cap T_{2}\right)=P_{r}\left(T_{1}\right) P_{r}\left(T_{2}\right)\left(1+r^{n-1}\right) . \tag{4}
\end{equation*}
$$

On the other hand the condition $P_{r}\left(A_{1} \cap A_{2} \cap B\right)=P_{r}\left(A_{1}\right) P_{r}\left(A_{2}\right) P(B)$ reduces to

$$
P_{r}\left(T_{1} \cap T_{2}\right)=P_{r}\left(T_{1}\right) P_{r}\left(T_{2}\right)\left(1+r^{n-1}\right)^{2} P_{r}(B)
$$

which is the same as (4). So, three sets $A_{1}, A_{2}$ and $B$ are independent if and only if (4) is satisfied. Let us notice that the condition (4) may be interpreted as a conditional probability independence relation:

$$
\begin{equation*}
P_{r}\left(T_{1} \cap T_{2} \mid B\right)=P_{r}\left(T_{1} \mid B\right) P_{r}\left(T_{2} \mid B\right) \tag{5}
\end{equation*}
$$

At this point the construction we have in Theorem 2 can be repeated. As a result, regardless of what $r$ is, we obtain an uncountable family of three events which are mutually independent in $\left(\mathbb{N}_{0}, \mathcal{A}, P_{r}\right)$.

Theorem 3. For a fixed $n \geq 3$, we consider $B$ as in (2), and pick $b \in$ $\{2, \ldots, n-1\}$ such that $2(b-1)$ divides $m=n-1(m=2(b-1) k)$. For
$F:=\{0, b-1\}$, we let

$$
B_{1}^{\prime}:=\{\underbrace{1,2, \ldots, b-1}_{b-1}, \underbrace{2 b-1,2 b, \ldots, 3 b-3}_{b-1}, \underbrace{4 b-3,4 b-2, \ldots, 5 b-5}_{b-1}
$$

$$
\begin{equation*}
\ldots, \underbrace{(2 k-2)(b-1)+1, \ldots,(2 k-1)(b-1)}_{b-1}\} \tag{6}
\end{equation*}
$$

$$
B_{1}:=B_{1}^{\prime} \cup\left(B_{1}^{\prime}+2 m\right) \cup\left(B_{1}^{\prime}+4 m\right) \cup\left(B_{1}^{\prime}+6 m\right) \cup \ldots
$$

and $T$ be a subset of $B_{1}$. Then $T_{1}:=F+T$ and $B_{1}$ are independent sets relative to the induced probability measure on $B$. Moreover, $A_{1}:=T_{1}+\{0, n-1\}, A_{2}:=$ $B_{1}+\{0, n-1\}$ and $B$ form a triple of mutually independent sets in $\left(\mathbb{N}_{0}, \mathcal{A}, P_{r}\right)$ for all $r$.

Proof. The second part of the theorem follows from the considerations we made before the theorem and from the first part. To show the first part we need to check (4) for $T_{1}$ and $T_{2}=B_{1}$. Let us remember that

$$
\begin{aligned}
B= & \{\underbrace{1,2, \ldots, n-1}_{n-1}, \underbrace{2 n-1,2 n, \ldots, 3 n-3}_{n-1}, \\
& \underbrace{4 n-3,4 n-2, \ldots, 5 n-5}_{n-1}, \ldots\}, \text { and } P_{r}(B)=\frac{1}{1+r^{m}} .
\end{aligned}
$$

We observe that $B_{1}^{\prime} \subset\{1,2, \ldots, n-1\}$ and so $B_{1} \subset B$. Let us first take into consideration the case $T=\{1\}$. Since $T_{1}=\{1, b\}$ we get $T_{1} \cap T_{2}=\{1\}$, $P_{r}\left(T_{1}\right)=(1-r)\left(1+r^{b-1}\right)$, and

$$
P_{r}\left(B_{1}\right)=P_{r}\left(B_{1}^{\prime}\right)\left(1+r^{2 m}+r^{4 m}+r^{6 m}+\ldots\right)=\frac{P_{r}\left(B_{1}^{\prime}\right)}{1-r^{2 m}}
$$

So, it remains to calculate $P_{r}\left(B_{1}^{\prime}\right)$ :

$$
\begin{gathered}
P_{r}\left(B_{1}^{\prime}\right)=\frac{1-r}{r}\left(r+r^{2}+\ldots r^{b-1}\right)\left(1+r^{2(b-1)}+r^{4(b-1)}+\ldots+r^{2(k-1)(b-1)}\right) \\
=\left(1-r^{b-1}\right) \frac{1-r^{2 k(b-1)}}{1-r^{2(b-1)}}=\frac{1-r^{m}}{1+r^{b-1}} \\
P_{r}\left(B_{1}\right)=\frac{1}{\left(1+r^{b-1}\right)\left(1+r^{m}\right)} .
\end{gathered}
$$

This shows that (4) is satisfied. In the general case, i.e. an arbitrary subset $T$ of $B_{1}$, we proceed as in the proof of Theorem 1.

## 4. Uncountable sequences of independent events

In [10], Szekely and Mori give an example of an infinite sequence of independent sets in $\left(\mathbb{N}_{0}, \mathcal{A}, P_{1 / 2}\right)$. Given an infinite sequence of independent sets $\left\{A_{n}\right\}_{n}$ we may assume that $P_{r}\left(A_{k}\right) \leq \frac{1}{2}$ and so by [10, Proposition 1.1] we must have

$$
\sum_{k=1}^{\infty} P_{r}\left(A_{k}\right)<\infty
$$

Let us observe that Theorem 2 can be applied to a different space now that can be constructed within $B$ given by (2) in terms of classes: $\widehat{\mathbb{N}}_{0}=\{\hat{0}, \hat{1}, \hat{2}, \ldots\}$ where $\hat{0}=$ $\emptyset, \hat{1}:=\{1,2, \ldots, n-1\}, \hat{2}:=\{2 n-1,2 n, \ldots, 3 n-3\}, \hat{3}:=\{4 n-3,4 n-2, \ldots, 5 n-5\}$, $\ldots$, and the probability on this space is the conditional probability as subsets of $B$.

Hence for $k \in \mathbb{N}$, one can check that

$$
P(\hat{k})=\frac{1-r^{2 m}}{r^{2 m}} r^{2 k m} \quad \text { with } \quad m=n-1
$$

It shows that this space is isomorphic to $\left(\mathbb{N}_{0}, \mathcal{A}, P_{s}\right)$ with $s=r^{2 m}$. One can check the following proposition by induction.
Proposition 1. Let $n \in \mathbb{N}, n \geq 2$. If $A_{1}, \ldots, A_{n}$ are independent in $\widehat{\mathbb{N}}_{0}$, then $A_{1}+T, A_{2}+T, \ldots, A_{n}+T$ and $B$ are indepenedent in $\left(\mathbb{N}_{0}, \mathcal{A}, P_{r}\right)$.

This construction can be then iterated indefinitely giving rise of a sequence $B$, $B_{1}, B_{2}, \ldots$, which is going to be independent and its construction is in terms of a sequence ( $n, n_{1}, n_{2}, \ldots$ ) with $n_{k} \geq 2$. As a result, we have a uncountable way of constructing sequences of independent sets. This construction coincides with the one in [10] if $n_{k}=2$ for all $k \in \mathbb{N}$.

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