

Randomness and Complexity of Sequences over Finite Fields

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- Introduction
- A hierarchy of complexities
- Complexity and random sequences
- ∞ -distribution
- Complexity and ∞ -distribution
- Expansion complexity

Introduction

Sequences of random (or pseudorandom) elements are needed for various applications (cryptography, simulation methods, probabilistic algorithms,...). To assess randomness, we can use [complexity-theoretic properties](#) and [statistical properties](#) of sequences.

We focus on sequences over a finite field \mathbb{F}_q with q an arbitrary prime power.

A hierarchy of complexities

We first consider finite sequences. Let S_N be a finite sequence over \mathbb{F}_q of length N .

Def. 1. The **linear complexity** $L(S_N)$ of S_N is the least order of a linear recurrence relation over \mathbb{F}_q that generates S_N (with $L(S_N) = 0$ if S_N is the zero sequence).

Equivalently, $L(S_N)$ is the length of the shortest linear feedback shift register (FSR) that generates S_N .

Def. 2. The **quadratic complexity** $Q(S_N)$ of S_N is the length of the shortest FSR with feedback function of degree ≤ 2 that generates S_N .

We always have $0 \leq Q(S_N) \leq L(S_N) \leq N$. Similarly, we can define the cubic complexity, quartic complexity,.... At the end of this chain of complexities is the following one.

Def. 3 (Jansen, 1989). The **maximum order complexity** $M(S_N)$ of S_N is the length of the shortest (arbitrary) FSR that generates S_N .

$M(S_N)$ is a lower bound for all polynomial complexities.

$M(S_N)$ can be computed by **Blumer's algorithm** in graph theory, by reformulating the problem of computing $M(S_N)$ as a problem for directed acyclic graphs.

Recall that $L(S_N)$ can be efficiently computed by the **Berlekamp-Massey algorithm**.

In a different vein, we have the Kolmogorov complexity which is important in theoretical computer science.

Def. 4 (Kolmogorov, 1965). The Kolmogorov complexity $K(S_N)$ of S_N is the length of the shortest Turing machine program (in an alphabet of size q , say) that generates S_N .

We always have $K(S_N) \leq N + c$ with an absolute constant c (just list the terms of S_N).

Actually, most results hold for the more refined self-delimiting Kolmogorov complexity. These two Kolmogorov complexities differ at most by an additive term of logarithmic order.

For an infinite sequence S over \mathbb{F}_q we define **complexity profiles**. Let S_N denote the finite sequence consisting of the first N terms of S . Write $L_N(S) = L(S_N)$, etc.

Def. 5. The sequence $L_1(S), L_2(S), \dots$ is called the **linear complexity profile** of S .

Similarly for the other complexities. Each complexity profile is a nondecreasing sequence of nonnegative integers.

Complexity and random sequences

Let \mathbb{F}_q^∞ denote the **sequence space** over \mathbb{F}_q and let μ_q be the **natural probability measure** on \mathbb{F}_q^∞ , i.e., the complete product measure of the uniform probability measure on \mathbb{F}_q (the latter measure assigns measure q^{-1} to each element of \mathbb{F}_q).

We say that a property of sequences S over \mathbb{F}_q is satisfied **μ_q -almost everywhere** if the property holds for all $S \in \mathcal{R} \subseteq \mathbb{F}_q^\infty$ with $\mu_q(\mathcal{R}) = 1$. Such properties can be viewed as typical properties of random sequences.

The behavior of the linear complexity profile of random sequences is well understood.

Theorem 1 (H.N., 1988). We have μ_q -almost everywhere

$$\lim_{N \rightarrow \infty} \frac{L_N(S)}{N} = \frac{1}{2}. \quad (1)$$

More precisely, we have μ_q -almost everywhere

$$\limsup_{N \rightarrow \infty} \frac{|L_N(S) - N/2|}{\log_q N} = \frac{1}{2}.$$

Problem 1. Determine the behavior of the quadratic (cubic,...) complexity profile of random sequences.

Theorem 2 (Jansen, 1989). For the expected value of the maximum order complexity we have

$$\lim_{N \rightarrow \infty} \frac{E(M_N(S))}{\log_q N} = 2.$$

Problem 2. Determine whether $\lim_{N \rightarrow \infty} M_N(S) / \log_q N$ exists for random sequences, i.e., μ_q -almost everywhere.

Theorem 3 (Martin-Löf, 1966). For every $\varepsilon > 0$ we have μ_q -almost everywhere

$$K_N(S) \geq N - \log_q N - (1 + \varepsilon) \log_q \log N$$

for all sufficiently large N . Moreover, for every sequence S there are infinitely many N for which

$$K_N(S) \leq N - \log_q N.$$

∞ -distribution

This concept was popularized by the book of Knuth, *The Art of Computer Programming, Vol. 2*, at least in the binary case.

Let S be the sequence s_1, s_2, \dots over \mathbb{F}_q . For integers $k \geq 1$ and $n \geq 1$, write

$$\mathbf{s}_n^{(k)} = (s_n, s_{n+1}, \dots, s_{n+k-1}) \in \mathbb{F}_q^k.$$

For an integer $N \geq 1$ and a given block $\mathbf{b} = (b_0, b_1, \dots, b_{k-1}) \in \mathbb{F}_q^k$, let

$$A(\mathbf{b}, S; N) = \#\{1 \leq n \leq N : \mathbf{s}_n^{(k)} = \mathbf{b}\}.$$

Def. 6. For an integer $k \geq 1$, the sequence S over \mathbb{F}_q is k -distributed in \mathbb{F}_q if

$$\lim_{N \rightarrow \infty} \frac{A(\mathbf{b}, S; N)}{N} = \frac{1}{q^k} \quad \text{for all } \mathbf{b} \in \mathbb{F}_q^k.$$

The sequence S over \mathbb{F}_q is ∞ -distributed (or completely uniformly distributed) in \mathbb{F}_q if it is k -distributed in \mathbb{F}_q for all $k \geq 1$.

Theorem 4. Sequences over \mathbb{F}_q are ∞ -distributed in \mathbb{F}_q μ_q -almost everywhere.

Complexity and ∞ -distribution

∞ -distribution is clearly a statistical randomness property. Does this property imply, or is it implied by, complexity-theoretic randomness properties?

Theorem 5 (Martin-Löf, 1971). If $K_N(S) \geq N - c$ for some constant c and infinitely many N , then S is ∞ -distributed.

Thus, Kolmogorov complexity is a sufficiently strong concept to yield ∞ -distribution.

How about linear complexity? Does the limit relation (1) imply ∞ -distribution? Answer: **no**.

Theorem 6 (H.N., 2012). We can construct a sequence S over \mathbb{F}_q which satisfies (1), but is not 1-distributed in \mathbb{F}_q .

Proof. Consider the sequence S whose generating function (in the variable x^{-1}) is the power series $\sigma \in \mathbb{F}_q[[x^{-1}]]$ with continued fraction expansion

$$\sigma = 1/(x + 1/(x + \cdots)),$$

where all partial quotients are equal to x .

Problem 3. If $M_N(S) \approx 2 \log_q N$, check whether this implies that S is ∞ -distributed.

For Theorem 6 there is also a result in the opposite direction.

Theorem 7 (H.N., 2012). We can construct a sequence S over \mathbb{F}_q which is ∞ -distributed, but for which $L_N(S) = O((\log N)^2)$.

Problem 4. If S is ∞ -distributed, then it is trivial that

$$\lim_{N \rightarrow \infty} L_N(S) = \infty.$$

Determine the minimal growth rate of $L_N(S)$ as $N \rightarrow \infty$.

Presumably it is smaller than $(\log N)^2$ in Theorem 7.

Expansion complexity

Let S be the sequence s_1, s_2, \dots over \mathbb{F}_q . Let its generating function be

$$\gamma(x) = \sum_{i=1}^{\infty} s_i x^{i-1} \in \mathbb{F}_q[[x]].$$

Def. 7 (Diem, preprint 2011). The **expansion complexity** $E_N(S)$ is the least degree of a nonzero polynomial $h(x, y) \in \mathbb{F}_q[x, y]$ with

$$h(x, \gamma) \equiv 0 \pmod{x^N}.$$

Remark. If $s_i = 0$ for $1 \leq i \leq N$, define $E_N(S) = 0$. Then $0 \leq E_N(S) \leq N$ for any sequence S .

We can also introduce the **expansion complexity profile** $E_1(S), E_2(S), \dots$ of S .

Problem 5. Determine the relationship between the maximum order complexity $M_N(S)$ and the expansion complexity $E_N(S)$.

Problem 6. Determine the behavior of the expansion complexity profile of random sequences. A partial result is that μ_q -almost everywhere $E_N(S)$ grows at least at the rate $N^{1/2}$.