

**An alternative presentation of the analysis
of Nisan’s pseudorandom generator of space-bounded machines**

The following description of the analysis of *Nisan’s construction* [3] is inspired by [1], and differs from the presentation in [2, Sec. 8.4.2.1]. Specifically, the construction is the same, but rather than being analyzed by looking at contracted versions of the distinguisher (see [2, p. 321]), we consider a sequence of distributions that this distinguisher may examine.

Our description is meant to replace the text in [2, pp. 320-321], which means that it relies on the definitions and notations of [2, Sec. 8.4].

Sketch of the proof of [2, Thm. 8.21]. The main technical tool used in this proof is the “mixing property” of pairwise independent hash functions (see [2, Apdx. D.2]). A family of functions H_n , which map $\{0, 1\}^n$ to itself, is called *mixing* if for every pair of subsets $A, B \subseteq \{0, 1\}^n$ for all but very few (i.e., $\exp(-\Omega(n))$ fraction) of the functions $h \in H_n$, it holds that

$$\Pr[U_n \in A \wedge h(U_n) \in B] \approx \frac{|A|}{2^n} \cdot \frac{|B|}{2^n} \quad (1)$$

where the approximation is up to an additive term of $\exp(-\Omega(n))$. (See the generalization of [2, Lem.D.4], which implies that $\exp(-\Omega(n))$ can be set to $2^{-n/3}$.)

We may assume, without loss of generality, that $s(k) = \Omega(\sqrt{k})$, and thus $\ell \stackrel{\text{def}}{=} \ell(k) \leq 2^{s(k)}$ holds. For any $s(k)$ -space distinguisher D_k as in [2, Def. 8.20], we consider its computation when fed with ℓ -long sequences that are taken from various distributions. The first distribution is the uniform distribution over $\{0, 1\}^n$; that is, $U_\ell \equiv U_n^{(1)}U_n^{(2)} \dots U_n^{(\ell)}$, where $\ell' = \ell/n$ and the $U_n^{(j)}$ ’s are independent random variables each uniformly distributed over $\{0, 1\}^n$. The last distribution will be the one produced by the pseudorandom generator, and a generic (hybrid) distribution will have the form

$$\mathcal{H}_i \stackrel{\text{def}}{=} G_i(U_n^{(1)})G_i(U_n^{(2)}) \dots G_i(U_n^{((\ell'/2^i)-1)})G_i(U_n^{(\ell'/2^i)})$$

where G_i is an arbitrary mapping of n -bit strings to $2^i \cdot n$ -bit strings (and $i \in \{0, 1, \dots, \log_2 \ell'\}$).¹ That is, the i^{th} hybrid is obtained by applying $G_i : \{0, 1\}^n \rightarrow \{0, 1\}^{2^i \cdot n}$ to a sequence of $\ell'/2^i$ independently and uniformly distributed n -bit long strings. Note that $\mathcal{H}_0 \equiv U_\ell$ (with G_0 being the identity function), whereas $\mathcal{H}_{\log_2 \ell'} = G_{\log_2 \ell'}(U_n)$ is a distribution that is obtained by stretching random n -bit long strings into ℓ -bit long strings.

The key observation is that, for every i , the automata D_k cannot distinguish between \mathcal{H}_i and a distribution obtained by selecting a typical $h \in H_n$ and outputting

$$G_i(U_n^{(1)})G_i(h(U_n^{(1)})) \dots G_i(U_n^{(\ell'/2^{i+1})})G_i(h(U_n^{(\ell'/2^{i+1})})).$$

Note that the foregoing distribution is similar to \mathcal{H}_i , except that the $2j^{\text{th}}$ block is set to $G_i(h(U_n^{(j)}))$ rather than to $G_i(U_n^{(2j)})$ as in \mathcal{H}_i .² On the other hand, the foregoing distribution has the form of \mathcal{H}_{i+1} (i.e., let $G_{i+1}(s) = G_i(s)G_i(h(s))$). To prove that this replacement has little effect on the movement of D_k , we consider an arbitrary pair of vertices, u and v in layers $(2j - 2) \cdot 2^i \cdot n$

¹Indeed, while at this point G_i is to be thought of as arbitrary, later we shall use specific choices of G_i .

²Setting the $(2j - 1)^{\text{st}}$ block to $G_i(U_n^{(j)})$ rather than to $G_i(U_n^{(2j-1)})$ as in \mathcal{H}_i is immaterial.

and $(2j - 1) \cdot 2^i \cdot n$, respectively, and denote by $L_{u,v} \subseteq \{0, 1\}^n$ the set of the n -bit long strings s such that the automaton moves from vertex u to vertex v upon reading $G_i(s)$ (from locations $(2j - 2) \cdot 2^i \cdot n + 1, \dots, (2j - 1) \cdot 2^i \cdot n$ in its input). Similarly, for a vertex w at layer $2j \cdot 2^i \cdot n$, we let $L'_{v,w}$ denote the set of the strings s such that D_k moves from v to w upon reading $G_i(s)$. By Eq. (1), for all but very few of the functions $h \in H_n$, it holds that

$$\Pr[U_n \in L_{u,v} \wedge h(U_n) \in L'_{v,w}] \approx \Pr[U_n \in L_{u,v}] \cdot \Pr[U_n \in L'_{v,w}], \quad (2)$$

where “very few” and \approx are as in Eq. (1). Thus, for all but $\exp(-\Omega(n))$ fraction of the choices of $h \in H_n$, replacing the coins in the second transition (i.e., the transition from layer $(2j - 1) \cdot 2^i \cdot n$ to layer $2j \cdot 2^i \cdot n$) with the value of h applied to the outcomes of the coins used in the first transition (i.e., the transition from layer $(2j - 2) \cdot 2^i \cdot n$ to $(2j - 1) \cdot 2^i \cdot n$), approximately maintains the probability that D_k moves from u to w via v . Using a union bound (on all triples (u, v, w) as in the foregoing), we note that, for all but $2^{3s(k)} \cdot \ell' \cdot \exp(-\Omega(n))$ fraction of the choices of $h \in H_n$, the foregoing replacement approximately maintains the probability that D_k moves through any specific triple of vertices that are $2^i \cdot n$ apart. (We stress that the same h can be used in all these approximations.)

Thus, at the cost of extra $|h|$ random bits, we can reduce the number of true random coins used in transitions on D_k by a factor of two, without significantly affecting the final decision of D_k (where again we use the fact that $\ell' \cdot \exp(-\Omega(n)) < \exp(-\Omega(n))$, which implies that the approximation errors do not accumulate to too much). That is, fixing a good h (i.e., one that provides a good approximation to the transition probability over all $2^{3s(k)} \cdot \ell'$ triples), we can replace the amount of randomness in the hybrid (from $\ell'/2^i \cdot n$ in \mathcal{H}_i to $\ell'/2^{i+1} \cdot n$ in \mathcal{H}_{i+1} , which is defined based on this h), while approximately preserving the acceptance probability of D_k (i.e., $\Pr[D_k(\mathcal{H}_i) = 1] \approx \Pr[D_k(\mathcal{H}_{i+1}) = 1]$).

Applying the foregoing process can for $i = 0, \dots, \log_2 \ell' - 1$, we repeatedly reduce the randomness of the hybrid by a factor of two, by randomly selecting (and fixing) a new hash function. Thus, repeating the process for a logarithmic (in ℓ') number of times, we obtain a distribution that depends on n random bits, at which point we stop. In total, we have used $t \stackrel{\text{def}}{=} \log_2 \ell' < \log_2 \ell(k)$ random hash functions, denoted $h^{(1)}, \dots, h^{(t)}$. This means that we can generate a (pseudorandom) sequence that fools the original D_k by using a seed of length $n + t \cdot \log_2 |H_n|$ (see [2, Fig. 8.3] and [2, Exer. 8.28]). Using $n = \Theta(s(k))$ and an adequate family H_n (e.g., [2, Const. D.3]), we obtain the desired $(s, 2^{-s})$ -pseudorandom generator, which indeed uses a seed of length $O(s(k) \cdot \log_2 \ell(k)) = k$. \square

References

- [1] Eshan Chattopadhyay, Pooya Hatami, Kaave Hosseini, and Shachar Lovett. Pseudorandom Generators from Polarizing Random Walks *ECCC*, TR18-015, 2018
- [2] Oded Goldreich. *Computational Complexity: A Conceptual Perspective*. Cambridge University Press, 2008.
- [3] Noam Nisan. Pseudorandom Generators for Space Bounded Computation. *Combinatorica*, Vol. 12 (4), pages 449–461, 1992. Preliminary version in *22nd STOC*, 1990.