#### YALE UNIVERSITY DEPARTMENT OF COMPUTER SCIENCE

CPSC 467b: Cryptography and Computer Security

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# **Pseudorandom Sequence Generation**

## **1** Distinguishability and Bit Prediction

Let D be a probability distribution on a finite set  $\Omega$ . Then D associates a probability  $P_D(\omega)$  with each each element  $\omega \in \Omega$ . We will also regard D as a random variable that ranges over  $\Omega$  and assumes value  $\omega \in \Omega$  with probability  $P_D(\omega)$ .

**Definition:** An  $(S, \ell)$ -pseudorandom sequence generator (PRSG) is a function  $f: S \to \{0, 1\}^{\ell}$ . (We generally assume  $2^{\ell} \gg |S|$ .) More properly speaking, a PRSG is a randomness amplifier. Given a random, uniformly distributed seed  $s \in S$ , the PRSG yields the pseudorandom sequence z = f(s). We use S also to denote the uniform distribution on seeds, and we denote the induced probability distribution on pseudorandom sequences by f(S).

The goal of an  $(S, \ell)$ -PRSG is to generate sequences that "look random", that is, are computationally indistinguishable from sequences drawn from the uniform distribution U on length- $\ell$ sequences. Informally, a probabilistic algorithm A that always halts "distinguishes" X from Y if its output distribution is "noticeably differently" depending whether its input is drawn at random from X or from Y. Formally, there are many different kinds of distinguishably. In the following definition, the only aspect of A's behavior that matters is whether or not it outputs "1".

**Definition:** Let  $\epsilon > 0$ , let X, Y be distributions on  $\{0, 1\}^{\ell}$ , and let A be a probabilistic algorithm. Algorithm A naturally induces probability distributions A(X) and A(Y) on the set of possible outcomes of A. We say that A  $\epsilon$ -distinguishes X and Y if

$$|P[A(X) = 1] - P[A(Y) = 1]| \ge \epsilon,$$

and we say X and Y are  $\epsilon$ -indistinguishable by A if A does not distinguish them.

A natural notion of randomness for PRSG's is that the next bit should be unpredictable given all of the bits that have been generated so far.

**Definition:** Let  $\epsilon > 0$  and  $1 \le i \le \ell$ . A probabilistic algorithm  $N_i$  is an  $\epsilon$ -next bit predictor for bit *i* of *f* if

$$P[N_i(Z_1, \dots, Z_{i-1}) = Z_i] \ge \frac{1}{2} + \epsilon$$

where  $(Z_1, \ldots, Z_\ell)$  is distributed according to f(S).

A still stronger notion of randomness for PRSG's is that each bit i should be unpredictable, even if one is given all of the bits in the sequence except for bit i.

**Definition:** Let  $\epsilon > 0$  and  $1 \le i \le \ell$ . A probabilistic algorithm  $B_i$  is an  $\epsilon$ -strong bit predictor for bit i of f if

$$P[B_i(Z_1, \dots, Z_{i-1}, Z_{i+1}, \dots, Z_\ell) = Z_i] \ge \frac{1}{2} + \epsilon$$

where  $(Z_1, \ldots, Z_\ell)$  is distributed according to f(S).

The close relationship between distinguishability and the two kinds of bit prediction is established in the following theorems.

**Theorem 1** Suppose  $\epsilon > 0$  and  $N_i$  is an  $\epsilon$ -next bit predictor for bit i of f. Then algorithm  $B_i$  is an  $\epsilon$ -strong bit predictor for bit i of f, where algorithm  $B_i(z_1, \ldots, z_{i-1}, z_{i+1}, \ldots, z_{\ell})$  simply ignores its last  $\ell - i$  inputs and computes  $N_i(z_1, \ldots, z_{i-1})$ .

Proof: Obvious from the definitions.

Let  $\mathbf{x} = (x_1, \dots, x_\ell)$  be a vector. We define  $\mathbf{x}^i$  to be the result of deleting the  $i^{\text{th}}$  element of  $\mathbf{x}$ , that is,  $\mathbf{x}^i = (x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_\ell)$ .

**Theorem 2** Suppose  $\epsilon > 0$  and  $B_i$  is an  $\epsilon$ -strong bit predictor for bit i of f. Then algorithm A  $\epsilon$ -distinguishes f(S) and U, where algorithm A on input  $\mathbf{x}$  outputs 1 if  $B_i(\mathbf{x}^i) = x_i$  and outputs 0 otherwise.

**Proof:** By definition of A,  $A(\mathbf{x}) = 1$  precisely when  $B_i(\mathbf{x}^i) = x_i$ . Hence,  $P[A(f(S)) = 1] \ge 1/2 + \epsilon$ . On the other hand, for  $\mathbf{r} = U$ ,  $P[B_i(\mathbf{r}^i) = r_i] = 1/2$  since  $r_i$  is a uniformly distributed bivalued random variable that is independent of  $\mathbf{r}^i$ . Thus, P[A(U) = 1] = 1/2, so  $A \epsilon$ -distinguishes f(S) and U.

For the final step in the 3-way equivalence, we have to weaken the error bound.

**Theorem 3** Suppose  $\epsilon > 0$  and algorithm  $A \epsilon$ -distinguishes f(S) and U. For each  $1 \le i \le \ell$  and  $c \in \{0, 1\}$ , define algorithm  $N_i^c(z_1, \ldots, z_{i-1})$  as follows:

- 1. Flip coins to generate  $\ell i + 1$  random bits  $r_i, \ldots, r_\ell$ .
- 2. Let  $v = \begin{cases} 1 \text{ if } A(z_1, \dots, z_{i-1}, r_i, \dots, r_{\ell}) = 1; \\ 0 \text{ otherwise.} \end{cases}$
- 3. Output  $v \oplus r_i \oplus c$ .

Then there exist m and c for which algorithm  $N_m^c$  is an  $\epsilon/\ell$ -next bit predictor for bit m of f.

**Proof:** Let  $(Z_1, \ldots, Z_\ell) = f(S)$  and  $(R_1, \ldots, R_\ell) = U$  be random variables, and let  $D_i = (Z_1, \ldots, Z_i, R_{i+1}, \ldots, R_\ell)$ .  $D_i$  is the distribution on  $\ell$ -bit sequences that results from choosing the first *i* bits according to f(S) and choosing the last  $\ell - i$  bits uniformly. Clearly  $D_0 = U$  and  $D_\ell = f(S)$ .

Let  $p_i = P[A(D_i) = 1], 0 \le i \le \ell$ . Since  $A \epsilon$ -distinguishes  $D_\ell$  and  $D_0$ , we have  $|p_\ell - p_0| \ge \epsilon$ . Hence, there exists  $m, 1 \le m \le \ell$ , such that  $|p_m - p_{m-1}| \ge \epsilon/\ell$ . We show that the probability that  $N_m^c$  correctly predicts bit m for f is  $1/2 + (p_m - p_{m-1})$  if c = 1 and  $1/2 + (p_{m-1} - p_m)$  if c = 0. It will follow that either  $N_m^0$  or  $N_m^1$  correctly predicts bit m with probability  $1/2 + |p_m - p_{m-1}| \ge \epsilon/\ell$ .

Consider the following experiments. In each, we choose an  $\ell$ -tuple  $(z_1, \ldots, z_\ell)$  according to f(S) and an  $\ell$ -tuple  $(r_1, \ldots, r_\ell)$  according to U.

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Experiment E_0: Succeed if A(z_1, \ldots, z_{m-1}, \boxed{z_m}, r_{m+1}, \ldots, r_{\ell}) = 1.

Experiment E_1: Succeed if A(z_1, \ldots, z_{m-1}, \boxed{\neg z_m}, r_{m+1}, \ldots, r_{\ell}) = 1.

Experiment E_2: Succeed if A(z_1, \ldots, z_{m-1}, \boxed{r_m}, r_{m+1}, \ldots, r_{\ell}) = 1.
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Let  $q_j$  be the probability that experiment  $E_j$  succeeds, where j = 0, 1, 2. Clearly  $q_2 = (q_0 + q_1)/2$  since  $r_m = z_m$  is equally likely as  $r_m = \neg z_m$ .

Now, the inputs to A in experiment  $E_0$  are distributed according to  $D_m$ , so  $p_m = q_0$ . Also, the inputs to A in experiment  $E_2$  are distributed according to  $D_{m-1}$ , so  $p_{m-1} = q_2$ . Differencing, we get  $p_m - p_{m-1} = q_0 - q_2 = (q_0 - q_1)/2$ .

We now analyze the probability that  $N_m^c$  correctly predicts bit m of f(S). Assume without loss of generality that A's output is in  $\{0, 1\}$ . A particular run of  $N_m^c(z_1, \ldots, z_{m-1})$  correctly predicts  $z_m$  if

$$A(z_1, \dots, z_{m-1}, \boxed{r_m}, \dots, r_\ell) \oplus r_m \oplus c = z_m$$
<sup>(1)</sup>

If  $r_m = z_m$ , (1) simplifies to

$$A(z_1, \dots, z_{m-1}, \boxed{z_m}, \dots, r_\ell) = c,$$
<sup>(2)</sup>

and if  $r_m = \neg z_m$ , (1) simplifies to

$$A(z_1, \dots, z_{m-1}, \boxed{\neg z_m}, \dots, r_\ell) = \neg c.$$
(3)

Let  $OK_m^c$  be the event that  $N_m^c(Z_1, \ldots, Z_{m-1}) = Z_m$ , i.e., that  $N_m^c$  correctly predicts bit m for f. From (2), it follows that

$$P[OK_m^c | R_m = Z_m] = \begin{cases} q_0 & \text{if } c = 1\\ (1 - q_0) & \text{if } c = 0 \end{cases}$$

for in that case the inputs to A are distributed according to experiment  $E_0$ . Similarly, from (3), it follows that

$$P[OK_m^c \mid R_m = \neg Z_m] = \begin{cases} q_1 & \text{if } \neg c = 1\\ (1 - q_1) & \text{if } \neg c = 0 \end{cases}$$

for in that case the inputs to A are distributed according to experiment  $E_1$ . Since  $P[R_m = Z_m] = P[R_m = \neg Z_m] = 1/2$ , we have

$$P[OK_m^c] = \frac{1}{2} \cdot P[OK_m^c | R_m = Z_m] + \frac{1}{2} \cdot P[OK_m^c | R_m = \neg Z_m]$$
$$= \begin{cases} q_0/2 + (1 - q_1)/2 = 1/2 + p_m - p_{m-1} & \text{if } c = 1\\ q_1/2 + (1 - q_0)/2 = 1/2 + p_{m-1} - p_m & \text{if } c = 0. \end{cases}$$

Thus,  $P[OK_m^c] = 1/2 + |p_m - p_{m-1}| \ge \epsilon/\ell$  for some  $c \in \{0, 1\}$ , as desired.

## 2 BBS Generator

We now give a PRSG due to Blum, Blum, and Shub for which the problem distinguishing its outputs from the uniform distribution is closely related to the difficulty of determining whether a number with Jacobi symbol 1 is a quadratic residue modulo a certain kind of composite number called a Blum integer. The latter problem is believed to be computationally hard. First some background.

A Blum prime is a prime number p such that  $p \equiv 3 \pmod{4}$ . A Blum integer is a number n = pq, where p and q are Blum primes. Blum primes and Blum integers have the important property that every quadratic residue a has a square root y which is itself a quadratic residue. We call such a y a principal square root of a and denote it by  $\sqrt{a}$ .

**Lemma 4** Let p be a Blum prime, and let a be a quadratic residue modulo p. Then  $y = a^{(p+1)/4} \mod p$  is a principal square root of a modulo p.

**Proof:** We must show that, modulo p, y is a square root of a and y is a quadratic residue. By the Euler criterion [Theorem 2, handout 15], since a is a quadratic residue modulo p, we have  $a^{(p-1)/2} \equiv 1 \pmod{p}$ . Hence,  $y^2 \equiv (a^{(p+1)/4})^2 \equiv aa^{(p-1)/2} \equiv a \pmod{p}$ , so y is a square root of a modulo p. Applying the Euler criterion now to y, we have

$$y^{(p-1)/2} \equiv \left(a^{(p+1)/4}\right)^{(p-1)/2} \equiv \left(a^{(p-1)/2}\right)^{(p+1)/4} \equiv 1^{(p+1)/4} \equiv 1 \pmod{p}.$$

Hence, y is a quadratic residue modulo p.

**Theorem 5** Let n = pq be a Blum integer, and let a be a quadratic residue modulo n. Then a has four square roots modulo n, exactly one of which is a principal square root.

**Proof:** By Lemma 4, a has a principal square root u modulo p and a principal square root v modulo q. Using the Chinese remainder theorem, we can find x that solves the equations

$$x \equiv \pm u \pmod{p}$$
$$x \equiv \pm v \pmod{q}$$

for each of the four choices of signs in the two equations, yielding 4 square roots of a modulo n. It is easily shown that the x that results from the +, + choice is a quadratic residue modulo n, and the others are not.

From Theorem 4, it follows that the mapping  $b \mapsto b^2 \mod n$  is a bijection from the set of quadratic residues modulo n onto itself. (A *bijection* is a function that is 1–1 and onto.)

**Definition:** The Blum-Blum-Shub generator BBS is defined by a Blum integer n = pq and an integer  $\ell$ . It is a  $(\mathbf{Z}_n^*, \ell)$ -PRSG defined as follows: Given a seed  $s_0 \in \mathbf{Z}_n^*$ , we define a sequence  $s_1, s_2, s_3, \ldots, s_\ell$ , where  $s_i = s_{i-1}^2 \mod n$  for  $i = 1, \ldots, \ell$ . The  $\ell$ -bit output sequence is  $b_1, b_2, b_3, \ldots, b_\ell$ , where  $b_i = s_i \mod 2$ .

Note that any  $s_m$  uniquely determines the entire sequence  $s_1, \ldots, s_\ell$  and corresponding output bits. Clearly,  $s_m$  determines  $s_{m+1}$  since  $s_{m+1} = s_m^2 \mod n$ . But likewise,  $s_m$  determines  $s_{m-1}$  since  $s_{m-1} = \sqrt{s_m}$ , the principal square root of  $s_m$  modulo n, which is unique by Theorem 5.

#### **3** Security of BBS

**Theorem 6** Suppose there is a probabilistic algorithm A that  $\epsilon$ -distinguishes BBS( $\mathbf{Z}_n^*$ ) from U. Then there is a probabilistic algorithm Q(x) that correctly determines with probability at least  $\epsilon' = \epsilon/\ell$  whether or not an input  $x \in \mathbf{Z}_n^*$  with Jacobi symbol  $(\frac{x}{n}) = 1$  is a quadratic residue modulo n.

**Proof:** From A, one easily constructs an algorithm  $\hat{A}$  that reverses its input and then applies A.  $\hat{A} \epsilon$ -distinguishes the reverse of  $BBS(\mathbf{Z}_n^*)$  from U. By Theorem 3, there is an  $\epsilon'$ -next bit predictor  $N_m$  for bit  $\ell - m + 1$  of BBS reversed. Thus,  $N_m(b_\ell, b_{\ell-1}, \ldots, b_{m+1})$  correctly outputs  $b_m$  with probability at least  $1/2 + \epsilon'$ , where  $(b_1, \ldots, b_\ell)$  is the (unreversed) output from BBS( $\mathbf{Z}_n^*$ ). We now describe algorithm Q(x), assuming  $x \in \mathbb{Z}_n^*$  and  $\left(\frac{x}{n}\right) = 1$ . Using x as a seed, compute  $(b_1, \ldots, b_\ell) = BBS(x)$  and let  $b = N_m(b_{\ell-m}, b_{\ell-m-1}, \ldots, b_1)$ . Output "quadratic residue" if  $b = x \mod 2$  and "non-residue" otherwise.

To see that this works, observe first that  $N_m(b_{\ell-m}, b_{\ell-m-1}, \ldots, b_1)$  correctly predicts  $b_0$  with probability at least  $1/2+\epsilon'$ , where  $b_0 = (\sqrt{x^2} \mod n) \mod 2$ . This is because we could in principle let  $s_{m+1} = x^2 \mod n$  and then work backwards defining  $s_m = \sqrt{s_{m+1}} \mod n$ ,  $s_{m-1} = \sqrt{s_m} \mod n$ ,  $\ldots$ ,  $s_0 = \sqrt{s_1} \mod n$ . It follows that  $b_0, \ldots, b_{\ell-m}$  are the last  $\ell - m + 1$  bits of BBS $(s_0)$ , and  $b_0$  is the bit predicted by  $N_m$ .

Now, x and -x are clearly square roots of  $s_{m+1}$ . We show that they both have Jacobi symbol 1. Since  $\left(\frac{x}{n}\right) = \left(\frac{x}{p}\right) \cdot \left(\frac{x}{q}\right) = 1$ , then either  $\left(\frac{x}{p}\right) = \left(\frac{x}{q}\right) = 1$  or  $\left(\frac{x}{p}\right) = \left(\frac{x}{q}\right) = -1$ . But because p and q are Blum primes, -1 is a quadratic non-residue modulo both p and q, so  $\left(\frac{-1}{p}\right) = \left(\frac{-1}{q}\right) = -1$ . It follows that  $\left(\frac{-x}{n}\right) = 1$ . Hence,  $x = \pm \sqrt{s_{m+1}}$ , so exactly one of x and -x is a quadratic residue.

Since n is odd,  $x \mod n$  and  $-x \mod n$  have opposite parity. Hence, x is a quadratic residue iff x and  $\sqrt{s_{m+1}}$  have the same parity. But  $N_m$  outputs  $\sqrt{s_{m+1}} \mod 2$  with probability  $1/2 + \epsilon'$ , so it follows that Q correctly determines the quadratic residuosity of its argument with probability  $1/2 + \epsilon'$ .