# Notes for Lecture 14 v0.9

These notes are in a draft version. Please give me any comments you may have, because this will help me revise them.

Today we will prove the Goldreich-Levin theorem:

**Theorem 1** If  $f: \{0,1\}^n \to \{0,1\}^n$  is a one-way permutation then  $B(x,r) = x \cdot r = \sum_i x_i r_i \pmod{2}$  is a hardcore predicate for f'(x,r) = f(x), r.

In other words, given only f(x) and r it is hard to compute  $x \cdot r$ .

The asymptotic version of the Theorem follows from the following finite version, that we state directly in the counterpositive direction.

**Theorem 2** Let  $f : \{0,1\}^n \to \{0,1\}^n$  be a bijection computable by a circuit of size t, and suppose that there is a circuit C of size S such that

$$\mathbf{Pr}_{x,r}[C(f(x),r) = x \cdot r] \ge \frac{1}{2} + \epsilon$$

Then there is a circuit C' of size  $O((S+t) \cdot poly(n, 1/\epsilon))$  such that

$$\mathbf{Pr}_x[C'(f(x)) = x] \ge \frac{\epsilon}{4}$$

In turn, Theorem 2 follows from the existence of the following algorithm.

**Lemma 3** There is a probabilistic algorithm that, given a parameter  $\epsilon$  and oracle access to a function  $g: \{0,1\}^n \to \{0,1\}$ , runs in time  $O(n^2 \epsilon^{-4} \log n)$ , makes  $O(n \epsilon^{-4} \log n)$  oracle accesses, and outputs a list of  $O(1/\epsilon^2)$  elements of  $\{0,1\}^n$ .

If x is such that  $\mathbf{Pr}_r[B(r) = x \cdot r] \ge \frac{1}{2} + \epsilon$ , then there is a probability at least 1/2 that x is in the output list.

### 1 Proof of Theorem 2 Using Lemma 3

From the assumption that

$$\mathbf{Pr}_{x,r}[C(f(x),r) = x \cdot r] \ge \frac{1}{2} + \epsilon$$

it follows that

$$\mathbf{Pr}_{x}\left[\mathbf{Pr}_{r}[C(f(x),r)=x\cdot r]\geq \frac{1}{2}+\frac{\epsilon}{2}\right]\geq \frac{\epsilon}{2}$$

Let us call an x such that  $\mathbf{Pr}_r[C(f(x), r) = x \cdot r] \ge 1/2 + \epsilon/2$  a good x, and, for a given x, let us denote by  $B_x$  the function defined as  $B_x(r) = C(f(x), r)$ .

For a good x, if we apply the algorithm of Lemma 3 to the function  $B_x$  with parameter  $\epsilon/2$ , we obtain a list that, with probability at least 1/2, contains x.

Consider now the following algorithm: given a string y, define B(r) := C(y, r) and run the algorithm of Lemma 3 with function B() and parameter  $\epsilon/2$ . Once the algorithm outputs a list  $x^1, x^2, \ldots, x^{O(1/\epsilon^2)}$ , compute  $f(x^i)$  for each i, and output the  $x^i$  such that  $f(x^i) = y$ , if any.

If we pick x at random and give f(x) to the above algorithm, there is a probability at least  $\epsilon/2$  that x is good and, if so, there is a probability at least 1/2 that x is in the list. Therefore, there is a probability at least  $\epsilon/4$  that the algorithm inverts f(), where the probability is over the choices of x and over the internal randomness of the algorithm. In particular, there is a fixed choice of the internal randomness of the algorithm that results in inverting f() on an  $\epsilon/4$  fraction of the inputs. Finally, we convert the algorithm into a circuit, and the resulting circuit is C'.

## 2 Proof of Lemma 3

### 2.1 A Special Case First

We start by considering the simpler case in which we are given oracle access to a function B() such that  $\mathbf{Pr}_r[B(r) = x \cdot r] \ge 7/8$  and we want to find x.

If we denote by  $e_i$  the vector that has a 1 in the *i*-th coordinate and 0s in other coordinate, we see that  $x_i = x \cdot e_i$ . Furthermore, for every r, we have  $x_i = x \cdot (r \oplus e_i) \oplus x \cdot r$  because of the linearity of the  $\cdot$  operator and of the fact that  $r \oplus r$  is the all zero vector.

Consider now the process of picking a random r and computing  $B(r \oplus e_i) \oplus B(r)$ . Except with probability at most 1/8,  $B(r \oplus e_i) = x \cdot (r \oplus e_i)$  and, except with probability at most 1/8,  $B(r) = x \cdot r$ . Therefore, with probability at least 3/4,

$$B(r \oplus e_i) \oplus B(r) = x \cdot (r \oplus e_i) \oplus x \cdot r = x_i$$

This suggests the following algorithm:

### Algorithm $A_{\frac{7}{6}}$ :

for i := 1 to n do pick  $k = O(\log n)$  random elements  $r^1, \ldots, r^k \in \{0, 1\}^n$ compute:  $B(r^1 \oplus e_1) \oplus B(r^1)$   $B(r^2 \oplus e_2) \oplus B(r^2)$   $\vdots$   $B(r^k \oplus e_n) \oplus g(r^n)$ assign to  $x_i$  the value occurring in the majority of these computations

return x

To analyze the algorithm, note that, for a particular value of i, we expect to get the right value of  $x_i$  in a fraction 3/4 of the k trials, and the algorithm derives the correct value of

 $x_i$  provided that more than half of the k trials are correct. Then, by a Chernoff bound, the probability of estimating  $x_i$  incorrectly is  $e^{-\Omega(k)}$ . We can then choose  $k = O(\log n)$  and make sure that the error probability is at most, say, 1/100n, and hence we conclude that the output of the algorithm is correcgt with probability at least 99/100.

We note that the running time of this program is  $O(n^2k) = O(n^2 \log n)$  and that it makes  $O(nk) = O(n \log n)$  oracle accesses.

### 2.2 The General Case

Consider now the general case. We are given an oracle B() such that  $B(r) = x \cdot r$  for an  $1/2 + \epsilon$  fraction of the r. Our goal will be to use B() to simulate an oracle that has agreement 7/8 with  $x \cdot r$ , so that we can use the algorithm of the previous section to find x. We perform this "reduction" by "guessing" the value of  $x \cdot r$  at a few points.

We first choose t random points  $r^1 \dots r^t \in \{0, 1\}^n$  where  $t = O(1/\epsilon^2)$ . For the moment, let us suppose that we have "magically" obtained the values  $x \cdot r^1, \dots, x \cdot r^k$ . Then define B'(r) as the majority value of:

$$x \cdot r^{j} \oplus B(r \oplus r^{j}) \qquad j = 1, 2, \dots, t \tag{1}$$

For each j, the above expression equals  $x \cdot r$  with probability at least  $\frac{1}{2} + \epsilon$  (over the choices of  $r^{j}$ ) and by choosing  $t = O(1/\epsilon^{2})$  we can ensure that

$$\mathbf{Pr}_{r,r^{1},...,r^{t}}\left[B'(r) = x \cdot r\right] \ge \frac{31}{32}.$$
(2)

from which it follows that

$$\mathbf{Pr}_{r^1,\dots,r^k}\left[\mathbf{Pr}_r\left[B'(r)=x\cdot r\right] \ge \frac{7}{8}\right] \ge \frac{3}{4}.$$
(3)

Consider the following algorithm.

**Algorithm** GL-First-Attempt:

**pick**  $r^1, \ldots, r^t \in \{0, 1\}^k$  where  $t = O(1/\epsilon^2)$ for all  $b_1, \ldots, b_t \in \{0, 1\}$ define  $B'_{b_1 \ldots b_t}(r)$  as majority of:  $b_j \oplus B(r \oplus r^j)$ apply Algorithm  $A_{\frac{7}{8}}$  to  $B'_{b_1 \ldots b_t}$ add result to list

The idea behind this program is that we do not in fact know the values  $x \cdot r^j$ , but we can "guess" them by considering all choices for the bits  $b_j$ . If B(r) agrees with  $x \cdot r$  for at least a  $1/2 + \epsilon$  fraction of the rs, then there is a probability at least 3/4 that in one of the iteration we invoke algorithm  $A_{\frac{7}{8}}$  with a simulated oracle that has agreement 7/8 with  $x \cdot r$ . Therefore, the final list contains x with probability at least 3/4 - 1/100 > 1/2.

The obvious problem with this algorithm is that its running time is exponential in  $t = O(1/\epsilon^2)$  and the resulting list may also be exponentially larger than the  $O(1/\epsilon^2)$  bound promised by the Lemma.

To overcome these problems, consider the following similar algorithm.

Algorithm GL: pick  $r^1, \ldots, r^l \in \{0, 1\}^k$  where  $l = \log O(1/\epsilon^2)$ define  $r_S := \bigoplus_{j \in S} r^j$  for each non-empty  $S \subseteq \{1, \ldots, l\}$ for all  $b_1, \ldots, b_l \in \{0, 1\}$ define  $b_S := \bigoplus_{j \in S} b_j$  for each non-empty  $S \subseteq \{1, \ldots, l\}$ define  $B'_{b_1 \ldots b_l}(r)$  as majority over non-empty  $S \subseteq \{1, \ldots, l\}$  of  $b_S \oplus B(r \oplus r_S)$ run Algorithm  $A_{\frac{7}{8}}$  with oracle  $B'_{b_1 \ldots b_l}$ add result to list

Let us now see why this algorithm works. First we define, for any nonempty  $S \subseteq \{1, \ldots, l\}, r_S = \bigoplus_{j \in S} r^j$ . Then, since  $r^1, \ldots, r^l \in \{0, 1\}^k$  are random, it follows that for any  $S \neq T$ ,  $r_S$  and  $r_T$  are independent and uniformly distributed. Now consider an x such that  $x \cdot r$  and B(r) agree on a  $\frac{1}{2} + \epsilon$  fraction of the values of r. Then for the choice of  $\{b_j\}$  where  $b_j = x \cdot r^j$  for all j, we have that

$$b_S = x \cdot r_S$$

for every non-empty S. In such a case, for every S and every r, there is a probability at least  $\frac{1}{2} + \epsilon$ , over the choices of the  $r^j$  that

$$b_S \oplus B(r \oplus r_S) = x \cdot r$$

and these events are pair-wise independent. Note the following simple lemma.

**Lemma 4** Let  $R_1, \ldots, R_t$  be a set of pairwise independent 0-1 random variables, each of which is 1 with probability at least  $\frac{1}{2} + \epsilon$ . Then  $\Pr[\sum_i R_i \ge t/2] \ge 1 - \frac{1}{4\epsilon^{2t}}$ .

PROOF: Let  $R = R_1 + \cdots + R_t$ . The variance of a 0/1 random variable is at most 1/4, and, because of pairwise independence,  $\mathbf{Var}[R] = \mathbf{Var}[R_1 + \ldots + R_t] = \sum_i \mathbf{Var}[R_i] \le t/4$ .

We then have

$$\mathbf{Pr}[R \le t/2] \le \mathbf{Pr}[|R - \mathbf{E}[R]| \ge \epsilon t] \le \frac{\mathbf{Var}[R]}{\epsilon^2 t^2} \le \frac{1}{4\epsilon^2 t}$$

Lemma 4 allows us to upper-bound the probability that the majority operation used to compute B' gives the wrong answer. Combining this with our earlier observation that the  $\{r_S\}$  are pairwise independent, we see that choosing  $l = 2\log 1/\epsilon + O(1)$  suffices to ensure that  $B'_{b_1...b_l}(r)$  and  $x \cdot r$  have agreement at least 7/8 with probability at least 3/4. Thus we can use Algorithm  $A_{\frac{7}{8}}$  to obtain x with high probability. Choosing l as above ensures that the list generated is of length at most  $2^l = O(1/\epsilon^2)$  and the running time is then  $O(\epsilon^{-4} \cdot n^2 \log n)$  with  $O(\epsilon^{-4} \cdot n \log n)$  oracle accesses, due to the  $O(1/\epsilon^2)$  iterations of Algorithm  $A_{\frac{7}{8}}$ , that makes  $O(n \log n)$  oracle accesses, and to the fact that one evaluation of B'() requires  $O(1/\epsilon^2)$  evaluations of B().

## **3** References

The results of this lecture are from [GL89]. Goldreich and Levin initially presented a different proof. They credit the proof with pairwise independence to Rackoff. Algorithm  $A_{\frac{7}{8}}$  is due to Blum, Luby and Rubinfeld [BLR93]. The use of Algorithm GL-First-Attempt as a motivating example might be new to these notes. (Or, actually, to the Fall 2001 notes for this class.)

# References

- [BLR93] M. Blum, M. Luby, and R. Rubinfeld. Self-testing/correcting with applications to numerical problems. Journal of Computer and System Sciences, 47(3):549–595, 1993. Preliminary version in Proc. of STOC'90. 5
- [GL89] O. Goldreich and L. Levin. A hard-core predicate for all one-way functions. In Proceedings of the 21st ACM Symposium on Theory of Computing, pages 25–32, 1989. 5