| Seminar on Sublinear Time Algorithms |                         |                         |
|--------------------------------------|-------------------------|-------------------------|
| Lecture 4                            |                         |                         |
| April 14, 2010                       |                         |                         |
| Lecturer: Robert Krauthgamer         | Scribe by: Merav Parter | Updated: May 10, $2010$ |

# 1 Testing Homomorphism of a Function

**Definition 1** A function  $f : \mathbb{Z}_n \to \mathbb{Z}_n$  is called homomorphism if  $\forall x, y \in \mathbb{Z}_n$ , f(x) + f(y) = f(x+y).

**Definition 2** A function  $f : \mathbb{Z}_n \to \mathbb{Z}_n$  is called  $\epsilon$ -close to homomorphism if it can be changed in at most  $\epsilon$ -fraction of places  $x \in \mathbb{Z}_n$  to become a homomorphism, otherwise it is  $\epsilon$ -far.

<u>Task Definition</u>: Given a function f, test whether it is a homomorphism or  $\epsilon$ -far from it.

**Theorem 3** [Ben-Or, Luby, Rubinfeld and Coppersmith]  $\forall 0 \leq \epsilon \leq \frac{1}{3}$  there is a tester for homomorphism that determines w.h.p if f is a homomorphism or  $\epsilon$ -far from it in time  $O(\frac{1}{\epsilon})$ .

Key Idea: Relate  $\epsilon$  to  $\delta(f) = P_{\forall x, y \in \mathbb{Z}_n}[f(x) + f(y) \neq f(x+y)].$ 

### Algorithm Test Homomorphism

- 1. Repeat  $\frac{4}{\epsilon}$  times:
  - (a) Choose  $x, y \in \mathbb{Z}_n$  at random and check if f(x) + f(y) = f(x+y).
- 2. Accept if all these hold with equality, otherwise Reject.

#### Analysis:

<u>Runtime</u>: Obvious.

<u>Correctness</u>: Since the algorithm always accepts a homomorphism function, it is a one-sided error algorithm. For the rest of the discussion, we therefore assume that f is  $\epsilon$ -far from homomorphism. We wish to show that  $P(algorithm \ accepts \ f) \leq \frac{1}{3}$ .

A simple but important observation in this context is that if f is  $\epsilon$ -close to homomorphism then there exists a "corrected" function  $g : \mathbb{Z}_n \to \mathbb{Z}_n$  which is both homomorphism and  $\epsilon$ -close to f. This function g is defined as follow:  $g(x) = plurality_y f(x+y) - f(y)$ . Intuitively, the value f(x+y) - f(y) can be thought of as the "vote" of y on x. If f is  $\epsilon$ -close to homomorphism, then most of the votes for a given x are the same, resulting in a homomorphism function g. To prove this intuition in a formal manner, we state two auxiliary claims. **Claim 4** f and g agree on at least  $1 - 2\delta$  values.

**Claim 5** If  $\delta(f) \leq \frac{1}{6}$  then g is homomorphism.

Assuming these claims to be correct, we turn to prove Theorem 3.

### **Proof** [of Theorem 3]

Recall that we consider the case where f is  $\epsilon$ -far from homomorphism. First, assume that  $\delta(f) \leq \frac{\epsilon}{2}$   $(<\frac{1}{6})$ . By Claim 4 and Claim 5 we have that f is  $2\delta(f) \leq \epsilon$  close to homomorphism, and we end with a contradiction. Next, assume that  $\delta(f) > \frac{\epsilon}{2}$ . We will see that in this case the algorithm will reject the function w.h.p.:  $P[alg \ accepts \ f] \leq (1 - \delta(f))^{\frac{4}{\epsilon}} \leq (1 - \frac{\epsilon}{2})^{\frac{4}{\epsilon}} < e^{-2} < \frac{1}{3} \text{ as required.} \blacksquare$ 

It is yet left to prove the supporting claims. We begin with Claim 4.

### **Proof** [of Claim 4]

Let  $\Delta(f,g)$  denote the fraction of disagreements between f and g.

Let  $B = \{x : Pr_y[f(x) + f(y) \neq f(x+y)] \ge \frac{1}{2}\}$ . Notice that B contains all the x's where f and g disagree. In addition,  $\delta(f) \geq \frac{|B|}{n} \cdot \frac{1}{2}$ , where  $\frac{|B|}{n}$  is the probability to choose a bad x, and  $\frac{1}{2}$  is a lower bound for the probability to chose a bad partner y. Overall we have that  $\Delta(f,g) \leq \frac{|B|}{n} \leq 2\delta(f)$  as required.

As a step toward proving Claim 5, we state the following claim.

**Claim 6** The function g is a strong plurality (if  $\delta$  is small) in the following sense,  $\forall x, P_{y}[q(x) = f(x+y) - f(y)] \ge 1 - \delta(f).$ 

## **Proof** [of Claim 6]

We first analyze for an arbitrary  $x \in \mathbb{Z}_n$  the "collision probability" of two votes and then relate it to the "plurality probability" as required by the claim.

Fix x and choose  $y_1, y_2$  at random and independently. Then we have that  $P_{y_1,y_2}[f(x+y_1) - f(y_1) = f(x+y_2) - f(y_2)] =$ 

$$P_{y_1,y_2}[f(x+y_1)+f(y_2)=f(x+y_2)+f(y_1)] \ge$$

$$P_{y_1,y_2}[f(x+y_1) + f(y_2) = f(x+y_2) + f(y_1) = f(x+y_1 + f(y_2)] \ge 1 - 2\delta(f)$$

Where the last inequality follows by the union-bound. To show that the "collision probability" is at most the "plurality probability", we consider an experiment A with n possible outcomes, where  $o(A) \in [1, n]$  denotes the outcome of A. Let  $p_i = P[o(A) = i]$ , i.e., the probability where that the experiment ended with outcome *i*., where  $p_i > 0$  for every  $1 \leq i \leq n$ , and  $\sum_{i=1}^{n} p_i = 1$ . The probability that two independent experiments A, B ended with the same outcome (i.e., collision occurred) is given by

$$P[o(A) = o(B)] = \sum_{i=1}^{n} p_i^2 \le \max_j (p_j) \cdot \sum_{i=1}^{n} p_i = \max_j (p_j) \cdot 1 = \max_j (p_j).$$

Since  $\max_j(p_j)$  is the "plurality probability" the claim is established.

We are now ready to complete the proof for Claim 5.

#### Proof [of Claim 5]

Fix x, z. By applying Claim 6 three times, first for x, then for z and finally for x + z, we get

- 1)  $P_y[g(x) \neq f(x+y-x) f(y-x)] \le 2\delta(f) < \frac{1}{3}$
- 2)  $P_y[g(z) \neq f(z+y) f(y)] \le 2\delta(f) < \frac{1}{3}$
- 3)  $P_y[g(x+z) \neq f(x+z+y-x) f(y-x)] \le 2\delta(f) < \frac{1}{3}$

With positive probability none of these events happen, implying that  $\exists y \text{ such that } g(x) + g(z) = [f(y) - f(y - x)] + [f(z + y) - f(y)] = g(x + z)$  where the first equality is followed by (1,2) and the second equality is followed by (3). The Claim follows.

## 2 Testing a Dense Graph for Bipartiteness

**Definition 7** Graph G = (V, E) is  $\epsilon$ -far from bipartite if it is necessary to remove more than  $\epsilon |V|^2$  edges so that it becomes bipartite.

<u>Task definition</u>: Given a dense graph G = (V, E), determine w.h.p if it is bipartite or  $\epsilon$ -far from it.

**Theorem 8** (Goldreich-Goldwasser-Ron) There is a tester for bipartiteness that determines whether G is bipartite or  $\epsilon$ -far from it in time  $(\frac{1}{\epsilon})^{O(1)}$ .

In particular, the tester we present always accepts bipartite graphs and rejects  $\epsilon$ -far instances with probability at least  $\frac{2}{3}$ .

Key Idea: Sampling small number of vertices is in fact representative.

### Algorithm Test-Bipartite

- 1. Uniformly and independently select  $m = \Theta(\frac{\log(\frac{1}{\epsilon})}{\epsilon^2})$  vertices.
- 2. Accept iff the subgraph induced on them is bipartite (by BFS)

### Analysis:

<u>Runtime</u>:  $\Theta(\frac{\log^2(\frac{1}{\epsilon})}{\epsilon^4})$ . Quadratic in the size of the sample due to construction of the induced subgraph.

<u>Correctness</u>: If G is bipartite then clearly so is every subgraph of it. Hence, this is a onesided error tester. We next assume that G is  $\epsilon$ -far from bipartite and wish to show it is rejected by the algorithm with probability greater than  $\frac{2}{3}$ .

Let R denote the set of sampled vertices. It is convenient to view R as composed of two parts that are sampled one after the other, namely U and S respectively. Let  $|U| = O(\frac{1}{\epsilon} \log \frac{1}{\epsilon})$  and  $|S| = O(\frac{|U|}{\epsilon})$ . One can check that indeed m = |U| + |S|. Note, that since the vertices are selected independently, repetitions may occur (e.g., U and S may overlap). We first provide some definitions.

**Definition 9** A vertex v is high-degree if its degree is greater than  $\frac{\epsilon n}{3}$ .

**Definition 10** Set  $U \in V$  is good if all but at most  $\frac{\epsilon n}{3}$  of the high-degree vertices of V are adjacent to U.

Let  $\Gamma(w) = \{v : (w, v) \in E\}$  (the neighbors of w). Let  $\Gamma(W) = \bigcup_{w \in W} \Gamma(w)$ .

**Claim 11** With probability of at least  $\frac{5}{6}$  over the choice of U, the set U is good.

### Proof

Let  $v \in V$  be a high-degree vertex. The probability that U contains none of v's neighbors is at most

$$(1 - \frac{\epsilon}{3})^{|U|} < e^{\frac{\epsilon}{3} \cdot |U|} \tag{1}$$

If we sample  $|U| = \frac{3}{\epsilon} \cdot \ln(\frac{18}{\epsilon})$ , we get that this probability is at most  $(\frac{\epsilon}{18})$ . By linearity of expectation, the expected number of such v's is  $\leq \frac{\epsilon n}{18}$ . Finally by Markov's inequality, the probability that there are more than  $\frac{\epsilon n}{3}$  such v's (high-degree vertices with no neighbor in U) is at most  $\frac{1}{6}$  as required.

**Definition 12** An edge is said to disturb a partition  $U = U_1 \bigcup U_2$  if its endpoints are in the same  $\Gamma(U_i)$  for  $i \in [1, 2]$ .

**Claim 13** If G is  $\epsilon$ -far from bipartite then for every good U and for every partition of  $U = U_1 \bigcup U_2$  there are at least  $\frac{\epsilon n^2}{3}$  disturbing edges.

### Proof

Assume U is indeed good and consider a fixed partition  $U = U_1 \bigcup U_2$ . Let  $N = \Gamma(U)$  and  $C = V \setminus N$ . Since U is good, we have that C contains at most  $\frac{\epsilon n}{3}$  high-degree vertices. We next use the partition of U to induce a partition of N and eventually on V in the following manner:

 $N_1 = \Gamma(U_1)$  and  $N_2 = \Gamma(U) \setminus N_1$ . Let  $C_1, C_2$  be any partition of C such that  $(C \cap U_1) \subseteq C_1$ 

and  $(C \cap U_2) \subseteq C_2$ . The final partition of  $V = V_1 \bigcup V_2$  is  $V_1 = N_1 \bigcup C_2$  and  $V_2 = N_2 \bigcup C_1$ . Observe that since G is  $\epsilon$ -far from bipartite, *every* partition of V has more than  $\epsilon n^2$  "disturbing" edges. In particular this is correct for the partition  $(V_1, V_2)$ . We next show that many of these "disturbing" edges are incident to vertices in U.

Q: How many disturbing edges at-most can be incident to C (i.e., not incident to N)? Ans.: C contains at-most n edges from each of at-most  $\frac{\epsilon n}{3}$  high-degree vertices. In addition, it contains at-most  $\frac{\epsilon n}{3}$  edges from each of at-most n non-high-degree vertices.

Overall, we get that the are at least  $\frac{\epsilon n^2}{3}$  disturbing edges that are incident to N.

We are now ready to complete the proof of Theorem 8.

### **Proof** [of Theorem 8]

Let G[R] be the graph induced by the selected set R. For G[R] to be bipartite we must have either:

1) U is not good (w.p  $\leq \frac{1}{6}$ )

2) U is good and  $\exists$  a partition  $U = U_1 \bigcup U_2$  such that none of its disturbing edges occur in G[R]. Applying the union-bound over the possible  $2^{|U|}$  partitions of |U| and combining Claim 13, we get that the probability for such an event is at most  $2^{|U|}(1-\frac{\epsilon}{3})^{\frac{|S|}{2}} = 2^{|U|} \cdot e^{\frac{-\epsilon|S|}{6}} < \frac{1}{6}.$ 

Overall, the probability to accept an  $\epsilon$ -far graph G is at-most  $\frac{1}{6} + \frac{1}{6} = \frac{1}{3}$  as required.