

## Lecture 11: October 3

*Lecturer: Eshan Chattopadhyay**Scribe: Sloan Nietert*

## 11.1 Overview

This lecture will cover the following topics:

- Probabilistic existence of seeded extractors
- Simulation of randomized algorithms using defective sources
- Universal hash functions (with explicit construction)
- Construct seeded extractors from hash functions

## 11.2 Existence of seeded extractors

Recall that we defined a seeded extractor to be a set of functions

$$\{\text{Ext}_i : \{0, 1\}^n \rightarrow \{0, 1\}^m\}_{i \in \{0, 1\}^d}$$

such that  $|\text{Ext}_{U_d}(X) - U_m| < \varepsilon$  for any  $(n, k)$ -source  $X$ .

Alternatively, we can view an extractor as a single function.

**Definition 11.1** *We say that*

$$\text{Ext} : \{0, 1\}^n \times \{0, 1\}^d \rightarrow \{0, 1\}^m$$

*is a  $(k, \varepsilon)$ -extractor if  $|\text{Ext}(X, U_d) - U_m| < \varepsilon$  for any  $(n, k)$ -source  $X$ .*

Sometimes, we will need an even stronger notion.

**Definition 11.2** *We say that*

$$\text{Ext} : \{0, 1\}^n \times \{0, 1\}^d \rightarrow \{0, 1\}^m$$

*is a strong-seeded extractor if  $|\text{Ext}(X, U_d), U_d - U_m, U_d| < \varepsilon$ .*

This is equivalent to requiring that

$$\mathbb{E}_{y \sim U_d} |\text{Ext}(X, y) - U_m| < \varepsilon.$$

In what follows, we will use the notation  $D_1 \approx_\varepsilon D_2$  to mean that  $|D_1 - D_2| < \varepsilon$ . Next, we show the existence of seeded extractors.

**Proof:**

Let  $\text{Ext} : \{0, 1\}^n \times \{0, 1\}^d \rightarrow \{0, 1\}^m$  be a random function, and let  $X$  be a flat  $(n, k)$ -source. Take  $T \subseteq \{0, 1\}^m$  to be a “test” subset for our extractor. The main idea of what follows is to take a union bound over all tests and flat sources. Note that all uppercase letters represent the base 2 exponential of their lowercase counterparts.

For any  $x \in \{0, 1\}^n$  and  $y \in \{0, 1\}^d$ , let

$$Z_{x,y} = \begin{cases} 1, & \text{if } \text{Ext}(x, y) \in T \\ 0, & \text{o.w.} \end{cases}$$

Next, define

$$W_T = \Pr_{x \sim X, y \sim U_d}[\text{Ext}(x, y) \in T] = \frac{1}{KD} \sum_{x \in \text{supp}(X), y \in \{0, 1\}^d} Z_{x,y}.$$

Clearly,  $\mathbb{E}[W_T] = |T|/M$ . Noting that the  $Z_{x,y}$  variables are i.i.d. and uniform, we can use Chernoff’s bound to find

$$\Pr[|W_T - \mathbb{E}(W_T)| > \varepsilon] \leq \exp(-\Omega(\varepsilon^2 KD)).$$

Thus, we can bound the probability of Ext performing poorly on any test set or flat source by

$$2^M \binom{N}{K} \exp(-\Omega(\varepsilon^2 KD)).$$

We choose  $m = k + d - 2 \log(1/\varepsilon) - O(1)$  and bound  $\binom{N}{K}$  by  $(Ne/K)^K$  to see that

$$d \geq \log(n - k) + 2 \log(1/\varepsilon) + O(1)$$

suffices to bring the error probability below 1 and guarantee the existence we desire. ■

### 11.3 Simulation of randomized algorithms

Suppose  $\mathcal{A}$  is a randomized algorithm such that  $\mathcal{A}(\cdot, U_m)$  is incorrect with probability at most  $\varepsilon$ , and suppose we have access to an  $(n, k)$ -source  $X$  with  $k > m$ .

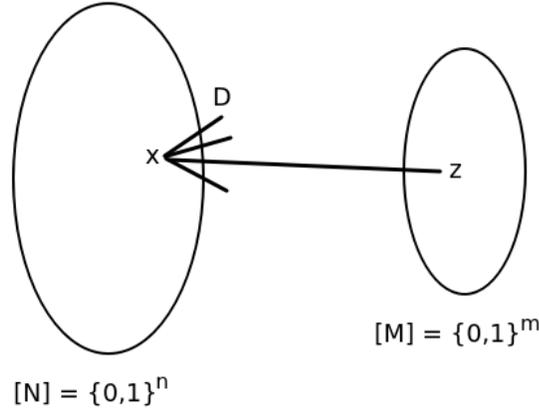
If Ext is a  $(k, \delta)$ -extractor, we can define a new algorithm

$$\mathcal{A}'(\cdot, X) = \text{maj}_{y \in \{0, 1\}^d} \{\mathcal{A}(\cdot, \text{Ext}(X, y))\}$$

Clearly, the time blow up is a factor of  $D$ .

**Claim 11.3**  $\mathcal{A}'$  has error at most  $2(\varepsilon + \delta)$ .

**Proof:** We now introduce a graph interpretation of the seeded extractor  $\text{Ext} : \{0, 1\}^m \times \{0, 1\}^n \rightarrow \{0, 1\}^m$ ,



where the above represents a  $D$  left-regular bipartite graph with  $x$  and  $z$  joined if and only if there exists  $y \in \{0, 1\}^d$  such that  $\text{Ext}(x, y) = z$ .

Define the set  $\text{Bad} = \{z \in \{0, 1\}^m : \mathcal{A}(\cdot, z) \text{ is wrong}\}$ . Clearly,  $|\text{Bad}| \leq \varepsilon M$ . Next, define

$$\text{Bad}_X = \{x \in \text{supp}(X) : |\Gamma(x) \cap \text{Bad}| > D/2\},$$

so that  $\mathcal{A}'(\cdot, x)$  is wrong if and only if  $x \in \text{Bad}_X$ . Then, because  $\text{Ext}$  is an extractor, we have that  $|\text{Bad}|/M \leq \varepsilon$ , and if  $x \in \text{Bad}_X$ , then

$$\Pr[\text{Ext}(x, U_d) \in \text{Bad}] > \frac{1}{2}.$$

Putting everything together, we have

$$\Pr_{x \sim X}[x \in \text{Bad}_X] \leq 2 \Pr[\text{Ext}(X, U_d) \in \text{Bad}] \leq 2 \left( \frac{|\text{Bad}|}{M} + \delta \right) \leq 2(\varepsilon + \delta).$$

Thus, the error of  $\mathcal{A}'$  is bounded by  $2(\varepsilon + \delta)$ , as desired. ■

## 11.4 Universal hash functions

**Definition 11.4** We say that  $\mathcal{H} = \{h_i\}_{i \in I}$ ,  $h_i : [N] \rightarrow [M]$ , is a universal hash function if for all  $x \neq y \in [N]$

$$\Pr_{h \in \mathcal{H}}[h(x) = h(y)] \leq \frac{1}{M}.$$

Let  $\mathbb{F}_q$  be a finite field with  $q = 2^n$ , and define

$$h_a : x \mapsto a \cdot x \pmod{2^m}$$

for each  $a \in \mathbb{F}_q$ .

**Claim 11.5**  $\mathcal{H} = \{h_a\}_{a \in \mathbb{F}_q}$  is a universal hash function.

**Proof:** Take  $x, y$  distinct in  $\mathbb{F}_q$ . Then,

$$\Pr_{h \in \mathcal{H}}[h(x) = h(y)] = \Pr_{a \in \mathbb{F}_q}[ax = ay \pmod{2^m}] = \Pr_{a \in \mathbb{F}_q}[az = 0 \pmod{2^m}] = 1/M,$$

where  $z = x - y \neq 0$ . ■

## 11.5 Constructing extractors from hash functions

**Definition 11.6** We define the collision probability

$$\text{cp}(D) = \Pr[D = D'],$$

where  $D$  and  $D'$  are independent copies of  $D$ .

Observe that

$$\text{cp}(D) = \sum_{x \in \Omega} D(x)^2.$$

**Lemma 11.7** If  $D$  is a distribution on  $\{0, 1\}^n$  with  $\text{cp}(D) \leq \frac{1+\varepsilon}{N}$ , then

$$|D - U_n| \leq \sqrt{\varepsilon}$$

**Proof:** We have

$$2|D - U_n| = \|D - U_n\|_1 \leq \sqrt{N} \|D - U_n\|_2$$

by Cauchy-Schwartz, and

$$\|D - U_n\|_2^2 = \sum_{x \in \{0,1\}^n} (D(x) - U(x))^2 = \sum_{x \in \{0,1\}^n} D(x)^2 - 2 \sum_{x \in \{0,1\}^n} D(x) \frac{1}{N} + \frac{1}{N^2} N = \text{cp}(D) - \frac{1}{N},$$

from which the desired inequality follows. ■

(to be continued in next lecture)