

## Lecture 4

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# 1 Expansion - General Notions

## 1.1 A Bad Code as a Useful Example

Construct a parity-check matrix  $H$  (from which an encoding matrix  $G$  can be derived) of size  $n \times (n - k)$ , so that  $H_{ij} = 1$  with a probability of  $\frac{10}{n}$ , 0 otherwise. The code,  $C$ , is  $C = \{y | yH = 0\}$ .

The probability for a row of zeros is  $(1 - \frac{10}{n})^{n-k}$ , which is fairly high. Therefore, 2 words -  $y_1, y_2$  which are differentiated by 1 bit, can both be legitimate words (because the 1 bit is multiplied by the zero row) -  $y_1H = y_2H = 0$ . This means the code distance is 1 - which is the worst distance possible.

## 1.2 Modeling a Graph

**Definition 1** Every matrix  $H$  of size  $n \times m$  defines a 2-sided graph  $G = (L \cup R, E)$ .  $|L| = n, |R| = m$ .  $H_{ij} = 1 \iff (i, j) \in E$ . From now on,  $L$  will denote the left vertices, and  $R$ , the right ones

**Definition 2** A bipartite graph  $G$  is called  $(c, d)$ -regular if all the vertices on the left have a rank of  $c$ , and all vertices on the right have a rank of  $d$ , (thus,  $c \cdot n = d \cdot m$ ).

Let  $S$  be a set of vertices. We denote the neighbors of  $S$  be  $\Gamma(S) = \{v \mid \exists u \in S, (u, v) \in E\}$ .

**Definition 3** A  $(c, d)$ -regular graph will be called a  $(c, d, \delta, \gamma)$ -expander for  $0 < \delta < 1, 0 < \gamma < 1$ , if for every  $S \subseteq L$ ,  $|S| \leq \delta n$  then  $|\Gamma(S)| \geq |S| \cdot c \cdot \gamma$ .

**Theorem 4** For every  $\varepsilon > 0$ , there are  $c, d, \delta > 0$  so there is an infinite family (so that  $n$  keeps growing) of  $(c, d, \delta, \gamma)$ -expanding graphs.

**Sketch of Proof** Choose a random matching between two sets of  $cn$  vertices. On the right, we combine every  $d$  vertices, and on the left, combine every  $c$  vertices. This gives a random  $(c, d)$ -regular graph, possibly with parallel edges.

For any fixed set  $S \subseteq L$ ,  $|S| \leq \delta n$  let us calculate the expected number of neighbors of  $S$ . A tail inequality is then used to show that with (very) high probability  $|\Gamma(S)|$  doesn't deviate much from the expectation. This holds even after taking a union over all sets  $S$  of cardinality at most  $\delta n$ . ■

**Definition 5** If  $S \subseteq L$ , then  $\Gamma_{\text{unique}}(S) = \{v \in R \mid |\Gamma(v) \cap S| = 1\}$  - the vertex has only one "neighbor" in  $S$ .

**Lemma 6** Let  $G$  is a  $(c, d, \delta, \gamma)$ -expander, and let  $S \subseteq L$ ,  $|S| \leq \delta n$ . Then  $|\Gamma_{\text{unique}}(S)| \geq (2\gamma - 1) \cdot c \cdot |S|$ .

**Proof** Consider a set of vertices  $S$ . Define  $A = \Gamma_{\text{unique}}(S)$  and  $B = \Gamma(S) \setminus A$ . Notice that  $|A| + |B| \geq c\gamma|S|$  (since it's a  $(c, d, \delta, \gamma)$ -regular graph). Also, recall that the total number of edges from  $S$  is  $c \cdot |S|$  (because it is a  $(c, d)$ -regular graph), and  $A$  receives  $|A|$  edges from  $S$ , while  $B$  receives  $2 \cdot |B|$  edges.

$$c \cdot \gamma \cdot |S| \leq |A| + |B| \leq |A| + 2 \cdot |B| \leq c \cdot |S|$$

$$\Rightarrow |B| \leq c|S|(1 - \gamma)$$

$$\Rightarrow |\Gamma_{\text{unique}}(S)| = |A| \geq c \cdot \gamma \cdot |S| - |B| \geq c \cdot \gamma \cdot |S| - c \cdot |S| \cdot (1 - \gamma) = c \cdot |S| \cdot (2\gamma - 1)$$

■

## 2 Sipser-Spielman/Gallager Code

### 2.1 What is the Code?

Let  $G$  be a  $(c, d, \delta, \gamma)$ -expander. The code  $C(G)$  contains all strings  $x \in \{0, 1\}^n$  such that:

$$\forall j \in R, \quad \sum_{i \in \Gamma(j)} x_i = 0 \pmod{2}.$$

One can think of it as if  $x$  is “laid upon”  $L$ , and the each vertex in  $R$  “sees”  $0 \pmod{2}$  “1”s in the vector.

**Lemma 7** *If  $\gamma > \frac{1}{2}$ , then the relative distance of the code  $C(G)$  is at least  $\delta$ .*

**Proof** Since the code is linear, it is enough to prove that  $wt(x) \geq \delta n$  for every  $0 \neq x \in C(G)$ .

Let's assume there is an  $x \in C(G)$  so that  $wt(x) < \delta n$ . Define  $S = \{i \in [n] \mid x_i = 1\}$ . According to Lemma 6,  $|\Gamma_{\text{unique}}(S)| \geq (2\gamma - 1) \cdot c \cdot |S| > 0$ . For  $j \in \Gamma_{\text{unique}}(S)$ , the  $j$  constraint (according to the definition -  $\sum_{i \in \Gamma(j)} x_i = 0 \pmod{2}$ ) isn't fulfilled, and therefore  $x \notin C(G)$ . ■

### 2.2 Sipser-Spielman Decoding Algorithm

Suppose  $x \in \{0, 1\}^n$ , and it is guaranteed that  $\text{dist}(x, C(G)) \leq \frac{\delta n}{2}$  (which is, as half the distance, an error-correcting distance). We want to find the unique word  $\tilde{x}$  which is the closest one to  $x$ .

**Definition 8** *For a given word  $x \in \{0, 1\}^n$ , a vertex  $j \in R$  is **satisfied** if*

$$\sum_{i \in \Gamma(j)} x_i = 0 \pmod{2}.$$

*Otherwise, it is **unsatisfied**.*

**The Algorithm:** If there is an  $i \in L$  that has more unsatisfied adjacent vertices than satisfied ones - it is flipped ( $x_i = 1 - x_i$ ).

It is not hard to implement the algorithm in *linear* time.

**Theorem 9** *If  $\gamma \geq \frac{3}{4}$ , then the algorithm corrects  $\frac{\delta}{2} \cdot n$  errors.*

**Proof** Denote by  $\tilde{x}$  the codeword closest to  $x$ , and let  $S = \{i \in L \mid x_i \neq \tilde{x}_i\}$ . By assumption,  $|S| \leq \delta n/2$ .

- The number of unsatisfied constraints is at most  $|S| \cdot c \leq \frac{\delta n c}{2}$ . This is since a constraint can only be unsatisfied if it is adjacent to a vertex in  $S$ .
- Each iteration of the algorithm decreases the number of unsatisfied constraints. This is because a bit is flipped when more of its neighbors are unsatisfied than satisfied. The flip causes all satisfied neighbors to become to unsatisfied and vice versa.
- As long as  $0 < |S| < \delta n$ , the algorithm doesn't stop. (In other words, there is some bit that has more unsatisfied neighbors than satisfied ones). This follows from Lemma 6 since

$$|\Gamma_{\text{unique}}(S)| \geq (2\gamma - 1)c|S| \geq \frac{c}{2}|S|$$

(since  $\gamma \geq \frac{3}{4}$ ). Therefore, there is at least one vertex  $v \in S$  that has more than  $\frac{c}{2}$  adjacent vertices in  $\Gamma_{\text{unique}}$ . Now, since all vertices in  $\Gamma_{\text{unique}}$  are unsatisfied, it follows that  $v$  has more unsatisfied neighbors than satisfied ones, and should be flipped.

- Throughout the algorithm,  $|S| < \delta n$ .

$S$  increases in jumps of 1. If  $|S| = \delta n$  then  $\Gamma_{\text{unique}}(S) > \frac{c}{2}|S| = \frac{c}{2}\delta n$  which lower-bounds the number of unsatisfied constraints at that point. But since we started with  $|S| = \frac{\delta n}{2}$ , the number of unsatisfied constraints was at most  $|S| \cdot c \leq \delta cn/2$  and each step only decreases the number of unsatisfied constraints.

It follows that the algorithm stops when  $|S| = 0$ , i.e. we have reached the word  $\tilde{x}$  as claimed. ■

For this code, the encoding is quadratic (basically, find a generator matrix for the code, and encoding amounts to vector-matrix multiplication. There is another code (also based on expanders), discovered by Spielman, which enables linear time encoding *and* decoding.

### 3 Expanders

We will focus on undirected  $d$ -regular graphs (every vertex has  $d$  edges).

#### 3.1 Definitions

**Definition 10**  $G = (V, E)$  is a  $d$ -regular graph.  $S, T \subset V$ ,  $S \cap T = \emptyset$ . We define  $E(S, T) = \{(u, v) | u \in S, v \in T, (u, v) \in E\}$ . Notice that the expected value of  $|E(S, T)| = |S| \cdot d \cdot \frac{|T|}{n}$ , because  $|S|d$  is the number of edges “leaving”  $S$ , and  $\frac{|T|}{n}$  is the chance that those edges will “reach” somewhere in  $T$ .

**Definition 11** For a set  $S \subseteq V$ , the edge boundary of  $S$  is  $\partial S = E(S, V \setminus S)$ .

**Definition 12** Graph  $G$ 's expansion is  $h(G) = \min_{S | |S| \leq \frac{|V|}{2}} \frac{|\partial S|}{|S|}$ .

**Definition 13** A family of  $d$ -regular graphs  $\{G_i\}_{i \geq 0}$  is an  $\varepsilon$ -expander family if:

- $|G_i| = n_i$  and  $n_i \rightarrow \infty$  (but not too fast - we want every  $n$  to have an expander of a size that's “close” to it).
- $h(G_i) > \varepsilon$ .

**Definition 14** A family  $\{G_i\}$  will be called explicit if there is an algorithm, that given  $1^i$  creates, in polynomial time, the graph  $G_i$ .

A family is very explicit if there is a polynomial algorithm that given  $(i, u, k)$  (for  $u \in 1 \dots n_i$  and  $k \in 1 \dots d$ ) gives the vertex  $v \in V$  that is the  $k$ th neighbor of  $u$  in  $G_i$ .

#### 3.2 Examples

- Random  $d$ -regular graphs.
- Margulis/Gaber - Galil:  
Take  $\mathbb{Z}_m$ , and define  $V = \mathbb{Z}_m \times \mathbb{Z}_m$ .  
 $(x, y) \in V$  is connected with an edge to  $(x + y, y)$ ,  $(x - y, y)$ ,  $(x, x + y)$ ,  $(x, x - y)$ .  
This is a 4-regular graph, and there is a  $\varepsilon$  so it is an  $\varepsilon$ -expander. We must remove  $(0, 0)$  in order to have this graph connected. Observe that there are parallel edges from vertices  $(0, a)$  or  $(a, 0)$ .
- LPS (Lubotzky - Phillips - Sarnak):  
Using a prime number  $p$ , we define  $V = \mathbb{Z}_p$ .  
 $x \in V$  is connected to  $x - 1, x + 1, x^{-1}$ . Again 0 needs to be treated separately, as no  $0^{-1}$  exists, so we add a self-loop.
- Zig-Zag graphs / RVW (Reingold - Vadhan - Wigderson).