

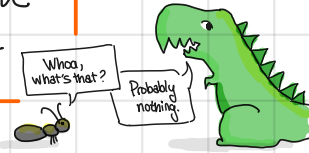
# CS250/EE387 - LECTURE 15 - LOCALITY AND LISTS

## AGENDA

- ① Motivation: Learning Boolean Fns
- ② Goldreich-Levin Algorithm
- ③ Local List Decoding.

## TODAY'S ANT FACT

Ants have been around for over 120 million years, which means they cohabitated with dinosaurs and have survived multiple mass extinctions.



① RECAP. Last time we talked about LOCALLY CORRECTABLE CODES.

The basic principle was illustrated by the Hadamard Code:

$$\mathcal{H} = \{ \langle \omega, \alpha_1 \rangle, \dots, \langle \omega, \alpha_{2^m} \rangle : \omega \in \mathbb{F}_2^m \}$$

The key was that  $\forall i, \langle \omega, \alpha_i + \beta \rangle + \langle \omega, \beta \rangle = \langle \omega, \alpha_i \rangle$ , so to locally recover  $\langle \omega, \alpha_i \rangle$  we query  $\langle \omega, \alpha_i + \beta \rangle, \langle \omega, \beta \rangle$ , HOPE they are not corrupted, and add them together.

## ① MOTIVATION: LEARNING BOOLEAN FNS.

Suppose you have some Boolean function  $G: \mathbb{F}_2^m \rightarrow \{-1, +1\}$ .

You have query access to  $G$  and you'd like to learn an approximation  $\hat{G}$  to  $G$ .

Throughout this lecture, capital letters mean the range is  $\{\pm 1\}$ .  
Lowercase  $g: \mathbb{F}_2^m \rightarrow \mathbb{F}_2$  have range  $\mathbb{F}_2$ .

DEF: For  $G: \mathbb{F}_2^m \rightarrow \{-1, +1\}$ , the FOURIER TRANSFORM of  $G$  over  $\mathbb{F}_2$  is  $\hat{G}: \mathbb{F}_2^m \rightarrow \mathbb{R}$  given by

$$\hat{G}(\omega) = \frac{1}{2^m} \sum_{x \in \mathbb{F}_2^m} G(x) \cdot (-1)^{\langle x, \omega \rangle}$$

If you haven't seen this before, but have seen the Fourier Transform over  $\mathbb{C}$ , all the same things hold. In particular:

$$G(x) = \sum_{\omega \in \mathbb{F}_2^m} \hat{G}(\omega) (-1)^{\langle x, \omega \rangle}$$

$$\text{and: } 1 = \frac{1}{2^m} \sum_{x \in \mathbb{F}_2^m} G(x)^2 \stackrel{\text{"Parseval's Thm"}}{=} \sum_{\omega \in \mathbb{F}_2^m} |\hat{G}(\omega)|^2$$

so in particular the number of Fourier coefficients  $\hat{G}(\omega)$  so that  $|\hat{G}(\omega)| > \tau$  is  $\leq 1/\tau^2$ .

Suppose we want to learn  $G$  from samples.

If the Fourier spectrum of  $G$  is "spiky," it suffices to estimate  $y_\omega \approx \hat{G}(\omega)$  for all  $\omega$  so that  $|\hat{G}(\omega)| > \tau$ . Indeed, then we'd have

$$G(x) \approx \sum_{\omega: |\hat{G}(\omega)| > \tau} \hat{G}(\omega) (-1)^{\langle x, \omega \rangle} \approx \sum_{\omega: |\hat{G}(\omega)| > \tau} y_\omega \cdot (-1)^{\langle x, \omega \rangle}.$$

Turns out, we can estimate any particular  $\hat{G}(\omega)$  from samples:

$$\hat{G}(\omega) := \frac{1}{2^m} \sum_x G(x) (-1)^{\langle x, \omega \rangle}, \quad \text{so choose a bunch of } x\text{'s at random, and estimate the sum.}$$

But we can't do this for all  $2^m$  coeffs  $\hat{G}(\omega)$ , or else that takes  $\Omega(2^m)$  samples - kinda dumb. Instead we'll just do it for the big ones... but we need to know which those are.

**GOAL.** Given query access to  $G(x)$  and a parameter  $\tau > 0$ , find a set  $S$  of size  $\text{poly}(m)$  so that  $\forall \omega \text{ w/ } |\hat{G}(\omega)| \geq \tau, \omega \in S$ .

NOTE: We'll lose the  $|\cdot|$  in the GOAL for simplicity. By repeating whatever we come up with for  $-G$ , it will be fine.

Now,  $\hat{G}(\omega) \geq \tau$

$$\Leftrightarrow \frac{1}{2^m} \sum_{x \in \mathbb{F}_2^m} G(x) \cdot (-1)^{\langle x, \omega \rangle} \geq \tau$$

remember,  $\epsilon \in \{\pm 1\}$

$$\Leftrightarrow \frac{1}{2^m} \left( |\{x: G(x) = (-1)^{\langle x, \omega \rangle}\}| - |\{x: G(x) \neq (-1)^{\langle x, \omega \rangle}\}| \right) \geq \tau$$

$$\Leftrightarrow \frac{1}{2^m} \left( 2 |\{x: G(x) = (-1)^{\langle x, \omega \rangle}\}| - 1 \right) \geq \tau$$

$$\Leftrightarrow \frac{1}{2^m} |\{x: G(x) = (-1)^{\langle x, \omega \rangle}\}| \geq \frac{1}{2} + \frac{\tau}{2}$$

$$\Leftrightarrow \frac{1}{2^m} |\{x: g(x) = \langle x, \omega \rangle\}| \geq \frac{1}{2} + \frac{\tau}{2} \quad \text{where } G(x) = (-1)^{g(x)}, \text{ aka, } g(x) = \begin{cases} 0 & G(x)=1 \\ 1 & G(x)=-1 \end{cases}$$

$$\Leftrightarrow d(g, l_\omega) \leq \frac{1}{2} - \frac{\tau}{2}, \quad \text{where } l_\omega(x) = \langle x, \omega \rangle \text{ and } (l_\omega(x_1), l_\omega(x_2), \dots, l_\omega(x_{2^n})) \text{ is a Hadamard codeword!}$$

**NEW GOAL.** Given query access to a received word  $g: \mathbb{F}_2^m \rightarrow \mathbb{F}_2$ , find all the Hadamard codewords  $(\langle w, x_1 \rangle, \dots, \langle w, x_{2^m} \rangle) = (l_w(x_1), \dots, l_w(x_{2^m}))$  so that  $\delta(g, l_w) \leq \frac{1}{2} - \varepsilon$ .

That is, we'd like to LIST DECODE the Hadamard Code... in SUBLINEAR TIME!

**NOTICE:**  $\text{Dist}(\text{Hadamard Code}) = \frac{1}{2}$ , so we can only uniquely decode up to radius  $1/4$ .  
 (relative) You showed this on HW1.

But we could hope to list-decode up to  $1/2$ . In this case, the Johnson radius is  $J_2(\frac{1}{2}) = \frac{1}{2}(1 - \sqrt{1 - 2 \cdot \frac{1}{2}}) = \frac{1}{2}$ , so we know that the list size isn't too big.

We also know this from the argument with Parsen's Thm earlier.

② GOLDREICH-LEVIN ALG.

To warm up, let's do it for  $\frac{1}{4}$ :

**ALG O.**

Input: query access to  $g: \mathbb{F}_2^m \rightarrow \mathbb{F}_2$ , a parameter  $\varepsilon$ .

Output: The  $w \in \mathbb{F}_2^m$  s.t.  $\delta(g, l_w) \leq \frac{1}{4} - \varepsilon$ , w/ prob  $\geq 99/100$ .

Draw  $\beta_1, \dots, \beta_T \in \mathbb{F}_2^m$  uniformly at random.

For  $i=1, \dots, m$ : Set  $T = O(m/\varepsilon^2)$

For  $t \in 1, \dots, T$ :

Set  $\tilde{w}_i(\beta_t) = g(e_i + \beta) + g(\beta)$

$\tilde{w}_i \leftarrow \text{MAJ}(\tilde{w}_i(\beta_t))$

RETURN  $\tilde{w} = (\tilde{w}_1, \tilde{w}_2, \dots, \tilde{w}_m)$

Notice this alg makes  $T(1+m)$  queries:  
 $g(\beta_t)$  for  $t=1, \dots, T$   
 $g(\beta_t + e_i)$  for  $t \in [T], i \in [m]$

We saw something like this last time.

Why does this work? As we've seen before:

$$\begin{aligned}\mathbb{P}\{\tilde{w}_i(\beta) \text{ is incorrect}\} &\leq \mathbb{P}\{\text{either } g(e_i + \beta) \text{ or } g(\beta) \text{ were in error}\} \\ &\leq \left(\frac{1}{4} - \epsilon\right) + \left(\frac{1}{4} - \epsilon\right) \\ &= \frac{1}{2} - 2\epsilon.\end{aligned}$$

$$\mathbb{P}\{\text{More than } \frac{1}{2} \text{ of the } \tilde{w}_i(\beta) \text{ are incorrect}\}$$

$$= \mathbb{P}\left\{\frac{1}{T} \sum_{t=1}^T \left[\mathbb{1}\{\tilde{w}_i(\beta) \text{ incorrect}\} - \left(\frac{1}{2} - 2\epsilon\right)\right] > 2\epsilon\right\}$$

$$\leq \frac{\frac{1}{T^2} \sum_{t=1}^T \mathbb{E} \left( \mathbb{1}\{\tilde{w}_i(\beta) \text{ incorrect}\} - \left(\frac{1}{2} - 2\epsilon\right) \right)^2}{(2\epsilon)^2} \quad \text{by Chebyshev}$$

$$= \frac{1}{T \cdot 4\epsilon^2} \cdot \left(\frac{1}{2} - 2\epsilon\right) \left(\frac{1}{2} + 2\epsilon\right)$$

$$= \frac{(1 - 16\epsilon^2)}{T \cdot 16 \cdot \epsilon^2}$$

$$\leq \frac{1}{100m} \text{ if we choose } T = \Theta(m/\epsilon^2).$$

Now union bound over all  $i$  and win.



OK, but now we want to do it up to  $\frac{1}{2} - \epsilon$ , not  $\frac{1}{4} - \epsilon$ .

Suppose we had access to a magic genie who will just tell us the correct value  $\langle \omega, \beta_i \rangle$ .  
But we can only ask the genie for  $T$  values.

### ALG 1.

Input: query access to  $g: \mathbb{F}_2^m \rightarrow \mathbb{F}_2$ , a parameter  $\epsilon$ , and a magic genie.

Output: An  $\omega \in \mathbb{F}_2^m$  s.t.  $\delta(g, \omega) \leq \frac{1}{2} - \epsilon$ , w/ prob  $99/100$ .

Draw  $\beta_1, \dots, \beta_T$  uniformly at random. set  $T = O(m/\epsilon^2)$

Ask the genie for  $b_1, \dots, b_T$  so that  $b_i = \langle \omega, \beta_i \rangle$

For each  $i = 1, \dots, m$ :

For  $t \in 1, \dots, T$ :

Set  $\tilde{\omega}_i(\beta_t) = g(e_i + \beta_t) + b_t$

$\tilde{\omega}_i \leftarrow \text{MAJ}(\tilde{\omega}_i(\beta_t))$

RETURN  $\tilde{\omega} = (\tilde{\omega}_1, \tilde{\omega}_2, \dots, \tilde{\omega}_m)$

This alg makes  $T \cdot m$  queries.

Now, the same argument works:

$$\begin{aligned} \mathbb{P}\{ \tilde{\omega}_i(\beta_t) \text{ is incorrect} \} &= \mathbb{P}\{ g(e_i + \beta_t) \text{ incorrect or the genie lied} \} \\ &= \mathbb{P}\{ g(e_i + \beta_t) \text{ incorrect} \} \quad (\text{because genies don't lie}). \\ &\leq \frac{1}{2} - \epsilon, \end{aligned}$$

so everything goes through as before.

The problem: WE DON'T HAVE A GENIE.

### ALG 2.

Input: query access to  $g: \mathbb{F}_2^m \rightarrow \mathbb{F}_2$ , a parameter  $\epsilon$ ,

Output: A list of  $\omega \in \mathbb{F}_2^m$  s.t.  $\delta(g, l_\omega) \leq \frac{1}{2} - \epsilon$ , w/ prob  $99/100$ .

Initialize  $S \leftarrow \emptyset$

For each  $(b_1, \dots, b_T) \in \mathbb{F}_2^T$ :

define  $\text{GENIE}_{b_1, \dots, b_T}(t) = b_t$

Run ALG 1. using this genie to obtain  $\omega$   
Add  $\omega$  to  $S$ .

RETURN  $S$

Why is this a good idea?

- If  $\delta(l_\omega, g) \leq \frac{1}{2} - \epsilon$ , then  $\exists b_1, \dots, b_T$  ( $= \langle \omega, \beta_1 \rangle, \dots, \langle \omega, \beta_T \rangle$ ) so that ALG 1 returns  $\omega$ . Thus  $\omega$  ends up in the list  $S$ .

Why is this a bad idea?

- $|S| = 2^T = 2^{O(m/\epsilon^2)} \geq |\mathbb{F}_2^m|$ .
- But  $S \subseteq \mathbb{F}_2^m$  was supposed to be a small subset.

To fix this, we will use a PSEUDORANDOM genie.

To see what this means, consider the following way of picking the  $\beta$ 's.

- Choose  $\beta_1, \dots, \beta_\ell$  randomly in  $\mathbb{F}_2^m$  [and let  $\ell = \log(T)$ ]
- For  $A \subseteq [\ell]$ , define  $\beta_A := \sum_{i \in A} \beta_i$
- Now I have  $2^\ell = T$  different values of  $\beta$ .

CLAIM.  $\{\beta_A : A \subseteq [\ell]\}$  are PAIRWISE INDEPENDENT.  
aka, for any  $A \neq A'$ ,  $\beta_A$  and  $\beta_{A'}$  are independent.

proof.

$$\beta_A = \beta_{A'} + \underbrace{\sum_{t \in A \Delta A'} \beta_t}_{\text{uniformly random and independent from } \beta_{A'}} = \text{something uniformly random and indep. from } \beta_{A'}$$

- Notice that our correctness argument before never used the fact that the  $\beta_i$  were fully independent: for Chebyshev we only needed pairwise independence.
- So ALG1. works just fine with these  $\beta$ 's!

### ALG 3.

Input: query access to  $g: \mathbb{F}_2^m \rightarrow \mathbb{F}_2$ , a parameter  $\epsilon$ , and a magic genie.

Output: An  $w \in \mathbb{F}_2^m$  s.t.  $\delta(g, l_w) \leq \frac{1}{2} - \epsilon$ , w/ prob  $99/100$ .

Draw  $\beta_1, \dots, \beta_\ell$  uniformly at random,  $\leftarrow \ell = \log(m/\epsilon^2) + O(1)$

Ask the genie for  $b_1, \dots, b_\ell$  so that  $b_i = \langle w, \beta_i \rangle$ .

For  $A \subseteq [\ell]$ , let  $\beta_A = \sum_{t \in A} \beta_t$ , let  $b_A = \sum_{t \in A} b_t$ .

For each  $i=1, \dots, m$ :

For  $A \subseteq [\ell]$ :

Set  $\tilde{w}_i(\beta_A) = g(e_i + \beta_A) + b_A$

$\tilde{w}_i \leftarrow \text{MAJ}(\tilde{w}_i(\beta_A))$

RETURN  $\tilde{w} = (\tilde{w}_1, \tilde{w}_2, \dots, \tilde{w}_m)$

This alg makes  $T \cdot m$  queries.

Notice that if the genie is correct about  $b_1, \dots, b_\ell$ , then  $\langle w, \beta_A \rangle = \sum_{t \in A} \langle w, \beta_t \rangle = \sum_{t \in A} b_t = b_A$ ,  
so the genie is correct about  $b_A \forall A \subseteq [\ell]$ .

This alg. is correct for exactly the same reason as before, since the  $\beta_A$  are pairwise independent.

## ALG 4 (GOLDREICH-LEVIN)

Input: query access to  $g: \mathbb{F}_2^m \rightarrow \mathbb{F}_2$ , a parameter  $\epsilon$ ,

Output: A list of  $w \in \mathbb{F}_2^m$  s.t.  $\delta(g, l_w) \leq \frac{1}{2} - \epsilon$ , w/ prob  $99/100$ .

Initialize  $S \leftarrow \emptyset$

For each  $(b_1, \dots, b_\ell) \in \mathbb{F}_2^\ell$  : ← set  $\ell = \log(m/\epsilon^2) + O(1)$

define  $\text{GENIE}_{b_1, \dots, b_\ell}(t) = b_t$

Run ALG 3 using this genie to obtain  $w$   
Add  $w$  to  $S$ .

RETURN  $S$

We have basically already proven:

THM. The Goldreich Levin algorithm makes  $\text{poly}(m/\epsilon)$  queries to  $g$  and returns a list  $S \subseteq \mathbb{F}_2^m$  of size at most  $\text{poly}(m/\epsilon)$  so that,  $\forall w \in \mathbb{F}_2^m$  with  $\delta(l_w, g) \leq \frac{1}{2} - \epsilon$ ,  $\mathbb{P}[w \in S] \geq 99/100$ .

Informal  
COR.

(KUSHILEVITZ- MANSOUR)

If  $G: \mathbb{F}_2^m \rightarrow \{\pm 1\}$  is a Boolean function, then we can estimate

$$\tilde{G}(x) \approx \sum_{w: |\hat{G}(w)| > \tau} \hat{G}(w) \cdot (-1)^{\langle x, w \rangle}$$

using  $\text{poly}(m/\tau)$  queries, whp.

### ③ LOCAL LIST DECODING.

What we just saw was a LOCAL LIST DECODING ALGORITHM.

DEF.  $C \subseteq \Sigma^n$  is  $(Q, \rho, L)$ -LOCALLY LIST DECODABLE if:

There is a randomized algorithm  $\mathcal{A}$  that outputs at most  $L$  other algs  $B_1, \dots, B_L$  so that:

- $\forall i \in [L], B_i$  takes an input  $j \in [n]$ , uses at most  $Q$  queries to  $g \in \Sigma^n$ .

- $\forall g \in \Sigma^n,$

$\forall c \in C$  w/  $\delta(c, g) \leq \rho, \exists i$  s.t.  $\forall j \in [n]:$

$$\Pr \{ B_i(j, \text{access to } g) = c_j \} \geq \frac{\rho}{L}$$

Think of each  $B_i$  as a different genie.

In the previous example, the  $B$ 's were indexed by  $(b_1, b_2, \dots, b_L) \in \mathbb{F}_2^L$ :

GENIE  $B_{(b_1, b_2, \dots, b_L)}$  (query access to  $g$ , eval pt  $\alpha$ ):

$$L \leftarrow \log(1/\epsilon^2) + O(1)$$

For  $A \subseteq [L]:$

$$\tilde{w}_\alpha(\beta_A) = g(\alpha + \beta_A) + \sum_{j \in A} b_j$$

$$\tilde{w}_\alpha \leftarrow \text{MAJ}(\tilde{w}_\alpha(\beta_A) : A \subseteq [L])$$

RETURN  $\tilde{w}_\alpha$

NOTE: This is not quite the same as in our Goldreich-Levin version, since that was supposed to recover all of  $w$ , and this just guesses  $\langle w, \alpha \rangle$ . But the idea is the same.

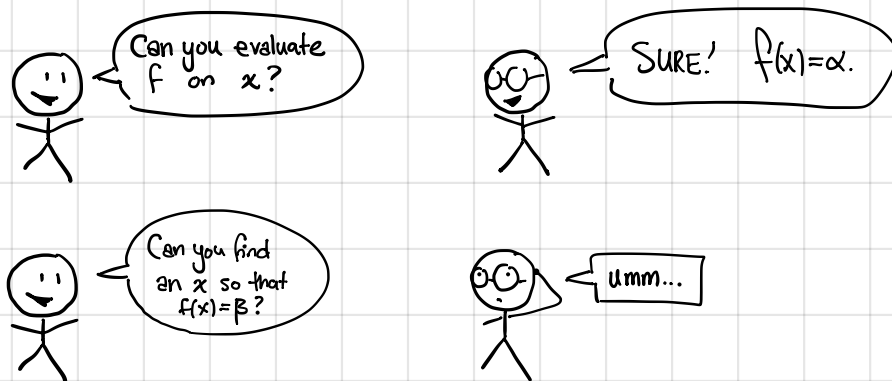
The reason we bother to give LOCAL LIST DECODING a name is because it has many applications. We've already seen one in learning theory, and here's another:

#### ④ PRGs from OWFs

(This is what Goldreich + Levin were interested in).

WARNING: This will be extra handwavy.

"DEF." A ONE-WAY FUNCTION (OWF) is a function that is easy to apply by hand to invert.



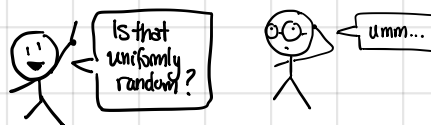
Intuitively, a OWF gives a problem that's hard to solve but easy to check, and that's what  $P \neq NP$  means.

- We don't know if OWFs exist. In fact,  $\exists \text{OWF} \Rightarrow P \neq NP$ .
- But there are several candidates: factoring, discrete log, etc.
- And if a OWF exists, we can do some cool things with it.

#### "DEF" PSEUDORANDOM GENERATOR.

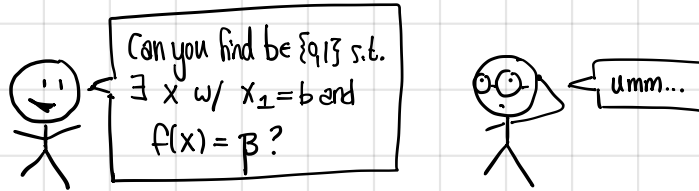
A PRG has output that is not very random, but is computationally difficult to distinguish from uniform.

short seed  $\rightarrow$  PRG  $\rightarrow$  loooooong pseudorandom sequence

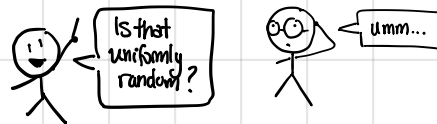
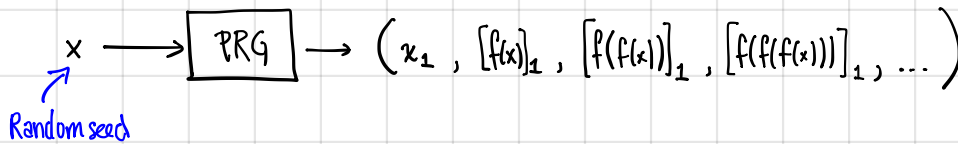


We might try to make a PRG from a OWF as follows:

- Say  $f$  is a OWF,  $f: \mathbb{F}_2^k \rightarrow \mathbb{F}_2^k$  ← Technically,  $f$  should be a ONE-WAY PERMUTATION.
- SUPPOSE that this also means that it's hard to guess  $x_1$  given  $f(x)$ . (\*)

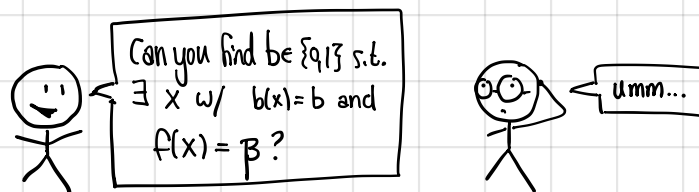


• Now consider the PRG:



- Turns out this is a good PRG, assuming (\*).
- But there is no reason (\*) should be true.

"DEF" A **HARDCORE PREDICATE**  $b(x)$  for  $f(x)$  is a function  $b: \mathbb{F}_2^k \rightarrow \mathbb{F}_2$  so that it's hard to guess  $b(x)$  given  $f(x)$ .



So in order to get PRGs from OWFs, we want a hardcore predicate for our OWF  $f$ .



In fact, we get this from the local list-decodability of the Hadamard code.

"CLAIM." Let  $f: \mathbb{F}_2^m \rightarrow \mathbb{F}_2^m$  be a ONE-WAY PERMUTATION.

Then it's hard to guess  $\langle \alpha, x \rangle$  given  $f(x)$  and  $\alpha$ .

aka, for all  $\alpha \in \mathbb{F}_2^m$ ,  $\langle \alpha, x \rangle$  is a hardcore predicate for  $\tilde{f}: (x, \alpha) \mapsto (f(x), \alpha)$ .

"Pf." Suppose there were some alg  $Q$  so that

$$\mathbb{P}_{\alpha} \{ Q(\alpha, f(x)) = \langle \alpha, x \rangle \} \geq \frac{1}{2} + \epsilon. \quad \text{Aka, } Q \text{ has just a slight advantage.}$$

Then I can get query access to  $g(\alpha) := Q(\alpha, f(x))$ , which is a very noisy version of a Hadamard codeword.

Now I can use my local list-decoding algorithm to obtain a list  $\mathcal{L}$  of  $O(1/\epsilon^2)$  possible  $x$ 's.

Then I compute  $\{f(x) : x \in \mathcal{L}\}$ , find  $x$  s.t.  $f(x) = y$ , and return it.

So  $f$  is easy to invert after all!

## QUESTIONS to PONDER

- ① Can you locally list decode  $RM_q(m, r)$  for  $r < q$ ?
- ② Can you learn Fourier-sparse  $f$ 's from  $\text{poly}(m/\epsilon)$  RANDOM queries?
- ③ Can you think of other applications of local list decoding?