Lecture 28: List Decoding Hadamard Code and Goldreich-Levin Hardcore Predicate

> Lecture 28: List Decoding Hadamard Code and Goldreich-I

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Recall

- Let $H: \{0,1\}^n \to \{+1,-1\}$
- Let:

 $L_{\varepsilon} = \{ S \colon \chi_S \text{ agrees with } H \text{ at } (1/2 + \varepsilon) \text{ fraction of points} \}$

- Given oracle access to H output a list $L \in 2^{[n]}$ such that: For all $S \in L_{\varepsilon}$, we have: $\Pr[S \in L] \ge 1/2$. The probability here is over the internal randomness of the algorithm generating L
- This procedure is identical to *list decoding* of Hadamard Code (Hadamard code is a linear code that maps the message S ⊆ [n] to χ_S ∈ {+1, −1}^{2ⁿ})

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• Let *H* be a oracle that agrees with χ_S at every $x \in \{0,1\}^n$

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function Basic-Decode(H)
for i from 1 to n do
a_i = H(e_i)
end for
Output (a_1, \ldots, a_n)
end function
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• Reconstruction of S: If $a_i = -1$ then $i \in S$; otherwise $i \notin S$

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 Let H be an oracle that agrees with χ_S (for some S ⊆ [n]) at some 3/4 + ε fraction of inputs

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function Unique-Decode(H)

for i from 1 to n do

for j from 1 to t do

Choose r_{i,j} \stackrel{\$}{\leftarrow} \{0,1\}^n

Let a_{i,j} = H(r_{i,j} + e_i) \cdot H(r_{i,j})

end for

Let a_i = Maj\{a_{i,1}, \dots, a_{i,t}\}

end for

Output (a_1, \dots, a_n)

end function
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• Let
$$E_{i,j} = \mathbf{1}_{(a_{i,j} = \chi_S(e_i))}$$

- Note that $\Pr[E_{i,j} = 0] \leq (1/4 \varepsilon) + (1/4 \varepsilon) = 1/2 2\varepsilon$ and, for each *i*, the $E_{i,j}$ s are i.i.d. variables
- So, $a_i = \chi_S(e_i)$ except with probability $\exp(-\Theta(t/\varepsilon^2))$. Using $t = \Theta\left(\frac{1}{\varepsilon}^2 \log n\right)$ we can make this failure probability $1/n^2$
- So, $a_i = \chi_S(e_i)$ for all $i \in [n]$, except with probability 1/n

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- Consider any $S \in L_{arepsilon}$
- Given H that agrees with χ_S at $1/2 + \varepsilon$ fraction of inputs, we want to *mimic* another *more precise* oracle \widetilde{H} that agrees with 7/8 fraction of inputs
- And we will successfully mimic \widetilde{H} with probability at least 3/4
- So, given access to \widetilde{H} , the unique decoder can recover S, except with probability 1/n
- So, we recover S with probability $3/4 1/n \ge 1/2$

Mimicking the More Precise Oracle in a Hypothetical World

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• Consider S \in L_{\varepsilon}
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```
function Mimic-Hypothetical(H)
    for i \in [\alpha] do
        Sample x_i \stackrel{\$}{\leftarrow} \{0,1\}^n
        Assume that we have magically obtained b_i = \chi_S(x_i)
    end for
    Define the following oracle H:
    function H(H, z)
        for i \in [\alpha] do
            a_i = H(z + x_i) \cdot b_i
        end for
        Return a = Mai\{a_1, \ldots, a_n\}
    end function
end function
```

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Analysis of Mimicking

• Note that $a_i = \chi_S(a)$ with probability $1/2 + \varepsilon$

- Since a_i s are i.i.d. we have that $a = \chi_S(z)$, except with probability $\exp(-\Theta(\alpha/\varepsilon^2))$
- So, choosing $\alpha = O(1/\varepsilon^2)$ we can achieve the correctness probability to be 31/32
- Formally:

$$\Pr_{z,x_1,\ldots,x_\alpha}[a=\chi_{\mathcal{S}}(z)] \geqslant 31/32$$

• Using averaging-argument:

$$\Pr_{x_1,\ldots,x_{\alpha}}\left[\Pr_{z}[a=\chi_{\mathcal{S}}(z)] \geqslant 7/8\right] \geqslant 3/4$$

• Summary: Over the random choices of x_1, \ldots, x_{α} we succeed with probability at least 3/4 in implementing an oracle \tilde{H} that agrees with χ_S at at least 7/8 fraction of inputs

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Partial Solution

- We enumerate all $2^{O(1/arepsilon^2)}$ possible b_1,\ldots,b_lpha bits
- And we execute unique decoding algorithm with the corresponding \widetilde{H} oracle
- Add the output of the unique decoding algorithm to the list L

• The list size is at most $2^{O(1/\varepsilon^2)}$ and for all $S \in L_{\varepsilon}$, with probability $\ge 1/2$ we have $S \in L$

• This procedure is inefficient if $1/\varepsilon$ is super-logarithmic in n

Changing the Analysis of the Mimicking Algorithm

- We do not need $\{a_1, \ldots, a_{\alpha}\}$ to be i.i.d.
- We just need them to be pairwise-independent
- In this case, we can apply Chebyshev's inequality
- The probability of $a \neq \chi_S(z)$ is defined as follows: Let $X_i = a_i \cdot \chi_S(z)$

$$\Pr\left[\sum_{i\in[\alpha]} X_i \leqslant \frac{\alpha}{2}\right] \leqslant \Pr\left[\left|\sum_{i\in[\alpha]} X_i - \left(\frac{1}{2} + \varepsilon\right)\alpha\right| \leqslant \varepsilon\alpha\right]$$
$$\leqslant \frac{\operatorname{Var}\left[\sum_{i\in[\alpha]} X_i\right]}{\varepsilon^2 \alpha^2} = \Theta\left(\frac{1}{\varepsilon^2 \alpha}\right)$$

• Choose $\alpha = O(1/\varepsilon^2)$ we can make the success probability $\geqslant 31/32$ as earlier

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Pairwise-Independent Distributions

- Let u_1, \ldots, u_{β} be uniform random strings from $\{0, 1\}^n$
- Let v_1,\ldots,v_eta be particular values of elements in $\{+1,-1\}$

• Let
$$\alpha = 2^{\beta} - 1$$

- Interpret every $i \in [\alpha]$ as the characteristic vector of the subset of $[\beta]$
- Define $x_i := \bigoplus_{k \in i} u_k$, for $i \in [\alpha]$
- Define $b_i := \prod_{k \in i} v_k$, for $i \in [\alpha]$
- Note that x_i and $x_{i'}$ are pairwise independent for $i \neq i'$
- Similarly, note that a_i and $a_{i'}$ are pairwise independent for $i \neq i'$
- To perform the "mimicking algorithm" choose random u_1, \ldots, u_β and enumerate all possible v_1, \ldots, v_β
- The number of possible enumerations is $2^eta=lpha+1={\it O}(1/arepsilon^2)$

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- We have $O(\frac{1}{\varepsilon^2})$ iterations for each setting of v_1, \ldots, v_β
- Each iteration of unique decoding takes $O(\frac{1}{r^2} n \log n)$ time
- Overall time-complexity: $O(\frac{1}{\varepsilon^4} n \log n)$
- The list size is $\leqslant 2^{eta} = O(rac{1}{arepsilon^2})$

Lemma (Hardcore Lemma)

Let $f: \{0,1\}^n \to \{0,1\}^m$ be a one-way function. Let X and R be a uniform random strings from $\{0,1\}^n$. Then, given (f(X), R) no polynomial time algorithm cannot predict $B := R \cdot X$ with $\varepsilon \ge 1/\text{poly}(n)$ advantage.

- $B = R \cdot X$ is known as the hardcore predicate
- Proof Idea: Proof by Contradiction. Given an adversary that predicts *B*, we use the adversary as an oracle to recover *x* using the list-decoding algorithm described previously

Proof

• Suppose there exists an adversary A that, given (f(X), R), can predict the random variable B with $\varepsilon = 1/\text{poly}(n)$ advantage

$$\Pr_{x,r}[A(f(x),r)=r\cdot x] \ge (1/2+\varepsilon)$$

• Using an averaging argument:

$$\Pr_{x}\left[\Pr_{r}[A(f(x),r)=r\cdot x] \ge (1/2+\varepsilon/2)\right] \ge \varepsilon/2$$

Call such an input x as a good input

- Conditioned on a good input x, the adversary A is an oracle that agrees with the function χ_x at $(1/2 + \varepsilon/2)$ fraction of inputs
- Using this oracle, recover x from the list L with probability 1/2 in $poly(m + n + 1/\varepsilon)$ time using Goldreich-Levin List-Decoding Algorithm
- With probability $(\varepsilon/2) \cdot (1/2)$ we successfully recover x in polynomial time and violate the one-way-ness of f