

Lecture 11

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1 Lecture Topic

Testing functions for the k-junta property.

1.1 Last Week

We found a property tester for dictator functions.

We defined k-junta functions and introduced a k-junta testing algorithm.

2 Definitions and Lemmas From Last Week

Definition 1. A function f is a k-junta function if its output depends only on k or less input variables.

Definition 2. $X_s \equiv \text{ordered-list}(X_i \mid i \in S)$

Definition 3. $X_s Y_{\bar{s}} \equiv Z$ s.t. $\forall_{i \in S} Z_i = X_i$ and $\forall_{i \in \bar{S}} Z_i = Y_i$

Definition 4. $\text{Inf}_f(s) \equiv 2Pr_{x,y|x_s=y_{\bar{s}}} [f(x) \neq f(y)]$

Last week we showed the following:

$$\text{Inf}_f(s) \equiv 2Pr_{x,y|x_s=y_{\bar{s}}} [f(x) \neq f(y)] = \sum_{T|S \cap T \neq \emptyset} \hat{f}(T)^2 = \sum_T \hat{f}(T)^2 - \sum_{T \subseteq \bar{S}} \hat{f}(T)^2 = 1 - \sum_{T \subseteq \bar{S}} \hat{f}(T)^2$$

where the last equality is derived from the boolean Parseval equality.

Lemma 5. If a function f is ϵ -far from being a k-junta then

$$\forall_{J:|J| \leq k} \text{Inf}_f([n] \setminus J) \geq \epsilon$$

3 Main Theorem

3.1 Algorithm For Testing The k-junta Property

given k, ϵ

- randomly partition $1..n$ into s parts $I_1 \dots I_s$ where $s = \text{poly}(k, \frac{1}{\epsilon})$
- $R \leftarrow \emptyset$
- repeat up to $r = O(\frac{k}{\epsilon})$ times
 - generate x, y randomly s.t. $x_R = y_R$
 - if $f(x) \neq f(y)$
 - * use binary search to find relevant I_j
 - * $R \leftarrow R \cup I_j$
 - if R has more than k relevant parts, REJECT
- ACCEPT

Note: The binary search here is performed on the partitions rather than the input bits themselves. That's why we obtain a running time, which is independent from n

3.2 Main Lemma

If f is ϵ -far from k -junta, and \mathcal{I} is a random partition into $s = \frac{10^{20}k^9}{\epsilon^5}$ parts,

we can get with probability $\geq \frac{5}{6}$:

$\forall J$ s.t. J is a union of at most k parts of \mathcal{I} , $\text{Inf}_f([n] \setminus J) \geq \frac{\epsilon}{2}$

We'll first see how we can use the lemma to get our desired result.

Claim 6. There exists an Algorithm \mathcal{T} that uses $O(\frac{k}{\epsilon} + k \log(k))$ queries such that:

- if a function f is a k -junta, \mathcal{T} will always pass on f
- if a function f is ϵ -far from being a k -junta, $\text{Pr}[\mathcal{T} \text{ will fail } f] \geq \frac{2}{3}$

Proof. Let f be a function.

- If f is a k -junta \implies any partition of $1..n$ can have at most k relevant partitions (in respect to f),

and hence the algorithm will ACCEPT.

- If f is ϵ -far from being a k -junta $\implies \text{Pr}_{x,y|x_R=y_R} [f(x) \neq f(y)] = \text{Inf}_f([n] \setminus R) \geq \frac{\epsilon}{2}$

where the first equality is simply the definition and the second inequality is derived from the **main-lemma**.

$\implies E[\text{time to find more than } k+1 \text{ relevant parts}] \leq (k+1)(\frac{4}{\epsilon})$

Now, using **Markov's inequality** we get $\text{Pr}[\text{don't find more than } k \text{ parts after } 6(\frac{4}{\epsilon})(k+1)] \leq \frac{1}{6}$

In addition to the error probability of $\frac{1}{6}$ that we get from **Markov's inequality**, we also get an error probability of $\frac{1}{6}$ from the **main-lemma**,

so altogether we get an error probability of $\frac{1}{3} \implies \text{Pr}[\mathcal{T} \text{ will fail } f] \geq \frac{2}{3}$

□

3.3 Proving The Main Lemma

What's left is to prove the following:

for most partitions \mathcal{T} & $\forall J|J$ is a union of k parts of \mathcal{T} s.t. J is a union of k parts of \mathcal{T} :

$$\text{Inf}_f([n] \setminus R) \stackrel{\text{by lemma}}{=} \sum_{S \subseteq [n]} \hat{f}(S) - \sum_{S \subseteq J} \hat{f}(S)^2 \stackrel{?}{\geq} \epsilon' \equiv \frac{\epsilon}{2}$$

We pick $\theta \equiv (\epsilon^2 \log(\frac{k}{\epsilon})) / 10^9 k^4$

Definition 7. $\text{Inf}_f^{\leq k}(S) \equiv \sum_{T|T \cap S \neq \emptyset \text{ and } |T| \leq 2k} \hat{f}(T)^2$

Definition 8. $\text{Inf}_f^{> k}(S) \equiv \sum_{T|T \cap S \neq \emptyset \text{ and } |T| > 2k} \hat{f}(T)^2$

Definition 9. $\text{Inf}_f(i) \equiv \text{Inf}_f(\{i\})$

3.3.1 Nice Property

$$Inf_f(S) \leq Inf_f(S \cup S') \leq Inf_f(S) + Inf_f(S')$$

The first inequality derives from the fact that $S \cup S'$ is possibly a bigger set than just S , and since $Inf_f(S)$ is a sum of positive elements ($Inf_f(S) = \sum_{T|T \cap S \neq \phi} \hat{f}(T)^2$).

The second inequality is true since:

$$\begin{aligned} \sum_{T|T \cap (S \cup S') \neq \phi} &= \sum_{T|T \cap S \neq \phi \text{ and } T \cap S' = \phi} + \sum_{T|T \cap S \neq \phi \text{ and } T \cap S' \neq \phi} + \sum_{T|T \cap S = \phi \text{ and } T \cap S' \neq \phi} \\ &= \text{merging the first two} \sum_s \sum_{T|T \cap S \neq \phi} + \sum_{T|T \cap S = \phi \text{ and } T \cap S' \neq \phi} \\ &\leq \text{adding another} \sum_{T|T \cap S \neq \phi} + \sum_{T|T \cap S = \phi \text{ and } T \cap S' \neq \phi} + \sum_{T|T \cap S \neq \phi \text{ and } T \cap S' \neq \phi} \\ &= \text{merging the last two} \sum_s \sum_{T|T \cap S \neq \phi} + \sum_{T|T \cap S' \neq \phi} \end{aligned}$$

and so we get that $Inf_f(S \cup S') \leq Inf_f(S) + Inf_f(S')$.

3.3.2 Lemma

$$\sum_{i \in [n]} Inf_f^{\leq 2k}(i) \leq 2k$$

$$\begin{aligned} \text{Proof. } \sum_{i \in [n]} Inf_f^{\leq 2k}(i) &= \sum_{i \in [n]} \sum_{T : i \in T \text{ and } |T| \leq 2k} \hat{f}(T)^2 \\ &= \sum_{T : |T| \leq 2k} |T| \hat{f}(T)^2 \leq 2k \sum_{T : |T| \leq 2k} \hat{f}(T)^2 \leq 2k \sum_T \hat{f}(T)^2 = 2k \end{aligned}$$

□

3.3.3 “Heavy” Index Groups

Definition 10. we define the group H to be the group of the “heavy” indices, in the sense that they have a large influence on f :

$$H = \left\{ i \in [n] \mid Inf_f^{\leq 2k}(i) \geq \Theta \right\}$$

Corollary 11. $|H| \leq \frac{2k}{\Theta}$

Proof. otherwise we get:

$$\sum_{i \in [n]} Inf_f^{\leq 2k}(i) \geq \sum_{i \in H} Inf_f^{\leq 2k}(i) \geq \sum_{i \in H} \Theta = |H| \Theta \geq \frac{2k}{\Theta} \Theta = 2k$$

in contradiction to the last lemma.

□

The previous **corollary** implies that there aren’t too many “heavy” indices.

(less than $\frac{2k}{\Theta}$, which is a lot smaller than s).

This is the reason that with a high probability (more than $\frac{17}{18}$) each of these “heavy” indices is on a different part I_i :

$$\text{Recall that } s = \frac{10^{20}k^9}{\epsilon^5} \text{ and that } \theta = \frac{\epsilon^2 \log(\frac{k}{\epsilon})}{10^9 k^4} \implies \frac{72k^2}{\Theta^2} = \frac{72k^2 \cdot 10^{18} k^8}{\epsilon^4 \log^2(\frac{k}{\epsilon})} \leq \frac{10^{20} k^{10}}{\epsilon^5} = s$$

since $s \geq \frac{72k^2}{\Theta^2}$,

$$Pr \left[\forall_{i \in [n]} |H \cap I_i| \leq 1 \right] \geq 1 - Pr \left[\exists_{i \in [n]} \text{s.t. } |H \cap I_i| > 2 \right] \geq 1 - \binom{|H|}{2} \cdot \frac{1}{s} \geq 1 - \left(\frac{2k}{\Theta} \right)^2 \cdot \frac{\Theta^2}{72k^2} \geq \frac{17}{18}$$

In other words, we have just shown that with high probability, each set of the partition gets at most one member of H .

3.3.4 Breaking Down The influence

We define Partition subsets of J :

$$\begin{aligned}\mathcal{H} &= \{S \subseteq J \mid S \subseteq J \cap H, \ |S| \leq 2k\} \\ \mathcal{L} &= \{S \subseteq J \mid S \not\subseteq J \cap H, \ |S| \leq 2k\} \\ \mathcal{B} &= \{S \subseteq J \mid |S| > 2k\}\end{aligned}$$

We now present the influence in terms of the above subsets:

$$Inf_f([n] \setminus R) = \sum_{S \subseteq [n]} \hat{f}(S) - \sum_{S \subseteq J} \hat{f}(S)^2 = \underbrace{\sum_{S \subseteq [n]} \hat{f}(S)^2 - \sum_{S \in \mathcal{H}} \hat{f}(S)^2}_{t_1} - \underbrace{\sum_{S \in \mathcal{L}} \hat{f}(S)^2}_{t_2} - \underbrace{\sum_{S \in \mathcal{B}} \hat{f}(S)^2}_{t_3}$$

We denote the above marked terms as t_1, t_2, t_3 respectively.

We'll prove that t_1 is sufficiently large and t_2, t_3 are sufficiently small to prove the main lemma.

1. Evaluating t_1

$$t_1 = \sum_{S \subseteq [n]} \hat{f}(S)^2 - \sum_{S \in \mathcal{H}} \hat{f}(S)^2 \geq \sum_{S \subseteq [n]} \hat{f}(S)^2 - \sum_{S \subseteq J \cap H} \hat{f}(S)^2 = Inf_f([n] - J \cap H)$$

From 3.3.3, we get that $|J \cap H| \leq k$, (because $Pr[\forall i \in [n] \mid H \cap I_i \leq 1] \geq \frac{17}{18}$)

and by last week's lemma we get:

$$t_1 \geq \epsilon$$

2. Evaluating t_2

Claim 12. If $s \geq \frac{16k^2}{\epsilon}$ and $\theta \leq \frac{\epsilon^2}{64} k^2 \log(18s)$, then with probability at least $\frac{17}{18}$

$$Inf_f^{\leq 2k}(I_i \setminus H) \leq \frac{\epsilon}{4k} \text{ for every } i \in [s]$$

First we'll use the claim to provide an upper bound on t_2 .

The claim implies that $\forall J$ s.t. J is a union of at most k parts of \mathcal{I} , the following is true:

$$\sum_{S \in \mathcal{L}} \hat{f}(S)^2 \leq Inf_f^{\leq 2k}(J \setminus H) \leq_{\text{nice property+previous claim}} k \cdot \frac{\epsilon}{4k} = \frac{\epsilon}{4}$$

Explanation of the first inequality:

$$\text{By definitions, } Inf_f^{\leq 2k}(J \setminus H) = \underbrace{\sum_{S \mid S \cap (J \setminus H) \neq \emptyset, \ |S| \leq 2k} \hat{f}(S)^2}_{\tau}$$

$$\text{and } \mathcal{L} = \{S \subseteq J \mid S \not\subseteq J \cap H, \ |S| \leq 2k\} = \{S \subseteq J \mid S \cap (J \setminus H) \neq \emptyset, \ |S| \leq 2k\}$$

$\mathcal{L} \subseteq \mathcal{T}$ since \mathcal{T} may contain elements that are not in J .

$$\sum_{S \in \mathcal{L}} \hat{f}(S)^2 \leq \sum_{S \in \mathcal{T}} \hat{f}(S)^2 \text{ and the inequality is derived.}$$

We'll now prove the claim by using Hoeffding's inequality:

$$Pr[s - E(s) \geq nt] \leq e^{\frac{-2n^2k^2}{\sum_{i=1}^n (b_i - a_i)^2}} \text{ where } s = \sum_{i=1}^n x_i \text{ and } a_i \leq x_i \leq b_i.$$

Proof. For each $i \in [s]$ we define: $\forall j \in [n] \ X_j \equiv \begin{cases} Inf_f^{\leq 2k}(j) & j \in I_i \setminus H \\ 0 & \text{otherwise} \end{cases}$

$$Inf_f^{\leq 2k}(I_i \setminus H) \leq_{\text{nice property}} \sum_{j \in I_i \setminus H} Inf_f^{\leq 2k}(j) \leq \sum_{j \in [n]} X_j$$

Thus, $E[\sum_{j \in [n]} X_j] \leq \sum_{j \in [n] \setminus H} \text{Inf}_{\bar{f}}^{\leq 2k}(j) \cdot \underbrace{E[1_{j \in I_i}] = \frac{1}{s}}_{\frac{1}{s}} \sum_{j \in [n] \setminus H} \text{Inf}_{\bar{f}}^{\leq 2k}(j) \stackrel{\text{from 3.3.2}}{\leq} \frac{2k}{s} \leq \frac{\epsilon}{8k}$.

We'll now use Hoeffding's inequality where $a_i = 0$ and $b_i = \text{Inf}_{\bar{f}}^{\leq 2k}(i)$

$$\sum_{i \in [n] \setminus H} (b_i - a_i)^2 = \sum_{i \in [n] \setminus H} \text{Inf}_{\bar{f}}^{\leq 2k}(i) \leq \max_{i \in [n] \setminus H} (\text{Inf}_{\bar{f}}^{\leq 2k}(i)) \cdot \sum \text{Inf}(i) \leq \theta \cdot 2k$$

$$\Pr[\underbrace{\sum_{j \in [n]} x_j}_{\frac{\epsilon}{8k}} - E[\sum_{j \in [n]} X_j] \geq \frac{\epsilon}{8k}] = \Pr[\sum x_i \geq \frac{\epsilon}{4k}] \leq e^{-\frac{(\frac{\epsilon}{4k})^2}{\theta \cdot 2k}} \leq \frac{1}{18s}$$

The last inequality applies for each index i , and since we have s indices, summing them up will produce probability of no more than $\frac{1}{18}$. \square

3. Evaluating t_3

All $S \in \mathcal{B}$ are subsets of J , which is a union of less than k parts

Claim 13. If $S > \frac{72ek}{\epsilon}$, then with probability of at least $\frac{17}{18}$,

$$\sum_{S \mid |S| \geq 2k \wedge s \in \text{less than } k \text{ parts of } \mathcal{I}} \hat{f}(S)^2 \leq \frac{\epsilon}{4}$$

Proof. Since $|S| \geq 2k$ We get,

$$\Pr[\text{All elements of } S \text{ sent to no more than } k \text{ parts in } \mathcal{I}] \leq \binom{|S|}{k} \left(\frac{k}{|S|}\right)^{2k+1} \leq \left(\frac{e|S|}{k}\right)^k \left(\frac{k}{|S|}\right)^{2k+1} \leq \frac{\epsilon}{72}$$

We can now put an upper bound on the term's expectation:

$$E \left[\sum_{S \mid |S| \geq 2k \wedge s \in \text{less than } k \text{ parts of } \mathcal{I}} \hat{f}(S)^2 \right] \leq \sum \hat{f}(S)^2 \cdot \Pr[1_{S \in k \text{ parts of } \mathcal{L}}] \leq \frac{\epsilon}{72} \sum \hat{f}(S)^2$$

=by boolean parseval $\frac{\epsilon}{72} \cdot 1 \leq \frac{k}{72}$ \square

To recap,

$$\text{Inf}_{\bar{f}}([n] \setminus R) = \underbrace{\sum_{S \subseteq [n]} \hat{f}(S)^2}_{t_1} - \underbrace{\sum_{S \in \mathcal{H}} \hat{f}(S)^2}_{t_2} - \underbrace{\sum_{S \in \mathcal{L}} \hat{f}(S)^2}_{t_3} - \underbrace{\sum_{S \in \mathcal{B}} \hat{f}(S)^2}_{t_4} \equiv t_1 - t_2 - t_3 \geq \epsilon - \frac{\epsilon}{4} - \frac{\epsilon}{4} \geq \frac{\epsilon}{2}$$

The last inequality proves our main lemma, and in turn, our main theorem.