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MIT 6.859 – Randomness and Computation, with Ronitt Rubinfeld Collaborators: Benjamin Rossman

Problem Set 2 – Solutions

Problem 1

(a) Assume f, h and g are linear consistent, then there is some linear $\phi(\cdot)$ such that:

$$f(x) = \phi(x) + a_f$$
$$g(y) = \phi(y) + a_g$$
$$h(x+y) = \phi(x+y) + a_h$$

where $a_f + a_g = a_h$, therefore:

$$f(x) + g(y) = \phi(x) + a_f + \phi(y) + a_g$$
$$= \phi(x + y) + a_h$$
$$= h(x + y)$$

In the other direction, define $\phi(x) = h(x) - h(0)$. Check that $\phi(\cdot)$ is linear:

$$\begin{aligned} \phi(x) + \phi(y) &= h(y) - h(0) + h(y) - h(0) \\ &= f(x) + g(0) - h(0) + f(0) + g(y) - h(0) \\ &= (f(x) + g(y)) + (f(0) + g(0)) - 2h(0) \\ &= h(x + y) + h(0) - 2h(0) \\ &= h(x + y) - h(0) \\ &= \phi(x + y) \end{aligned}$$

For f, define $a_f = h(0) - g(0)$, then:

$$f(x) = h(x) - g(0) = h(x) - h(0) + a_f = \phi(x) + a_f$$

For g, define $a_g = h(0) - f(0)$, then:

$$g(x) = h(x) - f(0) = h(x) - h(0) + a_g = \phi(x) + a_g$$

For h, define $a_h = h(0)$, then:

$$h(x) = h(x) - h(0) + h(0) = h(x) - h(0) + a_h = \phi(x) + a_h$$

Finally, verify that:

$$a_f + a_g = h(0) - g(0) + h(0) - f(0)$$

= $2h(0) - (f(0) + g(0))$
= $h(0)$
= a_h

(b) We know that $d(f, \chi_S) = (1 - \hat{f}(S))/2$, we also know that $d(f, -\chi_S) = 1 - d(f, \chi_S)$, therefore $d(f, -\chi_S) = (1 + \hat{f}(S))/2$. Hence, we can compute:

$$\min_{S} d(f, \pm \chi_{S}) = \min\left\{\min_{S} d(f, \chi_{S}), \min_{S} d(f, -\chi_{S})\right\}$$
$$= \min\left\{\frac{1}{2}\left(1 - \max_{S} \hat{f}(S)\right), \frac{1}{2}\left(1 + \min_{S} \hat{f}(S)\right)\right\}$$
$$= \frac{1}{2}\left(1 - \max_{S} |\hat{f}(S)|\right)$$

And therefore (we will need this later):

$$\min_{S} d(f, \pm \chi_S) \le \delta \Leftrightarrow \max_{S} |\hat{f}(S)| \ge 1 - 2\delta$$

(c) Begin by observing that:

$$\mathbf{I}\big[f(x)g(y) \neq h(xy)\big] = \frac{1 - f(x)g(y)h(xy)}{2}$$

Furthermore, (following the lecture notes):

$$\begin{aligned} \mathbf{E}_{x,y}\left[f(x)g(y)h(xy)\right] &= \mathbf{E}_{x,y}\left[\sum_{S}\hat{f}(S)\chi_{S}(x)\sum_{T}\hat{g}(T)\chi_{T}(y)\sum_{U}\hat{h}(U)\chi_{U}(xy)\right] \\ &= \sum_{S,T,U}\hat{f}(S)\hat{g}(T)\hat{h}(U)\mathbf{E}_{x,y}\left[\chi_{S}(x)\chi_{T}(y)\chi_{U}(xy)\right] \\ &= \sum_{S}\hat{f}(S)\hat{g}(S)\hat{h}(S) \end{aligned}$$

Notice:

$$\mathbf{Pr}_{x,y}[f(x)g(y) \neq h(xy)] = \mathbf{Pr}_{x,y}\left[\mathbf{I}[f(x)g(y) \neq h(xy)] = 1\right]$$
$$= \mathbf{E}_{x,y}\left[\mathbf{I}[f(x)g(y) \neq h(xy)]\right]$$
$$= \mathbf{E}_{x,y}\left[\frac{1 - f(x)g(y)h(xy)}{2}\right]$$

Therefore $\mathbf{Pr}_{x,y}[f(x)g(y) \neq h(xy)] \leq \delta$ amounts to:

$$\sum_{S} \hat{f}(S)\hat{g}(S)\hat{h}(S) \ge 1 - 2\delta$$

On the other hand, we derive that:

$$\begin{split} \max_{S} |\hat{f}(S)|^{2} &= \max_{S} |\hat{f}(S)|^{2} \left(\sum_{T} |\hat{g}(T)|^{2} \right) \left(\sum_{T} |\hat{h}(T)|^{2} \right) \text{ by Parseval} \\ &\geq \max_{S} |\hat{f}(S)|^{2} \left(\sum_{T} |\hat{g}(T)| \cdot |\hat{h}(T)| \right)^{2} \text{ by Cauchy-Schwarz} \\ &= \left(\max_{S} |\hat{f}(S)| \sum_{T} |\hat{g}(T)| \cdot |\hat{h}(T)| \right)^{2} \\ &\geq \left(\sum_{S} |\hat{f}(S)\hat{g}(S)\hat{h}(S)| \right)^{2} \\ &\geq \left(\sum_{S} \hat{f}(S)\hat{g}(S)\hat{h}(S) \right)^{2} \end{split}$$

Taking the square root gives us:

$$\max_{S} |\hat{f}(S)| \ge \sum_{S} \hat{f}(S)\hat{g}(S)\hat{h}(S) \ge 1 - 2\delta$$

Which further implies that:

$$\min_{S} d(f, \pm \chi_S) \le \delta$$

The proof for g and h is identical.

Problem 2

Let $f : {\pm 1}^n \to {\pm 1}$ be monotone, and for $x \in {\pm 1}^n$, let $x^+ \in {\pm 1}^n$ be x with the *i*-th entry set to +1, and define x^- accordingly. Then we have that:

$$\mathbf{I}[f(x^+) \neq f(x^-)] = \frac{1}{2} (f(x^+) - f(x^-))$$

= $\frac{1}{2} (f(x^+)\chi_{\{i\}}(x^+) + f(x^-)\chi_{\{0\}}(x^-))$

Apply this in the following:

$$\begin{aligned} \operatorname{Inf}_{i}(f) &= \mathbf{Pr}_{x}[f(x) \neq f(x \cdot u_{i})] \\ &= \frac{1}{2^{n}} \sum_{x \in \{0,1\}^{n}} \mathbf{I}[f(x^{+}) \neq f(x^{-})] \\ &= \frac{1}{2^{n}} \sum_{x \in \{0,1\}^{n}} \frac{1}{2} (f(x^{+})\chi_{\{i\}}(x^{+}) + f(x^{-})\chi_{\{0\}}(x^{-})) \\ &= \frac{1}{2^{n}} \sum_{x \in \{0,1\}^{n}} f(x)\chi_{\{i\}}(x) \\ &= \hat{f}(\{i\}) \end{aligned}$$

Problem 3

For any monotone function we have:

$$\inf(f) = \sum_{i \in [n]} \inf_i(f)$$
$$= \sum_{i \in [n]} \hat{f}(\{i\}) \text{ since } f \text{ monotone}$$
$$= \sum_{i \in [n]} \frac{1}{2^n} \sum_{x \in \{\pm 1\}^n} f(x) \cdot x_i$$
$$= \frac{1}{2^n} \sum_{x \in \{\pm 1\}^n} \sum_{i \in [n]} f(x) \cdot x_i$$

Maximizing over all (not just monotone) functions:

$$\max_{f} \inf(f) = \max_{f} \frac{1}{2^{n}} \sum_{x \in \{\pm 1\}^{n}} \sum_{i \in [n]} f(x) \cdot x_{i}$$
$$= \frac{1}{2^{n}} \sum_{x \in \{\pm 1\}^{n}} \max_{f(x)} \sum_{i \in [n]} f(x) \cdot x_{i}$$

The last quantity, when n is odd, is clearly maximized only when:

$$f(x) = \text{maj} \{x_1, \dots, x_n\}$$

Problem 4

Let the graph be random d-regular on the left (with each edge connected uniformly at random on the right). Then let X be a random variable equal to the number of subsets of the left vertices that shrink, i.e. whose neighbor sets on the right are equal or smaller in size. For a fixed left set of size k we have that the probability, p(k), that it shrinks is bounded by (using a union bound over all right subsets of size k):

$$p(k) \le \binom{2n/3}{k} \left(\frac{k}{2n/3}\right)^{kd}$$

Hence, for $\mathbf{E}[X]$ we have:

$$\mathbf{E}[X] \le \sum_{k=1}^{n/2} \binom{n}{k} \binom{2n/3}{k} \left(\frac{k}{2n/3}\right)^{kd}$$
$$\le \sum_{k=1}^{n/2} \left(e^2 \frac{3}{2} \left(\frac{3k}{2n}\right)^{d-2}\right)^k \quad \text{using } \binom{n}{k} \le \left(\frac{ne}{k}\right)^k$$
$$< \sum_{k=1}^{\infty} \left(e^2 \frac{3}{2} \left(\frac{3}{4}\right)^{d-2}\right)^k$$

When $d > \ln(3e^2) / \ln(4/3) + 2$, we have that $e^{2\frac{3}{2}} \left(\frac{3}{4}\right)^{d-2} < 1/2$, and therefore:

$$\mathbf{E}[X] < \sum_{k=1}^{\infty} (1/2)^k < 1$$

Therefore, there exists a graph of left degree $d = \lceil \ln(3e^2) / \ln(4/3) + 3 \rceil$ for which no left subset of size up to n/2 shrinks on the right.