

# Harmonic Analysis of Boolean Functions, and applications in CS

## Lecture 4

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At a previous lecture there was shown that for each function over a discrete cube there exists a single way to represent it as polynomial.

That's:  $\forall f: \{\pm 1\}^n \rightarrow \mathbb{R} \exists f(x) = \sum \hat{f} \prod_{x \in S} x$  That means that:  $I_i(f) = \left\| \frac{f - \sigma_i f}{2} \right\|_2^2$  - directly from computation Looking at the above formula makes a lot of sense. It generalizes the Influence formula of boolean function. In case of boolean function it is probability that  $f(x) \neq f(xe_i)$

$$\sigma_i x_s = \underbrace{x_s i \notin S, -x_s i \in S}_{\text{if } s=i} \quad (1)$$

$\sigma_i$  is linear operator. Thus,

$$I_i(f) = \left\| \frac{f - \sigma_i f}{2} \right\|_2^2 = \left\| \sum_{s: i \in S} f(s) \chi_s \right\|_2^2 = \sum_{i \in S} \hat{f}(s)^2 \quad (2)$$

Thus,

$$I_i(f) = \sum_{S \subseteq [n]} |s| \hat{f}(s)^2 \quad (3)$$

Thus, if  $V(f) = I(f)$  then  $f$  is linear:

$$f = \hat{f}(\phi) + \sum \hat{f}(i) \chi_i \quad (4)$$

From here it is understandable:

$$f(x) = a_0 + \sum_{i=1}^n a_i x_i \quad (5)$$

**Claim 1** *If  $f$  is Boolean linear function, then  $f$  is a dictatorship. This keeps also for almost linear case, and the proof will lead us to it.*

**Remark**

$$\chi_\phi = \prod_{x \in \phi} x_i = 1 \quad (6)$$

**Proof** Write

$$f = a_0 + \sum_{\chi_i(x)=x_i} a_i x_i \quad (7)$$

We want to pick one  $i$  item from it and prove that all other are zeros. We will choose for it  $i$  with the highest weight. Assume without loss of generality that:

$$|a_1| > |a_2| > \dots \geq |a_n| \quad (8)$$

We should prove that  $a_2 = 0$ , and then  $a_3, a_4, \dots = 0$  also. If  $a_2 \neq 0$  (otherwise we are done) then  $|a_2| \leq \frac{1}{\sqrt{2}}$  That's because:

$$\|f\|_2^2 = \sum a_i^2 + a_0^2 = 1 \quad (9)$$

(Because it's 2-norm of  $\mathbf{f}$  and  $\mathbf{f}$  is boolean. So according to Parseval's theorem the above equality keeps). In particular  $a_1^2 + a_2^2 \leq 1$  and  $a_1^2 > a_2^2$ . Thus,  $a_2^2 \leq \frac{1}{2}$ . Thus,  $\|a_2\| \leq \frac{1}{\sqrt{2}}$ . Hence, for every  $x \in \{\pm 1\}^n$  either  $f(x)$  is non-boolean, or  $f(x)$  is boolean, but then:  $f(xe_2) = f(x) - 2x_2a_2$  is not boolean. **Remark**  $f(x)$  is some  $a_i x_i$  For every  $x$  either  $f(x)$  or  $f(xe_i)$  is "far" from Boolean. The best hope is that  $a_2$  is small number, so that  $f(xe_2) \approx f(x)$   $\|f(x) - \text{sign}(f(x))\| \geq c \|a_2\|$  or  $\|f(xe_2) - \text{sign}(f(x)e_2)\| \geq c \|a_2\|$ . This is

a contradiction  $\Rightarrow a_2 = \dots = a_n = 0 \Rightarrow \chi_\phi = \prod_{x \in \phi} x_i = 1$

**Corollary 2** Thus, if  $a_2 \neq 0$ , then a distance from boolean function is at least  $a_2$ .

**Remark** Since  $f = \sum \hat{f}(s)\chi_s$ , we define  $f^{\leq k} = \sum_{s, |s| \leq k} \hat{f}(s)\chi_s$  - k-head of  $f$  and  $f^{>k} = \sum_{s, |s| > k} \hat{f}(s)\chi_s$  - k-tail of  $f$ . Hence,  $f = f^{\leq k} + f^{>k}$

$$\|f^{>k}\|_2^2 + \|f^{\leq k}\|_2^2 = \|f\|_2^2 \quad (10)$$

This holds because low-degree part and high-degree part are orthogonal. To verify it we can write Parseval's equality and check it. If k-tail is small, then most of the weight of  $\mathbf{f}$  is in k-head.  $\mathbf{f}$  is 'close' to deg-k if  $\|f^{>k}\|_2^2$  is small. ■

**Theorem 3** The FKN(Friedgut, Kalai, Naor) theorem: There exists global const  $C_0$  with the following property: If  $f : \{\pm 1\}^n \rightarrow \{\pm 1\}$  has  $\|f^{>1}\|_2^2 < \epsilon < 0.000001$  (i.e.  $\mathbf{f}$  is almost linear) then  $\mathbf{f}$  is close to dictatorship  $\|f - (\hat{f}(\phi) + \hat{f}(i^* \chi_{i^*}))\|_2^2 \leq C_0 \epsilon$ , when  $i^*$  is selected such that  $|\hat{f}(i^*)| = \max_i |\hat{f}(i)|$

**Remark** The question is if there is another dictatorship to which  $\mathbf{f}$  is close. The answer is negative, and it can be proved using Parseval. On the other hand,  $\|f\|_2^2 = 1$ . If we have 2 dictatorships close to  $\mathbf{f}$ , they will weight both close to 1. Thus, we have 2 different functions with weight close to 1. It can happen only if the constant coefficient ( $a_0$ ) is close to 1. But it means that both dictatorships are the same.

**Remark** Question: Is there a transitive boolean function  $\mathbf{f}$  with  $\|f^{=1}\|_2^2 > \text{const} > 0$ ? Is there really a threshold here? Answer: **Yes!**  $\|Maj^{=1}\|_2^2 = \text{const} > 0 \Rightarrow \|Maj^{>1}\|_2^2 = 1 - \text{const}$ .

**Remark** When  $\|f^{=1}\|_2^2 \nearrow 1$   $\|f^{>1}\|_2^2 \searrow 0$ . Small weight of  $f^{>1}$  forces us to particular structure of the function.

**Proof** Let  $M = \text{Maj}$ . Then:

$$\hat{M}(1) = \langle M, \chi_1 \rangle = E_x [M(x)\chi_1] = E_{x-1} [E_{x_1} [M(x)\chi_1]] \quad (11)$$

Thus,

$$E_{x-1} [E_{x_1} [M(x)\chi_1]] = E_{x-1} \left[ \underbrace{\frac{1}{2}M(1, x-1)}_{\text{case } x_1=1} - \underbrace{\frac{1}{2}M(-1, x-1)}_{\text{case } x_1=-1} \right] \quad (12)$$

Thus,

$$E_{x-1} \left[ \underbrace{\frac{1}{2}M(1, x-1)}_{\text{case } x_1=1} - \underbrace{\frac{1}{2}M(-1, x-1)}_{\text{case } x_1=-1} \right] = Pr_{x-1} [M(1, x=1) \neq M(-1, x-1)] \quad (13)$$

Thus,

$$Pr_{x-1} [M(1, x=1) \neq M(-1, x-1)] = Pr_x [M(x) \neq M(xe_1)] = I_1(M) \quad (14)$$

The above is correct because of monoteness of Maj function. Since Maj is monotone,  $\frac{1}{2}M(1, x-1) - \frac{1}{2}M(-1, x-1)$  is 0 when  $M(1, x-1) = M(-1, x-1)$ , and 1 otherwise.

**Corollary 4** We proved that for  $f$  Boolean and monotone,  $\hat{f}(i) = I_i(f)$   $\hat{M}(1) = I_1(M) = \frac{\text{const}}{\sqrt{n}}$  - From homework Therefore:  $\sum_i \hat{M}(i)^2 = \|M^{=1}\|_2^2 = (\text{const})^2 > 0$

**Proof** [Proof of closeness to boolean dictatorship] Will not be done fully in this lesson, but some parts of it will be collected. ■

**Claim 5**  $2 * |a| * |b| \leq a^2 + b^2$

**Proof** Trivial ■

$$\|f + g\|_2^2 = \langle f + g, f + g \rangle = \langle f, f \rangle + \langle g, g \rangle + 2 \langle f, g \rangle = \|f\|_2^2 + \|g\|_2^2 + 2 \langle f, g \rangle \quad (15)$$

By Cauchy-Schwarz:

$$\|f\|_2^2 + \|g\|_2^2 + 2 \langle f, g \rangle \leq \|f\|_2^2 + \|g\|_2^2 + 2 \|f\| \|g\| \quad (16)$$

By claim above:

$$\|f\|_2^2 + \|g\|_2^2 + 2 \|f\| \|g\| \leq 2 \|f\|_2^2 + \|g\|_2^2 \quad (17)$$

**Corollary 6**  $\|f + g\|_2^2 \leq 2\|f\|_2^2 + \|g\|_2^2$

Using this corollary we get:

$$f - \text{sign}\left(\left\|\hat{f}(\phi) + \hat{f}(i^*)\chi_{i^*}\right\|_2^2\right) \leq 2\left\|f - (\hat{f}(\phi) + \hat{f}(i^*)\chi_{i^*})\right\|_2^2 + 2\left\|\hat{f}(\phi) + \chi_{i^*} - \text{sign}(\hat{f}_{i^*}(\phi) + \hat{f}_{i^*}\chi_{i^*})\right\|_2^2 \quad (18)$$

If  $f$  Boolean then  $\forall g \|g - \text{sign}(g)\|_2^2 \leq \|g - f\|_2^2$

$$2\left\|f - (\hat{f}(\phi) + \hat{f}(i^*)\chi_{i^*})\right\|_2^2 + 2\left\|\hat{f}(\phi) + \chi_{i^*} - \text{sign}(\hat{f}_{i^*}(\phi) + \hat{f}_{i^*}\chi_{i^*})\right\|_2^2 \leq 4\left\|f - \hat{f}(\phi) + \hat{f}(i^*)\chi_{i^*}\right\|_2^2 \quad (19)$$

Forgetting  $f^{>1}$

$$\|f - (a_0 + a_i\chi_i)\|_2^2 \leq 2\|f - f^{\leq 1}\|_2^2 + 2\|f^{\leq 1} - (a_0 + a_i\chi_i)\|_2^2 \quad (20)$$

Since  $2\|f - f^{\leq 1}\|_2^2 = 2\|f^{>1}\|_2^2 < \epsilon$ , it is enough to show that  $2\|f^{\leq 1} - (a_0 + a_i\chi_i)\|_2^2 \leq C_0\epsilon$

$$\|f_{\leq 1} - \text{sign}(f_{\leq 1})\|_2^2 \leq \|f_{\leq 1} - f\|_2^2 = \|f\|_2^2 < \epsilon \quad (21)$$

To prove the theorem it's enough to prove the following sub-theorem:

**Theorem 7** *If  $f_{\leq 1} = a_0 + \sum a_i\chi_i$ ,  $\sum a_i^2 \leq 1$ , and  $\|f_{\leq 1} - \text{sign}(f_{\leq 1})\|_2^2 \leq \epsilon$ , then  $\|f_{\leq 1} - (a_0 + a_i^*\chi_{i^*})\|_2^2 < \epsilon(1 + \Theta(\epsilon))$*

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