## 1 Introduction

**Standard Definitions.** We identify the long code of  $x \in \{0,1\}^s$  by  $\mathsf{LC}(x) = \{f(x) \mid f : \{0,1\}^s \to \{0,1\}\}$ . Informally, we evaluate x on every Boolean function on s bits. Notice that every Boolean function on s bits may be represented by its truth table. In other words, by specifying its evaluation on all the  $2^s$  inputs. Alternatively, any string of length  $2^s$  may be interpreted as a Boolean function on s bits. We denote  $2^s$  by n. Since, there are  $2^n$  Boolean functions on s bits,  $\mathsf{LC}(\mathbf{x})$  is a string on length  $2^n$ . We use the letters a, b to denote Boolean functions. It is easy to check that given a table f,  $f \equiv \mathsf{LC}(\mathbf{x} : \{0,1\}^{2^s} \to \{0,1\})$ , f(a) + f(b) = f(a+b), for every  $a, b \in \{0,1\}^{2^s}$ .

For  $\alpha \subset [n]$ , define

$$\chi_{\alpha}: \{0,1\}^n \to \{0,1\}, \chi_{\alpha}(a) \triangleq \prod_{i \in \alpha} -1^{a(i)}$$

It is easy to check that the characters  $\{\chi_{\alpha}\}_{\alpha\subseteq[n]}$  form an orthonormal basis for the space of functions  $\{f:\{0,1\}^n\to\mathbb{R}\}$ , where inner product is defined by  $\langle f,g\rangle=\mathbb{E}_a[f(a)g(a)]=2^{-n}\sum_a f(a)g(a)$ . It follows that any function  $f:\{0,1\}^n\to\{0,1\}$  can be written as  $f=\sum_{\alpha}\hat{f}_{\alpha}\cdot\chi_{\alpha}$ , where  $\hat{f}_{\alpha}=\langle f,\chi_{\alpha}\rangle$ . We start by recalling a few important properties of the characters of the space of Boolean functions.

**Proposition 1.1.** For every character  $\chi_{\alpha}$  and any two vectors  $x, y, \chi_{\alpha}(x \cdot y) = \chi_{\alpha}(x) \cdot \chi_{\alpha}(y)$ .

**Proposition 1.2** (Orthonormality). For k > 1 and vector x, the following holds.

$$\exists i, j \in [k]: \alpha_i \neq \alpha_j \Leftrightarrow \mathbb{E}_x \Big[ \chi_{\alpha_1}(x) \cdot \chi_{\alpha_2}(x) \dots \chi_{\alpha_k}(x) \Big] = 0$$

**Proposition 1.3.** For every character  $\chi_{\alpha}$ , vector x and an integer y such that y mod 2 = 0,

$$\mathbb{E}_x\big[\left(\chi_\alpha\left(x\right)\right)^y\big] = 1$$

**The Long Code Test.** Let  $f: \{0,1\}^n \to \{0,1\}$ . We intend to test if  $f: \{0,1\}^n \to \{0,1\}$  is in fact the legal encoding of a value  $w \in [s]$ . In other words, if f(a) = a(w) for all  $a \in [2^n]$ .

Fix a parameter  $\rho \in [0, 1]$ . The test picks two uniformly random vectors  $a, b \in \{0, 1\}^n$  and then  $x \in \{0, 1\}^n$  according to the following distribution: for every coordinate  $i \in [n]$ , with probability  $1 - \rho$  we choose  $x_i = 0$  and  $x_i = 1$  otherwise. It is useful to imagine x as a noise vector. The test accepts iff f(a) + f(b) + f(a + b + x) = 0. The test accepts iff  $x_w = 0$ , which happens with probability  $1 - \rho$ . It follows from the construction that the test accepts any valid long code encoding with probability  $1 - \rho$ . We now state a certain converse of that, which was established by Håstad's lemma [Hås01].

**Lemma 1.4** (Corollary 22.25 in [AB09]). For every  $\delta, \epsilon > 0$ , if f passes the long code test with probability with  $1/2 + \delta$ , then for  $k = \frac{1}{2\rho} \log \frac{1}{\epsilon}$ , there exists  $\alpha$  with  $|\alpha| \le k$  such that  $\hat{f}_{\alpha} \ge 2\delta - \epsilon$ .

Say f is a purported long code table given to the verifier. We denote by  $T_r$  (the long code) test performed by the verifier on randomness r. The verifier accepts iff  $T_r$  evaluates to 0. We are interested in analyzing the soundness of a variant of  $T_{r_1} \cdot T_{r_2}$ , for  $r_1, r_2$  drawn from the uniform distribution. Our new verifier chooses a, b, c, d uniformly at random from  $\{0, 1\}^n$  and x, y are noise vectors. The new test may be expressed as the following.

$$[f(a) + f(b) + f(a+b+x)] \cdot [f(c) + f(d) + f(c+d+y)] = 0$$

We transform  $\{0,1\}$  to  $\{\pm 1\}$  via the mapping  $b \to (-1)^b$ . This also maps the usual XOR operation on GF(2) to a product operation and the multiplication operation to a new operation  $\otimes$ . We now recall a few basic properties of the operation  $\otimes$ .

**Proposition 1.5** ( $\otimes$  distributes over the product). For every m, n, k,

$$m \otimes (n \cdot k) = (m \otimes n) \cdot (m \otimes k)$$

Now, the "new" test may be written as the following.

$$f(a \otimes c) \cdot f(a \otimes d) \cdot f(a \otimes (c \cdot d \cdot y)) \cdot f(b \otimes c) \cdot f(b \otimes d) \cdot f(b \otimes (c \cdot d \cdot y)) \cdot f(a \cdot b \cdot x) \otimes c \cdot f((a \cdot b \cdot x) \otimes d) \cdot f((a \cdot b \cdot x) \otimes (c \cdot d \cdot y)) = 1$$

We would now analyze the soundness of the test discussed in the prequel. Say, the verifier accepts the test by probability  $1/2 + \delta$ , for some  $\delta > 0$ . Since the product terms like  $a \otimes c$  and others may not be uniformly distributed, we use a standard trick of replacing  $f(a \cdot c)$  by  $f(e \cdot a \oplus c) \cdot f(e)$ , where e is chosen uniformly at random from  $\{0,1\}^n$ .

$$2 \cdot \delta = \mathbb{E} \Big[ f((a \otimes c) \cdot e) \cdot f(e) \cdot f((a \otimes d) \cdot e) \cdot f(e) \dots \dots f\left(((a \cdot b \cdot x) \otimes (c \cdot d \cdot y) \cdot e\right) \cdot f(e) \Big]$$

$$2 \cdot \delta = \mathbb{E}_{a,b,c,d,e,x,y} \Big[ \left( \sum_{\alpha_1} \hat{f}_{\alpha_1} \chi_{\alpha_1} ((a \otimes c) \cdot e) \right) \cdot \left( \sum_{\alpha_2} \hat{f}_{\alpha_2} \chi_{\alpha_2}(e) \right) \cdot \left( \sum_{\alpha_3} \hat{f}_{\alpha_3} \chi_{\alpha_3} ((a \otimes d) \cdot e) \right) \dots$$

$$\dots \left( \sum_{\alpha_{17}} \hat{f}_{\alpha_{17}} \chi_{\alpha_{17}} ((a \cdot b \cdot x) \otimes (c \cdot d \cdot y) \cdot e) \right) \cdot \left( \sum_{\alpha_{18}} \hat{f}_{\alpha_{18}} \chi_{\alpha_{18}}(e) \right) \Big]$$

$$2 \cdot \delta = \mathbb{E}_{a,b,c,d,e,x,y} \Big[ \sum_{\alpha_1,\alpha_2,\dots\alpha_{18}} \hat{f}_{\alpha_1} \hat{f}_{\alpha_2} \dots \hat{f}_{\alpha_{18}} \chi_{\alpha_1} ((a \otimes c) \cdot e) \cdot \chi_{\alpha_2}(e) \cdot \chi_{\alpha_3} ((a \otimes d) \cdot e) \dots$$

$$\dots \chi_{\alpha_{17}} ((a \cdot b \cdot x) \otimes (c \cdot d \cdot y) \cdot e) \cdot \chi_{\alpha_{18}}(e) \Big]$$

By linearity of expectation, the above expression may be written as follows.

$$2 \cdot \delta = \sum_{\alpha_1, \alpha_2 \dots \alpha_{18}} \hat{f}_{\alpha_1} \hat{f}_{\alpha_2} \dots \hat{f}_{\alpha_{18}} \quad \mathbb{E}_{a, b, c, d, e, x, y} \left[ \chi_{\alpha_1} ((a \otimes c) \cdot e)) \cdot \chi_{\alpha_2} (e) \cdot \chi_{\alpha_3} ((a \otimes d) \cdot e)) \dots \right] \dots \chi_{\alpha_{17}} ((a \cdot b \cdot x) \otimes (c \cdot d \cdot y) \cdot e) \cdot \chi_{\alpha_{18}} (e)$$

We apply Proposition 1.1 to simplify the above expression to the following.

$$2 \cdot \delta = \sum_{\alpha_1, \alpha_2 \dots \alpha_{18}} \hat{f}_{\alpha_1} \hat{f}_{\alpha_2} \dots \hat{f}_{\alpha_{18}} \quad \mathbb{E}_{a, b, c, d, e, x, y} \left[ \begin{array}{c} \chi_{\alpha_1}(a \otimes c) \cdot \chi_{\alpha_1}(e) \cdot \chi_{\alpha_2}(e) \cdot \chi_{\alpha_3}(a \otimes d) \cdot \chi_{\alpha_3}(e) \dots \\ \dots \chi_{\alpha_{17}}((a \cdot b \cdot x) \otimes (c \cdot d \cdot y)) \cdot \chi_{\alpha_{17}}(e) \cdot \chi_{\alpha_{18}}(e) \end{array} \right]$$

Since e is mutually independent from a, b, c, d, x, y, we have

$$2 \cdot \delta = \sum_{\alpha_1, \alpha_2 \dots \alpha_{18}} \hat{f}_{\alpha_1} \hat{f}_{\alpha_2} \dots \hat{f}_{\alpha_{18}} \quad \mathbb{E}_{a, b, c, d, x, y} \left[ \chi_{\alpha_1}(a \otimes c) \cdot \chi_{\alpha_3}(a \otimes d) \dots \chi_{\alpha_{17}}((a \cdot b \cdot x) \otimes (c \cdot d \cdot y)) \right] \cdot \\ \mathbb{E}_e \left[ \chi_{\alpha_1}(e) \cdot \chi_{\alpha_2}(e) \cdot \chi_{\alpha_3}(e) \dots \chi_{\alpha_{18}}(e) \right]$$

We now invoke Proposition 1.2 to conclude that the expectation is 0 unless  $\alpha_1 = \alpha_2 \dots = \alpha_{18}$ . Therefore,

$$2 \cdot \delta = \sum_{\alpha} \hat{f}_{\alpha}^{18} \, \mathbb{E}_{a,b,c,d,x,y} \left[ \, \chi_{\alpha}(a \otimes c) \cdot \chi_{\alpha}(a \otimes d) \dots \chi_{\alpha}((a \cdot b \cdot x) \otimes (c \cdot d \cdot y)) \right] \cdot \mathbb{E}_{e} \left[ \, \left( \chi_{\alpha} \left( e \right) \right)^{18} \, \right]$$

By Proposition 1.3, 
$$\mathbb{E}_{e}\left[\left(\chi_{\alpha}\left(e\right)\right)^{18}\right]=1$$
. Hence,

$$2 \cdot \delta = \sum_{\alpha} \hat{f}_{\alpha}^{18} \quad \mathbb{E}_{a,b,c,d,x,y} \left[ \chi_{\alpha} \left( (a \otimes c) \cdot \chi_{\alpha}(a \otimes d) \dots \chi_{\alpha}((a \cdot b \cdot x) \otimes (c \cdot d \cdot y) \right) \right]$$

$$= \sum_{\alpha} \hat{f}_{\alpha}^{18} \quad \mathbb{E}_{a,b,c,d,x,y} \left[ \chi_{\alpha} \left( (a \otimes c) \cdot (a \otimes d) \cdot (a \otimes (c \cdot d \cdot y)) \cdot (b \otimes c) \cdot (b \otimes d) \cdot b \otimes (c \cdot d \cdot y) \cdot ((a \cdot b \cdot x) \otimes d) \cdot ((a \cdot b \cdot x) \otimes (c \cdot d \cdot y)) \right) \right]$$

$$((a \cdot b \cdot x) \otimes c) \cdot ((a \cdot b \cdot x) \otimes d) \cdot ((a \cdot b \cdot x) \otimes (c \cdot d \cdot y)) \right)$$

Invoking Proposition 1.5, we rewrite the above as.

$$2 \cdot \delta = \sum_{\alpha} \hat{f}_{\alpha}^{18} \mathbb{E}_{a,b,c,d,x,y} \left[ \chi_{\alpha} \left( (a \otimes y) \cdot (b \otimes y) \cdot (a \cdot b \cdot x) \otimes y \right) \right]$$

$$= \sum_{\alpha} \hat{f}_{\alpha}^{18} \mathbb{E}_{x,y} \left[ \chi_{\alpha}(x \otimes y) \right]$$

Denote  $x \otimes y$  by z. Notice that each coordinate of z in the above equation is drawn is chosen independently from the product distribution of  $x \otimes y$ . Since, each coordinate of x, y is independently set to 1 with probability  $1-\rho$  and -1 otherwise,  $\mathbb{E}\left[z_w\right] = \mathbb{E}\left[x_w \otimes y_w\right] = 1 \cdot ((1-\rho)^2 + \rho^2) - 1 \cdot 2 \cdot (1-\rho) \cdot \rho = 1 - 4\rho + 4\rho^2$ . Now, since each coordinate is chosen independently,  $\mathbb{E}_z\left[\chi_\alpha(z)\right] = \mathbb{E}_z\left[\prod_{w \in \alpha} z_w\right]$ . Hence,

$$2\delta = \sum_{\alpha} \hat{f}_{\alpha}^{18} \left( 1 - 4 \cdot \rho \cdot (1 - \rho) \right)^{|\alpha|}$$

## References

- [AB09] S. Arora and B. Barak. Computational Complexity A Modern Approach. Cambridge University Press, 2009. 1
- [Hås01] J. Håstad. Some optimal inapproximability results. *Journal of the ACM*, 48(4):798–859, 2001. 1