

Lectures on Linearity Testing

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1 Linearity Testing

Definition 1. Let G and H be finite groups. Let $f : G \rightarrow H$. Then f is said to be *linear* (i.e., is a *homomorphism*) if for all $x, y \in G$,

$$f(x) +_H f(y) =_H f(x +_G y).$$

For all $\varepsilon > 0$, f is said to be ε -linear if there exists a linear function $g : G \rightarrow H$ such that f and g agree on at least $1 - \varepsilon$ fraction of inputs in G , i.e.,

$$\mathbb{P}_{x \in G} [f(x) = g(x)] \geq 1 - \varepsilon,$$

or equivalently,

$$\frac{|\{x \in G : f(x) = g(x)\}|}{|G|} \geq 1 - \varepsilon.$$

Algorithm 1 is a natural test for the linearity of a function $f : G \rightarrow H$, where G and H are finite groups.

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1 repeat ? times
2   pick random  $x, y \in G$ 
3   if  $f(x) + f(y) \neq f(x + y)$  then
4     return "fail"
5 return "pass"
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Algorithm 1: A proposed test for the linearity of a function $f : G \rightarrow H$, where G and H are finite groups.

Observation 2. Let G be a finite group. For all $a, y \in G$, $\mathbb{P}_{x \in G} [y = a + x] = 1/|G|$. In other words, if x is chosen uniformly from G , then $a + x$ is also uniformly distributed in G .

Proof. Since only $x = y - a$ satisfies $y = a + x$, then $\mathbb{P}_{x \in G} [y = a + x] = \mathbb{P}_{x \in G} [x = y - a] = 1/|G|$. \square

2 Self-Correcting (Random Self-Reducibility)

Theorem 3. Let G be a finite group. Let $f : G \rightarrow G$ be a function such that there exists a linear function $g : G \rightarrow G$ and that $\mathbb{P}_{x \in G} [f(x) = g(x)] \geq 7/8$. Then for all $x \in G$, $g(x)$ can be computed with only $O(\log(1/\beta))$ calls to f (with at most β probability of error).

Given input $x \in G$ and black box access to f , we define a *self corrector* in Algorithm 2.

Proposition 4. $\mathbb{P}[\text{output} = g(x)] \geq 1 - \beta$.

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1 for  $i \leftarrow 1, \dots, C \cdot \log(1/\beta)$  do
2   pick  $y$  uniformly in  $G$ 
3    $answer_i \leftarrow f(y) + f(x - y)$ 
4 output the most common answer

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Algorithm 2: A self corrector for a $1/8$ -linear function $f : G \rightarrow G$ on input x , where G is a finite group.

Proof. Let y be chosen uniformly in G . By Observation 2, $x - y$ is also uniformly distributed in G . Therefore,

$$\mathbb{P}[f(y) \neq g(y)] \leq \frac{1}{8}, \quad \mathbb{P}[f(x - y) \neq g(x - y)] \leq \frac{1}{8}.$$

By the union bound,

$$\begin{aligned} \mathbb{P}[f(y) + f(x - y) = g(x)] &= \mathbb{P}[f(y) + f(x - y) = g(y) + g(x - y)] \\ &\geq \mathbb{P}[f(y) = g(y), f(x - y) = g(x - y)] \\ &\geq 1 - \left(\frac{1}{8} + \frac{1}{8}\right) = \frac{3}{4}. \end{aligned}$$

This implies that $\mathbb{P}[answer_i = g(x)] \geq 3/4$ for all i . The proof is hence complete. \square

3 Coppersmith's Example

Let $m \in \mathbb{N}$. Let $f : \mathbb{Z}_m \rightarrow \mathbb{Z}_m$ be defined by

$$f(x) = \begin{cases} 1, & \text{if } x \equiv 1 \pmod{3}, \\ 0, & \text{if } x \equiv 0 \pmod{3}, \\ -1, & \text{if } x \equiv 2 \pmod{3}. \end{cases}$$

The graph of f is plotted in Figure 1.

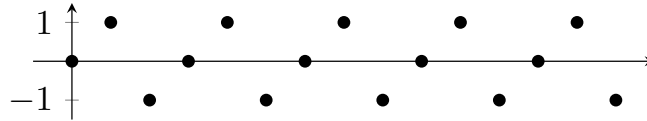


Figure 1: The graph of Coppersmith's example.

Note that the closest linear function $g : \mathbb{Z}_m \rightarrow \mathbb{Z}_m$ to f is given by $g(x) = 0$ for all $x \in \mathbb{Z}_m$, so f is $2/3$ -far from being linear. Note that f fails for $x, y \in \mathbb{Z}_m$ with $x \equiv y \equiv 1 \pmod{3}$ or $x \equiv y \equiv 2 \pmod{3}$, and passes for all other $x, y \in \mathbb{Z}_m$. Therefore, the *rejection probability of the linearity test* for f , denoted by δ_f , is given by

$$\delta_f = \mathbb{P}_{x, y \in \mathbb{Z}_m} [f(x) + f(y) \neq f(x + y)] = \frac{2}{9}.$$

Fortunately, $2/9$ is the threshold; in other words, Coppersmith's example is the worst example. If $\delta_f < 2/9$ for some function $f : G \rightarrow G$ and finite group G , then f must be δ_f -close to being linear.

4 Fourier Analysis for Boolean Functions

The n -dimensional Boolean hypercube $\{0, 1\}^n$ can be interpreted as having $n + 1$ layers, where the i^{th} layer consists of n -bit Boolean strings with i ones for each $i \in \{0, \dots, n\}$, and where two n -bit Boolean strings in consecutive layers are joined by an edge if they differ at exactly one bit. What are linear maps $\{0, 1\}^n \rightarrow \{0, 1\}$?

Definition 5. Given $x, y \in \{0, 1\}^n$, the *inner product* of x and y is defined to be

$$x \cdot y = \sum_{i=1}^n x_i y_i \pmod{2}.$$

Note that addition modulo 2 is the XOR operation. Linear functions on $\{0, 1\}^n$ are of the form

$$L_a(x) = a \cdot x, \quad \text{for fixed } a \in \{0, 1\}^n,$$

or, alternatively,

$$L_A(x) = \sum_{i \in A} x_i \pmod{2}, \quad \text{for fixed } A \subset [n].$$

Therefore, there are exactly 2^n linear functions on $\{0, 1\}^n$.

To simplify the presentation, we change the notation by letting $a \mapsto (-1)^a$ for $a \in \{0, 1\}$ and by changing addition $a + b$ to multiplication $(-1)^a (-1)^b = (-1)^{a+b}$. Hence, the condition of linearity $f(a) + f(b) = f(a \oplus b)$ for all $a, b \in \{0, 1\}^n$ is changed to $f(a) \cdot f(b) = f(a \odot b)$ for all $a, b \in \{1, -1\}^n$, where $(x_1, \dots, x_n) \oplus (y_1, \dots, y_n) = (x_1 + y_1, \dots, x_n + y_n)$ denotes the bitwise XOR (i.e., addition modulo 2) of two n -bit Boolean strings, and $(x_1, \dots, x_n) \odot (y_1, \dots, y_n) = (x_1 \cdot y_1, \dots, x_n \cdot y_n)$ denotes the bitwise multiplication of two n -bit $\{1, -1\}$ -valued strings. Moreover, linear functions on $\{1, -1\}^n$ are of the form

$$\chi_S(x) = \prod_{i \in S} x_i, \quad \text{for fixed } S \subset [n].$$

Now, the linearity test accepts if and only if $f(x) \cdot f(y) = f(x \odot y)$. Then

$$f(x)f(y)f(x \odot y) = \begin{cases} 1, & \text{if the test accepts,} \\ -1, & \text{if the test rejects.} \end{cases}$$

Therefore, the indicator variable for the event that the test rejects is given by

$$\mathbb{1}_{f(x)f(y) \neq f(x \odot y)} = \frac{1 - f(x)f(y)f(x \odot y)}{2} = \begin{cases} 0, & \text{if the test accepts,} \\ 1, & \text{if the test rejects.} \end{cases}$$

This allows us to express the rejection probability in terms of the indicator variable:

$$\delta_f = \mathbb{P}_{x, y \in \{1, -1\}^n} [f(x) \cdot f(y) \neq f(x \odot y)] = \mathbb{E}_{x, y \in \{1, -1\}^n} \left[\frac{1 - f(x)f(y)f(x \odot y)}{2} \right].$$