

Proof of weak PCP thm: $\text{PCP}(\text{poly}(n), \text{O(1)}) \geq \text{NP}$

def Walsh-Hadamard code. $\text{WH}: \{0,1\}^n \rightarrow \{0,1\}^{2^n}$

$$\text{WH}(u)(x) = \langle u, x \rangle \pmod{2}.$$

can think of $\text{WH}(u)$ as tt of function $\langle u, \cdot \rangle$.

claim: $u \neq v \Rightarrow \text{dist}(\text{WH}(u), \text{WH}(v)) \geq \frac{1}{2} \cdot 2^n$. $\sim \sim \sim \sim$ why?

How to know if w is a valid codeword? Obs $\text{WH}(\{0\}^n)$ is set of all linear functions, hence enough to test if f is linear.

def. Linearity Testing.

f, g ϵ -close if $\Pr[f(e) = g(e)] \geq p$.

f close to linear if $\exists g$ linear, f, g ϵ -close.

th: if $\Pr[f(e+x) = f(e) + f(x)] \geq p$, then f ϵ -close to linear.

given $u, \{v: \langle u, v \rangle = 0\}$ is linear space of dim $n-1$.

hence $\{x: \text{HW}(u) = \text{HW}(v)\} =$ end size 2^{n-1} .

$= \{x: \langle u-v, x \rangle = 0\}$ has size $\frac{1}{2} \cdot 2^{n-1}$.

we will use this extensively.

From now on, can assume we're $1-\delta$ -close to linear.

Can we recover real codeword \tilde{w} from w ?

Yes: $\begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{pmatrix} \xrightarrow{\text{local dec}} \begin{pmatrix} \tilde{w}_1 \\ \tilde{w}_2 \\ \vdots \\ \tilde{w}_n \end{pmatrix}$. But locally & efficiently?

Also yes. Want $\tilde{w}[x]$.

Sample x' ; set $x'' = x' + x$.

Query $y' = w[x']$, $y'' = w[x'']$.

Answer $y = y' + y''$.

obs $\Pr[y = \tilde{w}[x]] \geq 1 - \delta \quad \Pr[y = \tilde{w}[x]] \geq 1 - 2\delta$.

$\Pr[y'' = \tilde{w}[x'']] \geq 1 - \delta$.

We'll build a PCP for problem QuadEq.

i.e. find if system of quadratic eqs over \mathbb{F}_2 has solution.

claim: QuadEq is NP-complete.

assume wlog no terms of deg 1 (replace x_i by x_i^2).

reinterpret problem as $AU = b$,

where A is an $m \times n^2$ matrix and $U = u \otimes u = uu^\top$.

u is $\text{WH}(u) \text{WH}(uu^\top)$.

need to check: 1. $u = v_g$ is concatenation of two WH codes,

2. $u = uu^\top$.

3. $AU = b$.

correctness Ok.

soundness of 1: OK by linearity testing.

For 2 and 3 we'll assume that all queries done using local decoding protocol.

To test 2, let $v = \text{WH}^{-1}(A)$, $w = \text{WH}^{-1}(g)$; we test if

$$\langle (u \otimes u), r \otimes r' \rangle = \langle w, r \otimes r' \rangle = g(r \otimes r') \quad \text{for random } r, r'.$$

$$\begin{matrix} \langle u, r \rangle \cdot \langle u, r' \rangle \\ \parallel \\ \langle r, r' \rangle \end{matrix} \quad \begin{matrix} \uparrow \\ \text{obs can read directly} \\ \text{from } u. \end{matrix}$$

claim: if $w \neq u \otimes u$, then $\Pr_{r, r'}[\langle f(r) \cdot f(r'), g(r \otimes r') \rangle \neq g(r \otimes r')] \geq \frac{1}{4}$.

(apply $A \in \mathcal{B} \Rightarrow \Pr_x[A \neq B_x] \geq \frac{1}{2}$ twice).

To test 3, we'd like to test if $A; U = b$: b/c:

instead we'll check if $\sum_{i \in S} A_i; U = b_i$ for random S .

$$\text{iow: } \langle \mathbf{1}_S^\top A_i, U \rangle = \langle \mathbf{1}_S^\top, b_i \rangle.$$

$$\begin{matrix} \langle \mathbf{1}_S^\top A_i, U \rangle \\ \parallel \\ \text{can read directly from } u. \end{matrix}$$

This finishes proof of PCP thm.

Now let us see linearity test.

We'll represent functions in Fourier basis. $f: \{0,1\}^n \rightarrow \{0,1\}^n$.

i.e. usually we write $f = \sum f_x e_x \quad x \in \{0,1\}^n$

instead we'll write $f = \sum \hat{f}_S X_S \quad S \subseteq [n]$.

obs $x \cdot y$ in $\{0,1\}^n \equiv x \cdot y$ in $\{0,1\}^n$

linear functions over $\{0,1\}^n$ are characters X_S in $\{0,1\}^n$.

$$\langle f, g \rangle = \sum_x f(x) g(x) / 2^n$$

$$\hat{f}_S = \langle f, X_S \rangle = (\#x: f(x) = 1) / 2^n \Rightarrow f \text{ is } \frac{1}{2} + \epsilon \text{-close to } X_S$$

$$\text{iff } \hat{f}_S = \pm \epsilon.$$

Linear test thm is:

$$\Pr[e(f_\gamma) = f_\gamma] \geq \frac{1}{2} + \epsilon \Rightarrow \exists S \text{ s.t. } \hat{f}_S \geq \pm \epsilon.$$

$$\text{proof: obs } E[\langle f_\gamma, f_\gamma \rangle] = \Pr[\square] - \Pr[\square] \geq 2\epsilon.$$

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$$E[(\sum \hat{f}_S X_S(\gamma)) \cdot (\sum \hat{f}_S X_S(\gamma)) \cdot (\sum \hat{f}_S X_S(\gamma))] =$$

$$\begin{matrix} \parallel \\ \hat{f}_S(\gamma) \hat{f}_S(\gamma) \end{matrix}$$

$$= E[\sum \hat{f}_S \hat{f}_S^\top X_S(\gamma) X_S(\gamma) X_S(\gamma)^\top X_S(\gamma)] = \sum \hat{f}_S^2 E[$$

$$\begin{matrix} \parallel \\ \hat{f}_S, \hat{f}_S^\top \\ \parallel \\ \hat{f}_{S'} \hat{f}_{S'}^\top \end{matrix}$$

$$\begin{matrix} \parallel \\ \hat{f}_{S''} \hat{f}_{S''}^\top \end{matrix}$$

$$\leq \max_S \hat{f}_S^2 \cdot \sum_S \hat{f}_S^2$$

$$\leq 1 \text{ (Personal)}.$$