

6.S979 Lecture 16

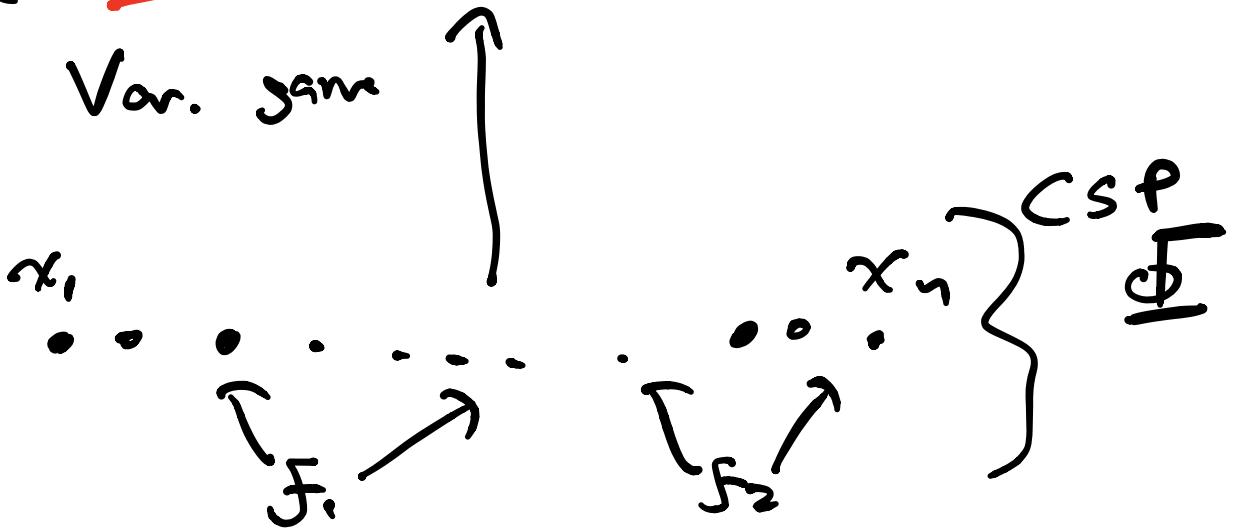
- This Friday Otl will be 3-4 pm Eastern

Goal: understand power of MIP

$$NP \subseteq MIP[\log(n)]$$

$$(\underline{NEXP} = MIP)$$

Clause Var. game



NP-hard:

$$YES: \omega(\Phi) = 1.$$

$$NO: \omega(\Phi) \leq 1/2.$$

fraction of constraints satisfiable

Idea: "encode" an NP-hard problem
(QUADEQ)

$$\Sigma_H(x) = \left(\underbrace{\langle x, 0\dots 0 \rangle, \langle x, 0\dots 01 \rangle, \dots}_{\in \{0,1\}^{2n}} \right)$$

$\in \{0,1\}^n$

$$\Sigma(x) = \left(\Sigma_H(x), \Sigma_H(x \otimes x) \right)$$

Last time: Given access to $\Sigma(x)$
can check whether x satisfies a
system of quad. eq. with just
2 queries to $\Sigma(x)$

YES: x satisfies \Rightarrow test passes
w/ certainty

NO: x doesn't satisfy \Rightarrow test pass
w/ prob $= 1/2$

Today: Given $y \in \{0, 1\}^{2^n}$,
check that y is close to $\Sigma_H(x)$
for some x , using 3 queries to y

- YES: If $y = \Sigma_H(x)$, then
test passes w/ certainty

- NO: If y differs from $\Sigma_H(x)$ on
at least δ -fr. of positions
(for all x), then test
succeeds w/ prob. $\leq 1 - \Omega(\delta)$

Note: This looks like self-testing

First: Some notational changes

Think of y as truth table of

$$f: \{0, 1\}^{2^n} \rightarrow \{0, 1\}$$

$$(f(0\dots 0), f(0\dots 01), f(0\dots 10), \dots)$$

2^n

$$\Sigma_H(x) = (\langle x, 0 \dots 0 \rangle, \langle x, 0 \dots 01 \rangle, \langle x, 0 \dots 10 \rangle, \dots)$$

$$g_x(a) = \langle x, a \rangle \quad \leftarrow \text{linear functions (mod 2)}$$

$a \in \{0, 1\}^n$

Goal: Given f , is it close to $g_x = \langle x, \cdot \rangle$ for some x

Test [BLR]:

Pick $a, b \in \{0, 1\}^n$ at random

$$\text{Check } f(a) + f(b) = f(a+b)$$

- YES: $f(a) = \langle x, a \rangle$

$$f(a) + f(b) = \langle x, a \rangle + \langle x, b \rangle$$

$$= \langle x, a+b \rangle$$

$$= f(a+b) \quad \checkmark$$

- NO: Suppose BLR test passes
 w/ prob. $1-\epsilon$
 $\Rightarrow \exists x, f$ is $O(\epsilon)$ -close
 to g_x

Notational change #2:

$\{0, 1\} \longrightarrow \{\pm 1\}$

Instead of f work with $(-1)^f = F$

BLR test: $f(a) \cdot f(b) \stackrel{?}{=} f(a \otimes b)$ entire product

$F(A) F(B) \stackrel{?}{=} F(A \otimes B)$
 $\in \{\pm 1\}^n \quad \in \{\pm 1\}^n \quad \in \{\pm 1\}^{2n}$

$g_x(a) = \langle x, a \rangle$

$G_x(A) = (-1)^{\langle x, a \rangle} = \prod_{i: x_i=1} A_i$
 $\parallel \quad \parallel$
 $(-1)^{a_i} \quad \parallel \quad (-1)^{a_i}$

Fanier characters

Capital = ± 1 , lower case = $\{0, 1\}$

Fourier analysis of Boolean functions:

$$F: \{\pm 1\}^n \rightarrow \{\pm 1\} \quad x_i^2 = 1$$

$$F(x_1, \dots, x_n)$$

$$= \hat{F}_\emptyset + \hat{F}_{\{1\}} x_1 + \hat{F}_{\{2\}} x_2 + \dots + \hat{F}_{\{1,2\}} x_1 x_2 + \dots + \hat{F}_{\{1,2,\dots,n\}} x_1 x_2 \dots x_n$$

Fourier decomposition of F

$$\hat{F}_S \cdot \prod_{i \in S} x_i$$

subset of $\{1, \dots, n\}$

$$x_S(x)$$

$$\hat{F}_S = \sum_{x \in \{\pm 1\}^n} F(x) \cdot \prod_{i \in S} x_i$$

$$= \sum_{x \in \{\pm 1\}^n} (-1)^{\langle S, x \rangle} F(x)$$

Fourier transform over $\{\pm 1\}^n$ $\langle F, F \rangle$

$$\langle F, F \rangle = \sum \hat{F}_s^2 = 1 = \mathbb{E} \sum_{x \in \{\pm 1\}^n} F(x)^2$$

Plancherel thm.

$$\langle F, G \rangle := \mathbb{E} \sum_{x \in \{\pm 1\}^n} F(x)G(x)$$

orthonormal basis

$$S \neq T \quad \langle \prod_{i \in S} x_i, \prod_{i \in T} x_i \rangle = 0$$

Suppose F that passes BLR test w/ prob. $1 - \epsilon$

$$\Leftrightarrow \Pr_{A, B} [\underbrace{F(A) \cdot F(B) \cdot F(A \circ B)}_{= 1 - \epsilon} = 1]$$

either 1 or -1

$$\prod_{A, B} F(A) \cdot F(B) F(A \circ B) = 1 - 2\varepsilon$$

$$F(A) = \sum_S \hat{F}_S \cdot \chi_S(A)$$

$$\prod_{A, B} \sum_{S, T, U} \hat{F}_S \cdot \hat{F}_T \cdot \hat{F}_U \cdot \chi_S(A) \chi_T(B)$$

$\chi_U(A \circ B)$
 indicator of S \rightarrow $\langle S, a \rangle$ \rightarrow (a_1, \dots, a_n)
 subsets of $\{1, \dots, n\}$

$$\chi_S(A) = \prod_{i \in S} A_i = \prod_{i \in S} (-1)^{a_i}$$

$$\prod_{a, b} \sum_{s, t, u} \hat{F}_s \cdot \hat{F}_t \cdot \hat{F}_u \cdot (-1)^{\langle s, a \rangle} \cdot (-1)^{\langle t, b \rangle} \cdot (-1)^{\langle u, a+b \rangle}$$

$$= \prod_{a, b} \sum_{s, t, u} \hat{F}_s \cdot \hat{F}_t \cdot \hat{F}_u \cdot (-1)^{\langle s+u, a \rangle} \cdot (-1)^{\langle t+u, b \rangle}$$

Nonzero iff
 $s+u=0$
 $s=u$

$$= \sum_{s,t,u} \hat{F}_s \cdot \hat{F}_t \cdot \hat{F}_u \left(\underbrace{\mathbb{E}_{(-1)}^{(s,t,u,a)}}_a \right)$$

Nonzero
iff $t=u$

$$\rightarrow \left(\underbrace{\mathbb{E}_{(-1)}^{(t,u,b)}}_b \right)$$

Recall: if $x \neq 0$, $P_n[(x,a)=0] = 1/2$
 $\in \{0,1\}^n$

$$\uparrow$$

$$\mathbb{E}_a^{(-1)} \langle x, a \rangle = 0$$

if $x=0$,

$$\langle x, a \rangle = 0$$

$$\mathbb{E}_a^{(-1)} \underbrace{\langle x, a \rangle}_{(-1) \cdot 0} = 1$$

$$= \sum_u (\hat{F}_u)^3 \geq 1 - 2\varepsilon$$

Recall: $\sum_u (\hat{F}_u)^2 = 1$
 Plancherel

$$\sum_u (\hat{F}_u)^3 = \sum_u \hat{F}_u \cdot (\hat{F}_u)^2 \geq 1 - 2\varepsilon$$

$$\leq \max_u \hat{F}_u$$

$$\max_u \hat{F}_u \cdot \sum_u (\hat{F}_u)^2 \geq \sum_u \hat{F}_u \cdot (\hat{F}_u)^2 \geq 1 - 2\varepsilon$$

$$\max_u \hat{F}_u \geq 1 - 2\varepsilon$$

$$f = \sum_s \hat{F}_s \chi_s \approx \hat{F}_{u^*} \chi_{u^*}$$

$\exists u^*$

linear function!!!

$$\approx \chi_{u^*}$$

\downarrow
 $\langle u^*, x \rangle$

$$= \prod_{i \in u^*} X_i = (-1)$$



Putting it all together:

CSP looks like this:

Variables $y^{(1)}, y^{(2)}$ Should be $= \sum_H(x) \sum_H(x \otimes x)$

Constraints:

— BLR constraints:

$$y_a^{(1)} + y_b^{(1)} = y_{a+b}^{(1)}$$

$$y_a^{(2)} + y_b^{(2)} = y_{a+b}^{(2)}$$

— "Tensor product" constraint confuse

$$y_a^{(1)} \cdot y_b^{(1)} = y_{a \otimes b + c}^{(2)} \quad \left(\begin{array}{l} \text{c. } y^{(2)} \\ \text{c} \end{array} \right)$$

$$y_{a \otimes b}^{(2)} = \langle a \otimes b, x \otimes x \rangle = (\langle a, x \rangle) (\langle b, x \rangle)$$

$y_a^{(1)} \cdot y_b^{(1)}$

$a \oplus b$ not uniformly random

But $a \oplus b + c$ is

c is

to learn $f(a)$, gives $f(a+b)$
 $f(b)$
"self-correction",
"interpolation"

- Quad. eq.

$$ax + bxx = c$$

$$y_a^{(1)} + y_b^{(2)} = c$$

↓

↓

$$y_{a+d}^{(1)} + y_d^{(1)} + y_{b+td}^{(2)} + y_d^{(2)} = c$$