

3/9/2015 Russell's course lec 3.

Linear, Mansour, Nissen

Algorithmic uses of switching lemma:

lower bound on approx using Fourier representation $\langle \chi, \mu \rangle$

Exercise show that a read-once setting, $2/3$ of inputs means bottom row in $O(\cos S)$ whp.

Consequences of switching lemma:

Let U be depth of tree of size S , $D \geq \cos S$, $p = \Omega(\frac{1}{D})$.

\exists a read-once setting whp $\text{Prob}[\exists \chi \text{ s.t. } |\langle \chi, \mu \rangle| \geq 2^{-p}] \leq 2^{-p}$

Prob χ is on unit distr.

Branch access to U - training: $\langle \chi_i, \mu \rangle$ $(\chi_i, \mu(\chi_i)) \dots (\chi_i, \mu)$ where χ_i are random n -bit inputs.

Learn hypothesis $h: \{0,1\}^n \rightarrow \{0,1\}$; approx of success $\geq 1 - \epsilon$

- know U is depth d and $S, D \leq S, n$ inputs

Fourier representation:

- represent true by -1 , and false by 1 .

Over $0,1$, $\sum \chi_i \bmod 2$ is parity. Over $\{-1,1\}$, parity is $\prod \chi_i$

- function as a vector: $F(\chi) = \langle F(\chi_1, \dots, \chi_1), F(\chi_1, \dots, \chi_1), \dots, F(\chi_1, \dots, \chi_1) \rangle$

- $F(1, \dots, 1)$. This is 2^n dim vector, embedded in 2^n -dim space. $\|F\| = \sqrt{\sum (F(\chi))^2} = 2^{n/2}$

Normalizing const $1/2^{n/2}$, $\chi_i = \pm 1$

\vec{F}, \vec{G} . $\vec{F} \cdot \vec{G} = \frac{1}{2^{n/2}} \sum_{\chi} F(\chi) \cdot G(\chi) =$

$= \frac{1}{2^n} \cdot (\sum_{\chi} F(\chi) \cdot G(\chi)) = \text{Prob}[F(\chi) = G(\chi)] - \text{Prob}[F(\chi) \neq G(\chi)] \in \{-1, 1\}$

measure of correlation between F and G

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Geometrically, work under diff. bases - one basis $(1, 0, \dots)$

Orthogonal basis: Set of n bases F_1, \dots, F_n s.t. $\langle F_i, F_j \rangle = 0$ if $i \neq j$

$\langle F_i, F_j \rangle = 0$. So $\text{Prob}[F_i = F_j] = \text{Prob}[F_i \neq F_j]$

Ideal subsets $S \subseteq \{1, \dots, n\}$. $\Theta_S = \{ \chi : \chi_i \in \{-1, 1\} \}$

$\Theta_S \cap \Theta_T = \Theta_{S \cap T}$. Let $S, T \subseteq [n]$, $S \neq T$. When does $\Theta_S(x) = \Theta_T(x)$?

$\Theta_S(x) = \Theta_T(x) \iff \text{Prob}_{\chi}(\Theta_{S \cap T}(x) = 1) \sim 1/2$

Φ_s are 2^n orthonormal vectors in 2^n dim space: basis.

Any bool fn $\vec{F} = \sum \hat{d}_s \Phi_s$

Fourier coeffs.

$\langle \vec{F}, \Phi_s \rangle = \sum \hat{d}_s \langle \Phi_s, \Phi_s \rangle = \sum \hat{d}_s = \text{Prob}_{x_i} [F(x) = 1] - \text{Prob}_{x_i} [F(x) \neq 1]$

$\frac{1}{2} \text{Prob}_{x_i} [F(x) = 1] = \sum \hat{d}_s \frac{1}{2} \text{Prob}_{x_i} [\Phi_s(x) = 1] = \sum \hat{d}_s \text{Prob}_{x_i} [F(x) = 1]$

- multilinear polys.

Dec. tree for \mathbb{R} : $\mathbb{R}|_{x_i=0} = 1$ $\mathbb{R}|_{x_i=1} = 0$

$\mathbb{R}(x) = \frac{1-x_i}{2} \cdot \mathbb{R}|_{x_i=0} + \frac{(1+x_i)}{2} \mathbb{R}|_{x_i=1}$ - multilinear poly of deg $\leq n$

\Rightarrow any Fourier coeff. of \mathbb{R} of size $\geq \frac{1}{2}$ is 0.

Idea [LNU] this means large Fourier coeffs or \vec{F} are small.

Parseval's identity: \forall bool fn, $\sum (\hat{d}_s)^2 = 1$

$\|\vec{F}\| \geq 1$. $\vec{F} = \langle \mathbb{R}, \mathbb{R} \rangle = \langle \sum \hat{d}_s \Phi_s, \sum \hat{d}_s \Phi_s \rangle$. $\|\vec{F}\| = \sqrt{1} = 1$

LNU claim: \mathbb{R} is computable by size s and ϵ .
Then $\sum_{|T| \geq \log s} (\hat{d}_T)^2 \leq \epsilon \text{val}(\mathbb{R})$, $\mathbb{R} \approx 0$ on \mathcal{D}^s .

$\mathbb{R} \approx \sum \hat{d}_s \mathbb{1}_{\mathcal{D}_s}$. $\mathbb{R}|_{x_i=1} = \sum (\hat{d}_s + \hat{d}_{s \cup \{i\}}) \mathbb{1}_{\mathcal{D}_s}$

$\mathbb{R}|_{x_i=1} \approx \sum_{s \in \mathcal{D}^s} (\hat{d}_s - \hat{d}_{s \cup \{i\}}) \mathbb{1}_{\mathcal{D}_s}$

$\mathbb{E} \left(\sum \hat{d}_s \right)^2 = \frac{1}{2} (\hat{d}_s + \hat{d}_{s \cup \{i\}})^2 + \frac{1}{2} (\hat{d}_s - \hat{d}_{s \cup \{i\}})^2 = \hat{d}_s^2 + \hat{d}_{s \cup \{i\}}^2$

So we set all x_i in T at random. Let $\mathbb{R}|_T = \mathbb{R}$. Let S be subset of vars. Then $\mathbb{E} \left(\sum_{T \cap S} \hat{d}_T \right)^2 \leq \sum_{T \cap S} \hat{d}_T^2$

Lemma if f is non-const unary poly of vars in S , then

$\mathbb{E} \left(\sum_{S \subseteq \text{subset}, T \cap S} \hat{d}_T \right)^2 \geq \sum_{S \subseteq \text{subset}} \hat{d}_S^2$

Work on the set T of set vars, u contributes \hat{d}_u to $\left(\sum \hat{d}_T \right)^2$. Part of sum is $|u-T| \geq 1$. $\mathbb{E}[|u-T|] = \text{Prob}[u \in T] = \frac{1}{2}$. $\mathbb{E}[|u-T|] = \text{Prob}[u \in T] = \frac{1}{2}$.