

Monotone Functions

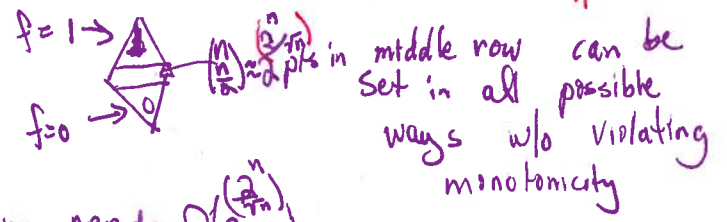
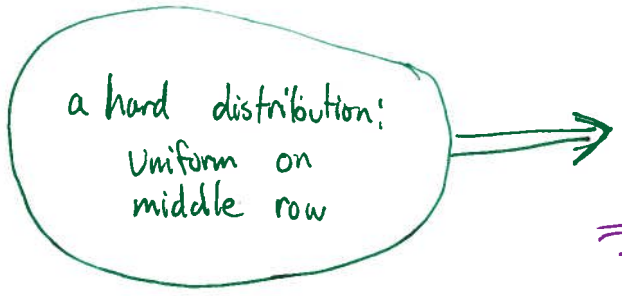
def partial order \leq
 $x \leq y$ iff $\forall i \quad x_i \leq y_i$
 monotone function f
 $x \leq y \Rightarrow f(x) \leq f(y)$

Learning algorithms for the class of monotone functions?
 in homework we see $2^{O(\sqrt{n})}$ random samples suffice for uniform distribution
 ← here $n = \text{dimension}$
 $n = \text{domain size}$

why is this nontrivial?
 we said poly samples is easy,
 the problem is computation time? poly in what?

but need $\text{poly}(\log |C|)$ samples
 all monotone fctns
 there are 2^{2^n} fctns, $\approx 2^{2^{1/n}}$ monotone fctns

why so many monotone fctns?
 consider slice fctns:
 $\# \text{fctns} \approx 2^{2^{(2^n/n)}}$



\Rightarrow learning needs $\Omega\left(\frac{2^n}{2^{1/n}}\right)$
 even with queries in PAC model

Today's question:

what about learning monotone distributions,
on uniform distribution,
with queries?

here we will get a very slight "win":

All monotone fctns have weak agreement
with some dictator fctn.

slightly better than random guess
↓
 $(\frac{1}{2} + \Theta(\frac{1}{\sqrt{n}}))$
(can get $\frac{1}{2} + \Theta(\frac{1}{\sqrt{n}}$ with majority of dictators)

slice fctns have weak agreement with all dictators on uniform dist

Thm $\forall f$ monotone, $\exists g \in \{\pm 1, x_1, x_2, \dots, x_n\}^{\mathcal{S}}$
s.t. $\Pr_x [f(x) = g(x)] \geq \frac{1}{2} + \Omega(\frac{1}{\sqrt{n}})$

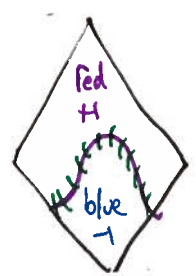
Why does this give weak learning algorithm? estimate agreement of f with all members of \mathcal{S} + output member with max agreement

Pf.

Case 1 $f(x)$ has weak agreement with $+1$ or -1
Case 2 otherwise $\Pr[f(x) = 1] \in [\frac{1}{4}, \frac{3}{4}]$

First a break,
before we prove case 2 ...

let's define influence of monotone fctns?



- # nodes = 2^n , # edges = $\frac{n \cdot 2^n}{2}$
- each level has $\binom{n}{j}$ weight j nodes
- monotone \Rightarrow no blue above any red
- slicing cube in roughly half cuts many edges + many in same direction
- $\text{Inf}(f) = \frac{\# \text{red-blue edges}}{2^{n-1}}$, $\text{Inf}_i(f) = \frac{\# \text{red edges in } i^{\text{th}} \text{ dir}}{2^{n-1}}$

Another Definition

influence of i^{th} var on $f: \{\pm 1\}^n \rightarrow \{\pm 1\}$

$$\text{Inf}_i(f) = \Pr_x [f(x) \neq f(x^{\oplus i})]$$

\uparrow i^{th} bit of x is flipped

Total Influence

$$\text{Inf}(f) = \sum_{i=1}^n \text{Inf}_i(f)$$

NOTE:
Exactly equal to
red-blue edges

Thm: f monotone $\Rightarrow \text{Inf}_i(f) = \hat{f}(\epsilon_i)$

Thm: majority fctn $f(x) \equiv \text{sign}\left(\sum_i x_i\right)$ (n odd)
maximizes influence among
monotone fctns.

Pfs on h.w.

recall H.W. :

Thm f monotone

$$\ln f_i(f) = \hat{f}(\{x_i\}) \stackrel{\substack{\uparrow \\ \text{Known}}}{=} 2\Pr[f(x) = \underbrace{X_i}_{x_i}] - 1$$

\uparrow H.W.

Plan:

Show $\ln f_i(f) \geq \Omega\left(\frac{1}{n}\right)$

$$\Rightarrow \Pr[f(x) = X_i] \geq \frac{1}{2} + \frac{\ln f_i(f)}{2}$$
$$\geq \frac{1}{2} + \Omega\left(\frac{1}{n}\right)$$

Will use following tool:

Canonical Path Argument

Plan 1) define canonical path for every red-blue pair of nodes (note such a path must cross at least one red-blue edge)

2) show upper bnd on # of c.p.'s passing through any edge (in particular, any red-blue edge)

3) conclude lower bnd. on # of red-blue edges

Part 1 of plan:

def. $\forall (x, y)$ s.t. x red $\neq y$ blue

"Canonical path from x to y " is:

scan bits left to right, flipping where needed
each flip \rightsquigarrow step in path

Example

	direction	<u>1</u>	<u>2</u>	<u>3</u>	<u>4</u>
$x =$		-1	+1	+1	+1
$w =$	\hookrightarrow	+1	+1	+1	+1
$z =$		+1	\hookrightarrow -1	+1	+1
$y =$		+1	-1	+ \hookrightarrow	-1

$x \rightarrow w \rightarrow z \rightarrow y$
each step is
Hamming distance 1

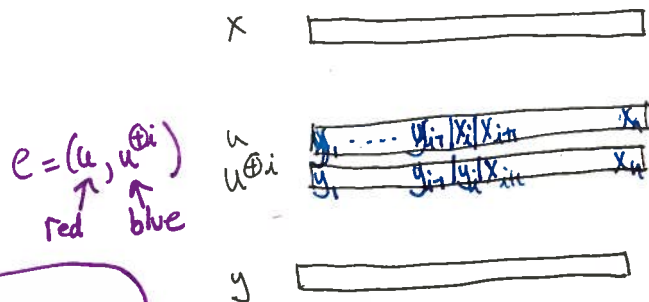
How many $\overset{\text{red-blue}}{\forall} x, y$ pairs have canonical paths?

recall, $\Pr[f(x)=1] \in [\frac{1}{4}, \frac{3}{4}]$

$$\# \text{ paths} \geq \frac{1}{4} \cdot 2^n \cdot \frac{1}{4} \cdot 2^n = \frac{1}{16} \cdot 2^{2n}$$

Part 2 of plan:

For any (red-blue) edge e , how many x, y pairs can cross it with canonical x, y -path?



edge in i^{th} direction $\leq 2^{i-1}$ settings for prefix of x

$y_1 \dots y_i x_i x_{i+1} \dots x_n$
 $y_1 \dots y_i y_i x_{i+1} \dots x_n$

$\leq 2^{n-i}$ settings for suffix of y

$\therefore \leq 2^n$ total settings of prefix x , suffix y consistent with this edge

Main point: all canonical paths crossing $u, u^{\oplus i}$ agree on $y_1 \dots y_i + x_{i+1} \dots x_n$

Part 3 of plan:

$(\# \text{ red-blue edges}) (\max \# \text{ canonical paths that use it}) \geq \# \text{ red-blue canonical paths}$

So

$$\# \text{ red-blue edges} \geq \frac{\frac{1}{16} 2^{2n}}{2^n} = \frac{1}{16} \cdot 2^n$$

\downarrow l.b. on $\#$ r-b pairs
 \uparrow u.b. on $\#$ Canonical paths crossing any edge

since each uses ≥ 1 red-blue edge

so $\exists i$ s.t. $\geq \frac{2^n}{16} \cdot \frac{1}{n}$ red-blue edges in direction i

so $\ln f_i(f) \geq \frac{2^n}{16n} = \frac{1}{8n} = \uparrow(\{i\}) = 2 \Pr[f(x) = x_i] - 1$

\uparrow
 total # edges
 in dir i

$$\therefore \Pr[f(x) = x_i] \geq \frac{1}{2} + \frac{1}{16 \cdot n}$$

□

Canonical Path argument also used in

- routing
- expansion / conductance of hypercube / other Markov Chains

What good is weak learning?

unclear

here only uniform distribution

if can learn in all distributions,

can do much more

(next result does not apply to monotone

function learning... \leftarrow i.e. $\frac{1}{n}$ agreement

in particular, this weak notion of learning \leftarrow i.e. const $> \frac{1}{2}$ agreement
 provably doesn't give anything for stronger learning)