

- Last time:
- "Chernoff bounds" / tail bounds (for hypth. testing)
 - learning by finding consistent hyp. from a fixed \mathcal{H} ^(CHF)
 - "Occam's Razor" ("cardinality" version)

- Today:
- Applic. of : learning sparse disj., using few examples, with hyp. that's a (pretty) sparse disj.
 - via "greedy set cover" heuristic
 - (start) proper vs improper learning:
 - "can efficiently improperly learn 3-term DNF;
 - "can't eff. properly learn 3-term DNF (unless $NP \subseteq RP$).

PS 2 due, PS 3 out.

Midterm Mon Oct 23

(on mat. thru Mon Oct 16)

Questions?

Thm
"Occam's razor, cardinality version":

Fix m, ϵ, \mathcal{H} . Suppose there's a subset

$\mathcal{H}_m \subseteq \mathcal{H}$ s.t. $m \geq \frac{1}{\epsilon} (\ln |\mathcal{H}_m| + \ln \frac{1}{\delta})$ \leftrightarrow $|\mathcal{H}_m| \leq \frac{\delta}{\epsilon} e^{\epsilon m}$

s.t. given any set S of m lab. ex. acc. to some $c \in \mathcal{C}$, there's an $h \in \mathcal{H}_m$ consistent with the m ex.

If L is any alg. that finds such an $h \in \mathcal{H}_m$ when given a sample of size m , then

L run on m ex from $EX(c, \mathcal{H})$ is an (ϵ, δ) -PAC learner.

Applic. : Using "greedy set cover" heuristic to eff. learn sparse disjunctions.

$\mathcal{C} = \{ \text{all mon. disj. of length } \leq k \text{ over } x_1, \dots, x_n \}$
 $X = \{0,1\}^n$

Window: OLMB $O(k \log n)$ mist.
 OLMB \rightarrow PAC conv:

$\frac{z}{\epsilon} \ln\left(\frac{z}{\epsilon}\right) = \# \text{ex}$
 is enough;

LTF hyp "

Now: use Occam stuff to learn with m samples,
 now using "pretty short" mon disj as hyp's.

Idea: we'll show how, given m ex,
 we can eff find a \cup (length s) mon disj consistent
 with our m ex.

$\mathcal{H}_m = \text{all length-}s \text{ mon disj}$

$|\mathcal{H}_m| \leq n^s$; so as long as

$n^s \leq \frac{1}{\epsilon} e^{\epsilon m}$, can apply Occam
 result.

"greedy
 set
 cover"
 heuristic

$$n^{k \ln m} \leq \frac{1}{\epsilon} e^{\epsilon m}$$

small m will work! 😊

Consider following problem:

SET COVER:

Input: family of r sets S_1, \dots, S_r ; each $S_i \subseteq [m]$.

Output: use min # of S_i 's to cover all of $[m]$.

OPT = size of optimal solution

Ex: $m=7$

$S_1 = \{1, 2, 3, 4, 5\}$	} opt sol: S_2, S_3
$S_2 = \{1, 3, 5, 7\}$	
$S_3 = \{2, 3, 4, 6\}$	
$S_4 = \{3, 4, 7\}$	

" NP-complete: almost certainly, there's no poly-time alg. that always finds opt. solution

" The greedy algorithm is efficient & provably finds an almost-optimal solution!

GH:

- First step: Pick S_i s.t. $|S_i|$ largest.
- Delete all elts in S_i from every S_j .
- Repeat using new S_j 's. Stop when all $[m]$ covered.

Ex of GH:

$S_1 = \{1, 2, 3, 4, 5\}$	← first set
$S_2 = \{\cancel{1}, 3, \cancel{5}, 7\}$	can choose S_2, S_3 or S_4
$S_3 = \{\cancel{2}, \cancel{3}, \cancel{4}, 6\}$	as second set; say S_2
$S_4 = \{\cancel{3}, \cancel{4}, 7\}$	$S_3 = 3^{\text{rd}}$ set

Thm: Greedy heuristic always finds a sol. that contains $\leq \text{OPT} \cdot \ln m$ sets.

Pf:

Fact: Consider any $U \subseteq [m]$. \rightarrow ("uncovered" elts)

There's some S_j , $1 \leq j \leq r$, covering $\geq \frac{|U|}{\text{OPT}}$ points in U .

[Some OPT many S_j 's cover $[m]$, hence cover U ;
so ^{at least} one of them covers $\geq \frac{|U|}{\text{OPT}}$ elts.]

Let $U_i \subseteq [m]$ be set of elts not yet covered after i stages.

FACT \Rightarrow next step covers $\geq \frac{|U_i|}{\text{OPT}}$ elts of U_i ;

(whichever S_j of orig. sets covered this many of U_i 's elts still covers them; uncovered elts don't get removed as we go.)

$$\text{So } |U_{i+1}| \leq |U_i| - \frac{|U_i|}{\text{OPT}} = |U_i| \left(1 - \frac{1}{\text{OPT}}\right).$$

$$\text{So } |U_i| \leq m \cdot \left(1 - \frac{1}{\text{OPT}}\right)^i \quad \left(\left(1 - \frac{1}{x}\right)^x < e^{-1} \text{ for } x > 1 \right)$$

$$i = \text{OPT} \cdot \ln m: \quad \rightarrow \leq m \cdot \left(1 - \frac{1}{\text{OPT}}\right)^{\text{OPT} \cdot \ln m}$$

$$\leq m \cdot e^{-\ln m} < 1$$

so $|\bigcup_i U_i| = 0$ ☺



Learning again: why is this useful for learning sp. non. disj.?

It'll let us find, given a set S of m ex. lab. by a non disj. of size k , a non disj. of length $\leq k \cdot \ln m$ that's consistent.

The alg:

① Run elim alg on sample.

(start w/ x_1, \dots, x_n ; for each neg ex in S , eliminate all x_i that are 1 in the ex.)

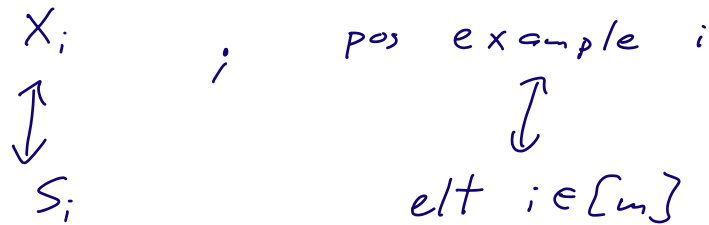
This gives a non disj. h , consistent w/ the sample S ; say it's x_1, \dots, x_r .
(Note: any sub-disj. of \hookrightarrow is consistent w/ neg ex.)

② We want to find a small sub-disj. of x_1, \dots, x_r consist. w/ all the (at most m) pos ex.

I.e. want to find small subset of x_1, \dots, x_r covering all m pos ex.

(x_i covers j^{th} pos ex $\Leftrightarrow x_i = 1$ in that ex.)

This is SET COVER!



Run greedy set cover heur. on the pos. ex.
 → Finds a coll. of $\leq \text{OPT} \cdot \ln m$ many vars x_i

Have $\text{OPT} \leq k$ by def. of \mathcal{C} .

So... alg has prop. that, given a set S of m ex. lab. by a mon disj of size k , it finds a mon disj of size $\leq k \cdot \ln m$ that's consistent.

So $\mathcal{H}_m =$ all mon disj over x_1, \dots, x_m of size $\leq k \cdot \ln m$.

$$\hookrightarrow |\mathcal{H}_m| = \binom{m}{0} + \dots + \binom{m}{k \cdot \ln m} \leq n^{k \cdot \ln m}.$$

So as long as $\underline{\underline{|\mathcal{H}_m| \leq \underline{\underline{\delta e^{\epsilon m}}}}}$ holds, $\ddot{\smile}$
 PAC learner using m ex.

How big does m have to be to give us

$$n^{k \cdot \ln m} \leq \delta e^{\epsilon m} ?$$

Taking

$$m \geq 2 \left(\frac{1}{\varepsilon} \cdot k \cdot \ln n \cdot \ln \frac{1}{\delta} \right) \cdot \ln \left(\frac{1}{\varepsilon} \cdot k \cdot \ln n \cdot \ln \frac{1}{\delta} \right)$$

is enough.

take \ln :

B/c:

$$k \cdot \ln m \cdot \ln n \leq \ln \delta + \varepsilon m$$

Need

$$k \cdot \ln m \cdot \ln n + \ln \frac{1}{\delta} \leq \varepsilon m$$

$A+B \leq AB$ for $A, B \geq 2$, so suff to have

$$k \cdot \ln m \cdot \ln n \cdot \ln \frac{1}{\delta} \leq \varepsilon m, \text{ i.e.}$$

suff to have

$$\frac{k \cdot \ln n \cdot \ln \frac{1}{\delta}}{\varepsilon} \leq \frac{m}{\ln m}$$

//
A

$$A \leq \frac{m}{\ln m} : m ?$$

$m = A \cdot \ln A$? not quite good enough:

$$A \stackrel{?}{\leq} \frac{A \cdot \ln A}{\ln(A \ln A)}$$

not true
frac < 1 //

But

$$m = 2A \cdot \ln A?$$

$$A \stackrel{?}{\leq} \frac{2A \ln A}{\ln(2A \ln A)}?$$

Yes:

$$= \frac{A \cdot \ln(A^2)}{\ln(2A \ln A)}$$

frac > 1
☺

Proper vs Improper Learning

Sometimes insisting on proper learning
makes learning hard!

\mathcal{C} = all 3-term DNFs:

$$\overline{T_1 \vee T_2 \vee T_3}$$

T_i = some
conj.

eff. PAC
Learnable?

Good news: \mathcal{Y} , using $\mathcal{H} = \text{all } 3\text{-CNFs}$.

Fact 1:

- Recall: The class of all mon. conj. over N vars is PAC learnable with

$$m = O\left(\frac{1}{\epsilon} \left(N + \ln \frac{1}{\delta}\right)\right) \text{ ex, in}$$

$\text{poly}\left(N, \frac{1}{\epsilon}, \ln \frac{1}{\delta}\right)$ time. (Elim alg, CHF).

Fact 2: Any 3-term DNF is logically \equiv to

{ a 3-CNF (each clause has ≤ 3 vars)

→ PF: De Morgan:

$$(abcd) \vee (ef) \vee (gh) \equiv$$

$$(a \vee e \vee g) \wedge (a \vee e \vee h) \wedge (a \vee f \vee g) \wedge (a \vee f \vee h) \wedge (b \vee e \vee g) \wedge \dots$$

Next time: