

Last time: part 2 of 3-part pf that if $VCOM(\mathcal{C})=d$, then any CHF for \mathcal{C} using \mathcal{C} as hyp class, run on $m \approx \frac{1}{\epsilon} (d \log \frac{1}{\epsilon} + \log \frac{1}{\delta})$ examples from $EX(\mathcal{C}, \mathcal{D})$, is an (ϵ, δ) -PAC learner.

Recall: ① Setup: Ponder $VCOM(\mathcal{C})$.
 We'll consider a function "growth function $\equiv \Pi_{\mathcal{C}}(m)$ of \mathcal{C} " } 2 lectures ago
 ↳ to get more info.

② Combinatorics arg: AMAZING THEOREM about how the growth fn behaves, for any \mathcal{C} .

last time:
 $\Pi_{\mathcal{C}}(m) \leq \left(\frac{em}{d}\right)^d$ if $m > d$

③ Learning argument: a lot like CHF thm, but using AMAZING THM to control # ways conc in \mathcal{C} can label a data set.

Today: • + finish pf of

- Application: efficient PAC learning of $\mathcal{C} = \text{LTFs}$ over $X = \mathbb{R}^n$ or $\{0,1\}^n$
- (maybe) Start unit on boosting confidence + accuracy.

Questions?

Recall: say concept/hyp. $h \in \mathcal{C}$ is bad if

$$\Pr_{x \sim \mathcal{D}} [h(x) \neq c(x)] > \epsilon, \quad c = \text{target concept in } \mathcal{C}$$

Recall our old CHF thm: (spec. to $\mathcal{H} = \mathcal{C}$)

Thm: Fix any finite c.c. \mathcal{C} , any target $c \in \mathcal{C}$, any dist \mathcal{D} .

Given m samples from $EX(c, \mathcal{D})$, where

$$m \geq \frac{1}{\epsilon} (\ln |\mathcal{C}| + \ln \frac{1}{\delta}), \quad \star$$

w.p. $\geq 1 - \delta$ all bad $h \in \mathcal{C}$ are inconsistent w/ the m samples.

(Key to pf: union bound over all $h \in \mathcal{C}$)

Want analogue for \mathcal{C} infinite.

We'll prove:

Thm: Fix any ^(may be ∞ !) c.c. \mathcal{C} , any target $c \in \mathcal{C}$, any dist \mathcal{D} .

Given m samples from $EX(c, \mathcal{D})$, where

$$m \geq \frac{2}{\epsilon} \left(\log \left(\prod_{c \in \mathcal{C}} (2m) \right) + \log \frac{1}{\delta} \right), \quad m \geq \frac{8}{\epsilon} \quad \star$$

w.p. $\geq 1 - \delta$ all bad $h \in \mathcal{C}$ are inconsistent w/ the m samples.

(Key to pf: union bound over all poss. labelings \mathcal{C} can induce on a 2^m -elt sample.)

So if m sat. \star , using a CHF on m ex. is an ϵ, δ -PAC learner.
 \nearrow for \mathcal{C} using \mathcal{C}

What's up with

$$m \geq \frac{2}{\epsilon} (\ln(\Pi_{\mathcal{C}}(2^m)) + \ln \frac{2}{\delta}), \quad m \geq \frac{8}{\epsilon} \quad \star \quad ?$$

if didn't have Amazing Thm, \nearrow we'd only know $\Pi_{\mathcal{C}}(2^m) < 2^{2^m}$. If it were

$\frac{1}{2} \cdot 2^{2^m}$, could never hold:

says $m \geq \frac{2}{\epsilon} \cdot (2^{m-1} + \ln \frac{2}{\delta})$

impossible!

But thanks to AT, have control on $\Pi_{\mathcal{C}}(2^m)$:

it's $\leq \left(\frac{e \cdot 2^m}{d}\right)^d$, so

\rightarrow holds for m "not too large".

Pf of Key Thm:

Consider doing the following:

Draw $2m$ samples from $EX(c, \mathcal{D})$.



Event A: there exists a bad $h \in \mathcal{C}$ that's $\leq \delta$ consistent w/ S_1 . (Goal: show $\Pr[A]$ low.)

Event B: there is some $h \in \mathcal{C}$ s.t.

- h is consist. w/ S_1 , but
- h makes $\geq \frac{\epsilon}{2} \cdot m$ mist. on S_2

Claim: For $m > 8/\epsilon$, have $\Pr[A] \leq 2 \cdot \Pr[B]$.
 \hookrightarrow given this: rest of effort is on showing $\Pr[B] \leq \delta/2$.

Pf: $\Pr[B] \geq \Pr[A+B] = \Pr[A] \cdot \Pr[B|A]$

So: enough to show $\Pr[B|A] \geq \frac{1}{2}$.

Since A happens, there's some bad $h^* \in \mathcal{C}$ cons. w/ S_1 .

$\Pr[\overset{\text{Bad!}}{h^*} \text{ makes } \geq \frac{\epsilon}{2} m \text{ mist. on } \overset{\text{fresh batch of ex}}{S_2}]$: high!

$\mathcal{C}B$: $\mathbb{E}[\# \text{ mist. of } h^* \text{ on } S_2] \geq \epsilon m$

so $\Pr[h^*$ makes fewer than $\frac{1}{2}$ its
exp. # of mist. on m pts]

$$\leq e^{-\epsilon m \gamma^2 / 2} \rightarrow \text{This is } \leq \frac{1}{2} \text{ as long as } m \geq 8/\epsilon. \quad \square \text{ (claim)}$$

Rest of pf: show $\Pr[B] \leq \delta/2$.



Can view B as follows (equiv. to original descrip.):

- Draw set S of $2m$ examples.
- Randomly split S into \approx sized S_1 & S_2 .

B = event that there's some $h \in \mathcal{C}$ • cons. with S_1 &
• wrong on $\geq \frac{\epsilon}{2} m$ pts in S_2 .

Fix a partic. labeling of the $2m$ pts in S
(which labels $\geq \frac{\epsilon}{2} m$ of them wrongly).

Now we'll argue

\Pr [this labeling puts all wrong pts in S_2]

 split of S

 into $S_1 + S_2$

very small:

Same as:

$2m$ balls in barrel, at least $k = \frac{\epsilon}{2}m$ are

 black, rest white.

Split balls in barrel into $m + m$ randomly.

\Pr [all k black balls end up in $2nd\ m$]

$$= \binom{m}{k} / \binom{2m}{k}$$

$$= \frac{m!}{(m-k)! k!} / \frac{(2m)!}{(2m-k)! k!}$$

$$= \frac{m(m-1)\dots(m-k+1)}{2m(2m-1)\dots(2m-k+1)} \leq \frac{1}{2^k} \quad k = \frac{\epsilon m}{2}$$

So each fixed lab. of $2m$ ex has $\leq \frac{1}{2^{\frac{\epsilon m}{2}}}$

chance of giving us B .

UB over all ($\leq \Pi_e(2m)$ ")

labelings of S : gives

$$\Pr[B] \leq \underbrace{\Pi_e(2m) \cdot \frac{1}{2^{\epsilon m/2}}}_{\leq \frac{\delta}{2}}$$

We're done provided that $\leq \frac{\delta}{2}$:

$$\Pi_e(2m) \cdot \frac{1}{2^{\frac{\epsilon m}{2}}} \leq \frac{\delta}{2} \quad : \text{ iff } (\log)$$

$$\log(\Pi_e(2m)) - \frac{\epsilon m}{2} \leq \log\left(\frac{\delta}{2}\right), \text{ i.e.}$$

$$\log(\Pi_e(2m)) + \log \frac{2}{\delta} \leq \frac{\epsilon m}{2}, \text{ i.e.}$$

$$m \geq \frac{2}{\epsilon} \left(\log(\Pi_e(2m)) + \log \frac{2}{\delta} \right).$$

End of pf of Key Thm!



Okay, have Key Thm:

$$m \geq \frac{2}{\epsilon} \left(\ln(\Pi_e(2m)) + \ln \frac{2}{\delta} \right), \quad m \geq \frac{8}{\epsilon}$$

is enough ex. for a CHF for e using e to

be a PAC learner.

Weren't we promised

$$m \approx \frac{1}{\epsilon} (d \log \frac{1}{\epsilon} + \log \frac{1}{\delta})$$

is enough ex. for a CHF for \mathcal{C} using \mathcal{C} to be a PAC learner?

We get \mathcal{P} using Amazing Thm.

$$\text{AT: } \mathbb{P}_e(Z_m) \leq \left(\frac{2em}{d}\right)^d \quad \text{for } m \geq d,$$

$$\text{so } \log \mathbb{P}_e(Z_m) \leq d \log\left(\frac{2em}{d}\right).$$

So suff to have m satisfy

$$m \geq \frac{2}{\epsilon} \left(d \log\left(\frac{2em}{d}\right) + \ln \frac{1}{\delta} \right), \quad m \geq \frac{8}{\epsilon}$$

Holds provided

$$m \geq K \left(\frac{d}{\epsilon} \cdot \log \frac{1}{\epsilon} + \frac{1}{\epsilon} \log \frac{1}{\delta} \right).$$

const like 8 or so

ignore δ part, constants this is \approx

$$m \geq \frac{2}{\epsilon} \left(d \log\left(\frac{m}{d}\right) \right), \quad \text{i.e.}$$

$$\underbrace{\frac{m}{d}}_B \geq \underbrace{\frac{2}{\epsilon}}_A \log \left(\frac{m}{d} \right)^B$$

$$B \geq A \log B$$

$$\frac{B}{\log B} \geq A : B = 2A \log A \text{ is enough.}$$

$$\frac{m}{d} = \frac{2}{\epsilon} \log \frac{2}{\epsilon} \text{ is enough.}$$

End of upper bd on sample exity
that suffices for PAC learning!

Ex / Application:

Consider $\mathcal{C} =$ all LTFs over \mathbb{R}^n .

Fact 1: $\text{VC DIM}(\mathcal{C}) = n+1$.

(recall: $n=2$, we argued $\text{VC DIM} = 3$).

Fact 2: There is a computationally efficient

CHF for $\mathcal{C} = (\text{all LTFs})$ using \mathcal{C}_0 .

"polynomial-time linear programming"

Say $w_1 x_1 + \dots + w_n x_n \geq \theta$ is an unknown LTF.

- Pos. ex. $(\underbrace{(3, 1, -5, \dots, 6)}_{\text{pt in } \mathbb{R}^n}; +)$: means

$$3w_1 + w_2 - 5w_3 \dots + 6w_n \geq \theta$$

- Neg. ex. $((4, -2, 7, \dots, 3.3); -)$ means

$$4w_1 - 2w_2 + 7w_3 \dots + 3.3w_n < \theta$$

LP

"linear programming feasibility": given a coll. of linear inequalities, is there a sol. sat. all of them? if so, find a sol.

Have Poly-time algs for this! 😊

This is a CHF for \mathcal{C} using \mathcal{C}_0 , & it's comput. efficient.

Summarizing, here's an eff PAC learning alg
for $\mathcal{C} = \text{LTFs over } \mathbb{R}^n$:

→ draw $m \approx \frac{1}{\epsilon} (n \cdot \log \frac{1}{\epsilon} + \ln \frac{1}{\delta})$ ex;

• run poly-time alg for LP to get a
consistent w_1, \dots, w_m, θ ;

output hyp $\text{sign}(w \cdot x - \theta)$.

Next time: unit on boosting.
