

- Last time:
- finish PAC learning  $\mathcal{C} = \text{intervals over } X = \mathbb{R}$
  - OLCMB  $\Rightarrow$  PAC conversion: get all our OLCMB results in PAC model  $\ddot{\smile}$
  - Revisit PAC learning definition:
    - "size" of concepts
    - efficiency of evaluating hyp.  $h$
  - Start tail bounds

- Today:
- "Chernoff bounds" (tail bounds on sums of indep. random variables) + applic. to hypothesis testing
  - learning by finding consistent hyp. from a fixed class  $\mathcal{H}$  "CHF"
  - "Occam's Razor" ("cardinality version")

Questions?

Sol to ps 1  $\ddot{\smile}$

don't use AI  $\ddot{\smile}$

Check Ed Discussion

## Tail Bounds on sums of indep. RV's

"Chernoff Bound": Let  $X_1, \dots, X_m$  be independent

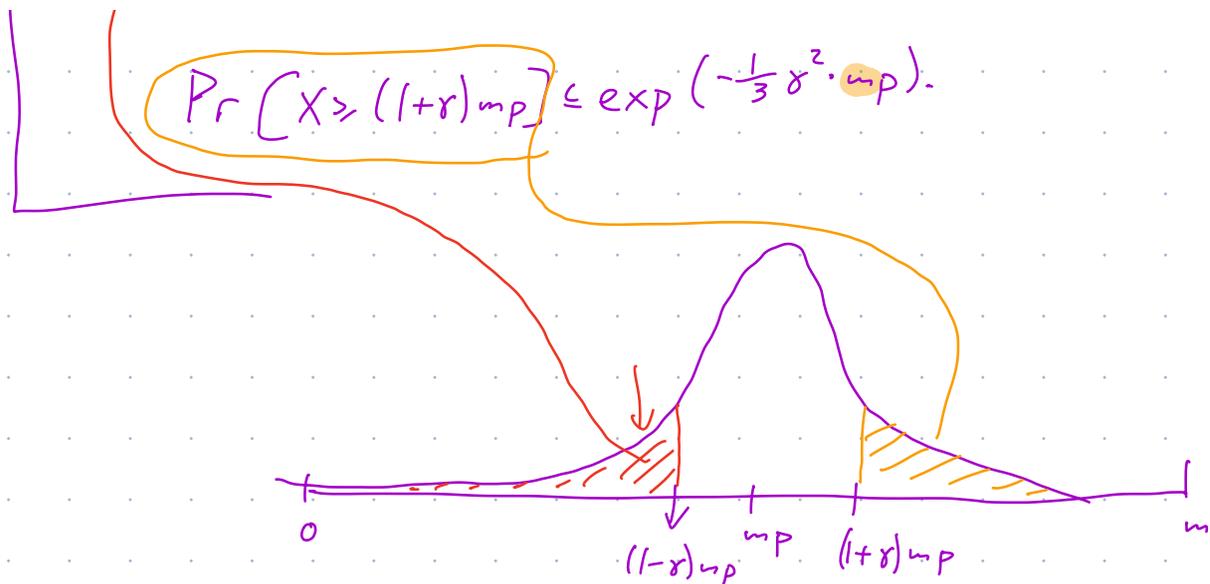
0/1-valued RV's with  $\Pr[X_i = 1] = p \quad \forall i.$

Let  $X = X_1 + \dots + X_m$ , so  $\mathbb{E}[X] = mp$

$\forall 0 \leq \gamma \leq 1,$

"relative error"

$$\Pr[X \leq (1-\gamma)mp] \leq \exp\left(-\frac{1}{2} \cdot \gamma^2 \cdot mp\right), \quad \forall$$



"absolute error" version:

Let  $X_1, \dots, X_m$  as above; let  $\hat{p} = \frac{X}{m}$  ( $=$  empirical estimate of  $p$ )

Then for  $\epsilon > 0$

$$\Pr[p - \hat{p} \geq \epsilon] \leq \exp(-2m\epsilon^2), \quad \forall$$

$$\Pr[\hat{p} - p \geq \epsilon] \leq \exp(-2m\epsilon^2).$$

Ex 1: Say team A wins each game w.p.  $\frac{1}{2}$ , indep. 162 games.

What's  $\Pr\{A \text{ wins} \leq 47 \text{ games}\}$ ?

Let  $X_i = \begin{cases} 1 & \text{win game } i \\ 0 & \text{o/w} \end{cases}$ .  $m = 162, p = \frac{1}{2}$

$$\Pr[X \leq 47] = \Pr[X \leq (1-r) \cdot \overset{81}{m} p] \leq \underbrace{\exp(-\frac{1}{2} \cdot r^2 \cdot mp)}$$

$(1-r)81 = 47.$

$$\gamma \approx 0.42 :$$

$$\exp\left(-\frac{1}{2}\gamma^2 \cdot m_p\right) \approx 0.0008$$

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goal:  
Ex: Design survey: estimate frac. of population  
who support Candidate B.  $\gamma/N$ ,  
How many people do we need to survey?  
want result acc. to  $\pm 3\%$   
with confidence 95%.

Additive bd:  $\epsilon = 0.03$ , want  $\exp(-2m\epsilon^2) = 2.5\%$

$$\text{so } e^{-0.0018m} = 0.025$$

$$\Downarrow m = \frac{1}{0.0018} \cdot \ln(40)$$

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!! Strong! Expon convergence.

!! Strong assumption:  $X = \text{sum of independent}$   
 $\wedge$  RV's.

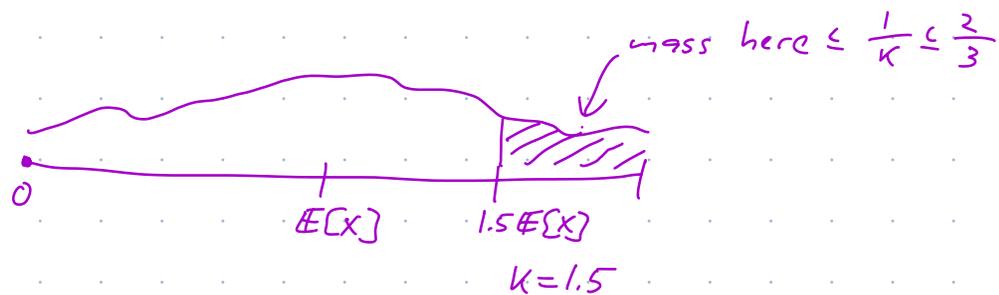
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What can we say for less structured  $X$ 's?

"Markov's inequality": Let  $X$  be non-negative RV.

Then for  $k > 1$ ,

$$\Pr[X \geq k \cdot \mathbb{E}[X]] \leq \frac{1}{k}.$$



Ex: Let  $X = \#$  children in random US household.

Say  $E[X] = 1.85$ .

Then  $P\{X \geq 10\}$  must be  $\leq \frac{1}{5}$   
 (duh avg would be  $\geq 2!$ )

## Learning by Finding Consistent Hyp's:

New way to PAC learn: find a CH from some a priori fixed  $\mathcal{H}$ .

Thm: Fix a  $\mathcal{C}$  + an  $\mathcal{H}$ . Fix any  $c \in \mathcal{C}$ .

Fix any dist  $\mathcal{D}$ .

Let  $(x^1, c(x^1)), \dots, (x^m, c(x^m))$  be iid draws from  $EX(c, \mathcal{D})$   
 $\downarrow$  in  $X$        $\downarrow$  0/1

where  $m \geq \frac{1}{\epsilon} \left( \ln |\mathcal{H}| + \ln \frac{1}{\delta} \right)$ .

Say an  $h: X \rightarrow \{0,1\}$  is bad if  $e_{\mathcal{D}}(h, c) > \epsilon$ .

Then  $\Pr\{\text{any bad } h \in \mathcal{H} \text{ is consistent w/ } \mathcal{L}^m\} \leq \delta.$

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Pf: Fix any bad  $h.$

Have

$$\Pr\{h \text{ cons. with } m \text{ lab. ex. from } EX(c, \mathcal{H})\} < (1-\epsilon)^m.$$

$\mathcal{H}$  has  $\leq |\mathcal{H}|$  bad  $h$ 's; so by union b.d.,

$$\Pr\{\text{any bad } h \in \mathcal{H} \text{ is consistent w/ } m \text{ lab. ex. from } EX(c, \mathcal{H})\} \leq |\mathcal{H}| \cdot (1-\epsilon)^m.$$

This is  $\leq \delta$ , b/c  $m = \frac{1}{\epsilon} \cdot (\ln |\mathcal{H}| + \ln \frac{1}{\delta}).$

$$\begin{aligned} |\mathcal{H}| \cdot (1-\epsilon)^m &= |\mathcal{H}| \cdot \left( (1-\epsilon)^{\frac{1}{\epsilon}} \right)^{\ln |\mathcal{H}| + \ln \frac{1}{\delta}} \\ &\leq |\mathcal{H}| \cdot e^{-\left( \ln \left( \frac{|\mathcal{H}|}{\delta} \right) \right)} \\ &= |\mathcal{H}| \cdot \frac{\delta}{|\mathcal{H}|}. \quad \blacksquare \end{aligned}$$

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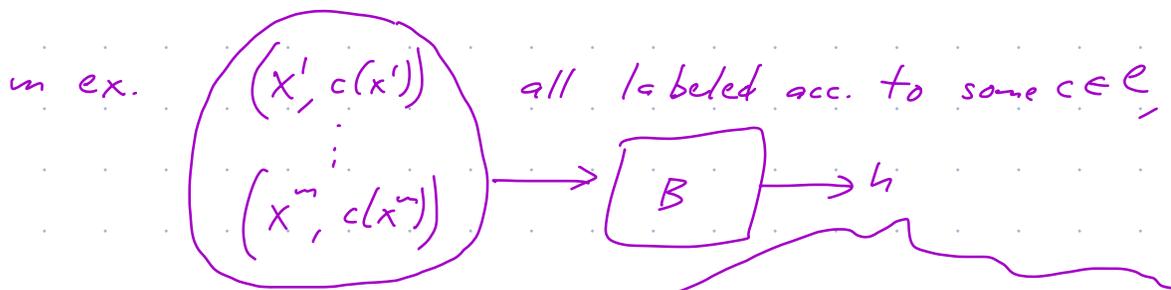
Interpreting / using it:

Fix  $\epsilon, \mathcal{H}.$

CHF

Def: A consistent hyp. finder for  $\mathcal{C}$  using  $\mathcal{H}$  is

an alg  $B: \forall m, \text{ if } B \text{ is given}$



then  $B$  outputs some  $h \in \mathcal{H}$  s.t.  $h(x^i) = c(x^i) \forall i \in [m]$

Means: if have CHF for  $\mathcal{C}$  using  $\mathcal{H}$ , can use it as  $(\epsilon, \delta)$  PAC learner for  $\mathcal{C}$  using  $\mathcal{H}$ : run it on  $m \geq \frac{1}{\epsilon} \cdot (\ln |\mathcal{H}| + \ln \frac{1}{\delta})$  ex from  $EX(c, \mathcal{D})$ .

Ex:  $\mathcal{C} =$  class of mon. disj.  
 $\mathcal{H} = \forall$

Recall our OLMB alg for  $\forall$ : elim.  
 This is a CHF for  $\mathcal{C}$  using  $\mathcal{H}$ .

So then says: run this on  $m = \frac{1}{\epsilon} \cdot \ln \left( \frac{|\mathcal{H}|}{\delta} \right)$  ex, it's a PAC learner.

Last time: our PAC sample complexity from OLMB  $\Rightarrow$  PAC conv. was  $M = n$   
 $M + \frac{M+1}{\epsilon} \cdot \ln \left( \frac{M+1}{\delta} \right)$   
 $= O \left( \frac{n}{\epsilon} \cdot \ln \left( \frac{n}{\delta} \right) \right)$

Now:  $m = \frac{1}{\epsilon} \cdot \ln \left( \frac{|\mathcal{H}|}{\delta} \right)$ ,  $|\mathcal{H}| = 2^n$   
 $|$

↳ so  $m = O\left(\frac{n}{\epsilon} \cdot \ln\left(\frac{1}{\delta}\right)\right)$ .

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Emphasize: if have CHF, get PAC learner

↳ big "if"

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↳ Does not mean building a lookup table to agree with  $m$  examples, is a succ. PAC learner:  
no fixed  $\mathcal{H}$ .

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## Refinement: "Occam's Razor"

"Entities should not be multiplied unnecessarily"

↳ there are few short explain: unlikely any of them which are not actually good, will do well.

-1320 William of Ockham

One view on CHF then:

$h \in \mathcal{H}$  is a "short explanation" of the data.

$(x^1, c(x^1))$   
⋮  
 $(x^m, c(x^m))$

•  $m$  bits of data to "explain" (labels).

• Any  $h \in \mathcal{H}$  can be encoded/described

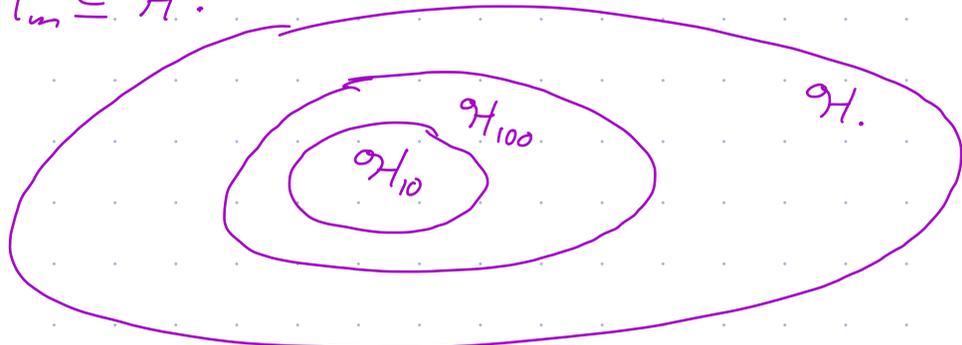
with  $\log |\mathcal{H}|$  bits of info.

So if  $m \gg \log |\mathcal{H}|$ :  $h$  is a "short explanation" of the data.

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Prev thm: for all  $m$ , CHF gives  $h \in \mathcal{H}$  consistent.

Maybe: for small  $m$ , can find a consistent  $h \in \mathcal{H}_m \subseteq \mathcal{H}$ .



If so, can take advantage.

Another version of prev. thm:

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"Occam's razor, cardinality version": Fix  $m$ .  
Suppose there's a  $\mathcal{H}_m \subseteq \mathcal{H}$  of hyp's, where

$$m \geq \frac{1}{\epsilon} \left( \ln |\mathcal{H}_m| + \ln \frac{1}{\delta} \right)$$

$$|\mathcal{H}_m| \leq \frac{1}{\delta} \cdot e^{\epsilon m}$$

st. given any set  $S$  of  $m$  ex. lab. by  $c \in \mathcal{C}$ ,  
there's an  $h \in \mathcal{H}_m$  consistent.

Let  $\mathcal{L}$  be an alg. which, given such an  $S$  of size  $m$ ,

outputs a cons.  $h \in \mathcal{H}_m$ .

Then  $L$  run on an ex from  $EX(c, \mathcal{D})$

is an  $(\epsilon, \delta)$  PAC learner. { Pf: same as before! }

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