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# Hoeffding's Bound & Applications

Hoeffding Bound:  $X_1, \dots, X_n$  independent

$$X = X_1 + \dots + X_n$$

$$X_i \in [a_i, b_i] \quad \forall i$$

For any  $\lambda > 0$

$$\Pr(X \geq \mathbb{E}X + \lambda) \leq \exp\left(-\frac{2\lambda^2}{\sum_{i=1}^n (b_i - a_i)^2}\right)$$

$$\Pr(X \leq \mathbb{E}X - \lambda) \leq \exp\left(-\frac{2\lambda^2}{\sum_{i=1}^n (b_i - a_i)^2}\right)$$

## Applications

① Estimating 'the average of a distribution.

Say we draw  $N$  rand samples

$$Y_1, \dots, Y_N$$

from distribution  $D$  on  $[0, M]$ .

Population average:

$$\mu = \mathbb{E}_{Y \sim D}[Y]$$

Empirical average:

$$\hat{\mu} = \frac{1}{N} (Y_1 + \dots + Y_N).$$

Q: How large must  $N$  be so that

$$\Pr(|\hat{y} - y| > \epsilon) < \delta?$$

Answer is a function of  $\epsilon, \delta, M$   
and is called the sample complexity  
of  $(\epsilon, \delta)$ -PAC mean estimation,  
"probably approximately correct"

$$\hat{y} = \frac{Y_1}{N} + \frac{Y_2}{N} + \dots + \frac{Y_N}{N}$$
$$\mathbb{E} \hat{y} = \frac{\mathbb{E} Y_1}{N} + \frac{\mathbb{E} Y_2}{N} + \dots + \frac{\mathbb{E} Y_N}{N} = y$$

$$X_i \in \left[ 0, \frac{M}{N} \right] \quad (b_i - a_i)^2 = \frac{M^2}{N^2}$$

Hoeffding:

$$\Pr(\hat{y} - y \geq \epsilon) \leq \exp\left(-\frac{2\epsilon^2}{N \cdot (M^2/N^2)}\right)$$

$$= \exp\left(-\frac{2\epsilon^2 N}{M^2}\right)$$

$$\Pr(\hat{y} - y \leq -\epsilon) \leq \exp\left(-\frac{2\epsilon^2 N}{M^2}\right)$$

$$\Pr(|\hat{y} - y| \geq \epsilon) \leq 2 \cdot \exp\left(-\frac{2\epsilon^2 N}{M^2}\right)$$

want this  $< \delta$  . . . . So make this  $< \delta$

$$2 \exp\left(-\frac{2\varepsilon^2 N}{M^2}\right) < \delta$$

$$\exp\left(\frac{2\varepsilon^2 N}{M^2}\right) > \frac{2}{\delta}$$

$$\frac{2\varepsilon^2 N}{M^2} > \ln\left(\frac{2}{\delta}\right)$$

$$N > \frac{1}{2} \left(\frac{M}{\varepsilon}\right)^2 \ln\left(\frac{2}{\delta}\right)$$

(2) Sample complexity of  $(\varepsilon, \delta)$ -PAC ERM.

Given  $N$  samples  $z_1, \dots, z_N$   
drawn iid. from  $D$  on  $\mathcal{Z}$ .

Given  $M$  hypotheses  
 $h_1, h_2, \dots, h_M \in \mathcal{H}$ .

Given loss function  
 $L: \mathcal{H} \times \mathcal{Z} \rightarrow [0, 1]$

$L(h, z)$  = "how badly does  $h$  explain  $z$ ?"

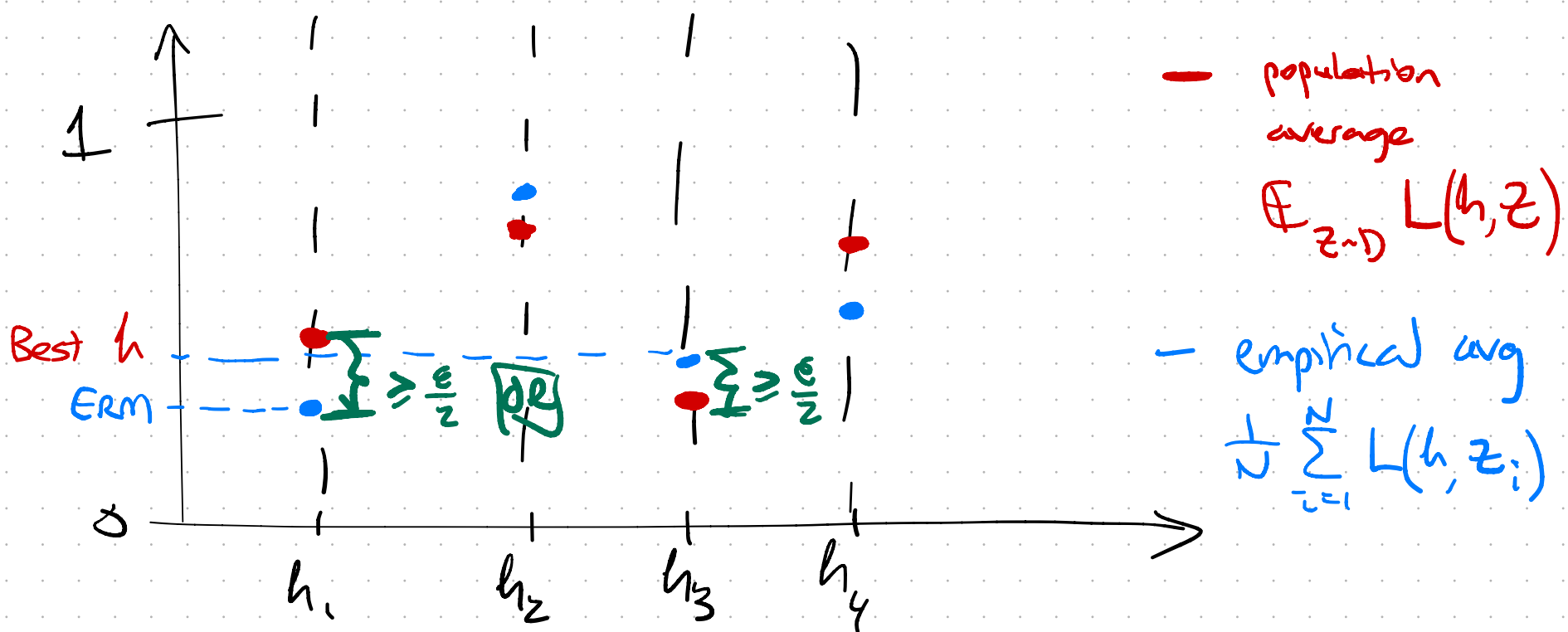
ERM is the algorithm that selects  
a hypothesis using the rule

$$h_{\text{ERM}} = \arg \min_{h \in \{h_1, \dots, h_M\}} \left[ \frac{1}{N} \sum_{i=1}^N L(h, z_i) \right]$$

i.e. select from  $\{h_1, \dots, h_m\}$  the one with least average loss on the data.

We say ERM is  $(\epsilon, \delta)$ -PAC if the following holds with probability  $> 1 - \delta$ ...

$$(*) \quad \mathbb{E}_{Z \sim D} [L(h_{\text{ERM}}, Z)] \leq \epsilon + \min_{i \in [m]} \mathbb{E}_{Z \sim D} [L(h_i, Z)]$$



Upshot:  $(*)$  holds as long as we're sure that  $\forall i \in [m]$

$$\left| \frac{1}{N} \sum_{j=1}^N L(h_i, z_j) - \mathbb{E}_{Z \sim D} (L(h_i, Z)) \right| \leq \frac{\epsilon}{2}$$

$m$  "bad events", one per  $i \in [m]$ .

Bad event: average of  $N$  iid variables in  $[0, 1]$  differs from pop avg by  $> \epsilon/2$ .

Goal:  $\Pr(\text{none of these happen}) < \delta$ .

$\Downarrow$   $\forall$  bad event  $\Pr(\text{bad event}) < \delta/m$ .



Plugging into earlier formula

$$N \geq \frac{1}{2} \left( \frac{2}{\varepsilon} \right)^2 \ln \left( \frac{2m}{\delta} \right)$$

$$N \geq \frac{2}{\varepsilon^2} \ln \left( \frac{2m}{\delta} \right).$$

sample complexity  
for  $(\varepsilon, \delta)$ -ERM  
with  
 $m$  hypotheses.

3) Reducing error rate of randomized algs.

Decision problem. Problem  $\Pi$  with  $\{0, 1\}$  answer.

We say  $\Pi$  belongs to BPP if

$\exists$  a poly-time algorithm

$A(x, r)$

random bits  $r \in \{0, 1\}^n$   
problem input

st.  $\forall x$  if  $\Pi(x) = 0$   $\Pr_{r \in \{0, 1\}^n} [A(x, r) = 0] \geq \frac{2}{3}$

if  $\Pi(x) = 1$   $\Pr_{r \in \{0, 1\}^n} [A(x, r) = 1] \geq \frac{2}{3}$ .

Goal. Boost success prob from  $\frac{2}{3}$  to  $1 - \delta$ .

Plan. Use random string  $R = (r_1, r_2, \dots, r_N)$ .

each  $r_i$  unif. rand  
in  $\{0, 1\}^n$ .

Algo  $B(x, R) = \text{MAJ}[A(x, r_1), A(x, r_2), \dots, A(x, r_N)]$ .

If  $B(x, R)$  is wrong,

$$\mathbb{E}[\# \text{ correct ans. among } A(x, r_1), \dots, A(x, r_N)]$$

$$\mathbb{E}X \geq \frac{2N}{3}$$

$$\text{Realized } \# \text{ correct ans.} = X_1 + \dots + X_N = X$$

$$\leq \frac{N}{2}$$

$$X_i = \begin{cases} 1 & \text{if } A(x, r_i) \text{ correct} \\ 0 & \text{if not} \end{cases}$$

$$(b_i - a_i)^2 = (1 - 0)^2 = 1$$

$$P_r(B(x, R) \text{ incorrect}) \leq P_r(X \leq \mathbb{E}X - \frac{N}{6})$$

$$\leq \exp\left(-\frac{2(N/6)^2}{N}\right)$$

$$= \exp\left(-\frac{N}{18}\right) < \delta$$

$$N > 18 \ln(1/\delta).$$

So if rand also can solve  $\Pi$  with  $\frac{1}{3}$  prob(err) using  $n$  rand bits, then  $18n \ln(1/\delta)$  rand bits suffice for  $\leq \delta$  prob(err).  $\Rightarrow \text{BPP} \subseteq \text{P/poly}$