
Today's Lecture

- Feasibility of learning (part 2)
 - Hoeffding inequality and multiple hypotheses
 - Noisy targets
- Generalization error (revisited)
- Toward an effective number of hypotheses } *next time*
 - Dichotomies

FEAS. OF LEARNING, part 2

FROM LAST TIME:

$$(1) \quad P[|E_{\text{in}}(h) - E_{\text{out}}(h)| > \epsilon] \leq 2e^{-2\epsilon^2 N} \quad \text{FOR ANY } \epsilon > 0.$$

~~PRE~~ Hoeffding Ineq. requires:

1. SPECIFY h
2. DRAW \mathcal{D}

3. USE (1) TO GET $E_{\text{out}}(h)$ BOUNDS

ML PARADIGM \Rightarrow

1. COLLECT \mathcal{D} .
2. CONSTRUCT \mathcal{H} .
3. TRAIN TO FIND h_g .
4. CALCULATE $E_{\text{in}}(h_g)$ AND/OR $E_{\text{Test}}(h_g)$.
5. WANT TO FIND OR ESTIMATE $E_{\text{out}}(h_g)$.

How to GET BOUND ON $E_{out}(h_g)$ WHEN WE NEED ~~to~~ TO CHOOSE h_g ?

LET $\mathcal{H} = \{h_1, h_2, \dots; h_M\}$.

LET $h_g =$ FINAL CHOSEN h_m .

AML SHOWS THAT:

$$P[|E_{in}(h_g) - E_{out}(h_g)| \geq \epsilon] \leq P\left[\bigcup_{m=1}^M (|E_{in}(h_m) - E_{out}(h_m)| > \epsilon)\right]$$

$$\leq \sum_{m=1}^M P[|E_{in}(h_m) - E_{out}(h_m)| > \epsilon]$$

$$\leq 2e^{-2\epsilon^2 N}$$

$$\leq 2Me^{-2\epsilon^2 N}$$

(2) $\therefore P[|E_{in}(h_g) - E_{out}(h_g)| > \epsilon] \leq 2Me^{-2\epsilon^2 N}$

$$M = |\mathcal{H}| = \# \text{ OF HYPOTHESES IN } \mathcal{H}.$$

→ A LOOSE UPPER BOUND.

Ex: PLA: $h(\underline{x}) = \text{sgn} \{ \underline{w}^T \underline{x} \},$

$$\underline{w} \in \mathbb{R}^{D+1}$$

$$\Rightarrow M = \infty.$$

2 ASPECTS OF LEARNING FEASIBILITY

FEASIBILITY OF
LEARNING:

can ~~we~~ we get
 $E_{\text{out}}(h_g)$ small enough?

Can ~~we~~ we get
 $|E_{\text{in}}(h_g) - E_{\text{out}}(h_g)|$
small enough?

Can we make
 $E_{\text{in}}(h_g)$ small
enough?

HOEFFDING INEQUALITY

ERROR OF h_g ON \mathcal{D} .

EFFECT OF:

$\mathcal{D} = \mathcal{D}_{\text{Tr}}, \mathcal{D}_{\text{val}}, \text{ or } \mathcal{D}_{\text{test}}.$

COMPLEXITY OF \mathcal{H}

$|\cdot|$ BOUND \uparrow , $E_{\mathcal{D}} \downarrow.$

NUMBER ~~OF~~
OF DATA POINTS, $N.$

$|\cdot|$ BOUND \downarrow , $E_{\mathcal{D}} \uparrow.$

NOISY TARGET FUNCTIONS

INSTEAD OF: $y = f(x)$, AND DATA POINTS
 \uparrow \nwarrow target function.
 o/p class
ARE FROM $p(x)$.

WE HAVE: $p(y|x)$, DATA POINTS
COME FROM:

$$p(x, y) = p(y|x) p(x)$$

e.g. : $y = f(x) + \text{noise}$ (REGR)

OR $y = \text{sgn}\{g(x) + \text{noise}\}$ (CLASS'N).

y_i 'S ARE DRAWN FROM $p(y|x)$, WITH

x_i DRAWN FROM $p(x)$

USE $p(y|x)$ AS THE MORE GENERAL MODEL.

e.g., $y = f(x)$ BY LETTING:

$$p(y|x) = \begin{cases} 1, & y = f(x) \\ 0, & \text{otherwise} \end{cases}, \text{ OR } p(y|x) = \delta[y - f(x)].$$

-- [FIG. 1.11 IN AML] --

GENERALIZATION ERROR [AML 2.1]

HOEFFDING INEQ:

$$P[|E_{\text{out}}(h_{\text{out}}) - E_{\text{in}}(h_{\text{out}})| > \epsilon] \leq 2Me^{-2\epsilon^2 N}$$

$\forall \epsilon > 0$.

RE-ARRANGE:

$$P[|E_{\text{out}}(h_{\text{out}}) - E_{\text{in}}(h_{\text{out}})| \leq \epsilon] > 1 - \delta$$

δ

$$\delta = 2Me^{-2\epsilon^2 N}$$

$$\Rightarrow \epsilon = \sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}}$$

$$\delta > 0 \Rightarrow E_{\text{out}} \leq E_{\text{in}} + \epsilon$$

$$\delta < 0 \Rightarrow E_{\text{out}} \geq E_{\text{in}} - \epsilon$$

(2.1)

$$P[E_{\text{out}}(h_{\text{out}}) \leq E_{\text{in}}(h_{\text{out}}) + \sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}}] > 1 - \delta$$

(2.1')

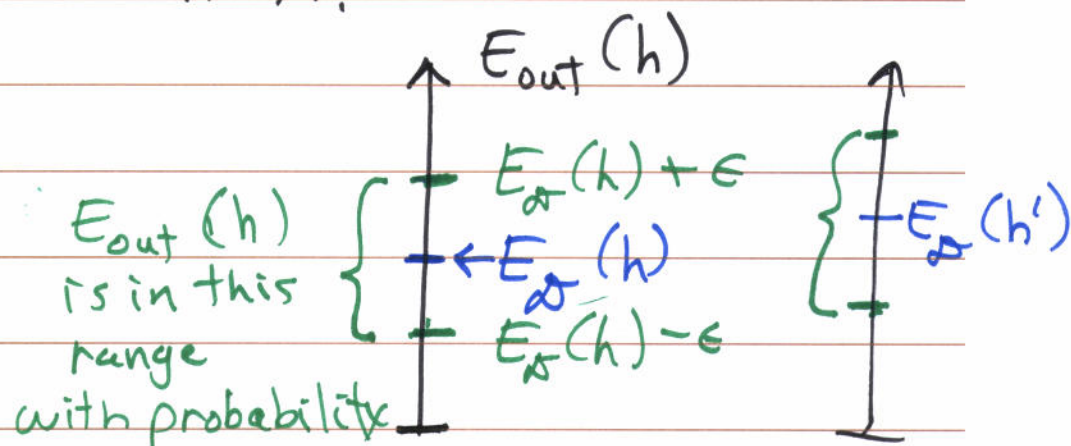
$$P[E_{\text{out}}(h_{\text{out}}) \geq E_{\text{in}}(h_{\text{out}}) - \sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}}] > 1 - \delta$$

δ IS OUR TOLERANCE.

(2.1) GIVES UPPER BOUND ON $E_{\text{out}}(h_g)$
 IN TERMS OF OUR TOLERANCE δ AND N, M .

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(2.1') GIVES LOWER BOUND ON $E_{\text{out}}(h)$,
 FOR ANY $h \in \mathcal{H}$.



$$P > 1 - \delta.$$