

Lecture - 1  
Supplementary Notes

Claim: Suppose loss  $l(f(x), y)$  is bounded i.e.  $a \leq l(f(x), y) \leq b$  for  $\mathcal{A}$  functions  $f \in \mathcal{F}$ , then the following is true:

$$P[|R_{\text{emp}}[f] - R[f]| > \epsilon] \leq 2e^{-\frac{2m\epsilon^2}{(b-a)^2}}$$

(Note: here we are given a fixed  $f \in \mathcal{F}$ )

Def 3 In words, for a given  $f \in \mathcal{F}$  the probability that the empirical risk deviates from true risk (by quantity  $> \epsilon$ ) falls exponentially w.r.t.  $m$ .

Proof:  $R_{\text{emp}}[f] = \frac{1}{m} \sum_{i=1}^m l(f(x_i), y_i)$  ;  $R[f] = E[l(f(x), y)]$ .

$$P[|R_{\text{emp}}[f] - R[f]| > \epsilon] = P[(R_{\text{emp}}[f] - R[f] > \epsilon) \cup (R[f] - R_{\text{emp}}[f] > \epsilon)] \\ \leq P[R_{\text{emp}}[f] - R[f] > \epsilon] + P[R[f] - R_{\text{emp}}[f] > \epsilon]$$

(This is called union bound  $P(\bigcup_{i=1}^n A_i) \leq \sum_{i=1}^n P(A_i)$ . This is also known as Boole's inequality  $\rightarrow$  infect  $P(\bigcup_{i=1}^{\infty} A_i) \leq \sum_{i=1}^{\infty} P(A_i) \rightarrow$  This is more generic result)

Now,  $P[R_{\text{emp}}[f] - R[f] > \epsilon] = P[\underbrace{\sum_{i=1}^m \{l(f(x_i), y_i) - E[l(f(x), y)]\}}_{\text{call it } Z_i} > m\epsilon]$  (I)

We know  $E[l(f(x_i), y_i)] = E[l(f(x), y)] \forall i$  (since  $(x_i, y_i)$  are iid)

$\therefore Z_i$  are mean zero. Moreover,  $a \leq l(f(x_i), y_i) \leq b$

In other words (I) is prob. of question  $\Rightarrow a - E[l(f(x), y)] \leq Z_i \leq b - E[l(f(x), y)]$   
sum of mean zero, iid rvs  $\Rightarrow$  length of interval in which  $Z_i$  moves is  $(b - E[l(f(x), y)]) - (a - E[l(f(x), y)]) = b - a$   
~~longer interval~~ all of which  
move in interval of length  $b - a$ !

bounds on probability regarding such sum of iid bounded r.v.s is efficiently given by Hoeffding bounds which in turn use Chernoff bounding techniques.

Now, (I) is  $P\left[\sum_{i=1}^m Z_i > m\epsilon\right] = P\left[e^{\sum_{i=1}^m \lambda Z_i} > e^{\lambda m\epsilon}\right] \quad \forall \lambda > 0$

Chernoff bounding scheme

$$\leq \frac{E\left[e^{\sum_{i=1}^m \lambda Z_i}\right]}{e^{\lambda m\epsilon}} \quad (\text{by Markov inequality})$$

$$= \frac{E\left[\prod_{i=1}^m e^{\lambda Z_i}\right]}{e^{\lambda m\epsilon}}$$

$$= \frac{\prod_{i=1}^m E\left[e^{\lambda Z_i}\right]}{e^{\lambda m\epsilon}} \quad (\because Z_i \text{ are iid})$$

$$\leq \frac{\prod_{i=1}^m e^{\lambda^2 (b-a)^2 / 8}}{e^{\lambda m\epsilon}} \quad (\text{by Hoeffding inequality given below})$$

Hoeffding Inequality: If  $X$  is a r.v. with mean  $\mu$  &  $a \leq X \leq b$ , then  $E[e^{\lambda X}] \leq e^{\lambda^2 (b-a)^2 / 8}$  (refer to internet for proof of this)

$$\leq e^{\frac{\lambda^2 m (b-a)^2}{8} - \lambda m\epsilon}$$

$$\leq e^{-2m\epsilon^2 / (b-a)^2} \quad (\text{by taking the } \lambda \text{ which minimizes the upper bound})$$

1) by we get,  $P[R(\mathcal{F}) - R_{\text{opt}}(\mathcal{F}) > \epsilon] \leq e^{-2m\epsilon^2 / (b-a)^2}$

$\therefore$  by union bound we get  $P[|R_{\text{opt}}(\mathcal{F}) - R(\mathcal{F})| > \epsilon] \leq 2e^{-2m\epsilon^2 / (b-a)^2}$

Hence Proved.