

1. a)  $\alpha_c = \alpha > 0$  independent of round  $t$

$\frac{1}{2} - \epsilon_c \geq \gamma > 0$ , find optimal  $\alpha$

if  $f(x) = \sum \alpha_i h_i(x)$

Distribution at  $t+1$ :

$$D_{t+1}(i) = \frac{e^{-\sum_{s=1}^t f_s(i)}}{\sum_{j=1}^n e^{-\sum_{s=1}^t f_s(j)}}$$

$$\alpha_n = \frac{1}{2} \ln \left( \frac{1+\gamma}{1-\gamma} \right) \quad (\text{given } \gamma \text{ and } \epsilon_c \text{ independent})$$

$$R_t \leq \frac{1}{n} \sum_i e^{-\sum_{s=1}^t f_s(i)}$$

$$R_t = \sum D_{t+1}(i) \pi Z_t = \pi (e^{-\alpha} (1 - \epsilon_c) + e^{\alpha} \epsilon_c) = \pi (e^{-\alpha} - \epsilon_c e^{-\alpha} + e^{\alpha} \epsilon_c)$$

$$= \pi (\epsilon_c (e^{\alpha} - e^{-\alpha}) + e^{-\alpha})$$

since  $\gamma \geq \frac{1}{2} - \epsilon_c \geq \gamma > 0$  then:  
 $\frac{1}{2} - \gamma \geq \epsilon_c$

$$R_t \leq ((\frac{1}{2} - \gamma) e^{\alpha} - (\frac{1}{2} - \gamma) e^{-\alpha} + e^{-\alpha})^T$$

$$((\frac{1}{2} - \gamma) e^{\alpha} - (\frac{1}{2} - \gamma) e^{-\alpha} + e^{-\alpha})^T = 0, \text{ then?}$$

$$(\frac{1}{2} - \gamma) e^{\alpha} \approx (\frac{1}{2} + \gamma) e^{-\alpha} \quad \text{where the log then:}$$

$$e^{\alpha} = \frac{\frac{1}{2} + \gamma}{\frac{1}{2} - \gamma} e^{-\alpha}$$

$$2\alpha = \ln \left( \frac{\frac{1}{2} + \gamma}{\frac{1}{2} - \gamma} \right) \Rightarrow \alpha = \frac{1}{2} \ln \left( \frac{1+\gamma}{1-\gamma} \right)$$

b) going off a):

$$p(\text{wrong}) = \epsilon_c e^{\alpha}$$

$$p(\text{correct}) = (1 - \epsilon_c) e^{-\alpha} \quad \& \quad \text{since } \epsilon_c = \frac{1}{2} - \gamma \quad \& \quad \alpha = \frac{1}{2} \ln \left( \frac{1+\gamma}{1-\gamma} \right)$$

Using mathematics:

$$p(\text{wrong}) - p(\text{correct}) = 0 \quad \text{so } p(\text{wrong}) = p(\text{correct}) \text{ for this } \alpha$$

c)  $R_t \leq ((\frac{1}{2} - \gamma) e^{\alpha} + (\frac{1}{2} + \gamma) e^{-\alpha})^T$  & substituting in  $\alpha = \frac{1}{2} \ln \left( \frac{0.5+\gamma}{0.5-\gamma} \right)$   
 in mathematics:  $\left( \sqrt{\frac{1+\gamma}{1-\gamma}} (\frac{1}{2} - \gamma) + \frac{\gamma + \frac{1}{2}}{\sqrt{\frac{1+\gamma}{1-\gamma}}} \right)^T$

$$\text{upper bound: } \exp(-2\gamma^2 T)$$

from alternate forms  
 in mathematics  
 when set to 0  
 initial  $\gamma$  bounds



# Machine Learning: Boosting

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2.  $t: h_t$ ,  $t+1: h_{t+1}$

Proof by Contradiction attempt 2.

$$\alpha_t = \frac{1}{2} \ln \left( \frac{1-\epsilon_t}{\epsilon_t} \right)$$

$$Z_t = 2 \sqrt{\epsilon_t (1-\epsilon_t)}$$

so assuming  $\epsilon_t = \frac{1}{4}$ :

$$\alpha_t = \frac{1}{2} \ln \left( \frac{1-\frac{1}{4}}{\frac{1}{4}} \right) = 0.594$$

$$Z_t = 2 \sqrt{\frac{1}{4} (1-\frac{1}{4})} = 0.866$$

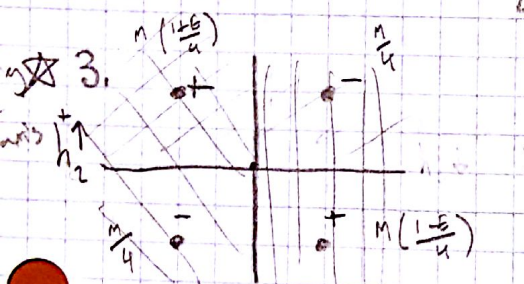
So for our distribution we have

$$D_t(x) = \frac{D_0(x) e^{-\alpha_t h_t(x)}}{Z_t} = \frac{1}{Z_t} = \frac{e^{-0.594}}{0.866} = \frac{2.09}{n}$$

$$\epsilon_{t+1} = \frac{1}{n} \cdot \frac{2.09}{n} = \frac{2.09}{n} = 0.5225$$

Since  $0.5225 > \frac{1}{2}$ , then we are

Say  $h_{t+1}$  is different by contradiction.



$$\frac{1}{2} \ln \left( \frac{1-\epsilon}{\epsilon} \right), \epsilon = \frac{1}{2}$$

$$\text{if } M=4 : (-1, 1) : \frac{1}{4} = \frac{1}{4} = 1$$

$$\frac{M(1-\epsilon)}{4} = \frac{4(0.5)}{4} = \frac{1}{2}$$

$$\frac{M(1+\epsilon)}{4} = \frac{4(1.5)}{4} = \frac{3}{2}$$

everything left of y-axis  
everything right of y-axis

$$\epsilon_1 = \frac{1}{n} \left( \frac{M}{4} + \frac{M}{4} (1-\epsilon) \right) = \frac{1}{4} + \frac{1}{4} (1-\frac{1}{2}) = \frac{1}{4} + \frac{1}{8} = \frac{3}{8}$$

$$Z_1 = 2 \sqrt{\epsilon_1 (1-\epsilon_1)} = 2 \sqrt{\frac{3}{8} (1-\frac{3}{8})} = 2 \sqrt{\frac{15}{64}} = 0.968$$

$$\alpha_1 = \frac{1}{2} \ln \left( \frac{1-\epsilon}{\epsilon} \right) = \frac{1}{2} \ln \left( \frac{1-\frac{3}{8}}{\frac{3}{8}} \right) = 0.255$$

$$D_2 = \frac{D_1 e^{-\alpha_1 h_1(x)}}{Z_1} = \frac{1.33}{n}, D_2 = \frac{0.18}{n}$$

when it gets bigger than  $\frac{1}{2}$  then we stop.

$$\epsilon_2 = \frac{1.75}{n} \left( \frac{M}{4} (1-\epsilon) \right) + \frac{0.8}{n} \left( \frac{M}{4} \right) = 0.535$$