



$$1. D_{t+1}(i) = \frac{D_t(i) \exp(-\alpha_t g_{if_t}(x_i))}{Z_t}$$

$$D_{T+1}(i) = \frac{1}{m} \frac{\exp(-y_i f_T(x_i))}{\prod_{t=1}^T Z_t}$$

$$\hat{f}(h) \leq \frac{1}{m} \sum_{i=1}^m [\langle y_i, f_t(x_i) \rangle] = \frac{1}{m} \sum_{i=1}^m e^{-y_i f_t(x_i)}$$

$$= \sum_{i=1}^m \frac{1}{\prod_{t=1}^T Z_t} D_{T+1}(i) = \frac{1}{\prod_{t=1}^T Z_t} \sum_{i=1}^m D_{T+1}(i) = \frac{1}{\prod_{t=1}^T Z_t}$$

$$\text{即证 } \prod_{t=1}^T Z_t \leq e^{-2 \sum_{t=1}^T (\frac{1}{2} - \xi_t)^2} =$$

$$\text{即证 } \sum_{t=1}^T \log Z_t \leq \sum_{t=1}^T -2(\frac{1}{2} - \xi_t)^2$$

$$d(\xi_t) = \log Z_t + 2(\frac{1}{2} - \xi_t)^2$$

$$= 4\xi_t^2 + 4\xi_{t+1} + \ln 2 + \frac{1}{2} \ln \xi_t + \frac{1}{2} \ln(1 - \xi_t)$$

$$\frac{\partial d(\xi_t)}{\partial \xi_t} = \left(8 - \frac{2}{\xi_t(1-\xi_t)} \right) \left(\xi_t - \frac{1}{2} \right)$$

$d(\xi_t)$ 在 $(0, \frac{1}{2})$ 上同增, $(\frac{1}{2}, 1)$ 上同减
($\frac{1}{2}$ 处取最大值)

$\leq (0, 1)$ 上 $d(\xi_t) \leq 0$.

$$\therefore \text{有 } \hat{f}(h) \leq \prod_{t=1}^T Z_t \leq e^{-\frac{1}{2} \sum_{t=1}^T (\frac{1}{2} - \xi_t)^2}, \text{ 即证}$$



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2. $H \subseteq \{-1, +1\}^X$

for $i \leq 1$ to m do

$D_t(i) \leftarrow w_i$

for $t \leq 1$ to T do

$h_t \leftarrow$ base classifier in H with $\epsilon_t = \Pr_{i \sim D_t} [h_t(x_i) \neq y_i]$

$\alpha_t \leftarrow \frac{1}{2} \log \frac{1 - \epsilon_t}{\epsilon_t}$

$Z_t \leftarrow 2 \left[\epsilon_t C(h_t) \right]^{\frac{1}{2}}$

for $i \leq 1$ to m do

$D_{t+1} \leftarrow \frac{D_t(i) \exp(-\alpha_t y_i h_t(x_i))}{Z_t}$

$f_t \leftarrow \sum_{s=1}^t \alpha_s h_s$

return $h = \text{sgn}(f_T)$

$$D_{t+1}(i) = \frac{D_t(i) e^{-\alpha_t y_i h_t(x_i)}}{Z_t} = \frac{w_i}{\sum_{s=1}^m Z_s} = \frac{w_i}{\sum_{s=1}^m Z_s} e^{-y_i \sum_{s=1}^t \alpha_s h_s(x_i)}$$

$$F(\bar{\alpha}) = \sum_{i=1}^n w_i e^{-y_i f(x_i)} = \sum_{i=1}^n w_i e^{-y_i \sum_{t=1}^T \alpha_t h_t(x_i)}$$

$$F(\bar{\alpha}_{t+1}, e) = \sum_{i=1}^n w_i e^{-y_i \sum_{s=1}^t \alpha_s h_s(x_i) - \eta y_i h_t(x_i)}$$

$$F'(\bar{\alpha}_{t+1}, e) = - \sum_{i=1}^n y_i h_t(x_i) w_i e^{-y_i \sum_{s=1}^t \alpha_s h_s(x_i)}$$

$$= - \sum_{i=1}^n y_i h_t(x_i) \bar{D}_t(i) \bar{Z}_t$$

$$= - \left[\sum_{i=1}^n \bar{D}_t(i) [y_i h_t(x_i) = 1] - \sum_{i=1}^n \bar{D}_t(i) [y_i h_t(x_i) = -1] \right] \bar{Z}_t$$

$$= - \left[(1 - \bar{\epsilon}_{t,1}) - \bar{\epsilon}_{t,1} \right] \bar{Z}_t = \left[2 \bar{\epsilon}_{t,1} \right] \bar{Z}_t$$

$$\frac{\partial F(\bar{\alpha}_{t+1})}{\partial \eta}$$

$$= \sum_{i=1}^n y_i h_t(x_i) \bar{D}_t(i) \bar{Z}_t e^{-\eta y_i h_t(x_i)}$$

$$= \sum_{i=1}^n \bar{D}_t(i) [y_i h_t(x_i) = 1] e^{-\eta} - \sum_{i=1}^n \bar{D}_t(i) [y_i h_t(x_i) = -1] e^{\eta}$$

$$\Rightarrow \left[(1 - \bar{\epsilon}_{t,1}) e^{-\eta} - \bar{\epsilon}_{t,1} e^{\eta} \right] =$$

$$\eta = \frac{1}{2} \log \frac{1 - \bar{\epsilon}_{t,1}}{\bar{\epsilon}_{t,1}}$$

