

Results so far

• Homework #3 due 03/08/2021

• Project Proposal Due 03/10/2021

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

② $\sum_{h_t(x_i) \neq y_i} D_{t+1}(i) = \frac{1}{2}$

③ Bounds on the training error

ensemble decision
↓
$$\left[\hat{\text{err}}(H) = \frac{1}{n} \sum_{i=1}^n \mathbb{I}[H(x_i) \neq y_i] \right] \leq \frac{1}{2} \text{sign} \left(\sum_{t=1}^T \alpha_t h_t(x) \right)$$

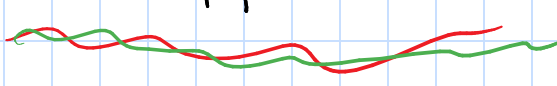
$$D_{t+1}(i) = \frac{D_t(i)}{Z_t} \exp(-\alpha_t h_t(x_i) y_i)$$

$$= \frac{D_{t-1}(i)}{Z_t Z_{t-1}} \exp(-\alpha_t h_t(x_i) y_i) \exp(-\alpha_{t-1} h_{t-1}(x_i) y_i)$$

...

$$D_{t+1}(i) = \frac{D_1(i)}{\prod_{r=1}^t Z_r} \exp \left(-y_i \sum_{r=1}^t \alpha_r h_r(x_i) \right)$$

$\xrightarrow{H(x)}$



From the previous lecture

$$\star D_1(i) = \frac{1}{n}$$

Let us examine what happens when we make a mistake on x_i ,

$$\text{sign} \left(\sum \alpha_t h_t(x_i) \right) \neq y_i \Rightarrow \text{sign} \left(y_i \sum_{t=1}^T \alpha_t h_t(x_i) \right) = -1$$

$$H(x_i) \neq y_i$$

$$\exp(-y_i \sum \alpha_t h_t(x_i)) \geq 1 \quad *$$

$$H(x_i) = y_i$$

$$\exp(-y_i \sum \alpha_t h_t(x_i)) \geq 0$$

$$\hat{\text{err}}(H) = \frac{1}{n} \sum_{i=1}^n \mathbb{I}[H(x_i) \neq y_i]$$

$$\leq \frac{1}{n} \sum_{i=1}^n \exp(-y_i \sum_{t=1}^T \alpha_t h_t(x_i))$$

$$= \frac{1}{n} \sum_{i=1}^n \cancel{n} D_{t+1}(i) \prod_{t=1}^T z_t$$

$$= \left(\sum_{i=1}^n D_{t+1}(i) \right) \prod_{t=1}^T z_t$$

$$= \prod_{t=1}^T 2 \sqrt{\epsilon_t (1 - \epsilon_t)} = 2^T \prod_{t=1}^T \sqrt{\epsilon_t (1 - \epsilon_t)}$$

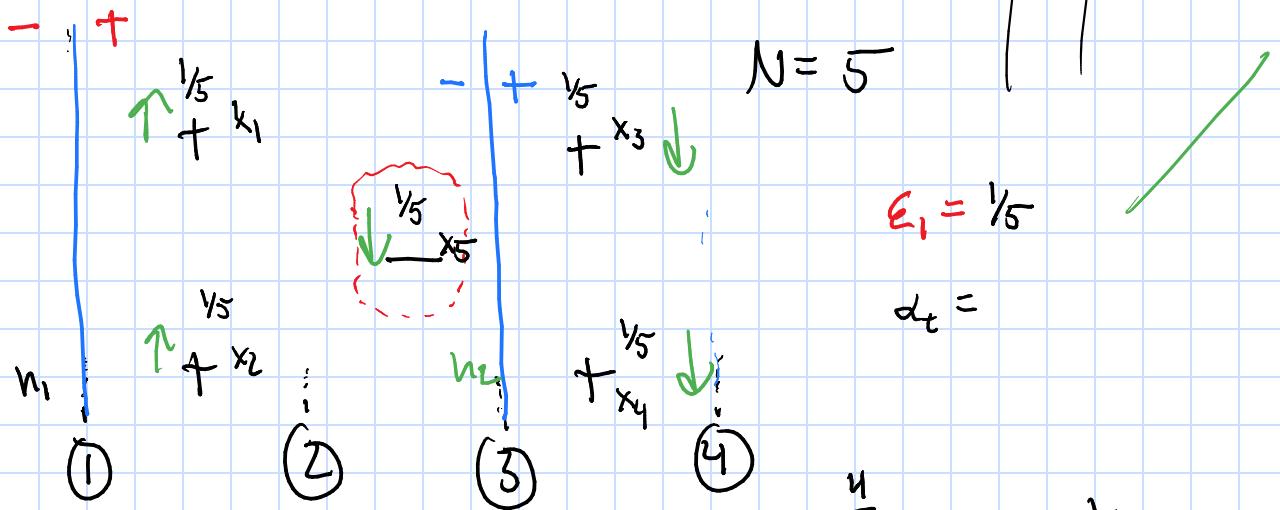
→ This error drops off exponentially fast

$$(2 \sqrt{\epsilon_1 (1 - \epsilon_1)}) (2 \sqrt{\epsilon_2 (1 - \epsilon_2)}) = 2^2 \prod_{t=1}^2 \sqrt{\epsilon_t (1 - \epsilon_t)}$$

$$i = \text{rand sample}(D, N)$$

$$X_{tr}, y_{tr} = X[i], y[i]$$

$$h_t = \text{CART}(X_{tr}, y_{tr}, \text{params})$$



$$\sum_{i=1}^n D_z(i) = \frac{1}{2}$$

$$D_z(5) = \frac{1}{2}$$

