## Multiplicative Weight Algorithm

Lecture 6.5

2. for 
$$t=1, \cdots$$
,

the committed probabily distribution is

 $P_{+}(a) \propto W_{+-1}(a)$ 

P<sub>t</sub> (a) 
$$\alpha$$
P<sub>t</sub> (b)  $\alpha$ 
P<sub>t</sub> am receives an whiting of

 $\pi_{t}(\alpha)$ 

5. After knowing  $T_{t}$ , update  $W_{t}(a) = W_{t-1}(a) \cdot (1+\xi)^{T_{t}(a)}$ He (A)  $A = \frac{1}{2} \sum_{k=1}^{n} A_{t}(k) \cdot (1+\xi)^{T_{t}(a)}$ 

Theorem: Let |A|=n. Then the MW algorithm has external regret  $O\left(\sqrt{\frac{\log n}{T}}\right)$ .

Proof:  $T_t := \sum_{\alpha \in A} w_t(\alpha)$  the sum of the weight value in the t-th iteration.

Expected white 
$$\frac{1}{2}$$
:  $\frac{1}{2}$   $\frac{1}{2}$ 

$$T_{T} \geqslant \frac{T}{T} (1+\epsilon)^{T_{t}} (\epsilon^{x})$$

$$= (1+\epsilon)^{T_{t}} \cdots \geqslant T_{t} \geqslant (1+\epsilon)^{T_{t}}$$

$$= (1+\epsilon)^{T_{t}} \cdots \geqslant T_{t} \geqslant (1+\epsilon)^{T_{t}} \qquad (1)$$

$$T_{T} = \sum_{\alpha \in A} \omega_{T}(\epsilon)$$

$$= \sum_{\alpha \in A} \omega_{T-1}(\epsilon) \cdot (1+\epsilon)^{T_{t}} \qquad (1+\epsilon)^{T_{t}} \approx (1+\epsilon)^{T_{t}} \approx$$

$$\Rightarrow \begin{array}{ll} \mathsf{OPT} - \sum_{t=1}^{T} \mathcal{I}_{t} & \leq & \mathsf{S} \ \mathsf{OPT} + \frac{\ln n}{\mathsf{E}} \\ & \leq & \mathsf{E} \mathsf{T} + \frac{\ln n}{\mathsf{E}} \\ & \leq & \mathsf{E} \mathsf{T} + \frac{\ln n}{\mathsf{E}} \end{array}$$

$$= \frac{2}{\mathsf{T}} \left( \begin{array}{ll} \mathsf{OPT} - \sum_{t=1}^{T} \mathcal{I}_{t} \\ & \leq & \mathsf{E} \mathsf{T} \end{array} \right) \left( \begin{array}{ll} \mathsf{Put} \ \mathsf{S} = \sqrt{\frac{\ln n}{\mathsf{T}}} \end{array} \right)$$

$$\Rightarrow \frac{1}{\mathsf{T}} \left( \begin{array}{ll} \mathsf{OPT} - \sum_{t=1}^{T} \mathcal{I}_{t} \\ & \leq & \mathsf{E} \mathsf{T} \end{array} \right) \left( \begin{array}{ll} \mathsf{In} \ \mathsf{n} \\ & \mathsf{T} \end{aligned} \right)$$

