

\* Today's lecture is optional (not on exam)

\* Exam 3 on Nov 20 (Thurs)

### Experts Problem:

- $n$  experts
  - Each round  $t \in \{1, \dots, T\}$
- (1) Alg chooses an expert / action  $a_t \in \{1, \dots, n\}$  random choice is allowed
- (2) Cost of each expert  $c_t(i) \in [-1, 1]$  revealed
- 8) Alg gets  $c_t(a_t)$

Objective: Min total Alg cost  $\sum_{t=1}^T c_t(a_t)$   
compared to best fixed expert  $\min_i \sum_{t=1}^T c_t(i)$

Remark: This problem is useful in Forecasting (weather, stocks),  
Spam detection, LP solving

a random alg with expected

Thm (lec 21): OGD implies  $\downarrow$  avg regret

$$\frac{1}{T} \left[ \mathbb{E} \left[ \sum_t c_t(a_t) \right] - \sum_t c_t(i) \right] \leq 2 \frac{\sqrt{n}}{\sqrt{T}}$$

Avg Regret

Thm (today): Hedge alg has  $\mathbb{E}[\text{Avg Regret}] \leq O\left(\sqrt{\frac{\log n}{T}}\right)$

Ques: Randomization is necessary to obtain  $O(1)$  avg regret.

$n=2$ experts	$c_t(i)$		
Cost	$c_1(1) = 1$	$c_2(a_2) = 1$	$c_3(a_3) = 1$
	$c_1(2) = 0$	$c_2(\bar{a}_2) = 0$	$c_3(\bar{a}_3) = 0$

For det alg, there is sequence where Alg incurs 1 cost per step  
 $\Rightarrow T$  cost overall

Benchmark  $\min_i \sum_t c_t(i) \leq T/2$

$$\Rightarrow \text{Total Regret} \geq T - T/2 = T/2$$

$$\Rightarrow \text{Avg regret} \geq \frac{T}{2}$$

**Thm (today):** Hedge alg has  $\mathbb{E}[\text{Avg Regret}] \leq O\left(\sqrt{\frac{\log n}{T}}\right)$

Alg

(1) Maintain wts  $w_t(i) \geq 0 \leftarrow \text{expert } i$

Initialized to 1 at  $t=1$

(2) In each round  $t$ :

choose expert  $i$  with prob  $\frac{w_t(i)}{\sum_{i=1}^n w_t(i)}$

$$(3) w_{t+1}(i) = w_t(i) \cdot \exp(-\eta_j c_t(i))$$

$$= \underbrace{w_t(i)}_{=1} \cdot \exp\left(-\eta \sum_{s \leq t} c_s(i)\right)$$

Step size  $\approx \sqrt{\frac{\log n}{T}}$

Pf: Define potential  $\phi_t = \sum_{i=1}^n w_t(i) \rightarrow \phi_1 = n$

Lemma:  $\phi_{t+1} \leq \phi_t \exp\left(-\eta \underbrace{\text{Alg}_t}_{t\text{-th step}} + \eta^2\right)$   $\mathbb{E}[\text{Cost of alg in}]$

Assuming lemma,

$$w_{t+1}(i) \leq \phi_{t+1} \leq \underbrace{\phi_1}_{=n} \exp\left(-\eta \sum_{t=1}^T \text{Alg}_t + \eta^2 T\right)$$

= Alg

$$\Rightarrow \log(w_{t+1}(i)) = \log\left(\exp\left(-\sum_{t=1}^T \eta c_t(i)\right)\right) \leq \log\left(n \cdot \exp(-\eta \text{Alg} + \eta^2 T)\right)$$

$$\Rightarrow -\eta \sum_t c_t(i) \leq \log n + (-\eta \text{Alg} + \eta^2 T)$$

$$\Rightarrow \text{Alg} \leq \sum_t c_t(i) + \underbrace{\log n}_{-\eta_1} + \eta T$$

$\sqrt{n}$

0  $\neq$   $t \sim$   $\underbrace{\text{avg}}$  + '0'

$$\underbrace{\eta}_{\leq \sqrt{T \log n}} \text{ for } \eta = \sqrt{\frac{\log n}{T}}$$

Lemma:  $\phi_{t+1} \leq \phi_t \exp(-\eta \text{Alg}_t + \eta^2)$

Pf:  $e^{-x} \leq 1 - x + x^2$  for  $x \leq 1$

$$e^{-x} \geq 1 - x \quad e^{-x} \approx 1 - x + \frac{x^2}{2} - \frac{x^3}{3!} \rightarrow \dots$$

$$\begin{aligned}\phi_{t+1} &= \sum_i w_{t+1}(i) = \frac{\phi_t \sum_i w_t(i) \exp(-\eta C_t(i))}{\phi_t} \\ &\leq \phi_t \sum_i \frac{w_t(i)}{\phi_t} \left( 1 - \eta C_t(i) + \eta^2 \cancel{\frac{C_t(i)^2}{\phi_t^2}} \right) \stackrel{\leq 1}{\circlearrowleft} \\ &\leq \phi_t \left( 1 - \eta \text{Alg}_t + \eta^2 \right) \\ &\leq \phi_t \exp(-\eta \text{Alg}_t + \eta^2) \quad \blacksquare\end{aligned}$$

Remark: Hedge alg can be captured as a generalization of OGD  
called Online Mirror Descent