

Adversarial Bandits.

For some fixed time horizon T imagine matrix

$$x_{t,i} \quad \begin{array}{l} t \in \{1, \dots, T\} \\ i \in \{1, \dots, n\} \end{array}$$

Your algorithm chooses I_t based on history

$$R_T = \max_i \sum_{t=1}^T x_{t,i} - x_{t,I_t}$$

Suppose I_t is a deterministic choice given history?

I_1 is deterministic. So define $x_{1,i} = \begin{cases} 0 & \text{if } i = I_1 \\ 1 & \text{o.w.} \end{cases}$

I_2 is deterministic given (I_1, X_{1,I_1})

so define $x_{2,i} = \begin{cases} 0 & \text{if } i = I_2 \\ 1 & \text{o.w.} \end{cases}$

Any deterministic algorithm gets $\Omega(T)$ regret against adversarial sequence.

At time t , draw $I_t \sim P_t$, observe X_{t,I_t}

and construct $\hat{x}_{t,i} = \frac{\mathbb{1}\{I_t=i\}}{P_{t,i}} X_{t,I_t}$

$$\mathbb{E}[\hat{x}_{t,i} | P_t] = \mathbb{E}\left[\frac{\mathbb{1}\{I_t=i\}}{P_{t,i}} X_{t,I_t} | P_t\right]$$

$$= \sum_{j=1}^n \underbrace{P(I_t=j)}_{=P_{t,j}} \frac{\mathbb{1}\{j=i\}}{P_{t,i}} X_{t,j}$$

$$= P_{t,i} \cdot \frac{1}{P_{t,i}} \cdot X_{t,i} = x_{t,i}$$

$$S_{t,i} = \sum_{s=1}^t \hat{x}_{t,i} \quad \mathbb{E}[S_{t,i}] = \sum_{s=1}^t x_{t,i} \quad \forall i$$

$$V(\hat{x}_{t,i}) = \mathbb{E}[\hat{x}_{t,i}^2] - \underbrace{\mathbb{E}[\hat{x}_{t,i}]^2}_{x_{t,i}^2}$$

$$= \sum_j P_{t,j} \frac{\mathbb{1}\{j=i\}}{P_{t,i}^2} x_{t,j}^2 - x_{t,i}^2$$

$$= \frac{x_{t,i}^2}{P_{t,i}} - x_{t,i}^2 = \frac{x_{t,i}^2 (1 - P_{t,i})}{P_{t,i}}$$

$$x_{t,i} =: 1 - l_{t,i} \quad \text{Assume } x_{t,i} \in [0,1]$$

$$\max_i \sum_{t=1}^T x_{t,i} - x_{t,i} = \max_i \sum_{t=1}^T l_{t,I_t} - l_{t,i}$$

$$\hat{l}_{t,i} = \frac{\mathbb{1}\{I_t=i\}}{P_{t,i}} l_{t,I_t}$$

EXP3

Adversary chooses $\{l_{t,i}\}_{\substack{i=1,\dots,n \\ t=1,\dots,T}}$

$W_0 = \mathbf{1} \in \mathbb{R}^n$, γ given

for $t=1, 2, \dots, T$

$$P_{t,i} = \frac{W_{t-1,i}}{\sum_j W_{t-1,j}} = \frac{W_{t-1,i}}{W_{t-1}} \quad (P_t \in \Delta_n)$$

$I_t \sim P_t$, observe l_{t,I_t}

$$\text{Set } \hat{l}_{t,i} = \frac{\mathbb{1}\{I_t=i\}}{P_{t,i}} l_{t,I_t}$$

$$W_{t,i} = \exp(-\gamma \hat{l}_{t,i}) W_{t-1,i}$$

$$= \exp\left(-\gamma \sum_{s=1}^t \hat{l}_{s,i}\right)$$

Theorem | Regret of EXP3 for any loss seq
in $[0,1]^{T \times n}$ is

$$\max_i \mathbb{E} \left[\sum_{t=1}^T l_{t,I_t} - l_{t,i} \right] \leq \sqrt{2nT \log(n)}.$$

$$\sum_{t=1}^T \log \left(\frac{W_t}{W_{t-1}} \right)$$

$$W_t = \sum_i W_{t,i}$$

$$\begin{aligned} \sum_{t=1}^T \log \left(\frac{W_t}{W_{t-1}} \right) &= \log \left(\frac{W_T}{W_{T-1}} \right) + \log \left(\frac{W_{T-1}}{W_{T-2}} \right) + \dots - \\ &= \log(W_T) - \log(W_{T-1}) + \log(W_{T-1}) - \log(W_{T-2}) + \dots - \\ &= \log(W_T) - \log(W_0) \\ &= \log \left(\sum_i \exp(-\lambda \sum_{t=1}^T \hat{\ell}_{t,i}) \right) - \log(n) \\ &\geq -\lambda \sum_{t=1}^T \hat{\ell}_{t,i} - \log(n) \quad \text{for any } i \in [n]. \end{aligned}$$

$$\log \left(\frac{W_t}{W_{t-1}} \right) = \log \left(\sum_{i=1}^n \frac{W_{t,i}}{W_{t-1}} \right)$$

$$= \log \left(\sum_{i=1}^n \frac{W_{t-1,i}}{W_{t-1}} \exp(-\lambda \hat{\ell}_{t,i}) \right)$$

$$= \log \left(\sum_{i=1}^n p_{t,i} \exp(-\lambda \hat{\ell}_{t,i}) \right)$$

$$e^x \geq 1+x \quad \forall x$$

$$e^{-x} \leq 1 - x + \frac{x^2}{2}, \quad x \geq 0$$

$$\log(1+x) \leq x$$

$$f(x) \leq f(0) + f'(0)x + \sup_{z \in (0,x)} f''(z) \frac{x^2}{2}$$

$$\leq \log \left(\sum_{i=1}^n P_{t,i} \left(1 - \zeta \hat{l}_{t,i} + \frac{\zeta^2 \hat{l}_{t,i}^2}{2} \right) \right)$$

$$= \log \left(1 - \zeta l_{t, \bar{I}_t} + \frac{\zeta^2}{2} \sum_{i=1}^n P_{t,i} \hat{l}_{t,i}^2 \right)$$

$$\leq -\zeta l_{t, \bar{I}_t} + \frac{\zeta^2}{2} \sum_{i=1}^n P_{t,i} \hat{l}_{t,i}^2$$

$$\sum_{i=1}^n P_{t,i} \hat{l}_{t,i} = \sum_i P_{t,i} \frac{\mathbb{1}\{\bar{I}_t = i\}}{P_{t,i}} l_{t, \bar{I}_t} = l_{t, \bar{I}_t}$$

$$-\zeta \sum_{t=1}^T \hat{l}_{t,i} - \log(n) \leq \sum_t \log \left(\frac{W_t}{W_{t-1}} \right) \leq -\zeta \sum_{t=1}^T \hat{l}_{t, \bar{I}_t} + \frac{\zeta^2}{2} \sum_{t,i} P_{t,i} \hat{l}_{t,i}^2$$

divide ζ on both sides and rearrange

$$\sum_{t=1}^T \hat{l}_{t, \bar{I}_t} - \hat{l}_{t,i} \leq \frac{\log(n)}{\zeta} + \frac{\zeta}{2} \sum_t \underbrace{\sum_i P_{t,i} \hat{l}_{t,i}^2}_{\text{green bracket}}$$

$$\mathbb{E} \left[\sum_{t=1}^T \hat{l}_{t, \bar{I}_t} - \hat{l}_{t,i} \right] = \mathbb{E} \left[\sum_{t=1}^T \hat{l}_{t, \bar{I}_t} - \hat{l}_{t,i} \right]$$

$$\leq \frac{\log(n)}{\zeta} + \frac{\zeta}{2} nT \quad \square$$

$$\begin{aligned}
\mathbb{E} \left[\sum_i P_{t,i} \hat{\ell}_{t,i}^2 \right] &= \mathbb{E} \left[\sum_i P_{t,i} \frac{\mathbb{1}_{\{I_t=i\}}}{P_{t,i}^2} \ell_{t,i}^2 \right] \\
&= \sum_j P_{t,j} \sum_i P_{t,i} \frac{\mathbb{1}_{\{j=i\}}}{P_{t,i}^2} \ell_{t,i}^2 \\
&= \sum_i \ell_{t,i}^2 \leq n
\end{aligned}$$