

Adversary picks $l_t \in [-1, 1]^n \quad \forall t=1, \dots, T$

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

Player chooses $I_t \in [n]$ and observes l_{t, I_t}

$$\text{Regret} = \max_i \mathbb{E} \left[\sum_{t=1}^T l_{t, I_t} - l_{t, i} \right]$$

Let $P_t \in \Delta_n = \{ z \in \mathbb{R}_+^n : \sum_i z_i = 1 \}$, $P(I_t = i) = P_{t, i}$

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

$$\begin{aligned} \mathbb{E}[\hat{l}_{t, i}] &= \sum_{j=1}^n P(I_t = j) \frac{\mathbb{1}\{j=i\}}{P_{t, i}} l_{t, j} \quad \leftarrow \text{only well-defined if } P_{t, i} > 0 \forall i \\ &= \sum_{j=1}^n P_{t, j} \frac{\mathbb{1}\{j=i\}}{P_{t, i}} l_{t, j} \\ &= l_{t, i} \end{aligned}$$

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

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

$$= \sum_{j=1}^n P(I_t = j) \frac{\mathbb{1}\{j=i\}^2}{P_{t, i}^2} l_{t, j}^2 - l_{t, i}^2$$

$$= \sum_{j=1}^n \frac{\mathbb{1}\{j=i\}}{P_{t, i}} l_{t, j}^2 - l_{t, i}^2$$

$$= l_{t,i}^2 \left(\frac{1}{p_{t,i}} - 1 \right)$$

EXP3(γ)

Adversary chooses $l_t \in [-1, 1]^n \forall t$

Init: Choose $\lambda \in \Delta_n$. Set $p_1 = (\frac{1}{n}, \dots, \frac{1}{n})$, $w_1 = (1, \dots, 1)$, $\beta > 0$

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

Draw I_t from $q_t = \frac{(1-\gamma)p_t + \gamma\lambda}{\sum_j w_{t,j}}$

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

$$w_{t+1,i} = w_{t,i} \exp(-\beta \hat{l}_{t,i}) \quad p_{t+1,i} = \frac{w_{t+1,i}}{\sum_j w_{t+1,j}}$$

$$\mathbb{E} \left[\sum_{t=1}^T l_{t, I_t} - l_{t,i} \right] = \mathbb{E} \left[\sum_{t=1}^T \sum_{j=1}^n q_{t,j} l_{t,j} - l_{t,i} \right]$$

$$\leq \gamma T + \mathbb{E} \left[\sum_{t=1}^T \sum_{j=1}^n (1-\gamma) p_{t,j} l_{t,j} - l_{t,i} \right]$$

$$\leq \gamma T - \gamma \mathbb{E} \left[\sum_{t=1}^T \sum_{j=1}^n p_{t,j} l_{t,j} \right] + \mathbb{E} \left[\sum_{t=1}^T \sum_{j=1}^n p_{t,j} l_{t,j} - l_{t,i} \right]$$

$$\leq 2\gamma T + \mathbb{E} \left[\sum_{t=1}^T \sum_{j=1}^n p_{t,j} l_{t,j} - l_{t,i} \right]$$

$$\leq 2\delta T + \mathbb{E} \left[\sum_{t=1}^T \sum_{j=1}^n P_{t,j} \hat{l}_{t,j} - \hat{l}_{t,i} \right]$$

$$W_t = \sum_{i=1}^n w_{t,i}$$

$$\log \left(\frac{W_{T+1}}{w_i} \right) \geq \log \left(\frac{w_{T+1,i}}{w_i} \right) = \log(w_{T+1,i}) - \log(n)$$

$$e^x \leq 1+x+x^2 \quad \forall x \leq 1 \quad = -\gamma \sum_{t=1}^T \hat{l}_{t,i} - \log(n)$$

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

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

$$= \sum_{t=1}^T \log \left(\frac{1}{W_t} \sum_{i=1}^n w_{t,i} \exp(-\gamma \hat{l}_{t,i}) \right)$$

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

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

$$\leq -\gamma \sum_{t=1}^T \sum_{i=1}^n P_{t,i} \hat{l}_{t,i} + \gamma^2 \sum_{t=1}^T \sum_{i=1}^n P_{t,i} \hat{l}_{t,i}^2$$

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

$$|\mathbb{E} \hat{\ell}_{t,i}| \leq 1$$

\Rightarrow

$$\max_j \mathbb{E} \left[\sum_{t=1}^T \ell_{t,j} - \min_{i \neq j} \sum_{t=1}^T \ell_{t,i} \right] \leq 2\gamma T + \frac{\log(n)}{\gamma} + \gamma \mathbb{E} \left[\sum_{t=1}^T \sum_{i \neq j} p_{t,i} \hat{\ell}_{t,i}^2 \right]$$

$$q_{t,i} = (1-\gamma) p_{t,i} + \gamma \lambda_i \geq (1-\gamma) p_{t,i} \Rightarrow p_{t,i} \leq \frac{q_{t,i}}{1-\gamma}$$

$$\geq \gamma \lambda_i = \gamma/n$$

$$\mathbb{E} \left[\sum_{t=1}^T \sum_{i \neq j} p_{t,i} \hat{\ell}_{t,i}^2 \right] = \mathbb{E} \left[\sum_{t=1}^T \sum_{i \neq j} p_{t,i} \frac{\mathbb{1}\{I_t=i\}}{q_{t,i}^2} \ell_{t,i}^2 \right] \quad \lambda = (\frac{1}{n}, \dots, \frac{1}{n})$$

$$\leq \frac{1}{1-\gamma} \mathbb{E} \left[\sum_{t=1}^T \sum_{i \neq j} \frac{\mathbb{1}\{I_t=i\}}{q_{t,i}} \ell_{t,i}^2 \right]$$

$$= \frac{1}{1-\gamma} \sum_{t=1}^T \sum_{i \neq j} \ell_{t,i}^2$$

$$\leq \frac{nT}{1-\gamma}$$

$$\text{Regret} \leq 2\gamma T + \frac{\log(n)}{\gamma} + \frac{\gamma n T}{(1-\gamma)}$$

Need $-\mathbb{E} \hat{\ell}_{t,i} < 1$

$$|\mathbb{E} \hat{\ell}_{t,i}| = \left| \mathbb{E} \frac{\mathbb{1}\{I_t=i\}}{q_{t,i}} \ell_{t,i} \right| \leq \gamma \cdot \frac{1}{\gamma/n} = \frac{n\gamma}{\gamma} \leq 1$$

if $\gamma = n\gamma$

$$\text{Regret} \leq 2n \zeta T + \frac{\log(n)}{\zeta} + \frac{\zeta n T}{1-n\zeta}$$

$$\leq C \sqrt{n T \log(n)}$$

$$\zeta = \sqrt{\frac{\log(n)}{\zeta n T}} \wedge \frac{1}{2n}$$

FTPL - follow the perturbed leader

$$\sum_{s=1}^t \hat{l}_{s,i}$$

$$I_t = \arg \max_i z_i + \zeta \sum_{s=1}^t \hat{l}_{s,i}$$

Adversarial Linear Bandits

Input arm set $\mathcal{X} \subset \mathbb{R}^d$

Adversary chooses $l_t \in \mathbb{R}^d \quad \forall t=1, \dots, T$ arbitrarily

$$\text{s.t.} \quad \max_{x \in \mathcal{X}} |\langle x, l_t \rangle| \leq 1 \quad \forall t$$

for $t=1, 2, \dots$

Player chooses $x_t \in \mathcal{X}$

gets loss $\langle x_t, l_t \rangle$

$$\text{Regret} \quad \max_{x \in \mathcal{X}} \sum_{t=1}^T \langle x - x_t, l_t \rangle$$

$$\mathcal{X} = (x_1, \dots, x_n)$$

$$I_t \sim \mathcal{Q}_t \quad x_{I_t} = x_t$$

$$\text{For } \underline{x_i} \in \mathcal{X}, \text{ define } \hat{l}_{t,i} = x_i^T A(\mathcal{Q}_t)^{-1} x_i x_t^T l_t$$

$$A(\mathcal{Q}_t) = \sum_{x \in \mathcal{X}} \mathcal{Q}_{t,x} x x^T \equiv \sum_{i=1}^n \mathcal{Q}_{t,i} x_i x_i^T$$

$$\mathbb{E}[\hat{l}_{t,i}] = x_i^T A(\mathcal{Q}_t)^{-1} \underbrace{\mathbb{E}[x_i x_i^T]}_{= A(\mathcal{Q}_t)} l_t$$

$$= x_i^T l_t = l_{t,i}$$

$$\mathbb{E}[\hat{l}_{t,i}^2] = x_i^T A(\mathcal{Q}_t)^{-1} \mathbb{E}[x_i x_i^T \underbrace{(x_i^T l_t)^2}] A(\mathcal{Q}_t)^{-1} x_i$$

$$\leq x_i^T A(q_t)^{-1} x_i$$

Want to minimize $\max_{x \in \mathcal{X}} x^T A(q_t)^{-1} x = \max_{x \in \mathcal{X}} \|x\|_{A(q_t)^{-1}}^2$

Recall $\min_{\lambda \in \Lambda_{\mathcal{X}}} \max_{x \in \mathcal{X}} x^T A(\lambda)^{-1} x = d$ by Kiefer-Wolfowitz.

To apply EXP3(γ) theorem we need to

show $|\mathbb{E} \hat{\ell}_{\epsilon, i}| \leq 1 \quad \forall \epsilon, i$

$$|\mathbb{E} \hat{\ell}_{\epsilon, i}| \leq \mathbb{E} |x_i^T A(q_t)^{-1} x_t| \underbrace{\|x_t\|_{A(q_t)^{-1}}}_{\leq 1}$$

$$q_t = \frac{(1-\gamma)P_\epsilon + \gamma \lambda}{\gamma}$$

G-opt design

$$\leq \mathbb{E} |x_i^T A(q_t)^{-1} x_t|$$

$$\leq \mathbb{E} \|x_i\|_{A(q_t)^{-1}} \cdot \|x_t\|_{A(q_t)^{-1}}$$

$$\leq \frac{\mathbb{E}}{\gamma} \|x_i\|_{A(\lambda)^{-1}} \cdot \|x_t\|_{A(\lambda)^{-1}}$$

$$\leq \frac{\mathbb{E} d}{\gamma} \Rightarrow \gamma = \mathbb{E} d \Rightarrow |\mathbb{E} \hat{\ell}_{\epsilon, i}| \leq 1 \quad \forall \epsilon, i$$

$$\mathbb{E} \left[\sum_{t=1}^T \sum_{i=1}^n P_{\epsilon, i} \hat{\ell}_{\epsilon, i}^2 \right] = \mathbb{E} \left[\sum_t \sum_i P_{\epsilon, i} x_i^T A(q_t)^{-1} x_i x_t^T A(q_t)^{-1} x_t \underbrace{(x_t^T x_t)}_{\leq 1} \right]$$

$$= \sum_{\epsilon} \sum_i p_{\epsilon,i} x_i^T A(q_{\epsilon})^T E[x_i x_i^T] A(q_{\epsilon})^{-1} x_i$$

$$\begin{aligned} \text{Trace}(ABC) \\ &= \text{Tr}(BCA) \\ &= \text{Tr}(CAB) \end{aligned}$$

$$= \sum_{\epsilon} \sum_i p_{\epsilon,i} x_i^T A(q_{\epsilon})^{-1} x_i$$

$$\leq \frac{1}{1-\gamma} \sum_{\epsilon} \sum_i p_{\epsilon,i} x_i^T A(p_{\epsilon})^{-1} x_i$$

$$= \frac{1}{1-\gamma} \sum_{\epsilon} \text{Tr} \left(\underbrace{\sum_i p_{\epsilon,i} x_i x_i^T}_{=A(p_{\epsilon})} A(p_{\epsilon})^{-1} \right)$$

$$= \frac{dT}{1-\gamma}$$

\Rightarrow (By above analogy to MAP)

$$\text{Result} \leq c \sqrt{dT} \log d.$$