# Week 3 Lecture 12

# Full information Regret $R(n, H) = \sup_{h \in \{x,y\}} \left[ \frac{\ddot{z}}{\xi} | \dot{y}_{\xi} - \dot{y}_{\xi}| - \int_{\xi} (h(x_{\xi}) - \dot{y}_{\xi}) \right]$

After prediction, when the information comes we get to know the output for all hypothesis in H.

Eq. (1) holds for any sequence, any hypothesis. The bound for Regret from eq. (1)  $\leq \sqrt{2\log|\mathcal{H}|}$  in If we write eq. (1) as:

$$R(n,H) = Sup Sup \left[ \sum_{t=1}^{n} |\hat{y}_t - y_t| - \sum_{t=1}^{n} |\hat{y}_t - y_t| \right]$$

The bound still holds but the regret now is dependent on the sequence.

Eq 1a can jurther be written as:

$$R(n, H) = \sup_{(x,y)} \left[ \sum_{t=1}^{n} |\hat{y}_t - y_t| - \inf_{t} \sum_{t=1}^{n} |\hat{y}_t - y_t| - \inf_{t=1}^{n} |\hat{y}_t - y_t| - \inf_{t} \sum_{t=1}^{n} |\hat{y}_t - y_t| - \inf_{t} |\hat{y}_t - y_t| - \inf_{t} \sum_{t=1}^{n} |\hat{y}_t - y_t| - \inf_{t} |\hat{y}_t| - \inf_{t=1}^{n} |\hat{y}_t - y_t| - \inf_{t} |\hat{y}_t - y_t| - \inf_{t} |\hat{y}_t| -$$

This was the case for Expert Advice setting with weighted majority.

## Bandit Setting

When we choose an arm, we incur loss, however we get information only about the arm we played.

Ex: Casino with various slot machines where we get to play only one machine at one time point

### Full information ex:

Share market: we get to choose one share to play: but at the end of the day we can get information about all the shares

$$K-actions$$
  $[K] = {1, 2, -... K}$ 

In round t Environment arrings loss vector  $x_t \in [0, 1]^k$ Player selects action  $I_t \in [K]$ Observes loss from  $I_t = X_{tI_t}$  History till round t

$$H_{\xi} = \{I_{1}, \chi_{1I_{1}}, I_{2}, \chi_{2I_{2}}, \dots I_{\xi-1}, \chi_{\xi-1}I_{\xi-1}\}$$

### Policy TT:

Choose an action in round t given the

Expected

Regret would then be:

$$\mathbb{E}\left[R(n,\pi)\right] = \mathbb{E}\left[\sum_{t=1}^{n} \chi_{tI_{t}} - \mathbb{E}\left[\min_{i}\sum_{t=1}^{n} \chi_{ti}\right]\right]$$

because R(n, n) is a random as It is random

Since  $R(n,\pi)$  is a random quantity hence we will bound the  $E(R(n,\pi))$ 

Regret bound:

$$\Sigma\psi$$
  $\mathbb{E}\left[\mathbb{R}(n,\pi)\right]$