# MLT Homework 8

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## Question 1

Let  $\psi(\lambda) = \frac{\lambda^2}{2}$ . The Legendre-Fenchel transform of  $\psi$  is given by

$$\psi^*(\epsilon) = \sup_{\lambda \in \mathbb{R}} \lambda \epsilon - \psi(\lambda).$$

## Subquestion 1.1

$$\psi^*(\epsilon) = \frac{\epsilon^2}{2}$$
.

### Solution

Let define  $\psi_1(\lambda)$ :

$$\psi_1(\lambda) = \lambda \epsilon - \frac{\lambda^2}{2}$$

Furthermore:

$$\psi_1'(\lambda) = \epsilon - \lambda$$

So, the maximum is reached at:

$$\lambda = \epsilon \quad \Rightarrow \quad \psi_1(\epsilon) = \epsilon \cdot \epsilon - \frac{\epsilon^2}{2} = \frac{\epsilon^2}{2}$$

We can conclude:

$$\psi^*(\epsilon) = \psi_1(\epsilon) = \frac{\epsilon^2}{2}$$

### Subquestion 1.2

$$(\psi^*)^{-1}(z) = \pm \sqrt{2z}.$$

#### Solution

From previous point we know:

$$\psi^*(\epsilon) = \frac{\epsilon^2}{2}$$

It follows:

$$z = \frac{\epsilon^2}{2}$$
$$2z = \epsilon^2$$
$$\epsilon = \pm \sqrt{2z}$$

So:

$$(\psi^*)^{-1}(z) = \pm \sqrt{2z}$$

## Question 2

## Question 3

We consider an adversarial bandit model with  $K^2$  arms indexed by  $i \in [K]$  and  $j \in [K]$ . For each arm (i,j), the loss at time t is  $a_t^i + b_t^i$ , where  $a_t^i \in [0,1]$  and  $b_t^i \in [0,1]$  are chosen by the adversary before the start of the interaction. Then each round the learner picks an arm  $(I_t, J_t) \in [K]^2$  and observes  $a_t^{I_t}$  and  $b_t^{J_t}$  separately (and incurs their sum as the loss).

## Subquestion 3.1

Consider running a single instance of EXP3 on all  $K^2$  arms (with loss range [0,2]). Show that the expected pseudo-regret compared to the best arm  $(i^*,j^*)$  is bounded by

$$\bar{R}_n \le 2\sqrt{2nK^2 \ln(K^2)}$$

#### Solution

### Subquestion 3.2

Now we will use the  $a_t^i$  and  $b_t^j$  observations separately. Consider running two K-arm instances of EXP3, one with  $i \to a_t^i$  as the loss and one with  $j \to b_t^j$  as

the loss. Have the first algorithm control  $I_t$  and the second  $J_t$ . Show that the overall expected pseudo-regret is bounded by

$$\bar{R}_n \le 2\sqrt{2nK\ln K}.$$

Solution