Exercise 4: Stochastic bandits with a continuum of arms (can be solved after Course #4)

Consider the setting of stochastic bandits with a continuum of arms indexed by $\mathcal{A} = [0, 1]$. A bandit problem is given by a mean-payoff function f, which we assume to be continuous, thus bounded; for simplicity we consider $f : [0, 1] \rightarrow [0, 1]$. When the player picks arm $X_t \in [0, 1]$ at round t, she gets a payoff $Y_t \in [0, 1]$ drawn at random according to a distribution with expectation $f(X_t)$, conditionally to X_t . The question is to upper bound the regret defined as

$$R_T = T \max_{[0,1]} f - \mathbb{E}\left[\sum_{t=1}^T Y_t\right].$$

Consider the following two-stage methodology. We divide [0, 1] into the $K \ge 2$ regular intervals [(i - 1)/K, i/K], for $i \in \{1, \ldots, K\}$. An auxiliary algorithm picks an interval index $I_t \in \{1, \ldots, K\}$. An arm X_t is then drawn at random within bin I_t and a payoff Y_t is obtained.

We already considered this setting and strategy in an exercise in the lecture notes, with Lipschitz meanpayoff functions and the UCB algorithm.

In the present exercise, we rather assume that the mean-payoff function f is α -Hölder, for some $\alpha > 0$: there exists L > 0 such that for all $x, x' \in [0, 1]$,

$$\left|f(x) - f(x')\right| \leq L|x - x'|^{\alpha}.$$

Also, we rather consider MOSS as the auxiliary algorithm; we recall that its distribution-free regret bound is $K + 45\sqrt{KT}$ against K-tuples of probability distributions over [0, 1].

- 1. Show that for a fixed number of bins $K \ge 2$ and a fixed horizon $T \ge 2$, the regret of the two-stage strategy above is upper bounded by $K + 45\sqrt{KT} + TL/K^{\alpha}$.
- **2.** Explain how to pick K and which regret bound is obtained when T and α are known.
- **3.** What could we do when α is known but T is unknown? Provide the (order of magnitude of the) corresponding regret bound.

Dealing with an unknown α is much more challenging, but was optimally solved by Hédi Hadiji in his PhD thesis, which he defended in December 2020.