

EXERCICES
BLACKWELL'S APPROACHABILITY
 UNIVERSITÉ PARIS–SACLAY



Consider the approachability framework from the course and corresponding notation. Let $\mathcal{C} \subset \mathbb{R}^d$ be a closed convex cone satisfying Blackwell's condition and $\alpha : \mathcal{C}^\circ \rightarrow \mathcal{A}$ an associated oracle such that

$$x' = \lambda x \quad \text{for some } \lambda > 0 \quad \implies \quad \alpha(x') = \alpha(x).$$

The goal is to define two families of parameter-free algorithms for approachability and apply them to regret minimization on the simplex.

Let $\beta > 0$.

1) Let h be a regularizer with domain \mathcal{C}° such that for all $x \in \mathbb{R}^d$ and $\lambda \geq 0$,

$$h(\lambda x) - \min h = \lambda^\beta \left(h(x) - \min h \right).$$

Consider the DA algorithm for approachability associated with regularizer h , constant parameter 1, and oracle α :

$$a_t = \alpha \left(\nabla h^* \left(\sum_{s=1}^{t-1} r_s \right) \right), \quad t \geq 1.$$

Let $\mathcal{X}_0 \subset \mathcal{C}^\circ$ be a nonempty closed set. Let $T \geq 1$.

a) Prove that for all $\lambda > 0$,

$$\max_{x \in \lambda \mathcal{X}_0} \left\langle \sum_{t=1}^T r_t, x \right\rangle \leq \lambda^\beta \left(\max_{\mathcal{X}_0} h - \min h \right) + \sum_{t=1}^T D_{h^*}(y_t + r_t, y_t),$$

where $y_t = \sum_{s=1}^{t-1} r_s$ for all $t \geq 1$.

b) Deduce that

$$\max_{x \in \mathcal{X}_0} \left\langle \sum_{t=1}^T r_t, x \right\rangle \leq 2 \left(\max_{\mathcal{X}_0} h - \min h \right)^{1/\beta} \left(\sum_{t=1}^T D_{h^*}(y_t + r_t, y_t) \right)^{1-1/\beta}.$$

- 2) Let H be a mirror map compatible with \mathcal{C}° . We assume that H admits a minimum on \mathbb{R}^d , that the minimizer x_1 belongs to \mathcal{C}° , and that for all $x \in \mathbb{R}^d$ and $\lambda \geq 0$,

$$H(\lambda x) - \min H = \lambda^\beta \left(H(x) - \min H \right).$$

Consider the OMD algorithm for approachability associated with regularizer H , constant step-size 1, oracle α , and initial action $a_1 = \alpha(x_1)$:

$$x_{t+1} = \arg \max_{x \in \mathcal{C}^\circ} \{ \langle \nabla H(x_t) + r_t, x \rangle - H(x) \} \quad \text{and} \quad a_{t+1} = \alpha(x_{t+1}), \quad t \geq 1.$$

Prove that for all $T \geq 1$,

$$\max_{x \in \mathcal{X}_0} \left\langle \sum_{t=1}^T r_t, x \right\rangle \leq 2 \left(\max_{\mathcal{X}_0} H - \min H \right)^{1/\beta} \left(\sum_{t=1}^T D_{H^*}(\nabla H_t(x_t) + r_t, \nabla H(x_t)) \right)^{1-1/\beta}.$$

- 3) Let $1 < p \leq 2$. Consider algorithms from the above families associated with ℓ_p regularizer on \mathcal{C}° and ℓ_p mirror map on \mathbb{R}^d respectively:

$$h_p = \frac{1}{2} \|\cdot\|_p^2 + \mathbf{I}_{\mathcal{C}^\circ} \quad \text{and} \quad H_p = \frac{1}{2} \|\cdot\|_p^2.$$

Using the fact that h_p and H_p are $(p-1)$ -strongly convex for $\|\cdot\|_p$, derive corresponding guarantees.

- 4) Let $L > 0$. In the context of regret minimization on the simplex, assume that payoff vectors $(u_t)_{t \geq 1}$ are bounded as $\|u_t\|_\infty \leq L$ for all $t \geq 1$. Then derive guarantees for the above algorithms corresponding to ℓ_p regularizer and mirror map. Which value of p minimizes the regret bounds thus obtained?

