

EVALUATION  
**ONLINE LEARNING**  
LINKS WITH OPTIMIZATION AND GAMES  
UNIVERSITÉ PARIS–SACLAY



ADAGRAD FOR APPROACHABILITY AND APPLICATION TO GAMES

Let  $d \geq 1$  and consider an approachability problem given by some actions sets  $\mathcal{A}, \mathcal{B}$  and an outcome function  $g : \mathcal{A} \times \mathcal{B} \rightarrow \mathbb{R}^d$ . Let  $\mathcal{C} \subset \mathbb{R}^d$  be a closed convex cone satisfying Blackwell's condition with respect to  $g$ , and let  $\alpha : \mathcal{C}^\circ \rightarrow \mathcal{A}$  be an associated oracle satisfying:

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

We define an adaptation of AdaGrad-Norm for this approachability problem as follows. Let  $(b_t)_{t \geq 0}$  a sequence in  $\mathcal{B}$ ,  $x_0 = 0$ ,  $a_0 = \alpha(x_0)$  and for  $t \geq 0$ ,

$$x_{t+1} = \Pi_{\mathcal{C}^\circ} \left( x_t + \frac{\gamma}{\sqrt{\sum_{s=0}^t \|r_s\|_2^2}} r_t \right), \quad a_{t+1} = \alpha(x_{t+1}),$$

where  $\gamma > 0$ ,  $\Pi_{\mathcal{C}^\circ}$  denotes the Euclidean projection onto  $\mathcal{C}^\circ$  and for  $t \geq 0$ ,  $r_t = g_t(a_t, b_t)$ .

- 1) Is this algorithm parameter-free?
- 2) Derive guarantees.

- 3) Write the algorithm in the special case of regret minimization on the simplex, and derive guarantees.
- 4) Apply to regret learning for finite two-player zero-sum games and derive guarantees. Perform numerical experiments to compare the convergence of the algorithm with RM, RM+ and exponential weights.
- 5) BONUS. — In numerical experiments, also consider optimistic variants of all algorithms.
- 6) BONUS. — Also consider an adaptation of AdaGrad-Diagonal.

