EVALUATION ONLINE LEARNING LINKS WITH OPTIMIZATION AND GAMES

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COMPOSITE OPTIMIZATION

We consider a convex "composite" optimization problem where the objective function writes $f+\phi$, with $f:\mathbb{R}^d\to\mathbb{R}$ being a convex differentiable function, and $\phi:\mathbb{R}^d\to\mathbb{R}\cup\{+\infty\}$ a proper, convex, lower semicontinuous function. We assume that the proximal operator associated with ϕ :

$$\operatorname{Prox}_{\boldsymbol{\varphi}}(x) = \operatorname*{arg\,min}_{x' \in \mathbb{R}^d} \left\{ \boldsymbol{\varphi}(x') + \frac{1}{2} \left\| x' - x \right\|_2^2 \right\}, \quad x \in \mathbb{R}^d.$$

is easily computable.

EXEMPLE. — If $\phi = \lambda ||x||_1$ for some $\lambda > 0$,

$$\operatorname{Prox}_{\operatorname{\varphi}}(x)_i = egin{cases} x_i - \lambda & \text{if } x_i \geqslant \lambda \ 0 & \text{if } |x_i| \leqslant \lambda \ x_i + \lambda & \text{if } x_i \leqslant -\lambda \end{cases}, \quad 1 \leqslant i \leqslant d, \; x \in \mathbb{R}^d.$$

The most popular algorithm for this setting is the *proximal gradient method*, aka *forward-backward splitting*: let $x_0 \in \mathbb{R}^d$ such that $\partial \phi(x_0) \neq \emptyset$, $(\gamma_t)_{t\geqslant 0}$ a positive sequence, and

$$x_{t+1} = \operatorname{Prox}_{\gamma, \Phi} \left(x_t - \gamma_t \nabla f(x_t) \right), \quad t \geqslant 0.$$

- 1) Prove that the proximal gradient method is an extension of projected gradient descent.
- 2) Prove that the proximal gradient method is an instance of UMD iterates.
- 3) L > 0. In the case where f is L-smooth for $\|\cdot\|_2$, using the tools from the course, establish for the proximal gradient method a convergence guarantee that extends the classical guarantee for projected gradient descent.
- 4) Let $\|\cdot\|$ be an arbitrary norm in \mathbb{R}^d and assume that f is L-smooth for $\|\cdot\|$. Propose for this context an algorithm that extends the proximal gradient method and its convergence guarantee.
- 5) Propose an alternative algorithm for the context of the previous question and derive corresponding guarantees.

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