Labs

**Optimization for Machine Learning** Spring 2021

**EPFL** 

School of Computer and Communication Sciences

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github.com/epfml/OptML\_course

# Problem Set 8, due April 30, 2021 (Frank-Wolfe)

## Convergence of Frank-Wolfe

#### Exercise 1:

Assuming  $h_0 \leq 2C$ , and the sequence  $h_0, h_1, \ldots$  satisfies

$$h_{t+1} \le (1 - \gamma)h_t + \gamma^2 C$$
  $t = 0, 1, \dots$ 

for  $\gamma = \frac{2}{t+2}$ , prove that

$$h_t \le \frac{4C}{t+2} \qquad t = 0, 1, \dots$$

### **Applications of Frank-Wolfe**

#### Exercise 2:

Derive the LMO formulation for matrix completion, that is

$$\min_{Y \in X \subseteq \mathbb{R}^{n \times m}} \sum_{(i,j) \in \Omega} (Z_{ij} - Y_{ij})^2$$

when  $\Omega \subseteq [n] \times [m]$  is the set of observed entries from a given matrix Z.

Where our optimization domain X is the unit ball of the trace norm (or nuclear norm), which is defined the convex hull of the rank-1 matrices

$$X := conv(\mathcal{A}) \quad \text{with} \quad \mathcal{A} := \left\{ \mathbf{u}\mathbf{v}^\top \ \middle| \ \substack{\mathbf{u} \in \mathbb{R}^n, \ \|\mathbf{u}\|_2 = 1 \\ \mathbf{v} \in \mathbb{R}^m, \ \|\mathbf{v}\|_2 = 1} \right\} \ .$$

- 1. Derive the LMO for this set X for a gradient at iterate  $Y \in \mathbb{R}^{n \times m}$ .
- 2. Dervie the *projection* step onto X. How does the computational operations (or costs) needed to compute the LMO and the projection step compare?

## **Practical Implementation**

Follow the Python notebook provided here:

github.com/epfml/OptML\_course/tree/master/labs/ex08/