Convex Optimization for Computer Vision Lecture: M. Möller and T. Möllenhoff Exercises: E. Laude Summer Semester 2016 Computer Vision Group Institut für Informatik Technische Universität München

Weekly Exercises 4

Room: 02.09.023 Friday, 13.5.2016, 09:00-11:00 Submission deadline: Wednesday, 11.5.2016, 14:00, Room 01.09.014

Theory: Projected gradient descent (12+4 Points)

Exercise 1 (4 Points). Let $A \in \mathbb{R}^{n \times n}$ be orthonormal, meaning that $A^{\top}A = AA^{\top} = I$. Let the convex set C be given as

$$C := \left\{ u \in \mathbb{R}^n : \|Au\|_{\infty} \le 1 \right\}.$$

Compute a formula for the projection onto C given as

$$\Pi_C(v) := \arg\min_{u \in \mathbb{R}^n} \|u - v\|_2^2, \quad \text{s.t. } u \in C.$$

Hint: Show that the ℓ_2 -norm of a vector is invariant under a multiplication with an orthonormal matrix A, meaning that $||u||_2 = ||Au||_2$.

Exercise 2 (8 Points). Let $f : \mathbb{R}^n \to \mathbb{R}$ be coercive, differentiable and let ∇f be locally Lipschitz continuous i.e. for each $x \in \mathbb{R}^n$ there exists $\epsilon > 0$ and a constant $L_{\epsilon} > 0$, such that ∇f is L_{ϵ} -Lipschitz on $B_{\epsilon}(x)$.

- 1. Give an example of a function $f : \mathbb{R}^n \to \mathbb{R}$ that meets the above assumptions, but for which ∇f is not Lipschitz continuous. (With a proof).
- Let f be an arbitrary function meeting the above assumptions. Show that for any α ≥ min_x f(x) there exists a constant L_α such that f is L_α-smooth on the sublevel set S_α := {x : f(x) ≤ α}. Hint: You can use the topological definition of compactness: Any subset X ⊂ ℝⁿ is called compact (closed and bounded in the Euclidean case that we consider) if each of its open covers has a finite subcover (see https://en.wikipedia.org/wiki/Compact_space). One possible open cover of X is the union of all ε-balls with any ε: ⋃_{x∈X} B_ε(x).
- 3. Show that $x^+ := x \tau \nabla f(x) \in S_\alpha$ if $\tau < \frac{1}{L_\alpha}$ and $x \in S_\alpha$.
- 4. Conclude that for each initialization x_0 , there exists a τ_{x_0} such that gradient descent with the constant step size τ_{x_0} converges.

Exercise 3 (4 Points). Let C_i , $1 \le i \le n$ be a family of closed convex sets such that

$$\bigcap_{1 \le i \le n} C_i \ne \emptyset.$$

Show that the problem of finding an element u^* in the intersection

$$u^* \in \bigcap_{1 \le i \le n} C_i$$

can be formulated as the following optimization problem:

$$u^* \in \arg\min_{u \in \bigcap_{i \in \mathcal{I}} C_i} \sum_{\substack{j \notin \mathcal{I} \\ 1 \le j \le n}} d^2(u, C_j),$$

where $\mathcal{I} \subseteq \{1, 2, ..., n\}$ can be arbitrary (including the empty set) and d(z, X) is the distance of a point z to the closed convex set X defined as

$$d(z, X) := \min_{x \in X} \|x - z\|_2.$$

Programming: SUDOKU

Exercise 4 (12 Points). Solve the SUDOKUs given in the files exampleSudoku1.mat and exampleSudoku2.mat with projected gradient descent. For that you need to find a point

$$u^* \in \bigcap_{1 \le i \le n+m+1} C_i$$

where the convex sets in the intersection are given as

$$C_{i} := \{ u \in \mathbb{R}^{729} : \langle a_{i}, u \rangle = 1 \}, \quad 1 \le i \le n,$$
$$C_{i} := \{ u \in \mathbb{R}^{729} : u_{j} = 1, \ j \in \mathcal{B} \}, \quad n+1 \le i \le n+m$$

and \mathcal{B} is the set of indexes corresponding to the known numbers and

$$C_{n+m+1} := \{ u \in \mathbb{R}^{729} : u_j \in [0,1], \, \forall \, 1 \le j \le 729 \}.$$

For a more precise definition of the constraint sets see lecture. Solve the programming assignment in the spirit of exercise 3 using the following two partitions of the indexes $\{1, 2, ..., n + m + 1\}$:

- 1. $\mathcal{I}_1 := \{n + m + 1\}$ and
- 2. $\mathcal{I}_2 := \{n+1, n+2, \dots, n+m+1\},\$

and plot the resulting energy decays.

Hint: Show that for the linear constraint sets C_i , $1 \le i \le n$ the distance $d(z, C_i)$ of a point z to the set C_i is equal to $|\langle a_i, z \rangle - 1|$.

(12 Points)