Convex Optimization for Machine Learning and Computer Vision

Lecture: Dr. Tao Wu

Exercises: Yuesong Shen, Zhenzhang Ye

Computer Vision Group
Institut für Informatik

Winter Semester 2018/19 Technische Universität München

Weekly Exercises 4

Room: 01.09.014 Wednesday, 21.11.2018, 12:15-14:00

Submission deadline: Monday, 19.11.2018, 16:15, Room 01.09.014

Convex conjugate

(14+6 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

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

Compute a formula for the projection onto C given as

$$\Pi_C(v) := \operatorname{argmin}_{u \in \mathbb{R}^n} \frac{1}{2} ||u - v||_2^2, \quad \text{s.t. } u \in C.$$

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

Solution. We begin proving the hint:

$$||Ax||_2^2 = \langle Ax, Ax \rangle = \langle A^{\mathsf{T}}Ax, x \rangle = \langle x, x \rangle = ||x||_2^2.$$

The idea is to rewrite the projection onto the set C in terms of the projection $\Pi_{\tilde{C}}$ onto the unit ball of the ℓ_{∞} -norm $\tilde{C} := \{x \in \mathbb{R}^n : ||x||_{\infty} \leq 1\}$. With the substitution

$$w := Au \iff u = A^{\top}w$$

and using the hint we obtain:

$$\begin{split} \Pi_{C}(v) &= \operatorname{argmin}_{\|Au\|_{\infty} \leq 1} \frac{1}{2} \|v - u\|^{2} \\ &= A^{\top} \operatorname{argmin}_{\|w\|_{\infty} \leq 1} \frac{1}{2} \|v - A^{\top}w\|^{2} \\ &= A^{\top} \operatorname{argmin}_{\|w\|_{\infty} \leq 1} \frac{1}{2} \|A(v - A^{\top}w)\|^{2} \\ &= A^{\top} \operatorname{argmin}_{\|w\|_{\infty} \leq 1} \frac{1}{2} \|Av - AA^{\top}w\|^{2} \\ &= A^{\top} \operatorname{argmin}_{\|w\|_{\infty} \leq 1} \frac{1}{2} \|Av - w\|^{2} \\ &= A^{\top} \Pi_{\tilde{C}}(Av). \end{split}$$

Exercise 2 (6 points). Assume $J: \mathbb{R}^n \to \mathbb{R}$, compute the convex conjugate of following functions:

- $J(u) = \frac{1}{q}||u||_q^q = \sum_{i=1}^n \frac{1}{q}|u_i|^q, \ q \in [1, +\infty].$
- $J(u) = \sum_{i=1}^{n} u_i \log u_i + \delta_{\triangle^{n-1}}(u)$.
- $J(u) = \begin{cases} \frac{1}{2} \|u\|_2^2, & \|u\|_2 \le \epsilon \\ +\infty, & \text{otherwise} \end{cases}$

Solution. • $J^*(v) = \sup_u \langle u, v \rangle - J(u)$. Since it is separable, we apply first-order optimality condition elementwisely:

$$\sup_{u_i} \langle u_i, v_i \rangle - \frac{1}{q} (|u_i|)^q \Rightarrow 0 = v_i - |u_i|^{q-1} \operatorname{sign}(u_i) \Rightarrow u_i = |v_i|^{1/(q-1)} \operatorname{sign}(v_i)$$

Substitute u_i back to the first equation, we have

$$J^*(v)_i = |v_i|^{q/(q-1)} - \frac{1}{q} |v_i|^{q/(q-1)}$$
$$= (1 - \frac{1}{q}) |v_i|^{q/(q-1)}$$
$$= (1 - \frac{1}{q}) |v_i|^{1/(1 - \frac{1}{q})}$$

Substituting $\frac{1}{p} = 1 - \frac{1}{q}$, we get $J^*(v) = \frac{1}{p}||v||_p^p$.

• Consider the convex conjugate elementwisely: $J^*(v) = \sup_u \sum_i^n u_i v_i - u_i \log u_i - \delta_{\triangle^{n-1}}(u)$. Let's consider the following minimization problem given v_i :

$$\min_{u} \sum_{i}^{n} u_{i} \log u_{i} - u_{i} v_{i}$$
s.t. $\mathbb{1}u = 1$

where $\mathbb{1} = [1, \dots, 1] \in \mathbb{R}^n$. It is obvious that this two problems share the same optimal variable u^* and the domain of log implies $u_i > 0$. Since the feasible set is compact and original energy function is continuous, the KKT condition holds on u^* . Therefore, we have certain $\lambda \in \mathbb{R}$ such that

$$\log u_i^* + 1 - v_i + \lambda = 0, \ \forall i = 1, \dots, n$$

which give $u_i^* = \exp\{-\lambda + v_i - 1\}$. Additionally, $\sum_{i=1}^n u_i^* = 1$. We can get

$$0 = \log(\sum_{i=1}^{n} \exp\{-\lambda + v_i - 1\}) = \log(\exp\{-\lambda - 1\}) \sum_{i=1}^{n} e^{v_i} = (-\lambda - 1) + \log(\sum_{i=1}^{n} e^{v_i})$$

Now, substitute u^* back into the convex conjugate and we can get

$$J(v)^* = \sum_{i=1}^{n} \exp\{-\lambda + v_i - 1\}v_i - \exp\{-\lambda + v_i - 1\}(-\lambda + v_i - 1)$$
$$= \sum_{i=1}^{n} -\exp\{-\lambda + v_i - 1\}(-\lambda - 1)$$
$$= -(-\lambda - 1) = \log(\sum_{i=1}^{n} e^{v_i})$$

• Rewrite the convex conjugate as $J^*(v) = \sup_{\|u\|_2 \le \epsilon} \langle u, v \rangle - \frac{1}{2} \|u\|_2^2$. We first try to find the corresponding u^* .

$$u^* = \operatorname{argmin}_{\|u\|_2 \le \epsilon} \frac{1}{2} \|u\|_2^2 - \langle u, v \rangle + \frac{1}{2} \|v\|_2^2$$
$$= \operatorname{argmin}_{\|u\|_2 \le \epsilon} \frac{1}{2} \|u - v\|_2^2$$

which is a projection problem i.e. project v into a convex set $\{u: \|u\|_2 \le \epsilon\}$. Therefore, if $\|v\|_2 \le \epsilon$, $u^* = v$. Otherwise, $u^* = \epsilon \frac{v}{\|v\|}$.

$$J^{*}(v) = \begin{cases} \frac{1}{2} \|v\|_{2}^{2}, & \|v\|_{2} \leq \epsilon \\ \epsilon \|v\|_{2}^{2} - \frac{1}{2}\epsilon^{2}, & \text{otherwise} \end{cases}$$

Exercise 3 (4 points). Assume $J: \mathbb{E} \to \mathbb{R}$, prove following facts of convex conjugate:

- $\bullet \ \ \tilde{J}(\cdot) = \alpha J(\cdot) \Rightarrow \tilde{J}^*(\cdot) = \alpha J^*(\cdot/\alpha), \ \alpha > 0.$
- $\tilde{J}(\cdot) = J(\cdot z) \Rightarrow \tilde{J}^*(\cdot) = J^*(\cdot) + \langle \cdot, z \rangle$.

Solution. • Using the definition of convex conjugate:

$$\tilde{J}(\cdot) = \sup_{u} \langle u, \cdot \rangle - \tilde{J}(u)$$

$$= \sup_{u} \langle u, \cdot \rangle - \alpha J(u)$$

$$= \alpha \sup_{u} \langle u, \cdot /\alpha \rangle - J(u)$$

$$= \alpha J^*(\cdot /\alpha)$$

• $\tilde{J}^*(\cdot) = \sup_u \langle u, \cdot \rangle - J(u-z)$. Define v = u - z and by substitution we have:

$$\begin{split} \tilde{J}^*(\cdot) &= \sup_{v} \langle v + z, \cdot \rangle + J(v) \\ &= \sup_{v} \langle v, \cdot \rangle + J(v) + \langle z, \cdot \rangle \\ &= J^*(\cdot) + \langle \cdot, z \rangle \end{split}$$

Exercise 4 (6 points). Show that projection onto a convex set is Lipschitz continuous with constant equals 1, i.e.

$$||\Pi_C(u) - \Pi_C(v)|| \le ||u - v||, \ \forall u, v \in \mathbb{E}$$

where C is a convex set.

Solution. Given a point u, recall the property of projection:

$$\langle u - \Pi_C(u), x - \Pi_C(u) \rangle \le 0, \ \forall x \in C.$$

Since $\Pi_C(v)$ is also an element in C, we get:

$$\langle u - \Pi_C(u), \Pi_C(v) - \Pi_C(u) \rangle \le 0.$$

As same as above, we can get the ineuqlity for point v:

$$\langle v - \Pi_C(v), \Pi_C(u) - \Pi_C(v) \rangle \le 0.$$

Sum above inequalities up, we have:

$$\langle u - \Pi_{C}(u) + \Pi_{C}(v) - v, \Pi_{C}(v) - \Pi_{C}(u) \rangle \leq 0$$

$$\Rightarrow \langle \Pi_{C}(v) - \Pi_{C}(u), \Pi_{C}(v) - \Pi_{C}(u) \rangle \leq \langle v - u, \Pi_{C}(v) - \Pi_{C}(u) \rangle$$

$$\Rightarrow \|\Pi_{C}(v) - \Pi_{C}(u)\|^{2} \leq \|v - u\| \|\Pi_{C}(v) - \Pi_{C}(u)\|$$

$$\Rightarrow \|\Pi_{C}(v) - \Pi_{C}(u)\| \leq \|v - u\|$$