

## 2 Projecting onto the L1-ball [Kevin, 25 points]

In this problem we will devise an efficient algorithm for projecting a vector  $y \in \mathbb{R}^n$  onto the unit  $L1$ -ball. That is, to solve the optimization problem

$$x^* = \arg \min_x \|x - y\|_2^2 / 2, \quad \text{subject to:} \quad (1)$$

$$\|x\|_1 \leq 1. \quad (2)$$

- (a) [2 points] Write out the Lagrangian,  $L(x, \lambda)$ , where  $\lambda \in \mathbb{R}$ .

This problem is convex and strictly feasible, so by Slater's condition strong duality holds and therefore the KKT conditions are necessary. Convexity implies the KKT conditions are sufficient.

- (b) [4 points] Write out the KKT conditions that an optimal primal/dual pair,  $(x^*, \lambda^*)$ , must satisfy.
- (c) [3 points] Show that if  $\|y\|_1 \leq 1$  then  $x^* = y, \lambda^* = 0$  satisfy the KKT conditions.

Assume from this point that  $\|y\|_1 > 1$ . Otherwise, by (c), we know  $x^*$ .

- (d) [4 points] Prove by contradiction and the KKT conditions that  $\lambda^* > 0$  and  $\|x^*\|_1 = 1$ .

For any  $\lambda \geq 0$ , let us define  $x(\lambda) = S_\lambda(y)$ . That is,

$$x_i(\lambda) = \begin{cases} y_i - \lambda & \text{if } y_i \geq \lambda \\ y_i + \lambda & \text{if } y_i \leq -\lambda \\ 0 & \text{otherwise} \end{cases} \quad (3)$$

- (e) [5 points] Prove that  $(x(\lambda), \lambda)$  satisfy the stationarity condition (*i.e.*, first-order optimality). That is,  $(x(\lambda), \lambda)$  satisfy all the KKT conditions, but perhaps not primal feasibility.

Note:  $(x(\lambda), \lambda)$  satisfy the stationarity condition uniquely.

Let  $f(\lambda) = \|x(\lambda)\|_1$ . We've now shown that if we find a  $\lambda^* \geq 0$  such that  $f(\lambda^*) = 1$ , then  $x(\lambda^*)$  satisfies the KKT conditions and solves our problem.

- (f) [2 points] Show that  $f(0) > 1$  and there is a  $\lambda^+$  such that  $f(\lambda^+) = 0$ .

It is apparent that  $f(\lambda)$  is continuous (it is also decreasing and convex). Therefore, by the intermediate value theorem, we have that  $\lambda^*$  exists. Now let's quickly compute  $\lambda^*$  exactly.

It is not hard to see that  $\lambda^*$  is invariant under permutations of the elements of  $y$ , or changes to their signs. Let's consider the vector  $\tilde{y}$ , which has the same elements as  $|y|$ , sorted in descending order. That is,  $\tilde{y}_1 \geq \tilde{y}_2 \geq \dots \geq \tilde{y}_n \geq 0$ . Given an index  $k \in \{1, 2, \dots, n-1\}$ , we know that choosing  $\lambda \in [\tilde{y}_k, \tilde{y}_{k+1}]$  will make all  $\tilde{x}_j(\lambda) = 0$ , for  $j \in \{k+1, k+2, \dots, n\}$ .

- (g) [5 points] Using no more than  $\Sigma_k = \sum_{i=1}^k \tilde{y}_i$ , and the values  $\tilde{y}_k, \tilde{y}_{k+1}$  give a simple formula for  $\lambda^* \in [\tilde{y}_k, \tilde{y}_{k+1}]$  if it exists, or a test that concludes  $\lambda^*$  is not in that range.

That is, to project  $\|y\|_1 > 1$  onto the  $L1$ -ball, we start by sorting,  $\tilde{y} = \text{sort}(|y|)$ , then find  $\lambda^*$  by testing each interval with part (h), and finish by computing  $x^* = S_{\lambda^*}(y)$  by soft-thresholding. This all takes only  $O(n \log n)$  time.

(By the way, a similar procedure can be derived for projecting onto the probability simplex—another frequently-encountered set.)