Nonlinear Optimization (18799 B, PP) IST-CMU PhD course Spring 2018 Instructor: João Xavier (jxavier@isr.ist.utl.pt) TA: Hung Tuan (hung.seadc@gmail.com)

The homework is due April 19.

## Homework 2

Instructions: read section 3.2 of [1].

**Problem A.** (Bhattacharyya distance) Let  $\Delta_n = \{p \in \mathbf{R}^n : p \ge 0, \mathbf{1}^T p = 1\}$  be the probability simplex of size n. That is, a point in  $\Delta_n$  is a vector  $p = (p_1, p_2, \dots, p_n)$  that can represent a probability mass function over an alphabet of size n:  $p_i \ge 0$  for all i, and  $p_1 + p_2 + \dots + p_n = 1$ . The Phottacharyya distance between two points p and q in  $\Delta_n$  is defined as

The Bhattacharyya distance between two points p and q in  $\Delta_n$  is defined as

BC 
$$(p,q) = -\log\left(\sum_{i=1}^{n} \sqrt{p_i}\sqrt{q_i}\right)$$

Let q be a fixed point in  $\Delta_n$ . Show that the function  $f : \Delta_n \to \mathbf{R}, f(p) = \mathrm{BC}(p,q)$  is convex.

- **Problem B.** (*VaR and CVaR*) In financial risk management, it's important to measure how risky is a given loss vector  $L = (L_1, L_2, \ldots, L_n)$ , where  $L_i$  is the loss we suffer when scenario *i* happens.
  - (a) A popular risk measure is the value-at-risk (VaR). Assume scenario *i* happens with probability  $\pi_i$ . The VaR of a given loss vector  $L = (L_1, L_2, \ldots, L_n)$  is defined as

$$\operatorname{VaR}(L) = \min\left\{l : \sum_{i=1}^{n} \pi_i \mathbb{1}_{L_i \le l} \ge \alpha\right\}$$

where  $\alpha$  is a fixed confidence level (usually,  $\alpha = 0.9$  or  $\alpha = 0.99$ ). The symbol  $1_{a \leq b}$  means 1 if  $a \leq b$ , and 0 if a > b.

In words,  $\operatorname{VaR}(L)$  is the least number l for which the probability of a loss happening in the interval [0, l] is greater than or equal to  $\alpha$ . For example, take n = 5,  $(\pi_1, \pi_2, \pi_3, \pi_4, \pi_5) = (0.1, 0.4, 0.2, 0.25, 0.05)$ , and  $\alpha = 0.9$ . Then  $\operatorname{VaR}(2, 5, 3, 2, 9) = 5$ and  $\operatorname{VaR}(2, 9, 3, 2, 5) = 9$ .

Is VaR a convex function?

(b) Another popular risk measure is the conditional value-at-risk (CVaR). The CVaR of a given loss vector  $L = (L_1, L_2, \dots, L_n)$  is defined as

CVaR(L) = min 
$$\left\{ l + \frac{1}{1-\alpha} \sum_{i=1}^{n} \pi_i (L_i - l)_+ : l \in \mathbf{R} \right\},\$$

where  $a_{+} = \max\{a, 0\}$ . Is CVaR a convex function?

## References

[1] S. Boyd and L. Vandenberghe. *Convex optimization*. Cambridge University Press, 2004.