Data Driven Robust Optimization Exam 22/03/2020

The exam is made of two independant parts. If necessary, you can admit the results of previous questions. Send your work (french or english) in pdf by tuesday 8:00 by mail to vincent.leclere@enpc.fr. Please make it easily readable (bonus point for latex). Contact me in case of difficulty.

Some usefull recalls.

- 1. An SOCP constraint take the form $a^T x + b + ||c^T x + d||_2 \le 0$.
- 2. We have

$$\sup_{\mathbb{Q}\in\mathbb{B}_{\varepsilon}(\hat{\mathbb{P}}_{N})} \mathbb{E}_{\mathbb{Q}}[\ell(\boldsymbol{\xi})] = \inf \lambda\varepsilon + \frac{1}{N}\sum_{i=1}^{N}s_{i}$$

s.t. $[-\ell_{k}]^{\star}(z_{ik} - \nu_{ik}) + \delta^{\star}(\nu_{ik}, \Xi) - \langle z_{ik}, \hat{\xi}_{i} \rangle \leq s_{i} \quad \forall i \in [N], \forall k \in [K]$
 $\|z_{ik}\|_{\star} \leq \lambda$

where

- $\hat{\mathbb{P}}_N$ is the uniform measure on $\{\hat{\xi}_i\}_{i \in [N]}$, which are iid datapoints sampled from $\boldsymbol{\xi}$, such that $\operatorname{supp}(\boldsymbol{\xi}) \subset \Xi$;
- $\mathbb{B}_{\varepsilon}(\hat{\mathbb{P}}_N)$ is the Wasserstein ball of radius ε , for norm $\|\cdot\|$, centered in $\hat{\mathbb{P}}_N$;
- $\ell := \max_{k \in [K]} \ell_k$, where each $-\ell_k$ is a proper convex function
- 3. A random variable $\boldsymbol{\xi} \in \mathbb{R}^m$, is light-tailed under \mathbb{P} if there exists a > 1 such that, $\mathbb{E}_{\mathbb{P}}[\|\boldsymbol{\xi}\|^a] < +\infty$.
- 4. If $\boldsymbol{\xi} \in \mathbb{R}^m$, $m \geq 3$, is light-tailed under \mathbb{P} , then we have, for $\varepsilon < 1$,

$$\mathbb{P}_S(d_W(\mathbb{P}, \hat{\mathbb{P}}_N) \ge \varepsilon) \le c_1 e^{-c_2 N \varepsilon^m}$$

1) Data-Driven SOCP constraint

Consider the following problem :

$$\min_{x \in \mathbb{R}^n} \quad q^\top x \tag{1}$$

s.t.
$$\|Ax + b\|_2 \leq (c + \Delta_c u)^\top x + d \tag{2}$$

$$\|x\|_{\infty} \le 1 \tag{3}$$

where $u \in \mathbb{R}^m$ and $\Delta_c \in M_{n,m}(\mathbb{R})$.

- 1. (2 points) Let $f(u, x) := || Ax + b ||_2 (c + \Delta_c u)^\top x d \le 0$. Compute $f_{\star}(v, x) := \inf_{u \in \mathbb{R}^m} v^\top u f(u, x)$.
- 2. (3 points) We assume that $u \in \mathbb{R}$ (i.e m = 1), and that we have 100 independent realizations of \tilde{u} . Leveraging the CS test write an SOCP problem, with the same objective function (see section 8.1 of DDRO), whose solution satisfies (2) with probability 95% with confidence 90% in the data.
- 3. (2 points) We now assume that $\tilde{u} \ge 0$. Refine your previous SOCP using this assumption. What can you say about the relation between both SOCP ?

2) Chance constraint

Consider a random vector $\boldsymbol{\xi}$, of unknown law, taking values in \mathbb{R}^n , $n \geq 3$. We assume that we have an iid sample $\{\hat{\xi}_i\}_{i \in [N]}$. We assume that $\operatorname{supp}(\boldsymbol{\xi}) \subset \Xi$, with $\Xi = \{\boldsymbol{\xi} \in \mathbb{R}^m \mid C\boldsymbol{\xi} \leq d\}$ a polytope (bounded polyhedron).

Let $A = \{\xi \in \mathbb{R}^m \mid a_k^\top \xi < b_k, \forall k \in [K]\}$, such that $A \cap C \neq \emptyset$. We are interested in the following problem

$$\bar{p} := \sup_{\mathbb{Q} \in \mathbb{B}_{\varepsilon}(\hat{\mathbb{P}}_N)} \quad \mathbb{Q}(\boldsymbol{\xi} \notin A)$$

Define $\ell_k(\xi) = \begin{cases} 1 & \text{if } a_k^\top \xi \ge b_k \\ -\infty & \text{otherwise.} \end{cases}$

- 1. (2 points) Show that $\mathbb{Q}(\boldsymbol{\xi} \in A) = \mathbb{E}_{\mathbb{Q}}(\ell(\boldsymbol{\xi}))$, for a function ℓ to be defined.
- 2. (1 point) Compute $[-\ell_k]^*(z)$
- 3. (1 point) Compute $\delta^{\star}(z|\Xi)$
- 4. (2 points) Give a convex minimization problem with a finite number of constraints whose value is equal to \bar{p} .
- 5. (2 points) Specify the above problem in the case where the Wasserstein norm chosen is $\|\cdot\| = \|\cdot\|_{\infty}$. What is the simplest class of optimization problem it belongs to ?
- 6. (2 points) Same question for $\|\cdot\| = \|\cdot\|_2$. Is it numerically simpler, more complex or impossible to say, than the previous one ?
- 7. (4 points) We are now interested in the following chance constrained problem

$$(CCP) \quad \min_{x \in \mathbb{R}^n} \quad f(x)$$

s.t.
$$\mathbb{P}(a_k^\top \xi \le x_k, \ \forall k \in [K]) \ge 1 - \varepsilon$$

 \mathbb{P} is unknown, but we have iid realisations of $\boldsymbol{\xi}$ under \mathbb{P} . Give a robust formulation of (*CCP*) implying a probabilistic guarantee of level $1 - \varepsilon$ with confidence $1 - \alpha$.