

Exercise session notes - Week 10

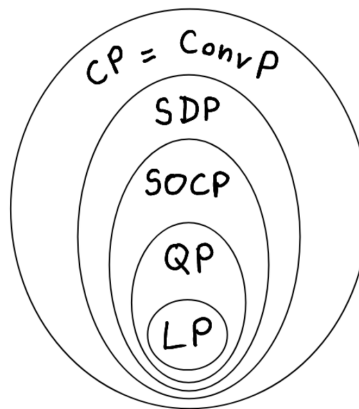
Second Order Cone Program + Robust Optimization

1 Convex Programs and Inclusions

Firstly we reviewed the types of convex programs that we studied so far.

$$\begin{array}{llll}
 \text{LP:} & \min f^\top x & \text{s.t. } Ax \leq b & ; \quad Fx = g \\
 \text{QP:} & \min f^\top x & \text{s.t. } x^\top Ax + b^\top x \leq c & ; \quad Fx = g \\
 \text{SOCP:} & \min f^\top x & \text{s.t. } \|Ax + b\|_2 \leq c^\top x + d & ; \quad Fx = g \\
 \text{SDP:} & \dots & & \\
 \text{CP:} & \min f^\top x & \text{s.t. } x \in K & ; \quad Fx = g \\
 \text{ConvP:} & \min f(x) & \text{s.t. } x \in C \text{ convex set} & ; \quad Fx = g
 \end{array}$$

We then pictured the diagram of all the inclusions between these sets.



The inclusions $\text{LP} \subseteq \text{QP}$, $\text{SOCP} \subseteq \text{CP}$ and $\text{CP} \subseteq \text{ConvP}$ are trivial. Then we proved $\text{ConvP} \subseteq \text{CP}$ and $\text{QP} \subseteq \text{SOCP}$, and next week we will consider also the Semi Definite Programs (SDP).

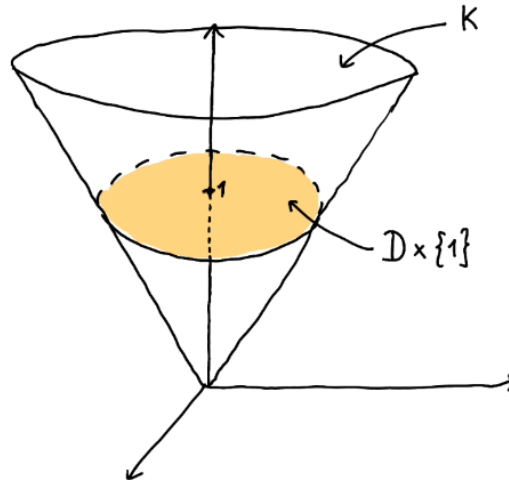
ConvP \subseteq CP. Consider a general convex program

$$\min f(x) \quad \text{s.t. } x \in C \text{ convex set} ; \quad Fx = g$$

Then we can write an equivalent program as

$$\begin{aligned}
 & \min t \quad \text{s.t. } t \geq f(x) ; x \in C \text{ convex set} ; \quad Fx = g \\
 \iff & \min t \quad \text{s.t. } (x, t) \in \text{epi}(f) ; (x, t) \in C \times \mathbb{R} ; \quad Fx = g
 \end{aligned}$$

Let $D := \text{epi}(f) \cap (C \times \mathbb{R})$ and note that this is a convex set. Then we pick the conic hull $K = \text{cone}(D \times \{1\})$ which is



Therefore $(x, t) \in D \iff (x, t, 1) \in E$. Finally we can write the equivalent conic program as

$$\min t \quad \text{s.t.} \quad (x, t, s) \in K ; \quad s = 1 ; \quad Fx = g$$

QP \subseteq SOCP. Consider a general quadratic program with A being a positive definite matrix

$$\min f^\top x \quad \text{s.t.} \quad x^\top Ax + b^\top x + c \leq 0 ; \quad Fx = g$$

Recall that by the Cholesky Decomposition there exists a matrix B with $A = B^\top B$. Then we can write an equivalent program as

$$\begin{aligned} \min f^\top x \quad \text{s.t.} \quad & x^\top B^\top Bx \leq -b^\top x - c ; \quad Fx = g \\ \min f^\top x \quad \text{s.t.} \quad & \|Bx\|_2^2 \leq -b^\top x - c ; \quad Fx = g \end{aligned}$$

Then recall the definition of the rotated Lorentz cone:

$$\mathcal{L}_{rot}^m = \left\{ (x, y, z) \mid \|x\|_2^2 \leq 2yz \right\}$$

Therefore we can model the program as the following SOCP

$$\min f^\top x \quad \text{s.t.} \quad (Bx, -b^\top x - c, \frac{1}{2}) \in \mathcal{L}_{rot}^m ; \quad Fx = g$$

2 Robust Optimization

Assume we are working for an airline company and we want to assign people to flights. We model this problem as an LP. In a perfect world we can write

$$\max \# \text{ people served} \quad \text{s.t.} \quad \text{assignment is feasible}$$

Since flight disruptions may happen (flight cancellations, delays) we want to find an assignment that works for any possible scenario. Such a solution is called a Robust Solution. A non-example would be:

Rome \rightarrow Zurich (wait 15 min)
Zurich \rightarrow London

This solution is clearly feasible but a small delay could change entirely our assignment, therefore it is not Robust. Let's formalize this problem (Ex. 6.6 from Lecture Notes). We want to solve an LP

$$\min \quad c^\top x \quad \text{s.t.} \quad a^\top x \leq b, \quad x \geq 0$$

but the vector $a \in \mathbb{R}^n$ is uncertain, it varies in some Ellipse

$$a \in E = \{a_0 + Pu \mid \|u\|_2 \leq 1\}$$

for a_0 the center and P the matrix corresponding to the stretch of the Ellipse. Our goal is to find a Robust solution, so a point x that is feasible for any choice of a and with minimal value. Therefore we may require

$$\begin{aligned} b &\geq \sup_{a \in E} a^\top x \\ &= a_0^\top x + \sup_{\|u\| \leq 1} u^\top P^\top x \\ &= a_0^\top x + \|P^\top x\|_2 \end{aligned}$$

Therefore we can model the robust LP as the following SOCP

$$\min \quad c^\top x \quad \text{s.t.} \quad a_0^\top x + \|P^\top x\|_2 \leq b, \quad x \geq 0$$