

# Intersection of Convex Sets

Lee, Sidford, and Wong 2015, pp. 60–70

Andreas Haupt

June 27, 2016, Graduate Seminar on Discrete Optimization, University of Bonn

## 1 The Intersection Problem

### Conventions

- **bold** lowercase letters are column vectors in  $\mathbb{R}^n$
- uppercase **bold** letters are in  $\mathbb{R}^{n \times n}$
- $K, K_1, K_2 \subseteq \mathbb{R}^n$  convex.
- $\|\cdot\|$  is the  $\ell^2$ -norm.

**Definition 1.1** (Oracles).  $\text{EO}_f$ : Input  $\mathbf{x}$ , Output  $f(\mathbf{x})$

$\text{OO}_\varepsilon(\mathbf{K})$ : Input  $\mathbf{c}$ , Output  $\max_{\mathbf{x}} \mathbf{c}^T \mathbf{x}$

$\text{SO}_{\varepsilon, \delta}(f)$ : Input  $\mathbf{x}$ , Output: Assertion  $f(\mathbf{x}) \leq \min_{\mathbf{y}} f(\mathbf{y}) + \eta$  or  $\mathbf{c} \neq 0$ ,  $b$  with  $b \leq \delta \|\mathbf{c}\|$  and  $\{\mathbf{z} \mid f(\mathbf{z}) \leq f(\mathbf{x})\} \subseteq \{\mathbf{z} \mid \mathbf{c}^T \mathbf{z} \leq \mathbf{c}^T \mathbf{x} + b\}$

$\text{SGO}_\delta$ : Input  $\mathbf{x}$ , Output:  $\mathbf{y} \in \{g \in \Omega \mid f(\mathbf{y}) + \delta \geq f(\mathbf{x}) + \mathbf{g}^T(\mathbf{y} - \mathbf{x}), \forall \mathbf{y} \in \Omega\}$ .

**Definition 1.2** (Strong Concavity).  $f: K \rightarrow \mathbb{R}$  is  $\alpha$ -strongly concave iff  $f(\mathbf{x}) + \alpha \|\mathbf{x}\|$  is concave.

**Lemma 1.3.** If  $f$  is  $\alpha$ -strongly concave with minimizer  $\mathbf{x}^*$ ,  $f(\mathbf{y}) \leq f(\mathbf{x}^*) + \varepsilon$ , then  $\frac{1}{2}\alpha \|\mathbf{x}^* - \mathbf{y}^*\|^2 \leq \varepsilon$ ,  $\forall \varepsilon > 0$ . Strongly concave functions have unique maximizers.

**Theorem 1.4** (Our Cutting-Plane Method). Let  $f: \mathbb{R}^n \rightarrow \mathbb{R}$ ,  $\alpha \in (0, 1)$ ,  $\Omega \subseteq B_\infty(\mathbf{0}, R)$  convex containing a minimizer of  $f$ . Then we can compute  $\mathbf{x} \in \mathbb{R}^n$  with

$$f(\mathbf{x}) - \min_{\mathbf{y} \in \Omega} f(\mathbf{y}) \leq \eta + \alpha (\max_{\mathbf{y} \in \Omega} f(\mathbf{y}) - \min_{\mathbf{y} \in \Omega} f(\mathbf{y}))$$

in  $O(n \text{SO}_{\eta, \delta}(f) \log(\frac{n\kappa}{\alpha}) + n^3 \log^{O(1)}(\frac{n\kappa}{\alpha}))$ ,  $\delta \in \Theta(\frac{\alpha \text{MinWidth}(\Omega)}{n^{\frac{3}{2}} \ln(\kappa)})$ ,  $\kappa = \frac{R}{\text{MinWidth}(\Omega)}$ .

### MATROID INTERSECTION PROBLEM

Instance: Matroids  $(E, \mathcal{I}_1)$ ,  $(E, \mathcal{I}_2)$  given via independence (or rank) oracles of complexity  $\mathcal{T}_{\text{ind}}$  (or  $\mathcal{T}_{\text{rank}}$ ),  $\mathbf{w} \in \mathbb{R}^E$ ,  $\|\mathbf{w}\|_\infty \leq M$ .

Task: Find  $S \in \arg \min_{S \in \mathcal{I}_1 \cap \mathcal{I}_2} w(S)$ .

## 2 Solving the Intersection Problem

### Assumptions

$$K_1, K_2 \subseteq B_2(0, M), M \geq 1, \quad \|\mathbf{c}\|_2 \leq M, \quad K_1 \cap K_2 \neq \emptyset \quad (\text{A})$$

### Regularize the Problem

$$f_\lambda(\mathbf{x}, \mathbf{y}) := \frac{1}{2} \mathbf{c}^T \mathbf{x} + \frac{1}{2} \mathbf{c}^T \mathbf{y} - \frac{\lambda}{2} \|\mathbf{x} - \mathbf{y}\|^2 - \frac{1}{2\lambda} \|\mathbf{x}\|^2 - \frac{1}{2\lambda} \|\mathbf{y}\|^2$$

**Lemma 2.1.** Assuming (A),  $f_\lambda$  has a unique maximizer  $(\mathbf{x}_\lambda, \mathbf{y}_\lambda)$  on  $K_1 \times K_2$  such that  $\|\mathbf{x}_\lambda - \mathbf{y}_\lambda\|^2 \leq \frac{6M^2}{\lambda}$  and

$$\max_{\mathbf{x} \in K_1 \cap K_2} \mathbf{c}^T \mathbf{x} \leq f_\lambda(\mathbf{x}_\lambda, \mathbf{y}_\lambda) + \frac{M^2}{\lambda}.$$

**Lemma 2.2** (Sion 1958, Corollary of 3.3). Let  $f: X \times Y \rightarrow \mathbb{R}$ ,  $(x, y) \mapsto f(x, y)$  continuous, convex in  $x$ , concave in  $y$ ,  $X$  or  $Y$  compact. Then

$$\sup_{x \in X} \inf_{y \in Y} f(\mathbf{x}, \mathbf{y}) = \inf_{y \in Y} \sup_{x \in X} f(\mathbf{x}, \mathbf{y}).$$

### Max-Min Problem

$$\begin{aligned} \Omega &:= \{(\boldsymbol{\theta}_1, \boldsymbol{\theta}_2, \boldsymbol{\theta}_3) \mid \|\boldsymbol{\theta}_1\| \leq 2M, \|\boldsymbol{\theta}_2\|, \|\boldsymbol{\theta}_3\| \leq M\} \\ g_\lambda(\mathbf{x}, \mathbf{y}, \boldsymbol{\theta}) &:= \left(\frac{\mathbf{c}}{2} + \lambda \boldsymbol{\theta}_1 + \frac{1}{\lambda} \boldsymbol{\theta}_2\right)^T \mathbf{x} \\ &\quad + \left(\frac{\mathbf{c}}{2} - \lambda \boldsymbol{\theta}_1 + \frac{1}{\lambda} \boldsymbol{\theta}_3\right)^T \mathbf{y} \\ &\quad + \frac{\lambda}{2} \|\boldsymbol{\theta}_1\|^2 + \frac{1}{2\lambda} (\|\boldsymbol{\theta}_2\|^2 + \|\boldsymbol{\theta}_3\|^2) \\ h_\lambda(\boldsymbol{\theta}) &:= \max_{(\mathbf{x}, \mathbf{y}) \in K_1 \times K_2} g_\lambda(\mathbf{x}, \mathbf{y}, \boldsymbol{\theta}), \boldsymbol{\theta} = (\boldsymbol{\theta}_1, \boldsymbol{\theta}_2, \boldsymbol{\theta}_3) \end{aligned}$$

### Getting an Approximately Optimal Solution

**Lemma 2.3.** Assuming (A),  $\lambda \geq 2$ ,

$$f_\lambda(\mathbf{x}, \mathbf{y}) = \min_{\boldsymbol{\theta} \in \Omega} g_\lambda(\mathbf{x}, \mathbf{y}, \boldsymbol{\theta}).$$

If

$$h_\lambda(\theta') \leq \min_{\theta \in \Omega} h_\lambda(\theta) + \varepsilon,$$

then  $\mathbf{z} = -\frac{1}{2}(\theta'_2 + \theta'_3)$  satisfies

$$\max_{\mathbf{x} \in K_1 \cap K_2} \mathbf{c}^T \mathbf{x} \leq \mathbf{c}^T \mathbf{z} + \frac{20M^2}{\lambda} + 20\lambda^3 \varepsilon$$

and  $\|\mathbf{z} - \mathbf{x}_\lambda\| + \|\mathbf{z} - \mathbf{y}_\lambda\| \leq 4\sqrt{2\lambda\varepsilon} + \sqrt{\frac{6M^2}{\lambda}}$ ,  $(\mathbf{x}_\lambda, \mathbf{y}_\lambda) \in \arg \max_{(\mathbf{x}, \mathbf{y}) \in K_1 \times K_2} f_\lambda(\mathbf{x}, \mathbf{y})$ .

### Getting a Separation Oracle for $h_\lambda$

**Lemma 2.4.**  $\text{SO}_{O(\sqrt{\varepsilon\lambda D}), O(\sqrt{\varepsilon\lambda D})}(h_\lambda)$  on  $\{\theta \mid \|\theta\| \leq D\}$  has complexity  $O(\text{OO}_\varepsilon(K_1) + \text{OO}_\varepsilon(K_2))$

### Solving the Intersection Problem

**Theorem 2.5** (Main Theorem). *Assuming (A). Then for any  $0 < \delta < 1$ , we can find  $\mathbf{z}$ ,  $d(\mathbf{z}, K_1) + d(\mathbf{z}, K_2) \leq \delta$  such that*

$$\max_{\mathbf{x} \in K_1 \cap K_2} \mathbf{c}^T \mathbf{x} \leq \mathbf{c}^T \mathbf{z} + \delta$$

in time

$$O(n(\text{OO}_\eta(K_1) + \text{OO}_\eta(K_2)) \log(\frac{nM}{\delta}) + n^3 \log^{O(1)}(\frac{nM}{\delta})),$$

$$\eta \in \Omega((\frac{\delta}{nM})^{O(1)}).$$

**Lemma 2.6** (Klivans and Spielman 2001, Lemma 4). *Let  $C$  be a set of linear forms in variables  $z_1, \dots, z_\ell$  with coefficients in  $[K]$ . For  $z_1, \dots, z_\ell \sim \text{Unif}_{[\frac{K\varepsilon}{\varepsilon}]}$ , with probability greater than  $1 - \varepsilon$ , there is a unique form of minimal value at  $z_1, \dots, z_\ell$ .*

- Note  $\min_{\mathbf{x} \in P} \mathbf{z}^T \mathbf{x} = \min_{\mathbf{x} \in P} \mathbf{x}^T \mathbf{z} = \min_{\mathbf{x} \in P} x(\mathbf{z})$ , where  $x \in \mathcal{L}(\mathbb{R}^n), \mathbf{z} \mapsto \mathbf{x}^T \mathbf{z}$ .
- This is the reason why algorithms in the applications give guarantees only with a certain probability.

### Unique Solutions and an Error Estimate

**Lemma 2.7** (Uniqueness Lemma). *Let  $P := \{\mathbf{x} \mid \mathbf{A}\mathbf{x} \geq \mathbf{b}\} \subseteq B_M^\infty(\mathbf{0})$  be integral,  $\mathbf{A} \in \mathbb{Z}^{m \times n}$ ,  $\mathbf{b} \in \mathbb{Z}^n$ ,  $\mathbf{c} \in \mathbb{Z}^n$ .*

- Then we can find  $\mathbf{z} \in \mathbb{Z}^n, \|\mathbf{z}\|_\infty \leq 100n^2M^2\|\mathbf{c}\|_\infty + 10nM$  such that with probability at least  $\frac{9}{10}$ ,  $\min_{\mathbf{x} \in P} \mathbf{z}^T \mathbf{x}$  has a unique minimizer  $\mathbf{x}^*$  and  $\mathbf{x}^* \in \arg \min_{\mathbf{x} \in P} \mathbf{c}^T \mathbf{x}$ .
- If  $\mathbf{y} \in P, \mathbf{z}^T \mathbf{y} \leq \min_{\mathbf{x} \in P} \mathbf{z}^T \mathbf{x} + \delta$ , then  $\|\mathbf{y} - \mathbf{x}^*\| \leq 2nM\delta$ .

## 3 Applications

### Weighted Matroid Intersection

**Theorem 3.1.** **WEIGHTED MATROID INTERSECTION** with  $\mathbf{w} \in \mathbb{Z}^n, \|\mathbf{w}\|_\infty \leq M$  can be solved in

$$O(n \text{GO} \log(nM) + n^3 \log^{O(1)}(nM))$$

with probability at least  $\frac{9}{10}$ , where  $\text{GO}$  is the complexity of optimizing  $\min_{S \in \mathcal{I}_i} w(S)$ .

For expensive  $\text{GO}$  and/or large  $r$  speedup.

### Submodular Flow

#### SUBMODULAR FLOW PROBLEM

Instance:  $(G, c)$  weighted digraph,  $l, u: V(G) \rightarrow \mathbb{R}$ ,  $f: 2^{V(G)} \rightarrow \mathbb{R}, f(\emptyset) = f(V) = 0$   
 Task: Find a flow  $\varphi: E(G) \rightarrow \mathbb{R}, l(e) \leq \varphi(e) \leq u(e), \forall e \in E(G), \sum_{v \in S} \text{ex}_\varphi(v) \leq f(S)$ .

LP formulation ( $\mathbf{A}$  incidence matrix of  $G$ )

$$\max_{\mathbf{x} \in P} \mathbf{c}^T \mathbf{x}, P := \{\varphi \mid \mathbf{l} \leq \varphi \leq \mathbf{u}, \mathbf{x} = \mathbf{A}\varphi, \mathbf{x}(S) \leq f(S), \forall S \subseteq V\}$$

Consider set  $P_\varepsilon$  of solution of value at most  $\text{OPT} + \varepsilon$ .

$$P_\varepsilon = \{\mathbf{x} \mid \exists \varphi: \mathbf{l} \leq \varphi \leq \mathbf{u}, \mathbf{x} = \mathbf{A}\varphi\} \\ \cap \{\mathbf{x} \mid \mathbf{x}(S) \leq f(S), \forall S \subseteq V, \mathbf{x}(V) = f(V)\}$$

### Submodular Flow

**Theorem 3.2.** The SUBMODULAR FLOW PROBLEM can be solved in

$$O(n^2 \text{EO}_f \log(mCU) \log(n) + n^3 \log^{O(1)}(mCU))$$

with probability at least  $\frac{9}{10}$ .

Previously best:  $\tilde{O}(n^6 \text{EO}_f + n^7)$  Fleischer, Iwata, and McCormick 2002 and  $O(mn^5 \log(nU) \text{EO})$  Fleischer and Iwata 2000

## References

- Fleischer, Lisa and Satoru Iwata (2000). “Improved algorithms for submodular function minimization and submodular flow”. In: *Proceedings of the thirty-second annual ACM symposium on Theory of computing*. ACM, pp. 107–116.
- Fleischer, Lisa, Satoru Iwata, and S Thomas McCormick (2002). “A faster capacity scaling algorithm for minimum cost submodular flow”. In: *Mathematical Programming* 92.1, pp. 119–139.
- Klivans, Adam R. and Daniel Spielman (2001). “Randomness Efficient Identity Testing of Multivariate Polynomials”. In: *Proceedings of the Thirty-third Annual ACM Symposium on Theory of Computing*. STOC ’01. Hersonissos, Greece: ACM, pp. 216–223. ISBN: 1-58113-349-9. DOI: 10.1145/380752.380801. URL: <http://doi.acm.org/10.1145/380752.380801>.
- Lee, Yin Tat, Aaron Sidford, and Sam Chiu-wai Wong (2015). “A Faster Cutting Plane Method and its Implications for Combinatorial and Convex Optimization”. In: *CoRR* abs/1508.04874. URL: <http://arxiv.org/abs/1508.04874>.
- Sion, Maurice (1958). “On general minimax theorems.” In: *Pacific J. Math.* 8.1, pp. 171–176. URL: <http://projecteuclid.org/euclid.pjm/1103040253>.