SUMMER SCHOOL ON ASPECTS OF OPTIMIZATION Discrete Optimization September 15th, 2022

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### Tailored cutting-plane methods

- Introduction
- Strong inequalities
- Polytope Dimension
- Facet and Convex Hull Proofs
- Lifting
- Separation procedure

## 2 Research Talk

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### 1 Tailored cutting-plane methods

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• standard vs. tailored solving method

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- Potentially enormous number of inequalities vs. only a relevant subset

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#### A priori vs. on the fly addition

- standard vs. tailored solving method
- Potentially enormous number of inequalities vs. only a relevant subset

#### **Decomposition** :

 $X = X^1 \cap X^2 \rightarrow$  focus on a simpler set, e.g.  $X^2$  because inequalities valid for  $X^2$  will be valid for X.

• Definition of strong inequalities

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- Polytope dimension and how to find it

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- Polytope dimension and how to find it
- Face, facets, extreme points

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# Strong valid inequalities

#### Remark

Inequalities  $\pi^{\top}x \leq \pi^{0}$  and  $\lambda \pi^{\top}x \leq \lambda \pi^{0}$  are identical for any  $\lambda > 0$ .

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#### Definition

If  $\pi^{\top}x \leq \pi^{0}$  and  $\mu^{\top}x \leq \mu^{0}$  are two valid inequalities for  $X \subseteq \mathbb{R}^{n}_{+}$ ,  $\pi^{\top}x \leq \pi^{0}$  dominates  $\mu^{\top}x \leq \mu^{0}$  if there exists u > 0 such that  $\pi > u\mu$  and  $\pi^{0} \leq u\mu^{0}$ .

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#### Definition

An inequality  $\pi^{\top}x \leq \pi^{0}$  is redundant in the description of  $X \subseteq \mathbb{R}^{n}_{+}$ , if there exist  $k \geq 2$  valid inequalities  $(\pi^{i})^{\top}x \leq \pi^{i0}$  for i = 1, ..., k for X and weights  $u_{i} > 0$  for i = 1, ..., k such that  $(\sum_{i=1}^{k} u_{i}\pi^{i})^{\top}x \leq \sum_{i=1}^{k} u_{i}\pi^{i0}$ dominates  $\pi^{\top}x \leq \pi^{0}$ .

#### • Checking redundancy may be very difficult

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- Checking redundancy may be very difficult
- **Theoretically**: it is important to know which inequalities are nonredundant

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- **Theoretically**: it is important to know which inequalities are nonredundant
- **Practically** : it is important to avoid using an inequality when one that dominates it is available

Classical formulation

$$K = \left\{ x \in \{0,1\}^n \mid \sum_{j=1}^n w_j x_j \le c \right\}$$

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Cover formulation

$$\mathcal{K}^{\mathcal{C}} = \left\{ x \in \{0,1\}^n \mid \sum_{j \in \mathcal{C}} x_j \leq |\mathcal{C}| - 1 \; \forall \text{ minimal cover } \mathcal{C} \text{ for } \mathcal{K} \right\}$$

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#### Proposition

The sets K and  $K^C$  coincide.

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#### Proposition

The sets K and  $K^C$  coincide.

#### Which formulation is stronger?

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$$n = 3, w = (3, 3, 3), c = 5$$
, thus  
 $K = \{x \in \{0, 1\}^n \mid 3x_1 + 3x_2 + 3x_3 \le 5\}$   
 $K^C = \{x \in \{0, 1\}^n \mid x_i + x_j \le 1 \ \forall i, j = 1, 2, 3; i \ne j\}$ 

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Summing up the 3 inequalities of  $K^C$  we have  $2x_1 + 2x_2 + 2x_3 \le 3$ 

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Summing up the 3 inequalities of  $K^{c}$  we have  $2x_1 + 2x_2 + 2x_3 \le 3$  which implies  $3x_1 + 3x_2 + 3x_3 \le 5$ ,

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Summing up the 3 inequalities of  $K^C$  we have  $2x_1 + 2x_2 + 2x_3 \le 3$  which implies  $3x_1 + 3x_2 + 3x_3 \le 5$ , thus  $K^C$  is stronger than K (e.g.  $(1, \frac{2}{3}, 0)$ ).

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Summing up the 3 inequalities of  $K^C$  we have  $2x_1 + 2x_2 + 2x_3 \le 3$  which is implied by  $x_1 + x_2 + x_3 \le 1$ , thus K is stronger than  $K^C$  (e.g.  $(\frac{1}{2}, \frac{1}{2}, \frac{1}{2})$ ).

• Can we improve  $K^C$ ?  $\rightarrow$  strong valid inequalities

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- $K^C$  has an exponential number of constraints  $\rightarrow$  no a-priori addition
- Use "some" constraints of  $K^C$  to improve  $K \rightarrow$  separation procedure

Aim: how to (try to) identify the best possible cuts?

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#### Theorem

If P is a full-dimensional polyhedron, it has a unique minimal description

$$P = \{x \in \mathbb{R}^n \mid (a^i)^\top x \le b_i \text{ for } i = 1, \dots, m\}$$

where each inequality is unique within a positive multiple.

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where each inequality is unique within a positive multiple.

Thus, any other valid inequality is redundant.

# Polyhedra, Face, Facets

### Definition

The points  $x^1, x^2, \ldots, x^k \in \mathbb{R}^n$  are affinely independent if the only solution to the linear system

$$\sum_{j=1}^{n} \lambda_j x^j = 0$$
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#### Definition

The dimension of P, dim(P), is one less than the maximum number of affinely independent points in P.

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• *F* defines a face of the polyhedron *P* if  $F = \{x \in P \mid \pi^{\top}x = \pi^{0}\}$  for some valid inequality  $\pi^{\top}x \leq \pi^{0}$  of *P*. The valid inequality  $\pi^{\top}x \leq \pi^{0}$  is said to represent or define a face.

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- Alternatively, F is a face of the polyhedron P if there is a vector c for which F is the set of vectors attaining max{c<sup>⊤</sup>x | x ∈ P} provided that this maximum is finite.

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- Alternatively, F is a face of the polyhedron P if there is a vector c for which F is the set of vectors attaining max{c<sup>⊤</sup>x | x ∈ P} provided that this maximum is finite.
- F is a facet of P if F is a face of P of  $\dim(F) = \dim(P) 1$ .

An extreme point of P is a 0-dimensional face of P.

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### Proposition

If *P* is full-dimensional, a valid inequality  $\pi^{\top} x \leq \pi^{0}$  is necessary in the description of *P* if and only if it defines a facet of *P*.

An extreme point of P is a 0-dimensional face of P.

### Proposition

If *P* is full-dimensional, a valid inequality  $\pi^{\top} x \leq \pi^{0}$  is necessary in the description of *P* if and only if it defines a facet of *P*.

Thus, for full-dimensional polyhedra,  $\pi^{\top} x \leq \pi^0$  defines a facet of *P* if and only if there are *n* affinely independent points of *P* satisfying it at equality.

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# Polytope dimension

How to prove that the set S has dimension k?

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#### Method 1:

• Find a system Ax = b of equations such that  $S \subseteq \{x \in \mathbb{R}^n \mid Ax = b\}$ and rank(A) = n - k

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#### Method 2:

- Find a system Ax = b of equations such that  $S \subseteq \{x \in \mathbb{R}^n \mid Ax = b\}$ and rank(A) = n - k
- Prove that every equation α<sup>T</sup>x = β satisfied by all x ∈ S is a linear combination of Ax = b (i.e., there exists u s.t. α = uA and β = ub). Thus, Ax = b is the affine hull of S, thus S has dimension k.

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Example: 01-KP

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Example: 01-KP

$$S = \text{conv}(\{x \in \{0,1\}^n \mid w^{\top}x \le c\})$$

#### Proposition

The dimension of S is n - |J|, with  $J = \{j \mid w_j > c \text{ for } j = 1, \dots n\}$ .

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• Given  $X \in \mathbb{Z}_+^n$  and a valid inequality  $\pi^\top x \le \pi^0$  for X, how to prove that  $\pi^\top x \le \pi^0$  is a facet of conv(X)?

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#### Approach 1:

• Find *n* points in *X* that satisfy  $\pi^{\top}x = \pi^{0}$  and prove they are affinely independent.

## Facet and Convex Hull Proofs

Hp. conv(X) is bounded and full-dimensional.

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# Facet and Convex Hull Proofs

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#### Approach 2:

• Select  $t \ge n$  points  $x^1, \ldots, x^t \in X$  satisfying  $\pi^\top x = \pi^0$ . Suppose that all these points lie on a generic hyperplane  $\mu^\top x = \mu^0$ .

• Solve the linear equation system

$$\sum_{j=1}^n \mu_j x_j^k = \mu_0 \quad \text{ for } k = 1, \dots, t$$

in the n+1 unknowns  $(\mu, \mu^0)$ 

• If the only solution is  $(\mu, \mu^0) = \alpha(\pi, \pi^0)$  for  $\alpha \neq 0$ , then the inequality  $\pi^\top x = \pi^0$  is facet-defining.

• Given  $X \in \mathbb{Z}_+^n$  and a polyhedron  $P = \{x \in \mathbb{R}^n \mid Ax \le b\}$ , show P describes the conv(X).

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#### Approach 1:

• Show that matrix A have a special structure guaranteeing  $P = \operatorname{conv}(X)$ , like TUM.

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• Show that matrix A have a special structure guaranteeing  $P = \operatorname{conv}(X)$ , like TUM.

### Approach 2:

 Show that points x<sup>\*</sup> ∈ P with a fractional element are not extreme points of P.

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#### Approach 3:

• Show that for all  $c \in \mathbb{R}^n$  the linear program min $\{c^\top x \mid Ax \leq b\}$  has an optimal solution  $x^* \in X$ .

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#### Approach 4:

Show that if π<sup>⊤</sup>x ≤ π<sup>0</sup> defines a facet of conv(X) then it must be identical to one of the inequalities a<sup>⊤</sup><sub>i</sub>x ≤ b<sub>i</sub> defining P.

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#### Approach 5:

Show that for any c ∈ ℝ<sup>n</sup>, c ≠ 0 the set of optimal solutions to the problem min{c<sup>T</sup>x | x ∈ X} lies in {x | a<sub>i</sub><sup>T</sup>x = b<sub>i</sub>} for some i = 1,..., m.
Hp. conv(X) is bounded and full-dimensional.

• Given  $X \in \mathbb{Z}_+^n$  and a polyhedron  $P = \{x \in \mathbb{R}^n \mid Ax \le b\}$ , show P describes the conv(X).

#### Approach 5:

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Etc.

Consider the decomposition  $X = X^1 \cap X^2 \rightarrow$  focus on a simpler set, e.g.  $X^2$  because inequalities valid for  $X^2$  will be valid for X.

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Computational results to show method effectiveness.

# Strengthening Cover Inequalities: Extended Covers

Yesterday we have seen the Extended Cover Inequalities for the 01-KP problem.

#### Proposition

If C is a cover for X, the feasible set of the 01-KP problem, the extended cover inequality

$$\sum_{\in E(C)} x_j \le |C| - 1$$

is valid for X, where  $E(C) = C \cup \{j \mid w_j \ge w_i \text{ for all } i \in C\}$ .

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• The (Extended) Cover inequality can be strengthened by lifting

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#### Proposition

Let C be a cover for K. The cover inequality associated with C is facet-defining for  $P_C = \operatorname{conv}(K) \cap \{x \in \mathbb{R}^n \mid x_j = 0 \ \forall j \in \{1, \ldots, n\} \setminus C\}$  if and only if C is a minimal cover.

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Given a cover C for the 01-KP problem, find the best possible values for  $\alpha_j$  for  $j \in \{1, ..., n\} \setminus C$  such that

$$\sum_{j\in C} x_j + \sum_{j\in\{1,\ldots,n\}\setminus C} \alpha_j x_j \le |C| - 1$$

is valid for X.

Given a cover *C* for the 01-KP problem, find the best possible values for  $\alpha_j$  for  $j \in \{1, \ldots, n\} \setminus C$  such that  $\sum_{j \in C} x_j + \sum_{j \in \{1, \ldots, n\} \setminus C} \alpha_j x_j \leq |C| - 1$  is valid for *X*.

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Iteratively solve the problem

$$\xi_t = \max \sum_{i=1}^{t-1} \alpha_{j_i} x_{j_i} + \sum_{j \in C} x_j$$
$$\sum_{i=1}^{t-1} \alpha_{j_i} x_{j_i} + \sum_{j \in C} w_j x_j \le b - w_{j_t}$$
$$x \in \{0, 1\}^{|C|+t-1}$$

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The resulting Lift Cover Inequalities are facet-defining

Generalization

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### Generalization

#### Proposition

Consider a set  $S \subseteq \{0,1\}^n$  such that  $S \cap \{x \mid x_n = 1\} \neq \emptyset$  and let  $\sum_{j=1}^{n-1} \alpha_j x_j \leq \beta$  be a valid inequality for  $S \cap \{x \mid x_n = 0\}$ . Then

$$\alpha_n = \beta - \max\left\{\sum_{j=1}^{n=1} \alpha_j x_j \mid x \in S, x_n = 1\right\}$$

is the largest coefficient such that  $\sum_{j=1}^{n-1} \alpha_j x_j + \alpha_n x_n \leq \beta$  is valid for S. Furthermore, if  $\sum_{j=1}^{n-1} \alpha_j x_j \leq \beta$  defines a d-dimensional face of  $\operatorname{conv}(S) \cap \{x_n = 0\}$ , then  $\sum_{j=1}^n \alpha_j x_j \leq \beta$  defines a face of  $\operatorname{conv}(S)$  of dimension at least d + 1.

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Potentially as difficult as solving the 01-KP problem  $\rightarrow$  heuristics!

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