# Secondary simplex method for 2-Stage Stochastic Linear Problem

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SMAL MODE





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- Linear programming and polyhedral geometry
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  - Chamber complex
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$$\min_{x \in \mathbb{R}^n} c^{\top} x$$
s.t.  $Ax \leqslant b$ 

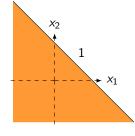
$${\sf A}=\left(egin{array}{ccc}1&&1\\&&\end{array}
ight)\,b=\left(egin{array}{ccc}1\\&&\end{array}
ight)$$

$$x_1 + x_2 \leqslant 1$$









$$\min_{x \in \mathbb{R}^n} c^{\top} x$$
s.t.  $Ax \leqslant b$ 

$$A = \begin{pmatrix} 1 & 1 \\ 1 & -1 \\ & & \end{pmatrix} b = \begin{pmatrix} 1 \\ 1 \\ & \end{pmatrix}$$

$$b = \begin{pmatrix} 1 \\ 1 \\ & \end{pmatrix}$$

$$(1)$$

$$x_1 + x_2 \leqslant 1$$

$$(2)$$

$$(3)$$

$$(4)$$

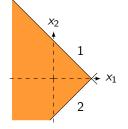
$$(5)$$

$$(6)$$

$$x_1+x_2\leqslant 1$$

$$-x_2\leqslant 1 \qquad (2)$$





$$\min_{x \in \mathbb{R}^n} c^\top x$$
  
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$$A = \begin{pmatrix} 1 & 1 \\ 1 & -1 \\ -1 & -1 \end{pmatrix} b = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} \begin{cases} x_1 + x_2 \leqslant 1 & (1) \\ x_1 - x_2 \leqslant 1 & (2) \\ -x_1 - x_2 \leqslant 1 & (3) \\ (4) & (5) \\ (6) & 3 & 2 \end{cases}$$

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$$A = \begin{pmatrix} 1 & 1 \\ 1 & -1 \\ -1 & -1 \\ -1 & 1 \end{pmatrix} b = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} \begin{cases} x_1 + x_2 \leqslant 1 & (1) \\ x_1 - x_2 \leqslant 1 & (2) \\ -x_1 - x_2 \leqslant 1 & (3) \\ -x_1 + x_2 \leqslant 1 & (4) \end{cases}$$

$$(5)$$

$$(6)$$

$$(7)$$

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$$A = \begin{pmatrix} 1 & 1 \\ 1 & -1 \\ -1 & -1 \\ -1 & 1 \\ 1 & 0 \end{pmatrix} b = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 0.5 \end{pmatrix} \begin{pmatrix} x_1 + x_2 \leqslant 1 & (1) \\ x_1 - x_2 \leqslant 1 & (2) \\ -x_1 - x_2 \leqslant 1 & (3) \\ -x_1 + x_2 \leqslant 1 & (4) \\ x_1 \leqslant 0.5 & (5) \end{pmatrix} \begin{pmatrix} x_2 \\ -x_1 - x_2 \leqslant 1 & (2) \\ -x_1 - x_2 \leqslant 1 & (3) \\ -x_1 + x_2 \leqslant 1 & (4) \\ x_1 \leqslant 0.5 & (5) \end{pmatrix}$$

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$$A = \begin{pmatrix} 1 & 1 \\ 1 & -1 \\ -1 & -1 \\ 1 & 0 \\ 0 & 1 \end{pmatrix} b = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 0.5 \\ 0.5 \end{pmatrix} \begin{pmatrix} x_1 + x_2 \leqslant 1 & (1) \\ x_1 - x_2 \leqslant 1 & (2) & x_2 \\ -x_1 - x_2 \leqslant 1 & (3) & 6 \\ -x_1 + x_2 \leqslant 1 & (4) & 4 \\ x_1 \leqslant 0.5 & (5) & x_2 \leqslant 0.5 & (6) \\ x_2 \leqslant 0.5 & (6) & 3 \end{pmatrix} \begin{pmatrix} x_1 + x_2 \leqslant 1 & (1) \\ x_2 \leqslant 0.5 & (6) & 3 \end{pmatrix}$$

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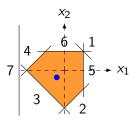
$$A = \begin{pmatrix} 1 & 1 \\ 1 & -1 \\ -1 & -1 \\ -1 & 1 \\ 1 & 0 \\ 0 & 1 \\ -1 & 0 \end{pmatrix} b = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 0.5 \\ 0.5 \\ -1.2 \end{pmatrix} \begin{cases} x_1 + x_2 \leqslant 1 & (1) \\ x_1 - x_2 \leqslant 1 & (2) \\ -x_1 - x_2 \leqslant 1 & (3) \\ -x_1 + x_2 \leqslant 1 & (4) \\ x_1 \leqslant 0.5 & (5) & 7 \\ x_2 \leqslant 0.5 & (6) \\ x_1 \geqslant -1.2 & (7) \end{cases}$$

#### **Definition**

We denote by  $\mathcal{I}(A,b)$ , the collection of sets of active constraints as :

$$\mathcal{I}(A,b) = \{I_{A,b}(x) \mid Ax \leqslant b\}$$

with 
$$I_{A,b}(x) := \{i \in [q] \mid A_i x = b_i\}$$



$$P = \{x \in \mathbb{R}^n \mid Ax \leqslant b\}$$

$$I_{A,b}(\mathbf{x}) = \emptyset$$

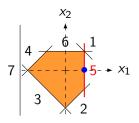
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$$I_{A,b}(x) = \{5\}$$

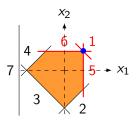
$$\mathcal{I}(A, b) = \{\emptyset, 5,$$

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$$I_{A,b}(x) = \{1,5,6\}$$

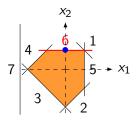
$$\mathcal{I}(A,b) = \{\emptyset, 5, 156,$$

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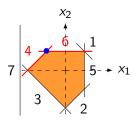
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$$P = \{x \in \mathbb{R}^n \mid Ax \leqslant b\}$$

$$I_{A,b}(x) = \{4,6\}$$

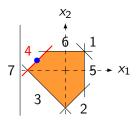
$$\mathcal{I}(A, b) = \{\emptyset, 5, 156, 6, 46, \}$$

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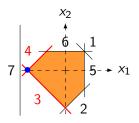
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$$I_{A,b}(x) = \{3,4\}$$

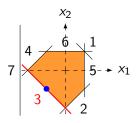
$$\mathcal{I}(A, b) = \{\emptyset, 5, 156, 6, 46, 4, 34, \}$$

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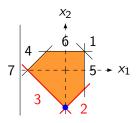
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$$I_{A,b}(x) = \{2,3\}$$

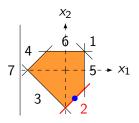
$$\mathcal{I}(A, b) = \{\emptyset, 5, 156, 6, 46, 4, 34, 3, 23, \}$$

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$$I_{A,b}(x) = \{2\}$$

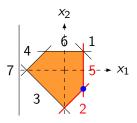
$$\mathcal{I}(A,b) = \{\emptyset, 5, 156, 6, 46, 4, 34, 3, 23, 2, \dots\}$$

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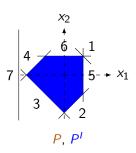
$$\mathcal{I}(A,b) = \{\emptyset, 5, 156, 6, 46, 4, 34, 3, 23, 2, 25\}$$

#### Definition

Let  $I \in \mathcal{I}(A, b)$ , we denote by  $P^I$  the face of P such that:

$$P^I = \{x \in P \mid A_I x = b_I\}$$

We have  $\dim(P^I) = n - \operatorname{rg}(A_I)$ Example for  $I = \emptyset$ 

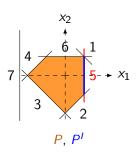


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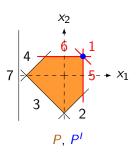


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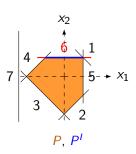


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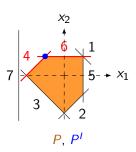


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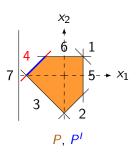


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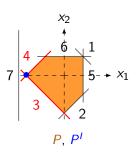


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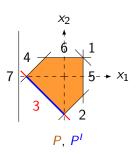


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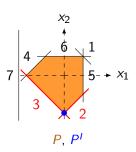


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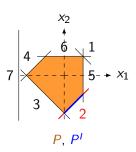


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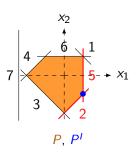


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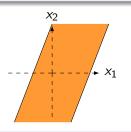
We have  $dim(P^I) = n - rg(A_I)$ Example for  $I = \{2, 5\}$ 



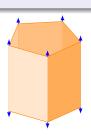
# Lineality space, vertices and bases

# Definition (Lineality space)

$$Lin(C) := \{ u \in C \mid \forall t \in \mathbb{R}, \ \forall x \in C, \ x + tu \in C \}.$$



If
$$P = \{x \in \mathbb{R}^n | Ax \leq b\},$$
then Lin(P) = Ker(A)



# Definition (Bases and vertices)

A basis B is a subset of [p] such that  $A_B = (A_{i,j})_{i \in B, 1 \le j \le n}$  is invertible. A vertex of P is a face of dimension 0. Vert(P) is the set of vertices.

 $Vert(P) \neq \emptyset \Leftrightarrow A \text{ admits at least one basis } \Leftrightarrow rg(A) = n \Leftrightarrow Lin(P) = \{0\}$ 

We make this assumption without loss of generality.

Geometrically: follow a path on the polyhedron from pivoting from basis to basis vertex to vertex

 $x_2$ 

$$\mathit{B}_1 = \{1,5\}$$

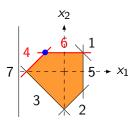
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$$B_1 = \{1, 5\}$$

$$B_2=\{1,6\}$$

Geometrically: follow a path on the polyhedron from pivoting from basis to basis vertex to vertex

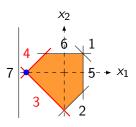


$$B_1=\{1,5\}$$

$$B_2 = \{1, 6\}$$
  
 $B_3 = \{4, 6\}$ 

$$B_3 = \{4, 6\}$$

Geometrically: follow a path on the polyhedron from pivoting from basis to basis vertex to vertex

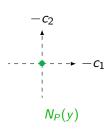


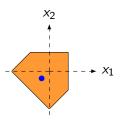
$$B_1 = \{1, 5\}$$
  
 $B_2 = \{1, 6\}$   
 $B_3 = \{4, 6\}$   
 $B_2 = \{3, 4\}$ 

#### Definition

The normal fan of the polyhedron P is

$$\mathcal{N}(P) := \{ N_P(x) \, | \, x \in P \}$$



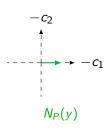


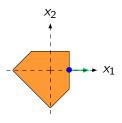
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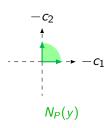


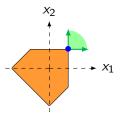
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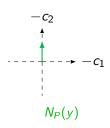


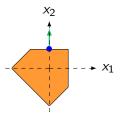
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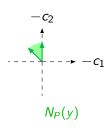


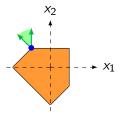
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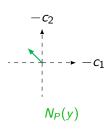


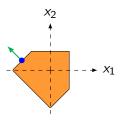
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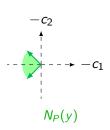


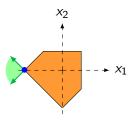
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The normal fan of the polyhedron P is

$$\mathcal{N}(P) := \{ N_P(x) \, | \, x \in P \}$$



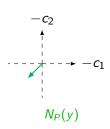


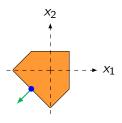
 $P \times \text{and } N_P(x)$ 

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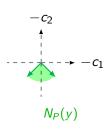


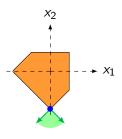
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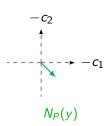


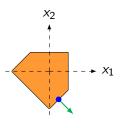
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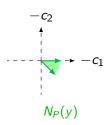


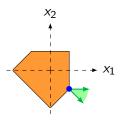
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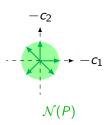
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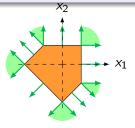
$$\mathcal{N}(P) := \{ N_P(x) \mid x \in P \}$$

with  $N_P(x) = \{c \mid \forall x' \in P, \ c^\top(x'-x) \leqslant 0\}$  the normal cone of P on x.

### Proposition

 $\{ri(N) | N \in \mathcal{N}(P)\}$  is a partition of supp  $\mathcal{N}(P)$  (=  $\mathbb{R}^m$  if P is bounded).





P and  $\mathcal{N}(P)$ 

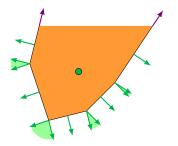
#### Definition (Recession cone)

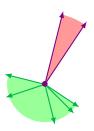
$$rc(C) := \{u \in C \mid \forall t \in \mathbb{R}_+, \ \forall x \in C, \ x + tu \in C\}.$$

Let  $P = \{x \mid Ax \leq b\}$ 

$$\mathsf{rc}(P) = \{u \,|\, Au \leqslant 0\}$$

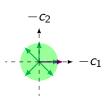
$$-\infty < \begin{cases} \inf_{x \in \mathbb{R}^n} & c^\top x \\ \text{s.t.} & Ax \leqslant b \end{cases} \iff -c \in \operatorname{rc}(P)^* = \operatorname{Cone}(A^\top) = \operatorname{supp}\left(\mathcal{N}(P)\right)$$



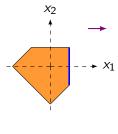


For any  $N \in \mathcal{N}(P)$  and  $-c \to \arg\min_{x \in P} c^{\top}x$  is constant for all  $-c \in ri(N)$ .

 $\arg\min_{x\in P} c^{\top}x$  is a face of P.

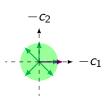


Cost -c and  $\mathcal{N}(P)$ 

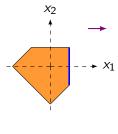


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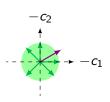
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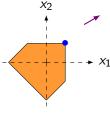
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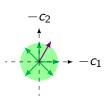
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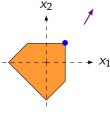
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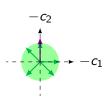
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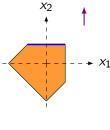
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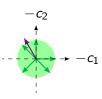
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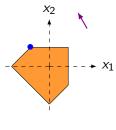
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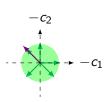
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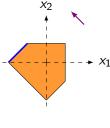
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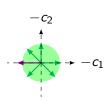


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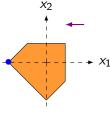


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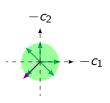
Cost -c and  $\mathcal{N}(P)$ 



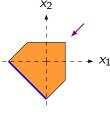
D

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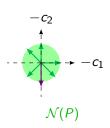


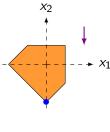
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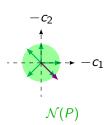
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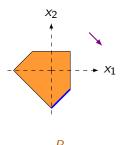
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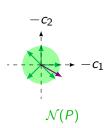


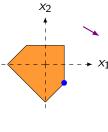
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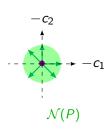


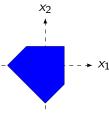
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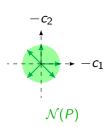


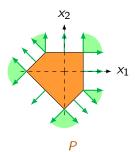
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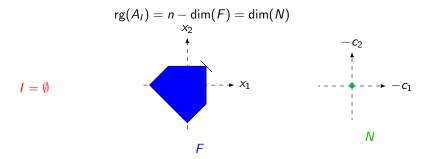
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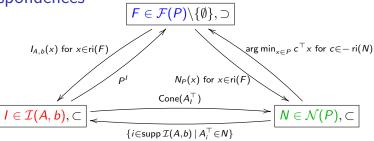


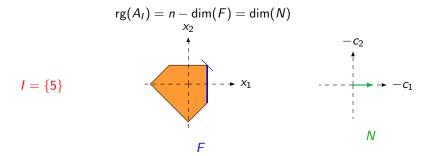


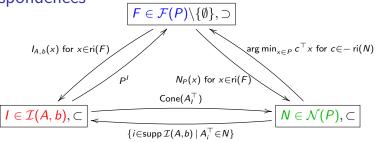


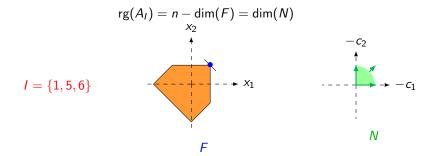


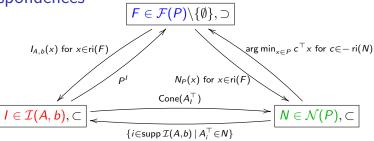


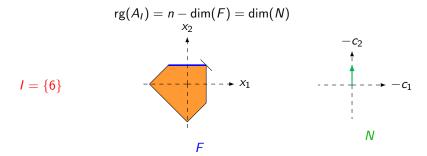


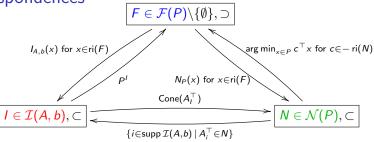


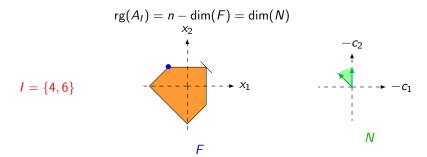


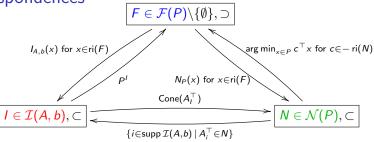


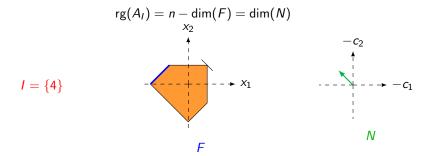


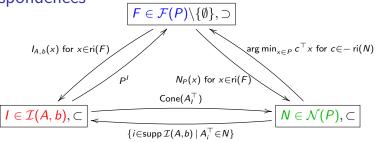


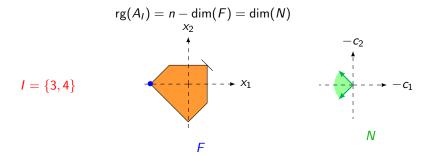


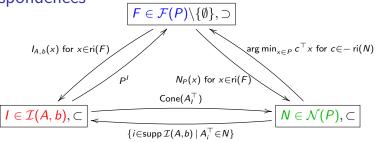


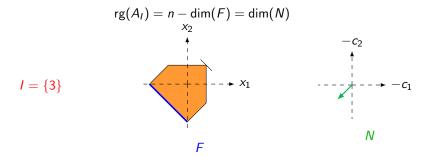


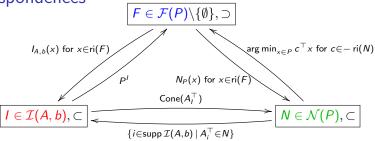


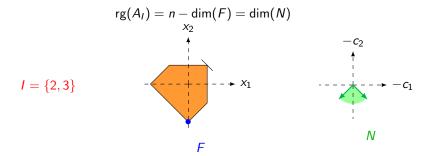




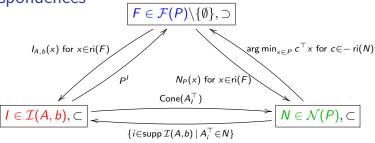








### Correspondences



$$rg(A_{I}) = n - dim(F) = dim(N)$$

$$X_{2}$$

$$C_{2}$$

$$X_{1}$$

$$X_{2}$$

$$X_{1}$$

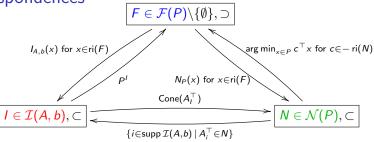
$$X_{2}$$

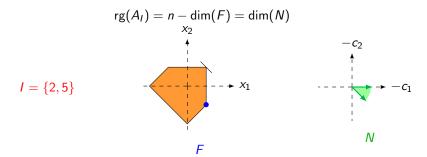
$$X_{1}$$

$$X_{2}$$

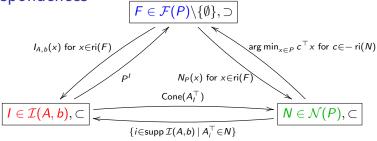
$$X_{1}$$

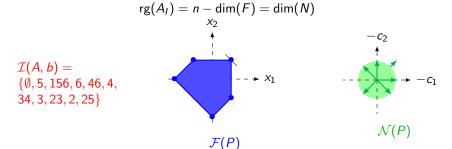
### Correspondences

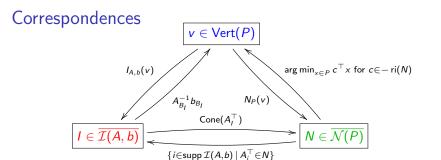


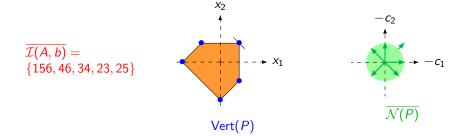


#### Correspondences









### Link with regular subdivisions

### Definition (DLRS10)

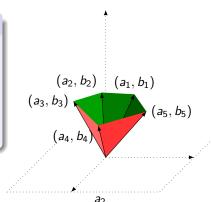
$$\mathcal{S}(A^\top,b) := \{ I_F \, | \, F \in \mathcal{F}_{\mathrm{low}}\big(LC_{A^\top,b}\big) \}$$

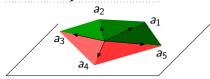
$$LC_{A^{\top},b} := \operatorname{Cone}\left(\begin{pmatrix} a_i \\ b_i \end{pmatrix}_{i \in [q]}\right)$$

$$I_F := \{i \in [q] | (a_i, b_i) \in F\}.$$

$$S(A^{\top},b) = \mathcal{I}(A,b)$$







$$\mathcal{I}(W^{\top},q) = \mathcal{I}_{com} \cup \big\{ \{5\}, \{4,5\}, \{1,5\} \big\}$$

### Link with regular subdivisions

### Definition (DLRS10)

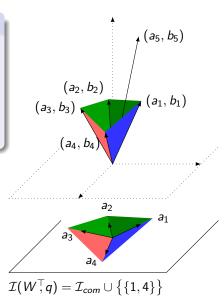
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### Link with regular subdivisions

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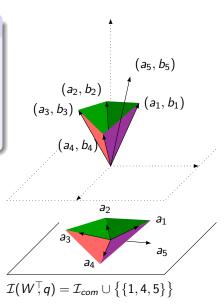
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#### Contents

- Linear programming and polyhedral geometry
  - Active constraints
  - Normal fan
  - Correspondences
- 2-Stage Stochastic Linear Programming
  - Reduction to finite sum
  - Chamber complex
  - Simplex for 2SLP

$$\min_{\mathbf{x} \in \mathbb{R}^{n}} c^{\top} \mathbf{x} + \mathbb{E} \begin{bmatrix} \min_{\mathbf{y} \in \mathbb{R}^{m}} \mathbf{q}^{\top} \mathbf{y} \\ \text{s.t.} & T\mathbf{x} + W\mathbf{y} \leqslant \mathbf{h} \end{bmatrix}$$
s.t.  $A\mathbf{x} \leqslant \mathbf{b}$  (2SLP)

where  $T \in \mathbb{R}^{p \times n}$ ,  $W \in \mathbb{R}^{p \times m}$  and  $h \in \mathbb{R}^p$ .

We can assume A = 0 and b = 0:

$$\widetilde{T} := \begin{pmatrix} T \\ A \end{pmatrix}, \quad \widetilde{W} := \begin{pmatrix} W \\ 0 \end{pmatrix} \quad \text{and } \widetilde{h} = \begin{pmatrix} h \\ b \end{pmatrix}$$

$$\min_{\mathbf{x} \in \mathbb{R}^n} \quad c^{\top} \mathbf{x} + \mathbb{E} \begin{bmatrix} \min_{\mathbf{y} \in \mathbb{R}^m} & \mathbf{q}^{\top} \mathbf{y} \\ \text{s.t.} & T\mathbf{x} + W\mathbf{y} \leqslant \mathbf{h} \\ & A\mathbf{x} & \leqslant \mathbf{b} \end{bmatrix}$$

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where  $T \in \mathbb{R}^{p \times n}$ ,  $W \in \mathbb{R}^{p \times m}$  and  $h \in \mathbb{R}^p$ .

We can assume A = 0 and b = 0:

$$\widetilde{T} := \begin{pmatrix} T \\ A \end{pmatrix}, \quad \widetilde{W} := \begin{pmatrix} W \\ 0 \end{pmatrix} \quad \text{and } \widetilde{h} = \begin{pmatrix} h \\ b \end{pmatrix}$$

$$\min_{x \in \mathbb{R}^n} c^\top x + V(x) \tag{2SLP}$$

where

$$V(x) := \mathbb{E} \begin{bmatrix} \min_{y \in \mathbb{R}^m} \mathbf{q}^{\top} y \\ \text{s.t.} \quad Tx + Wy \leqslant h \end{bmatrix}$$

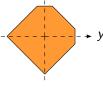
### Fiber $P_{\star}$

$$V(x) = \mathbb{E}\left[\min_{y \in P_X} \mathbf{q}^\top y\right]$$
 where  $P_X := \{y \in \mathbb{R}^m \mid Tx + Wy \leqslant h\}$ 

We assume supp( $\mathbf{q}$ )  $\subset$  - Cone( $W^{\top}$ ) i.e.  $V(x) > -\infty$ . Example:

$$T = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ -1 \\ -1 \\ 1 \end{pmatrix} W = \begin{pmatrix} 1 & 1 \\ 1 & -1 \\ -1 & 1 \\ 1 & 0 \\ 0 & 1 \\ -1 & 0 \end{pmatrix} h = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

$$P_{x} \text{ for } x = 0.8$$



 $P_{x}$  for x = 0.8

### Fiber $P_x$

$$V(x) = \mathbb{E} \left[ \min_{y \in P_x} \mathbf{q}^\top y \right]$$
 where  $P_x := \{ y \in \mathbb{R}^m \mid Tx + Wy \leqslant h \}$ 

We assume  $\operatorname{supp}(\mathbf{q}) \subset -\operatorname{Cone}(W^{\top})$  i.e.  $V(x) > -\infty$ . Example:

$$y_1 + y_2 \leqslant 1 \tag{1}$$

$$y_1 - y_2 \leqslant 1 \tag{2}$$

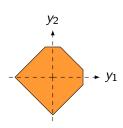
$$-y_1 - y_2 \leqslant 1 \tag{3}$$

$$-y_1 + y_2 \leqslant 1 \tag{4}$$

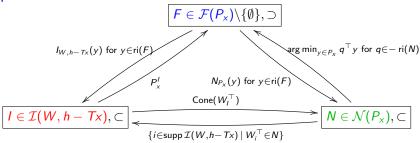
$$y_1 \leqslant x$$
 (5)

$$y_2 \leqslant x$$
 (6)

$$x \leqslant 1.5 \tag{7}$$

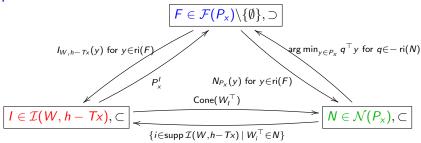


$$P_x$$
 for  $x = 0.8$ 

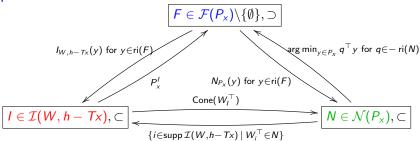


$$V(x) = \mathbb{E}\left[\min_{y \in P_x} \mathbf{q}^\top y\right]$$

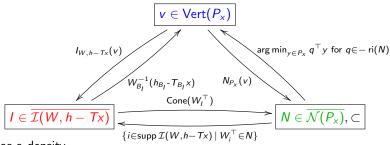
$$= \sum_{N \in \mathcal{N}(P_x)} \mathbb{E}\left[\mathbf{q}^\top \mathbb{1}_{\mathbf{q} \in -ri \, N}\right] y_N(x) \quad \text{with } y_N(x) \in \cap_{q \in -N} \arg\min_{y \in P_x} q^\top y$$



$$\begin{split} V(x) &= \mathbb{E} \big[ \min_{y \in P_x} \mathbf{q}^\top y \big] \\ &= \sum_{N \in \mathcal{N}(P_x)} \mathbb{E} \big[ \mathbf{q}^\top \mathbb{1}_{\mathbf{q} \in - \operatorname{ri} N} \big] y_N(x) \quad \text{with } y_N(x) \in \cap_{q \in -N} \arg\min_{y \in P_x} q^\top y \\ &= \sum_{F \in \mathcal{F}(P_x)} \mathbb{E} \big[ \mathbf{q}^\top \mathbb{1}_{\mathbf{q} \in - \operatorname{ri} N_{P_x}(F)} \big] y_F \quad \text{with } y_F \in F \end{split}$$



$$\begin{split} V(x) &= \mathbb{E}\big[\min_{y \in P_x} \mathbf{q}^\top y\big] \\ &= \sum_{N \in \mathcal{N}(P_x)} \mathbb{E}\big[\mathbf{q}^\top \mathbb{1}_{\mathbf{q} \in -\operatorname{ri} N}\big] y_N(x) \quad \text{with } y_N(x) \in \cap_{q \in -N} \arg\min_{y \in P_x} q^\top y \\ &= \sum_{F \in \mathcal{F}(P_x)} \mathbb{E}\big[\mathbf{q}^\top \mathbb{1}_{\mathbf{q} \in -\operatorname{ri} N_{P_x}(F)}\big] y_F \quad \text{with } y_F \in F \\ &= \sum_{I \in \mathcal{I}(W, h - Tx)} \mathbb{E}\big[\mathbf{q}^\top \mathbb{1}_{\mathbf{q} \in -\operatorname{ri} \operatorname{Cone}(W_I^\top)}\big] y_I(x) \quad \text{with } y_I(x) \in P_x^I \end{split}$$



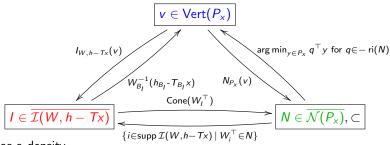
If q has a density,

$$V(x) = \mathbb{E}\left[\min_{y \in P_{x}} \mathbf{q}^{\top}y\right]$$

$$= \sum_{N \in \overline{\mathcal{N}(P_{x})}} \mathbb{E}\left[\mathbf{q}^{\top}\mathbb{1}_{\mathbf{q} \in -N}\right] y_{N}(x) \quad \text{with } y_{N}(x) \in \cap_{q \in -N} \arg\min_{y \in P_{x}} q^{\top}y$$

$$= \sum_{v \in \text{Vert}(P_{x})} \mathbb{E}\left[\mathbf{q}^{\top}\mathbb{1}_{\mathbf{q} \in -N_{P_{x}}(F)}\right] v$$

$$= \sum_{I \in \overline{\mathcal{I}(W, h - T_{x})}} \mathbb{E}\left[\mathbf{q}^{\top}\mathbb{1}_{\mathbf{q} \in -\text{Cone}(W_{I}^{\top})}\right] y_{I}(x) \quad \text{with } y_{I}(x) \in P_{x}^{I}$$

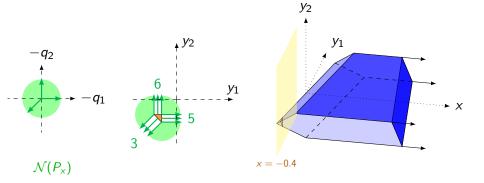


If q has a density,

$$\begin{split} V(x) &= \mathbb{E} \big[ \min_{y \in P_x} \mathbf{q}^\top y \big] \\ &= \sum_{N \in \overline{\mathcal{N}(P_x)}} \mathbb{E} \big[ \mathbf{q}^\top \mathbb{1}_{\mathbf{q} \in -N} \big] y_N(x) \quad \text{with } y_N(x) \in \cap_{q \in -N} \arg\min_{y \in P_x} q^\top y \\ &= \sum_{v \in \text{Vert}(P_x)} \mathbb{E} \big[ \mathbf{q}^\top \mathbb{1}_{\mathbf{q} \in -N_{P_x}(F)} \big] v \\ &= \sum_{l \in \overline{\mathcal{I}(W,h-T_x)}} \mathbb{E} \big[ \mathbf{q}^\top \mathbb{1}_{\mathbf{q} \in -\text{Cone}(W_l^\top)} \big] W_{B_l}^{-1}(h_{B_l} - T_{B_l} x) \text{ with basis } B_l \subset I \end{split}$$

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

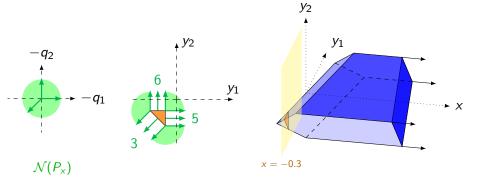
For 
$$x = -0.4$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{53, 36, 65\}$ 



 $P_{\times}$  and  $\mathcal{N}(P_{\times})$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

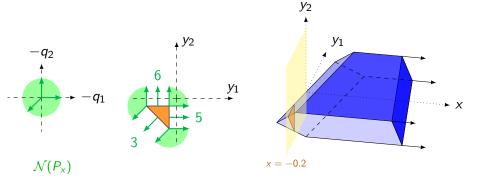
For 
$$x = -0.3$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{53, 36, 65\}$ 



 $P_{\times}$  and  $\mathcal{N}(P_{\times})$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

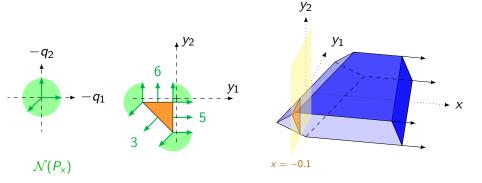
For 
$$x = -0.2$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{53, 36, 65\}$ 



 $P_{x}$  and  $\mathcal{N}(P_{x})$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

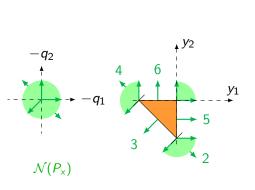
For 
$$x = -0.1$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{53, 36, 65\}$ 

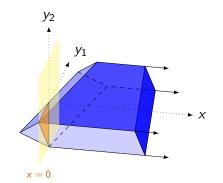


 $P_{\times}$  and  $\mathcal{N}(P_{\times})$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 0$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{523, 346, 65\}$ 



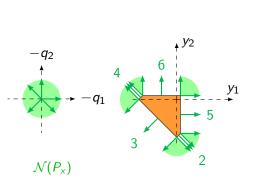


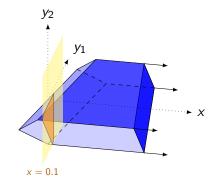
 $P_{x}$  and  $\mathcal{N}(P_{x})$ 

P and  $P_x$ 

$$P := \{(x,y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 0.1$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 65\}$ 



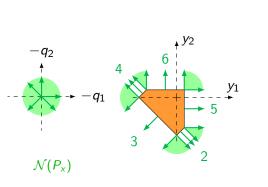


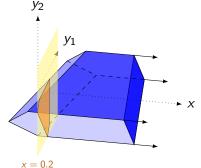
 $P_{x}$  and  $\mathcal{N}(P_{x})$ 

P and  $P_x$ 

$$P := \{(x,y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 0.2$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 65\}$ 



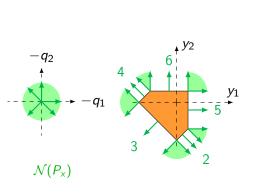


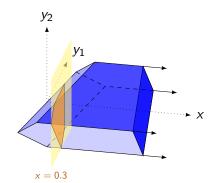
 $P_{x}$  and  $\mathcal{N}(P_{x})$ 

P and  $P_x$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 0.3$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 65\}$ 



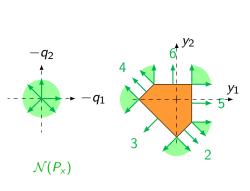


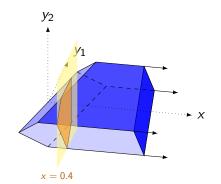
 $P_{x}$  and  $\mathcal{N}(P_{x})$ 

P and  $P_x$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 0.4$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 65\}$ 



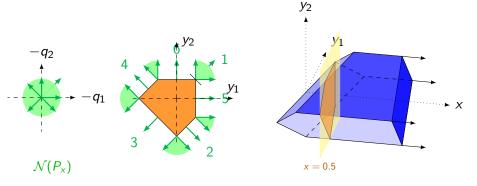


 $P_{x}$  and  $\mathcal{N}(P_{x})$ 

P and  $P_x$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 0.5$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 615\}$ 

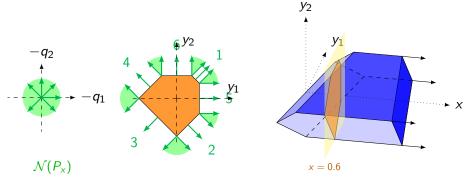


 $P_{x}$  and  $\mathcal{N}(P_{x})$ 

P and  $P_x$ 

$$P := \{(x,y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

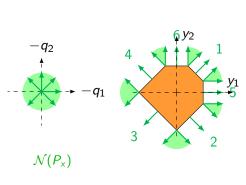
For 
$$x = 0.6$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 61, 15\}$ 

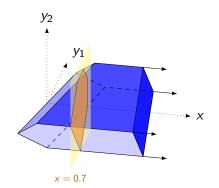


 $P_{\times}$  and  $\mathcal{N}(P_{\times})$ 

$$P := \{(x,y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 0.7$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 61, 15\}$ 



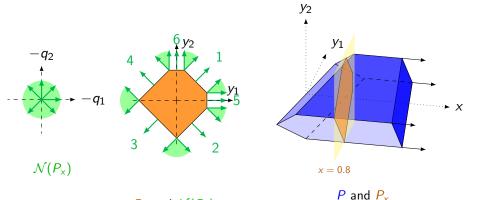


 $P_{x}$  and  $\mathcal{N}(P_{x})$ 

P and  $P_x$ 

$$P := \{(x,y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 0.8$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 61, 15\}$ 

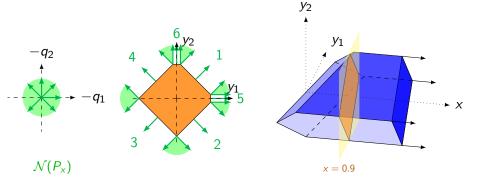


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 $P_{x}$  and  $\mathcal{N}(P_{x})$ 

$$P := \{(x,y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

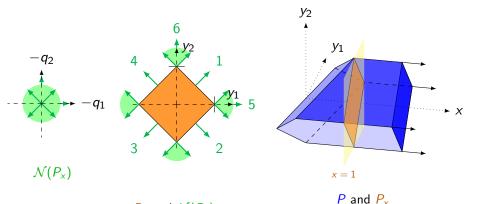
For 
$$x = 0.9$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 61, 15\}$ 



 $P_{x}$  and  $\mathcal{N}(P_{x})$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 1$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{152, 23, 34, 461\}$ 

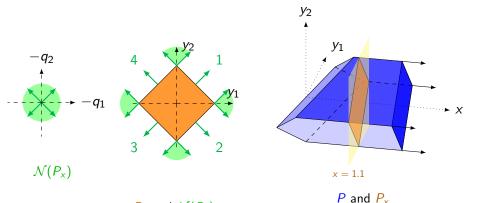


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 $P_{x}$  and  $\mathcal{N}(P_{x})$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 1.1$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{12, 23, 34, 41\}$ 

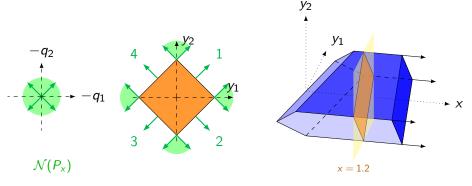


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 $P_{\times}$  and  $\mathcal{N}(P_{\times})$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 1.2$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{12, 23, 34, 41\}$ 

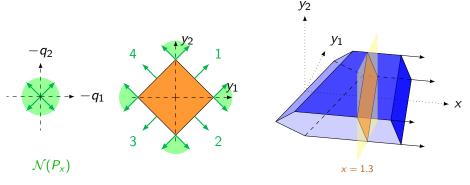


 $P_{\times}$  and  $\mathcal{N}(P_{\times})$ 

P and  $P_x$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 1.3$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{12, 23, 34, 41\}$ 

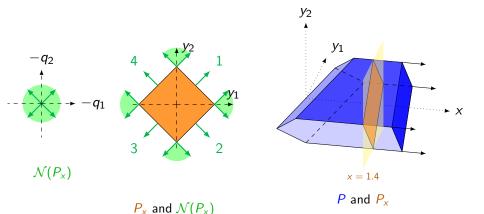


 $P_{\times}$  and  $\mathcal{N}(P_{\times})$ 

P and  $P_{x}$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
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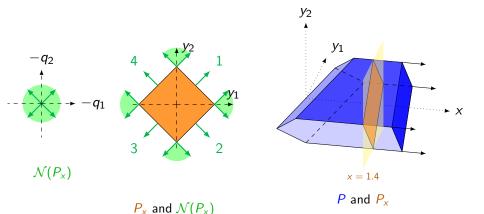
For 
$$x = 1.4$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{12, 23, 34, 41\}$ 



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$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

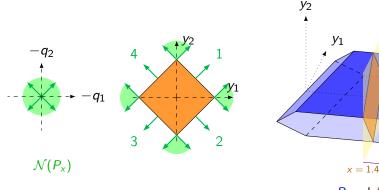
For 
$$x = 1.4$$
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For 
$$x = 1.4$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{12, 23, 34, 41\}$ 

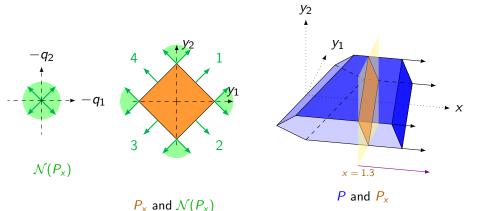


 $P_{x}$  and  $\mathcal{N}(P_{x})$ 

P and  $P_x$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

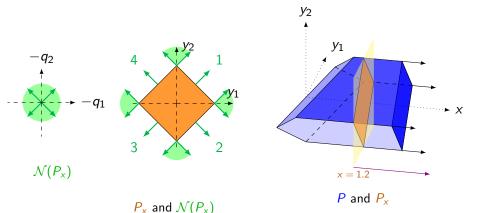
For 
$$x = 1.3$$
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$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

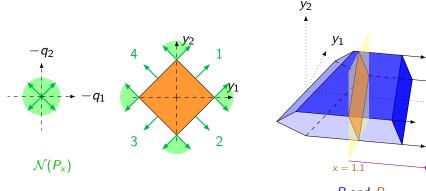
For 
$$x = 1.2$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{12, 23, 34, 41\}$ 



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$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

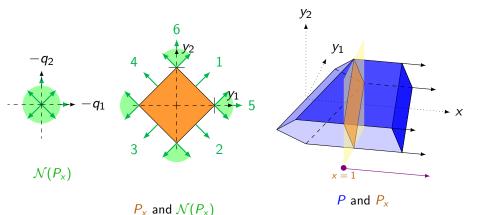
For 
$$x = 1.1$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{12, 23, 34, 41\}$ 





$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

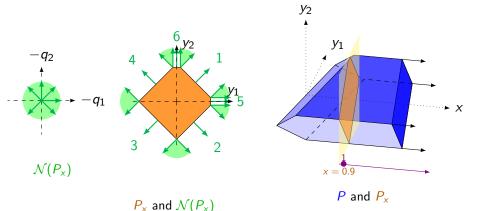
For 
$$x = 1$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{152, 23, 34, 461\}$ 



Maël Forcier

$$P := \{(x,y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

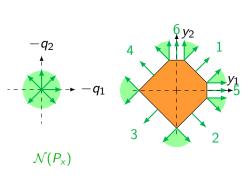
For 
$$x = 0.9$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 61, 15\}$ 

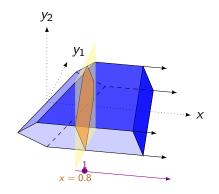


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$$P := \{(x,y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 0.8$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 61, 15\}$ 

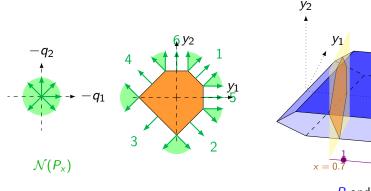




P and  $P_x$ 

$$P := \{(x,y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

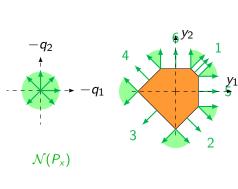
For 
$$x = 0.7$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 61, 15\}$ 

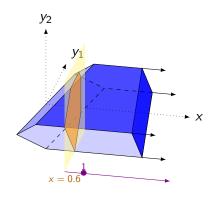




$$P := \{(x,y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 0.6$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 61, 15\}$ 

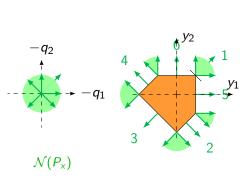


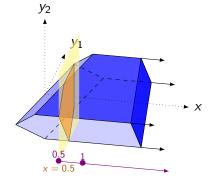


 $extcolor{black}{P}$  and  $extcolor{black}{P_{ imes}}$ 

$$P := \{(x,y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 0.5$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 615\}$ 



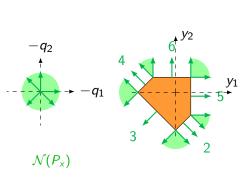


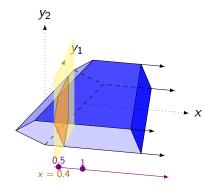
 $P_{x}$  and  $\mathcal{N}(P_{x})$ 

P and  $P_x$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 0.4$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 65\}$ 



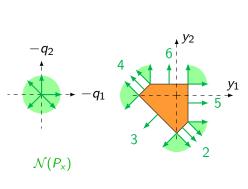


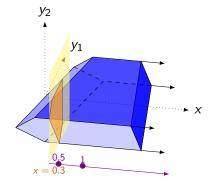
 $P_{x}$  and  $\mathcal{N}(P_{x})$ 

P and  $P_x$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 0.3$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 65\}$ 



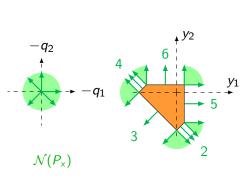


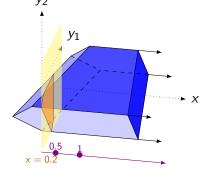
 $P_{\times}$  and  $\mathcal{N}(P_{\times})$ 

P and  $P_x$ 

$$P := \{(x,y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 0.2$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 65\}$ 



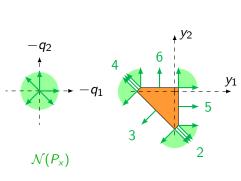


 $P_{\times}$  and  $\mathcal{N}(P_{\times})$ 

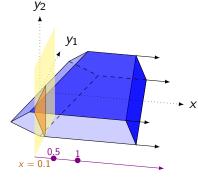
P and  $P_x$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\} \text{ and } P_x := \{y \mid Tx + Wy \leqslant h\}$$

For 
$$x = 0.1$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{52, 23, 34, 46, 65\}$ 



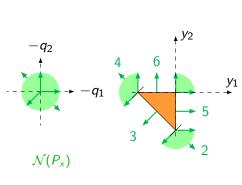


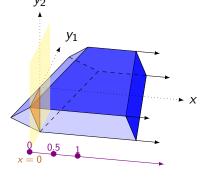


P and  $P_x$ 

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = 0$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{523, 346, 65\}$ 



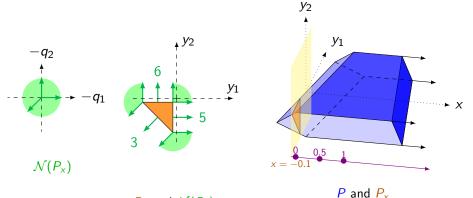


 $P_{\times}$  and  $\mathcal{N}(P_{\times})$ 

P and  $P_x$ 

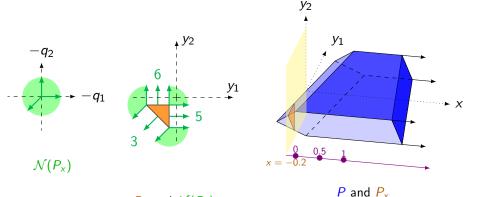
$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = -0.1$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{53, 36, 65\}$ 



$$P := \{(x,y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

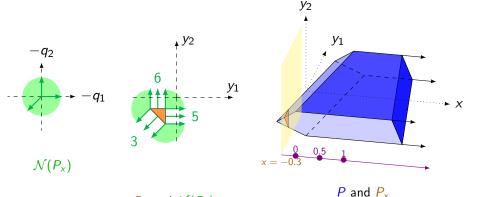
For 
$$x = -0.2$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{53, 36, 65\}$ 



Maël Forcier

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

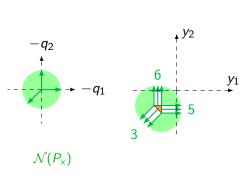
For 
$$x = -0.3$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{53, 36, 65\}$ 

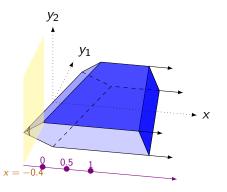


Maël Forcier

$$P := \{(x, y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = -0.4$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{53, 36, 65\}$ 



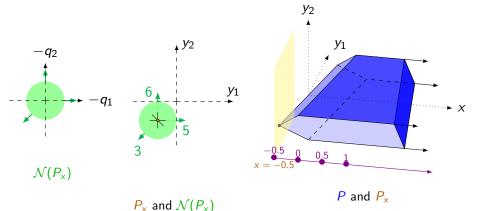


 $P_{\times}$  and  $\mathcal{N}(P_{\times})$ 

P and  $P_{x}$ 

$$P := \{(x,y) \mid Tx + Wy \leqslant h\}$$
 and  $P_x := \{y \mid Tx + Wy \leqslant h\}$ 

For 
$$x = -0.5$$
,  $\overline{\mathcal{I}(W, h - Tx)} = \{536\}$ 



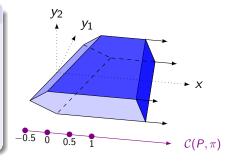
Maël Forcier

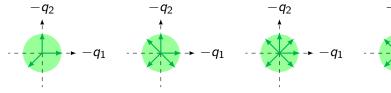
## What are the constant regions of $\mathcal{N}(P_x)$ , $\mathcal{I}(W, h - Tx)$ ?

#### Lemma

There exists a collection  $C(P,\pi)$  whose relative interior of cells are the constant regions of  $x \to \mathcal{N}(P_x)$  and  $x \to \mathcal{I}(W, h - Tx)$ .

For 
$$\sigma \in \mathcal{C}(P,\pi)$$
 and  $x, x' \in ri(\sigma)$ ,  $\mathcal{N}(P_x) = \mathcal{N}(P_{x'}) = \mathcal{N}_{\sigma}$   $\mathcal{I}(W, h - Tx) = \mathcal{I}(W, h - Tx') = \mathcal{I}_{\sigma}$ 





 $\mathcal{N}_{\sigma}$  for  $\sigma = [0, 0.5]$ 

 $\mathcal{N}_{\sigma}$  for  $\sigma = [-0.5, 0]$ 

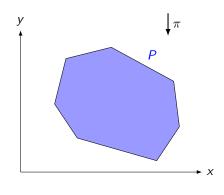
#### Definition

The chamber complex  $\mathcal{C}(P,\pi)$  of P along  $\pi$  is

$$\mathcal{C}(P,\pi) := \{ \sigma_{P,\pi}(x) \mid x \in \pi(P) \}$$

where

$$\sigma_{P,\pi}(x) := \bigcap_{F \in \mathcal{F}(P) \text{ s.t. } x \in \pi(F)} \pi(F)$$



$$\pi(E) := \{ x \in \mathbb{R}^n \mid \exists y \in \mathbb{R}^m, \ (x, y) \in E \}$$

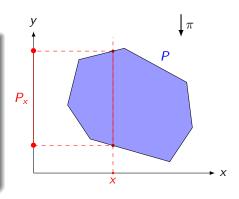
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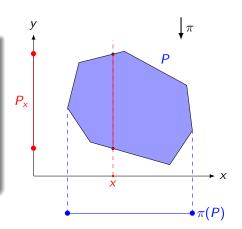
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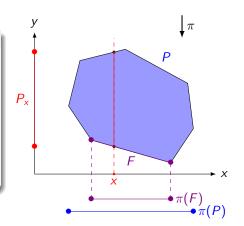
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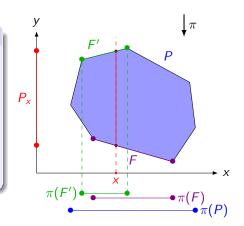
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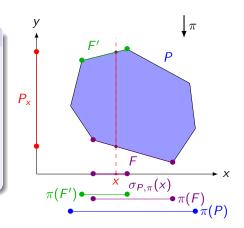
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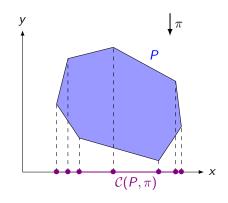
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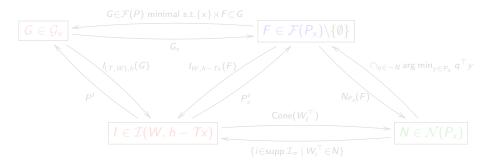
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### Proof of normal equivalence

$$\mathcal{G}_{\mathsf{X}} := \{ G \in \mathcal{F}(P) \, | \, \mathsf{X} \in \mathsf{ri}\left(\pi(G)\right) \}$$

Let  $\sigma \in \mathcal{C}(P, \pi)$ , for all  $x, x' \in ri(\sigma)$ , we have

$$\mathcal{G}_{\sigma}:=\mathcal{G}_{\mathsf{X}}=\mathcal{G}_{\mathsf{X}'}$$



By the correspondences,

$$\mathcal{I}_{\sigma} := \mathcal{I}(W, h - Tx) = \mathcal{I}(W, h - Tx')$$

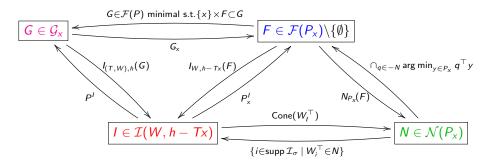
$$\mathcal{N}_{\sigma} := \mathcal{N}(P_{\mathsf{X}}) = \mathcal{N}(P_{\mathsf{X}'})$$

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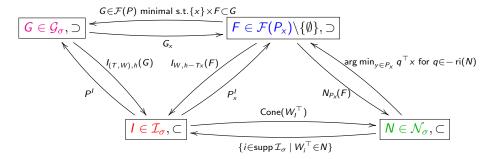
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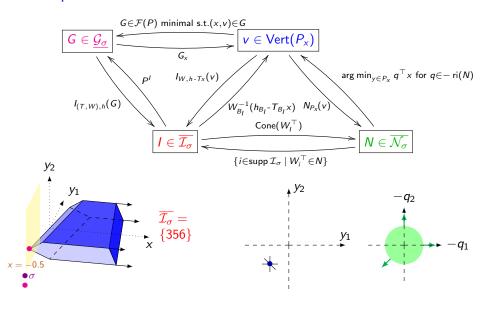
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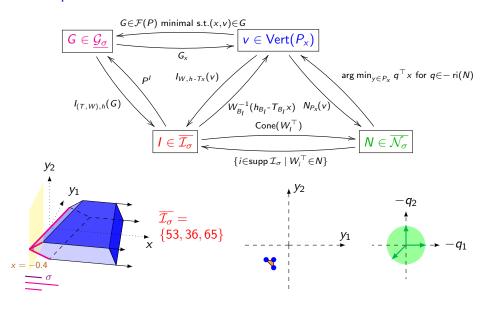
$$\mathcal{I}_{\sigma} := \mathcal{I}(W, h - Tx) = \mathcal{I}(W, h - Tx')$$
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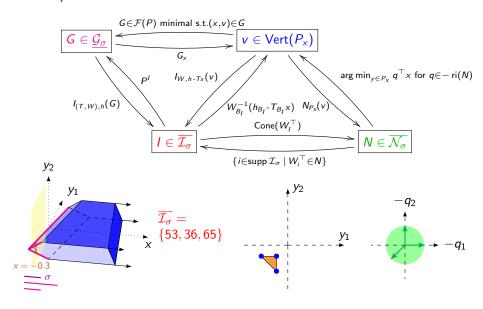
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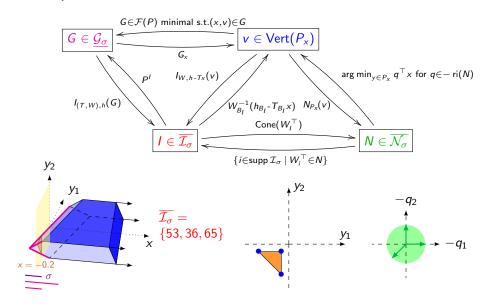


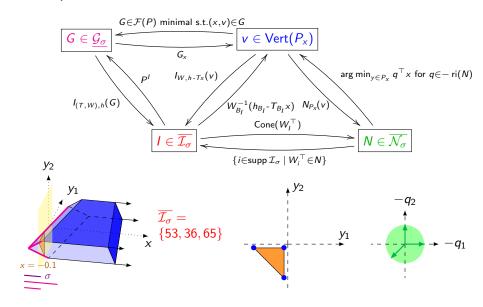
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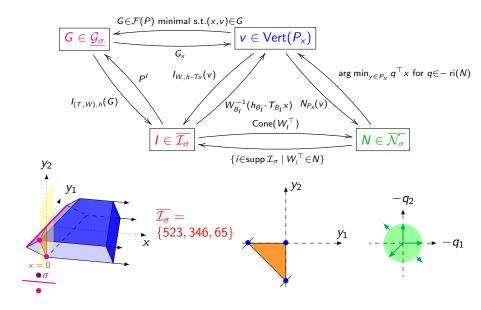


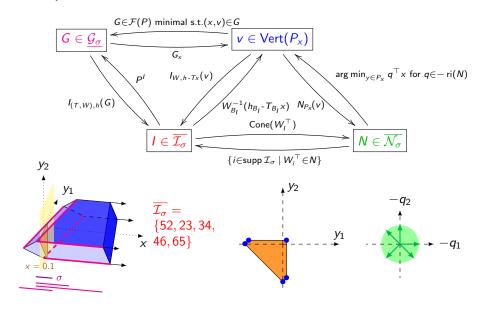


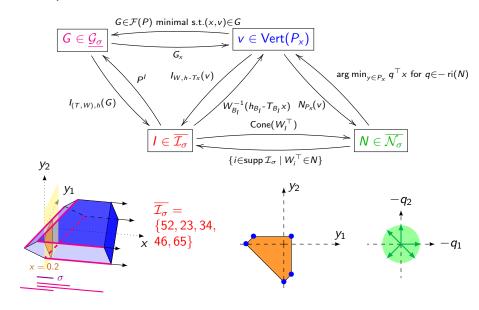


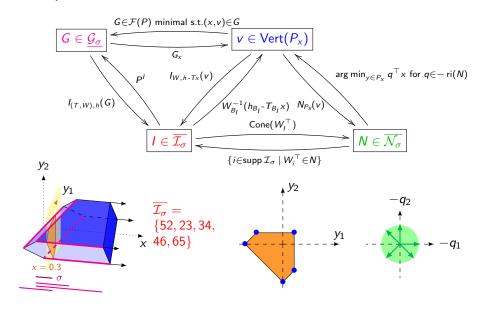


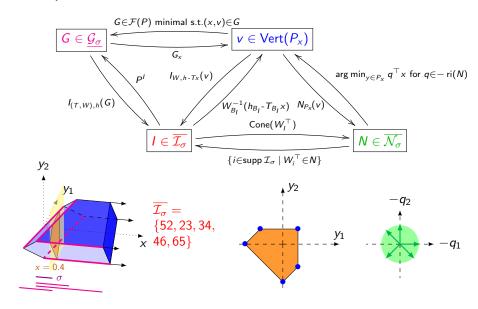


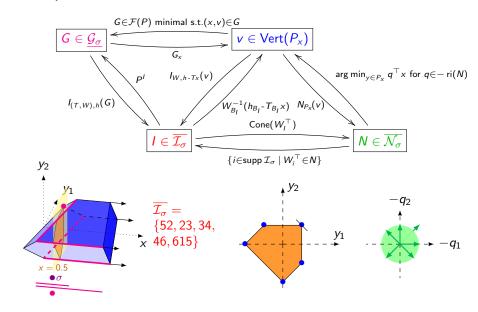


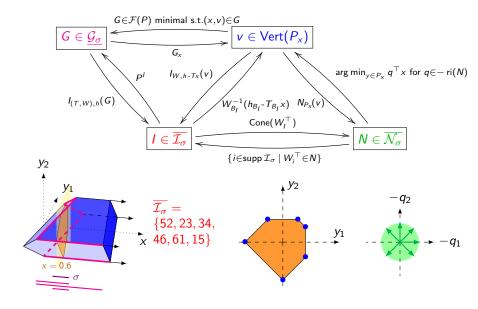


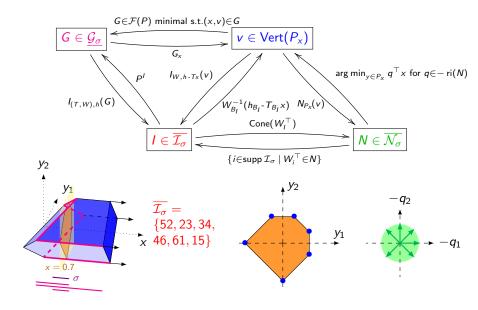


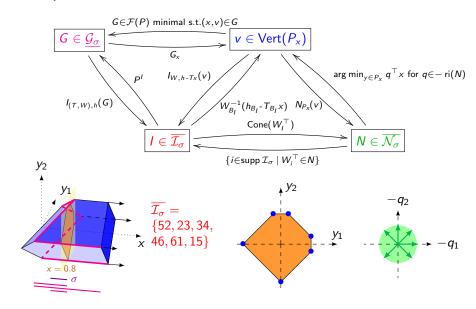


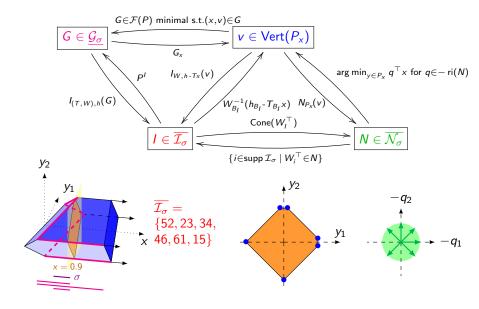


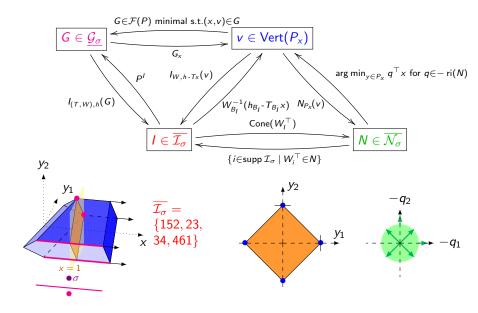


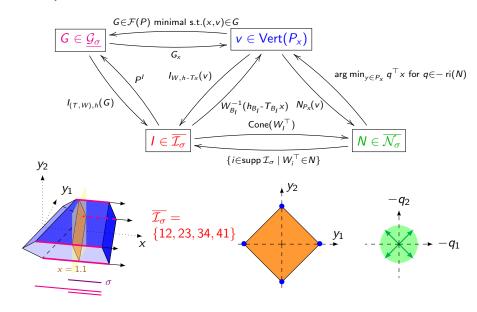


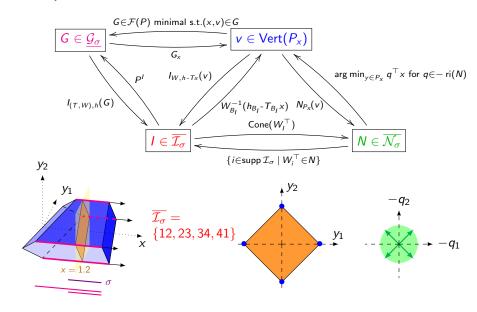


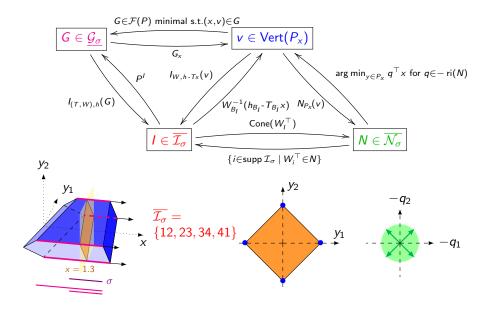


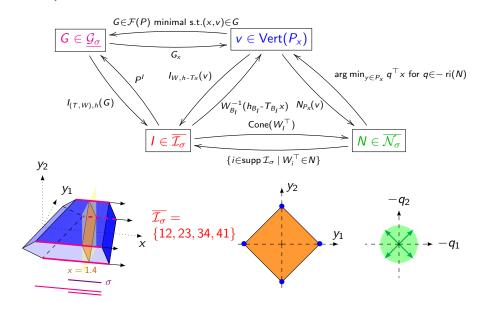












### $\mathcal{I}_{\sigma}$ contains all needed informations

Recall that, for all  $x \in ri(\sigma)$ 

$$V(x) = \sum_{I \in \overline{\mathcal{I}_{\sigma}}} \mathbb{E}\left[\mathbf{q}^{\top} \mathbb{1}_{\mathbf{q} \in -\mathsf{Cone}(W_{I}^{\top})}\right] W_{B_{I}}^{-1}(h_{B_{I}} - T_{B_{I}}x) \quad \text{ with } B_{I} \text{ basis } \subset I$$

Moreover, we can show

$$x \in ri(\sigma) \iff \begin{cases} \forall I \in \overline{\mathcal{I}_{\sigma}}, & \text{where} \\ \forall i \in I \backslash B_{I}, & v_{i}^{B_{I}} x = u_{i}^{B_{I}} & v_{i}^{B} := T_{i} - W_{i} W_{B}^{-1} T_{B} \\ \forall j \in [q] \backslash I, & v_{j}^{B_{I}} x < u_{j}^{B_{I}} & u_{i}^{B} := h_{i} - W_{i} W_{B}^{-1} h_{B} \end{cases}$$

If  $\sigma$  and  $\tau$  are adjacent chambers in  $\mathcal{C}(P,\pi)$ Then,  $\mathcal{I}_{\sigma}$  and  $\mathcal{I}_{\tau}$  do not differ at lot.

ightharpoonup Idea: Pivot between vertices in the chamber complex and update  $\mathcal{I}_{\sigma}$ 

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ightharpoonup Idea: Pivot between vertices in the chamber complex and update  $\mathcal{I}_{\sigma}$ 

## Secondary simplex algorithm: pivot procedure

Compute every edges directions d adjacent to x and  $\overline{\mathcal{I}_d} := \mathcal{I}(W, h - T(x + \varepsilon d))$  for  $\varepsilon > 0$  small enough;

if If there exists an edge with direction d such that,  $c^{\top}d + \sum_{I \in \overline{\mathcal{I}_{d}}} \mathbb{E}\left[\mathbf{q} \mathbb{1}_{\mathbf{q} \in -\mathsf{Cone}(W_{I}^{\top})}\right] W_{B_{I}}^{-1} T_{B_{I}} d_{B_{I}} < 0 \text{ then}$ Choose d such a direction and set  $\overline{\mathcal{I}} := \overline{\mathcal{I}_d}$ ; Let  $\lambda = \min_{l \in \overline{\mathcal{I}}, j \in [p] \backslash I | v_j^{B_l} d > 0} \frac{u_j^{B_l} - v_j^{B_l} x}{v_*^{B_l} d}$  ; Let  $Sat := \{(I,j) \mid I \in \overline{\mathcal{I}}, \ j \in [p] \setminus I, \lambda = \frac{u_j^{B_I} - v_j^{B_I} x}{v_j^{B_I} d}\};$ if  $\lambda = +\infty$  then Return "The value of (2SLP) is  $-\infty$ " else Let  $\mathcal{I}_{sat} = \{I \in \overline{\mathcal{I}} \mid \exists j, (I, j) \in Sat\};$ Let  $\mathcal{J}_{new} = \{I \cup \bigcup_{i \mid (I,i) \in Sat} \{j\} \mid I \in \mathcal{I}_{sat}\}$ ; Compute  $\overline{\mathcal{J}} = (\overline{\mathcal{I}} \setminus \mathcal{I}_{sat}) \cup \mathcal{J}_{new}$ ; Return  $(x + \lambda d, \overline{\mathcal{J}})$ end

else

Return "x is an optimal solution"

$$y_{1} + y_{2} \leq 1$$

$$y_{1} - y_{2} \leq 1$$

$$-y_{1} - y_{2} \leq 1$$

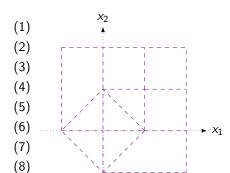
$$-y_{1} + y_{2} \leq 1$$

$$y_{1} \leq x_{1}$$

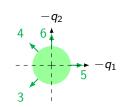
$$y_{2} \leq x_{2}$$

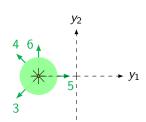
$$x_{1} \leq 2$$

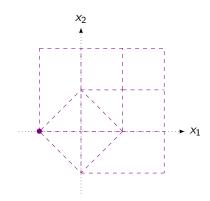
$$x_{2} \leq 2$$



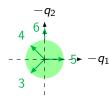
$$\overline{\mathcal{I}} = \{3456\}$$

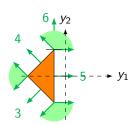


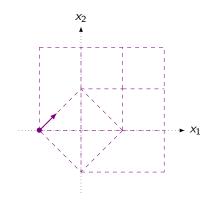




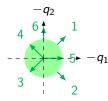
$$\overline{\mathcal{I}}=\{34,35,456\}$$

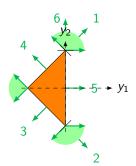


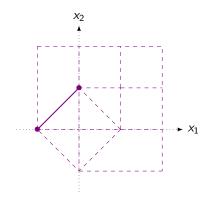




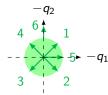
$$\overline{\mathcal{I}} = \{34, 235, 1456\}$$

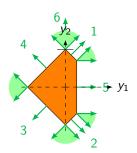


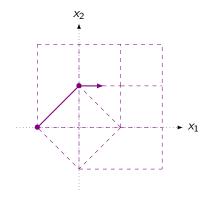




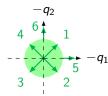
$$\overline{\mathcal{I}} = \{34, 23, 25, 146, 15\}$$

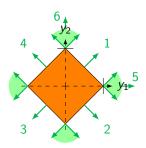


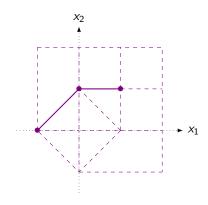




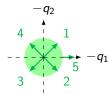
$$\overline{\mathcal{I}} = \{34, 23, 125, 146\}$$

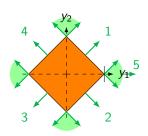


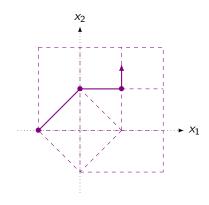




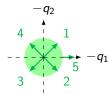
$$\overline{\mathcal{I}} = \{34, 23, 125, 14\}$$

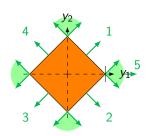


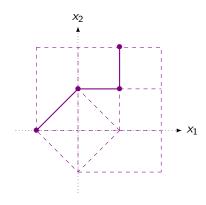




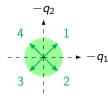
$$\overline{\mathcal{I}} = \{348, 238, 1258, 148\}$$

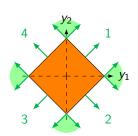


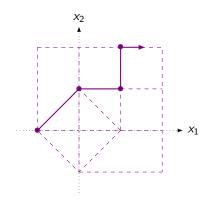




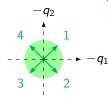
$$\overline{\mathcal{I}} = \{348, 238, 128, 148\}$$

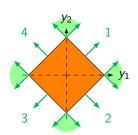


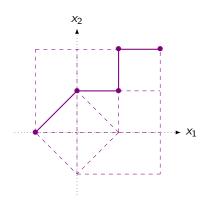




 $\overline{\mathcal{I}} = \{3478, 2378, 1278, 1478\}$ 







M. Forcier, S. Gaubert, V. Leclère

Exact quantization of multistage stochastic linear problems.

arXiv preprint arXiv:2107.09566 (2021).

M. Forcier, V. Leclère
Generalized adaptive partition-based method for two-stage stochastic linear programs: convergence and generalization.

arXiv preprint arXiv:2109.04818 (2021).

M. Forcier, V. Leclère Convergence of Stochastic Dual Dynamic Programming algorithms for non-finitely supported distributions soon.

Jesús A De Loera, Jörg Rambau, and Francisco Santos. *Triangulations Structures for algorithms and applications*. Springer, 2010.

# Thank you for listening! Any question?

