# The Minkowski-Weyl Theorem

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Defense of Bachelor's Thesis, 2019



#### Outline

- Goal and Outcome
  - Goal / Outcome
  - Strengths and Weaknesses
- The Work
  - Not Original
  - Implementation
  - Pointed / Full-Dimensional Polyhedra
- Closing Remarks
  - If I could do it again
  - Summary



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- Prove the Minkowski-Weyl Theorem
- Implement the proof in C++

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- Farkas Lemmas
- General idea for polyhedral reductions

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# V/H Polyhedra/Cones

Let 
$$U \in \mathbb{R}^{d \times l}$$
,  $V \in \mathbb{R}^{d \times m}$ ,  $A \in \mathbb{R}^{m \times d}$ 

V-Cone:

$$\{U\mathbf{t} \mid \mathbf{t} \geq \mathbf{0}\} \equiv \mathsf{cone}(U)$$

V-Polytope:

$$\{V\lambda \mid \lambda \geq \mathbf{0}, \mathbf{1}^T\lambda = 1\} \equiv \operatorname{conv}(V)$$

V-Polyhedron:

$$\{U\mathbf{t} + V\lambda \mid \mathbf{t}, \lambda \geq \mathbf{0}, \mathbf{1}^T\lambda = 1\} \equiv \mathsf{cone}(U) + \mathsf{conv}(V)$$

H-Cone:

$$\{x \mid Ax \leq 0\}$$

H-Polyhedron:

$$\{x \mid Ax \leq b\}$$

# Minkowski-Weyl Theorem

- General Statement:
   V-Polyhedra and H-Polyhedra are different representations of the same objects
- For Cones:
   V-Cones and H-Cones are different representations of the same objects

First, the proof is done for cones, then polyhedra are reduced to cones.

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#### Picture of Proof

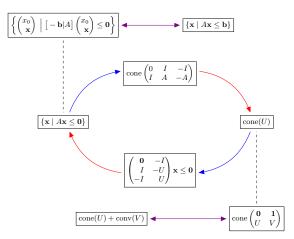


Figure 1: Diagram of the proof  $P_H \leftrightarrow P_V$ 

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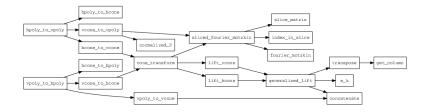
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### Files and Includes

file	includes	
linear_algebra.h	<c++ library="" standard=""></c++>	
fourier_motzkin.h	linear_algebra.h	
polyhedra.h	fourier_motzkin.h	
main.cpp	polyhedra.h	
test_functions.h	linear_alebra.h	
test.cpp	test_functions.h,polyhedra.h	

# Callgraph



### Matrix fourier\_motzkin(Matrix,k)

```
23    const auto z_end = partition(M.begin(), M.end(),
24         [k](const Vector &v) { return v[k] == 0; });
25    const auto p_end = partition(z_end, M.end(),
26         [k](const Vector &v) { return v[k] > 0; });
```

#### Partition M into logical sets Z, P, N that satisfy the following:

set	range	property
Z	[M.begin(),z_end)	$it \in Z \Leftrightarrow (\star it)[k] = 0$
Р	[z_end, p_end)	$it \in P \Leftrightarrow (*it)[k] > 0$
Ν	[p_end, M.end())	$it \in N \Leftrightarrow (*it)[k] < 0$

### Matrix fourier\_motzkin(Matrix,k)

This function creates the sets which correspond to:

$$\left\{B_i^k B_j - B_j^k B_i \mid i \in P, j \in N\right\}, \quad \left\{Y_k^i Y^j - Y_k^j Y^i \mid i \in P, j \in N\right\}$$

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$$AV' \leq \mathbf{0}$$
  $\Rightarrow \operatorname{cone}(V') \subseteq \{\mathbf{x} \mid A\mathbf{x} \leq \mathbf{0}\}$   
 $V \subseteq V'$   $\Rightarrow \operatorname{cone}(V) \subseteq \operatorname{cone}(V')$   
 $\operatorname{cone}(V') \subseteq \{\mathbf{x} \mid A\mathbf{x} \leq \mathbf{0}\} \Rightarrow AV' \leq \mathbf{0}$   
 $\operatorname{cone}(V) \subseteq \operatorname{cone}(V')$   $\stackrel{?}{\Rightarrow} V \subseteq V'$ 

- The last item would create an equivalence
  - Must relax "⊆" (vectors vs rays)
  - Requires notion of "essentially unique" representation



$$\begin{array}{ll} AV' \leq \mathbf{0} & \Rightarrow \mathsf{cone}(V') \subseteq \{\mathbf{x} \mid A\mathbf{x} \leq \mathbf{0}\} \\ V \subseteq V' & \Rightarrow \mathsf{cone}(V) \subseteq \mathsf{cone}(V') \\ \mathsf{cone}(V') \subseteq \{\mathbf{x} \mid A\mathbf{x} \leq \mathbf{0}\} \Rightarrow AV' \leq \mathbf{0} \\ \mathsf{cone}(V) \subseteq \mathsf{cone}(V') & \stackrel{?}{\Rightarrow} V \subseteq V' \end{array}$$

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# Pointed/ Full-Dimensional Polyhedra

- These are basically non-degeneracy constraints.
  - Pointed Polyhedra have at least one vertex
  - Full dimensional polyhedra contain an affine independent set
- Pointed V-Polyhedra and Full-Dimensional H-Polyhedra have "essentially unique" sets of generators / contraints
- These "essentially unique" sets make it easy to test for equivalence
- The characterizations are similar to "linear independence"



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- These "essentially unique" sets are "minimal"
- A set V is called minimal for cone(V) if

$$(\forall \mathbf{v} \in V) \operatorname{cone}(V \setminus \{\mathbf{v}\}) \subsetneq \operatorname{cone}(V)$$

• A set A is called *minimal* for  $\{x \mid Ax \leq 0\}$  if

$$(\forall A_i \in A) \ \{\mathbf{x} \mid A \setminus \{A_i\} \ \mathbf{x} \leq \mathbf{0}\} \supsetneq \{\mathbf{x} \mid A\mathbf{x} \leq \mathbf{0}\}$$

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#### Pointed V-Cones

- The following statements are equivalent.
  - one(V) is pointed.
  - **2**  $t \ge 0$ ,  $[Vt = 0 \Rightarrow t = 0]$
- Suppose that cone(V) is pointed. The following two statements are equivalent:
  - V is minimal
  - 2  $\mathbf{t} \geq \mathbf{0}, [V\mathbf{t} = V\mathbf{e}_i \Rightarrow \mathbf{t} = \mathbf{e}_i]$
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#### Farkas Lemma

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$$(\exists t \ge 0) : \mathbf{x} = Ut$$
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#### Full-Dimensional H-Cones

A set V of vectors is full-dimensional if

$$\forall \mathbf{y} \neq \mathbf{0}, \forall \mathbf{c} \in \mathbb{R}, \exists \mathbf{v} \in \mathbf{V}: \mathbf{y}^T \mathbf{v} \neq \mathbf{c}$$

- The following two statements are equivalent:
  - $\{x \mid Ax \leq 0\}$  is full dimensional and A is minimal
  - $\bigcirc$  cone( $A^T$ ) is pointed and A is minimal
- The Farkas Lemma can be used to prove the following:

$$\{\mathbf{x}\mid A\mathbf{x}\leq \mathbf{0}\}=\left\{\mathbf{x}\mid A'\mathbf{x}\leq \mathbf{0}\right\}\Leftrightarrow \mathsf{cone}(A^T)=\mathsf{cone}(A'^T)$$

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- Suppose that  $P = \{\mathbf{x} \mid A\mathbf{x} \leq b\} = \operatorname{cone}(U) + \operatorname{conv}(V)$ , and let  $\mathbf{r}$  be a vector. The following are equivalent:
  - $\bigcirc$  Ar < 0
  - 2  $(\forall \mathbf{x} \in P)(\forall \alpha > 0) \mathbf{x} + \alpha \mathbf{r} \in P$
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- Note that (2) in the proof above is independent of *A* and *U*.

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- If **v** is a vertex of cone(U) + conv(V), then  $[\mathbf{v} = U\mathbf{t} + V\lambda] \Rightarrow \mathbf{t} = \mathbf{0}$  (I call **v** here *U-free*)
- If v is a vertex of cone(U) + conv(V), then v is a vertex of conv(V)
- Let P = cone(U) + cone(V) be pointed. The following are equivalent
  - $\bigcirc$  (U, V) is minimal for P
  - $\bigcirc$  U is minimal for cone(U), V is the vertex set of P
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#### Farkas Lemma 2

• We need another form of the Farkas Lemma:

$$(\exists \mathbf{t} \geq \mathbf{0})\mathbf{t}^{\mathsf{T}} A = \mathbf{y}, \ \mathbf{t}^{\mathsf{T}} \mathbf{b} \leq c \Leftrightarrow \begin{cases} (\forall \mathbf{x}) A \mathbf{x} \leq \mathbf{0} \Rightarrow \mathbf{y}^{\mathsf{T}} \mathbf{x} \leq \mathbf{0} \text{ and } \\ (\forall \mathbf{x}) A \mathbf{x} \leq \mathbf{b} \Rightarrow \mathbf{y}^{\mathsf{T}} \mathbf{x} \leq c \end{cases}$$

 Basically, a constraint is valid for a polyhedron if and only if it is a non-negative combination of rows of constraints (plus some change)

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- The following are equivalent
  - $\{ \mathbf{x} \mid A\mathbf{x} \leq \mathbf{b} \}$  is full-dimensional
- $\mathbf{y}^T A = \mathbf{0}$ ,  $\mathbf{y}^T \mathbf{b} > \mathbf{0}$  occurs when two of the bounding hyperplanes are parallel
- If {x | Ax ≤ b} is full-dimensional, the following are equivalent
  - (A, b) is minimal
  - 2  $\mathbf{y} \geq \mathbf{0}, \mathbf{y} \neq \mathbf{e}_i, \mathbf{y}^T A = A_i \Rightarrow \mathbf{y}^T \mathbf{b} > b_i$
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- $\mathbf{y}^T A = \mathbf{0}$ ,  $\mathbf{y}^T \mathbf{b} > \mathbf{0}$  occurs when two of the bounding hyperplanes are parallel
- If {x | Ax ≤ b} is full-dimensional, the following are equivalent
  - (A, b) is minimal
  - $\mathbf{\hat{2}} \ \mathbf{\hat{y}} \geq \mathbf{\hat{0}}, \mathbf{y} \neq \mathbf{e}_i, \mathbf{y}^T A = A_i \Rightarrow \mathbf{y}^T \mathbf{b} > b_i$
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# **Dual Homogenization Cone**

- For H-Polyhedra switching to the dual setting is slightly more complicated
- The following two statement are equivalent:

- If  $\{x \mid Ax \leq b\}$  is minimal and full-dimensional, then either
  - one  $\begin{pmatrix} -\mathbf{b}^T & -1 \\ A^T & \mathbf{0} \end{pmatrix}$  is minimal and pointed, or
  - ② cone  $\begin{pmatrix} -\mathbf{b}^T \\ A^T \end{pmatrix}$  is minimal and pointed, and

$$\operatorname{cone}\begin{pmatrix} -\mathbf{b}^T \\ A^T \end{pmatrix} = \operatorname{cone}\begin{pmatrix} -\mathbf{b}^T & -1 \\ A^T & \mathbf{0} \end{pmatrix}$$



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  - cone  $\begin{pmatrix} -\mathbf{b}^T & -1 \\ A^T & \mathbf{0} \end{pmatrix}$  is minimal and pointed, or
  - 2 cone  $\binom{-\mathbf{b}^T}{A^T}$  is minimal and pointed, and

$$\operatorname{cone}\begin{pmatrix} -\mathbf{b}^T \\ A^T \end{pmatrix} = \operatorname{cone}\begin{pmatrix} -\mathbf{b}^T & -1 \\ A^T & \mathbf{0} \end{pmatrix}$$

# Pointed H-Polyhedra and Full-Dimensional V-Polyhedra

• **t** is a non-negative vector,  $V \neq \emptyset$ , and abbreviate linear-independent as LI.  $\bar{V}$  denotes  $\{\mathbf{v} - \mathbf{v}' \mid \mathbf{v}, \mathbf{v}' \in V\}$ .

	Pointed	Full-Dimensional
cone( <i>U</i> )	$U$ t = 0 $\Rightarrow$ t = 0	d LI vectors in U
cone(U) + conv(V)	$U$ t = 0 $\Rightarrow$ t = 0	<i>d</i> LI vectors in $U \cup \bar{V}$
$\{\mathbf{x} \mid A\mathbf{x} \leq 0\}$	d LI row vectors in A	$\mathbf{t}^T A = 0 \Rightarrow \mathbf{t} = 0$
$\{\mathbf{x} \mid A\mathbf{x} \leq \mathbf{b}\}$	d LI row vectors in A	$\mathbf{t}^T A = 0 \Rightarrow \mathbf{t}^T \mathbf{b} > 0$

#### Not much to see here

- The proofs are more tedious than enlightening
- The implementation is straightforward

#### Outline

- Goal and Outcome
  - Goal / Outcome
  - Strengths and Weaknesses
- 2 The Work
  - Not Original
  - Implementation
  - Pointed / Full-Dimensional Polyhedra
- Closing Remarks
  - If I could do it again
  - Summary



#### **Better MWT for Cones**

- H-Cone → V-Cone. Use same transform and "tensor-notation"
- Farkas Lemma using the following facts:
  - An intersection of closed sets is closed
  - 2 A projection of a closed subset of  $\mathbb{R}^n$  is closed
  - (Heine-Borel) A subset of  $\mathbb{R}^n$  is compact if and only if it is closed and bounded
  - lack 4 A continuous function on a compact subset of  $\mathbb{R}^n$  achieves its maximum
  - **5** Linear functions on  $\mathbb{R}^n$  are continuous
  - The Cauchy-Shwartz inequality
- Better dual cone:

$$\{\mathbf{x} \mid A\mathbf{x} \leq \mathbf{0}\} = \operatorname{cone}(U) \Leftrightarrow \operatorname{cone}(A^T) = \{\mathbf{x} \mid U^T\mathbf{x} \leq \mathbf{0}\}$$



#### Tensor-like notation

First, we write 
$$\{\mathbf{x} \mid A\mathbf{x} \leq \mathbf{0}\} = \left\{\mathbf{x} \mid \begin{pmatrix} \mathbf{x} \\ \mathbf{0} \end{pmatrix} \in \operatorname{cone} \begin{pmatrix} I & -I & \mathbf{0} \\ A & -A & I \end{pmatrix}\right\}$$
  
Now we need to do our intersections:

- Let  $\mathbf{u} \in \text{cone}(U)$  with  $u_l = 0$
- Let  $\mathbf{p}^i t_i$  sum over elements of U with positive *I*-th elements

$$\bullet \mathbf{u} = \mathbf{p}^i t_i + \mathbf{n}^j t_j + \mathbf{z}^k t_k$$

• 
$$u_l = 0 \Rightarrow p_l^i t_i + n_l^j t_j = 0 \dots \quad \sigma := p_l^i t_i = -n_l^j t_j$$

$$\bullet \ \mathbf{u} = \frac{\mathbf{n}^j t_j \rho_l^i t_l - \mathbf{p}^i t_l n_l^j t_j}{\sigma} + \mathbf{z}^k t_k = (\mathbf{n}^j \rho_l^i - \mathbf{p}^i n_l^j) \frac{t_l t_j}{\sigma} + \mathbf{z}^k t_k$$

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#### Summary

- Finite generation
- The Farkas Lemma (separation)
- Pointed and Full-Dimensional polyhedra are "non-degenerate"

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#### **Further Considerations**

- Better algorithms (Dual-Description)
- Other interpretations...
  - ...Systems of logical deduction: Hahn-Banach vs Incompleteness
  - ...Systems where "lift and drop" creates a dual representation