

# Polyhedral Computations

## An Introduction to Porta and polymake

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00A1, 00A2: Lecture and Exercises

Block Course: “Combinatorial Optimization at Work”

Zuse Institute Berlin



This is an **integrated lecture and exercise**.

## Goals:

- ▷ Repetition of polyhedral theory
- ▷ Introduction to polyhedral combinatorics
- ▷ Usage of software
  - ▶ Porta
  - ▶ polymake
  - ▶ JavaView
  - ▶ (Zimpl)
- ▷ Preparation for forthcoming lectures

## Credits for the software we use:

▷ **Porta**

Thomas Christof, Andreas Löbel (ZIB)  
originally developed at U Heidelberg, maintained at ZIB

▷ **polymake**

Ewgenij Gawrilow (TU Berlin) and  
Michael Joswig (TU Darmstadt)  
developed at TU Berlin

▷ **JavaView**

Konrad Polthier, Klaus Hildebrandt, Eike Preuss, and  
Ulrich Reitebuch  
originally developed at TU Berlin – now ZIB

▷ **Zimpl**

Thorsten Koch (ZIB)

- 1 Basic Definitions
- 2 Visualization of Polyhedra
- 3 Polyhedral Combinatorics
- 4 Simplex Algorithm
- 5 V- and H-Description
- 6 Stable Set Polytopes

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▷ **Polyhedron:**

Intersection of finitely many halfspaces:

$$P = \{x \in \mathbb{R}^d : Ax \leq b\}, \text{ where } A \in \mathbb{R}^{m \times d}, b \in \mathbb{R}^m.$$

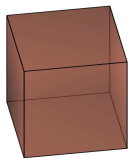
Halfspaces:

$$\{x \in \mathbb{R}^d : (a^i)^T x \leq b_i\}, \quad a^i = i\text{th row of } A.$$

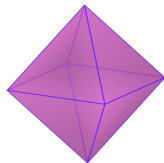
▷ **Polytope:**

bounded polyhedron

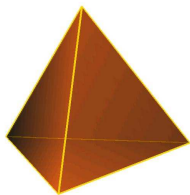
# Examples: Platonic Solids



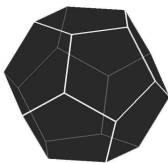
Cube



Octahedron



Tetrahedron



Dodecahedron



Icosahedron

Some easy examples:

▷  $\mathbb{R}^d$

▷  $\emptyset$

▷  $\mathbb{R}_+^d := \{x \in \mathbb{R}^d : x \geq 0\}$

▷ For  $v \in \mathbb{R}^d$ :

$$\{v\} = \{x \in \mathbb{R}^d : x_i \leq v_i, -x_i \leq -v_i, i = 1, \dots, d\}$$



Some easy examples:

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▷  $\mathbb{R}_+^d := \{x \in \mathbb{R}^d : x \geq 0\}$

▷ For  $v \in \mathbb{R}^d$ :

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Affine space:

$S \subseteq \mathbb{R}^d$  is affine if there exists  $A \in \mathbb{R}^{m \times d}$ ,  $b \in \mathbb{R}^m$  such that

$$S = \{x \in \mathbb{R}^d : Ax = b\},$$

i.e.,  $S$  is the solution of a linear equation system.

As polyhedron:  $S = \{x \in \mathbb{R}^d : Ax \leq b, -Ax \leq -b\}$ .

$P$  polyhedron in  $\mathbb{R}^d$ .

▷ **Valid Inequalities:**

inequality  $c^T x \leq c_0$ ,  $c \in \mathbb{R}^d$ ,  $c_0 \in \mathbb{R}$ , is **valid** if

$$P \subseteq \{x \in \mathbb{R}^d : c^T x \leq c_0\}.$$

▷ **Face of polyhedron  $P$ :**

$$F = \{x \in P : c^T x = c_0\},$$

for valid inequality  $c^T x \leq c_0$ .

$P$  polyhedron in  $\mathbb{R}^d$ .

▷ **Valid Inequalities:**

inequality  $c^T x \leq c_0$ ,  $c \in \mathbb{R}^d$ ,  $c_0 \in \mathbb{R}$ , is **valid** if

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▷ **Face of polyhedron  $P$ :**

$$F = \{x \in P : c^T x = c_0\},$$

for valid inequality  $c^T x \leq c_0$ .

▷ **Examples:**

▶  $P$  (take  $0^T x \leq 0$ )

▶  $\emptyset$  (take  $0^T x \leq -1$ )

“trivial faces”.

- ▷ Affine hull:

$$\text{aff}(S) = \cap \{S' : S' \supseteq S, S' \text{ affine space}\}.$$

- ▷ Dimension of  $S \subseteq \mathbb{R}^d$ :

$$\dim S = \dim \text{aff}(S).$$

Let  $P$  be a polyhedron in  $\mathbb{R}^d$ .

Face  $F$  is a

- ▷ **vertex** if  $\dim F = 0$
- ▷ **edge** if  $\dim F = 1$
- ▷ **ridge** if  $\dim F = \dim(P) - 2$
- ▷ **facet** if  $\dim F = \dim(P) - 1$

## Theorem

Every *polytope*  $P \subset \mathbb{R}^d$  is the convex hull of finitely many vectors  $v_1, \dots, v_k \in \mathbb{R}^d$ , i.e.,  $P = \text{conv}\{v_1, \dots, v_k\}$ .

Here

$$\begin{aligned} \text{conv}\{v_1, \dots, v_k\} = \{x \in \mathbb{R}^d : x = \lambda_1 v_1 + \dots + \lambda_k v_k, \\ \lambda_1 + \dots + \lambda_k = 1, \\ \lambda_1, \dots, \lambda_k \geq 0\}. \end{aligned}$$

## Theorem

Every *polyhedron*  $P \subseteq \mathbb{R}^d$  can be written as

$$P = \text{conv}\{v_1, \dots, v_k\} + \text{cone}\{r_1, \dots, r_\ell\}$$

with points  $v_1, \dots, v_k \in \mathbb{R}^d$  and rays  $r_1, \dots, r_\ell \in \mathbb{R}^d$ .

Here

$$\text{cone}\{r_1, \dots, r_\ell\} = \{x \in \mathbb{R}^d : x = \mu_1 r_1 + \dots + \mu_\ell r_\ell, \\ \mu_1, \dots, \mu_\ell \geq 0\}.$$

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# Exercise 1: Illustration of Definitions

With `polymake` ▷ ...



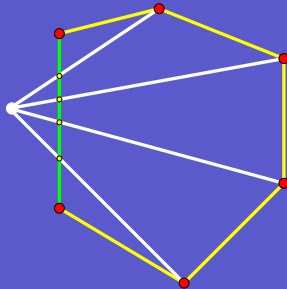
# Exercise 1: Illustration of Definitions

- ▷ Start JavaView.
- ▷ Delete geometry: **Inspector**→**Display**, “Clean all”.
- ▷ Load file cube3.jvx: **File**→**Open**→**Browse Disk**.
- ▷ Keys:
  - ▶ o: rotate
  - ▶ t: translate
  - ▶ s: scale
  - ▶ Shift-z: z-buffer on/off
- ▷ Open **Inspector**→**Geometry**→**Material**.
- ▷ Turn off facets (“Element”).
- ▷ Change size of vertices and edges.
- ▷ Turn on facets; turn on Transparency, use slider below.
- ▷ Delete geometry. Try tetrahedron.jvx, octahedron.jvx, dodecahedron.jvx, icosahedron.jvx.

# Schlegel Diagrams

One visualization technique are **Schlegel diagrams**:  
Projection on a facet reduces dimension by 1.

Example:



Same idea for 3- and 4-dimensional polytopes.

## Exercise 2: Schlegel Diagrams

- ▷ Start JavaView.
- ▷ Delete geometry: **Inspector**→**Display**, “Clean all”.
- ▷ Load file cube3.jvx: **File**→**Open**→**Browse Disk**.
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- ▷ Turn off facets (“Element”).
- ▷ Change size of vertices and edges.
- ▷ Rotate until you “see everything through a facet”.

## Exercise 2: Schlegel Diagrams

- ▷ Start JavaView.
- ▷ Delete geometry: **Inspector**→**Display**, “Clean all”.
- ▷ Load file cube3.jvx: **File**→**Open**→**Browse Disk**.
- ▷ Open **Inspector**→**Geometry**→**Material**.
- ▷ Turn off facets (“Element”).
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- ▷ Delete geometry. Load cube4.jvx.
- ▷ Delete geometry. Load 24-cell.jvx or cross4.jvx.

## Exercise 2: Schlegel Diagrams

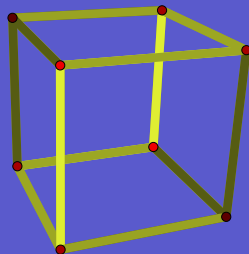
- ▷ Start JavaView.
- ▷ Delete geometry: **Inspector**→**Display**, “Clean all”.
- ▷ Load file cube3.jvx: **File**→**Open**→**Browse Disk**.
- ▷ Open **Inspector**→**Geometry**→**Material**.
- ▷ Turn off facets (“Element”).
- ▷ Change size of vertices and edges.
- ▷ Rotate until you “see everything through a facet”.
  
- ▷ Delete geometry. Load cube4.jvx.
- ▷ Delete geometry. Load 24-cell.jvx or cross4.jvx.
  
- ▷ Load cube4-e.jvx/cross4-e.jvx/24-cell-e.jvx.
- ▷ **Method**→**Effect**→**Explode Group of Geometries**

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$d$ -cube:

$$\begin{aligned} C_d &:= \{x \in \mathbb{R}^d : -1 \leq x_i \leq 1, i = 1, \dots, d\} \\ &= \text{conv}\{\{-1, 1\}^d\} \end{aligned}$$

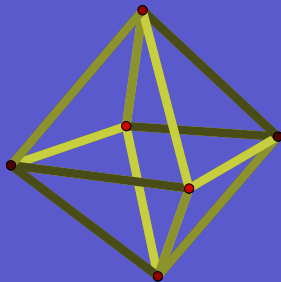
- ▷ 1 face of dimension  $d$
- ▷  $2 \cdot d$  facets
- ▷  $d \cdot 2^{d-1}$  edges
- ▷  $2^d$  vertices
- ▷ 1 face of dimension  $-1$



$d$ -crosspolytope:

$$\begin{aligned} C_d^\Delta &:= \text{conv}\{e_1, -e_1, \dots, e_d, -e_d\} \\ &= \{x \in \mathbb{R}^d : a^\top x \leq 1, a \in \{-1, 1\}^d\} \end{aligned}$$

- ▷ 1 face of dimension  $d$
- ▷  $2^d$  facets
- ▷  $d \cdot 2^{d-1}$  ridges
- ▷  $2 \cdot d$  vertices
- ▷ 1 face of dimension  $-1$





How many facets can a polytope with  $n$  vertices have?

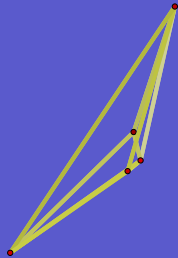
Theorem (McMullen 1970)

*The maximum number of facets that a  $d$ -polytope with  $n$  vertices can have is at most*

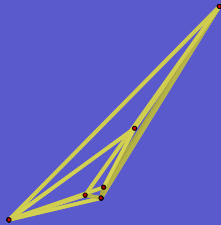
$$\sum_{i=0}^{\lfloor \frac{d}{2} \rfloor} 2 \binom{n-d-1+i}{i} \approx c \cdot n^{\lfloor \frac{d}{2} \rfloor} \quad (d \text{ even}).$$

The upper bound is attained by **cyclic polytopes**.

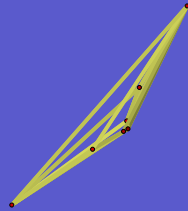
Can be explicitly constructed. Examples for  $d = 3$ :



$n = 5$



$n = 6$



$n = 7$

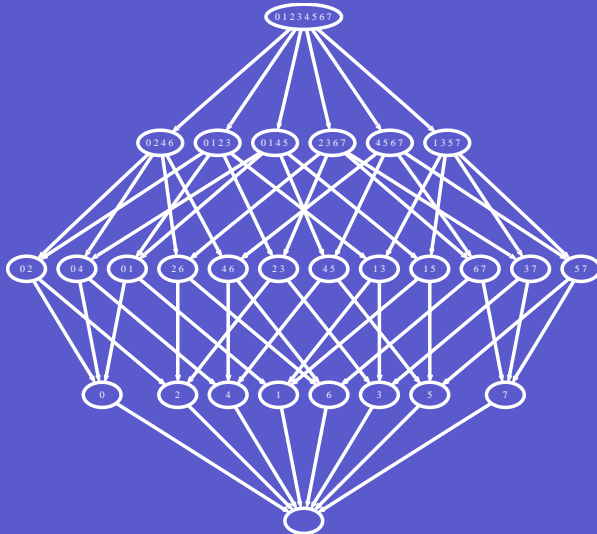
For  $d \geq 4$ , cyclic polytopes are **neighborly**, i.e., every two vertices are connected by an edge.

## Hasse diagram of polytope $P$ :

- ▷ directed graph
- ▷ nodes = faces of  $P$
- ▷ arc  $(G, F)$  if  $F \subseteq G$  and  $\dim(F) = \dim(G) - 1$
- ▷ has  $\dim(P) + 1$  many layers

# Example of Hasse Diagram

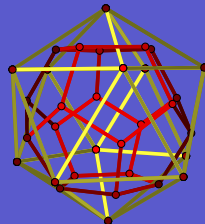
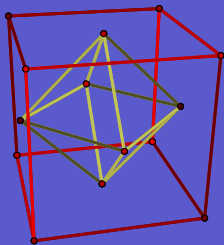
Hasse diagram of the 3-cube:



## Definition

Polytopes  $P$  and  $Q$  are **dual polytopes**, if the Hasse diagram of  $P$  is isomorphic to the Hasse diagram of  $Q$  turned upside down.

$k$ -faces  $\leftrightarrow$   $(d - k - 1)$ -faces  
vertices  $\leftrightarrow$  facets  
edges  $\leftrightarrow$  ridges

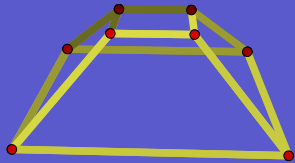
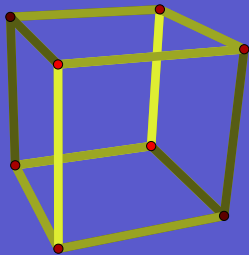


# Combinatorial Equivalence

## Definition

Two polytopes are **combinatorially equivalent** if their Hasse diagrams are isomorphic.

Isomorphism: re-order nodes in each layer such that the resulting graphs are the same.



1. A polytope  $P$  is a **0/1-polytope** if its vertices have coordinates 0 and 1 only.

Find all 3-dimensional 0/1-polytopes that are not combinatorially equivalent.

2. *Voluntary:*

Find all 3-dimensional polytopes with 4, 5, and 6 vertices that are not combinatorially equivalent.

3. Let  $P$  be a 3-polytope and  $v$  be a vertex of  $P$ .

Show that the number of facets equals the number of edges containing  $v$ .

Show or provide a counterexample for the claim in higher dimensions.

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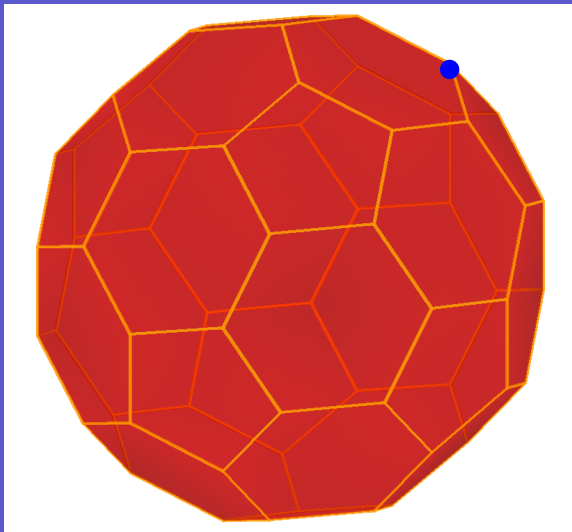
## Linear programming problem:

- ▷ Given:  $A \in \mathbb{R}^{m \times n}$ ,  $b \in \mathbb{R}^m$ ,  $c \in \mathbb{R}^n$
- ▷ Find:  $x^* \in P := \{x : Ax \leq b\}$  maximizing  $c^T x^*$

## Useful facts:

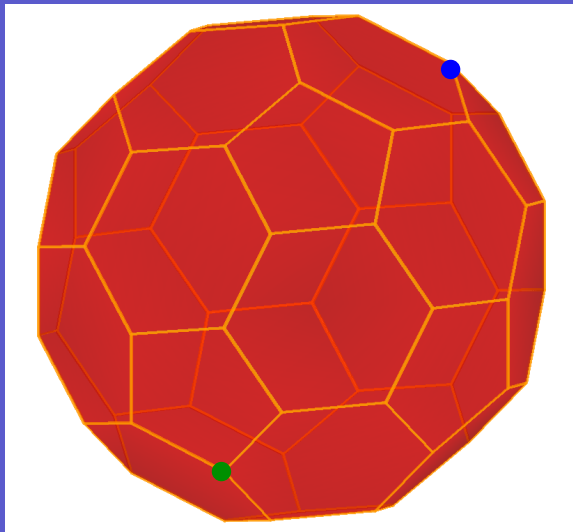
- ▷ Optimum is attained at **vertex** of  $P$ .
- ▷ Suffices to go from vertex to vertex  $\rightsquigarrow$  **Simplex-Algorithm**.

# Geometric View on the Simplex Algorithm



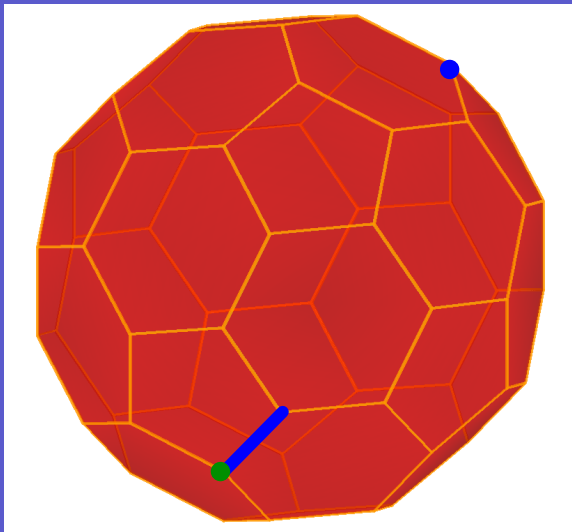
Objective function oriented to right top

# Geometric View on the Simplex Algorithm



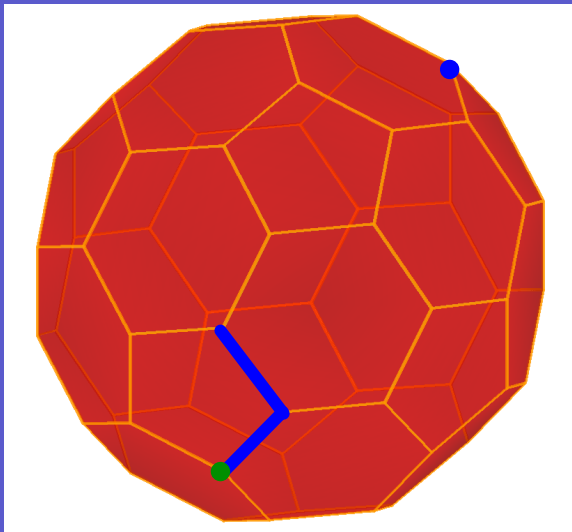
Start at vertex

# Geometric View on the Simplex Algorithm



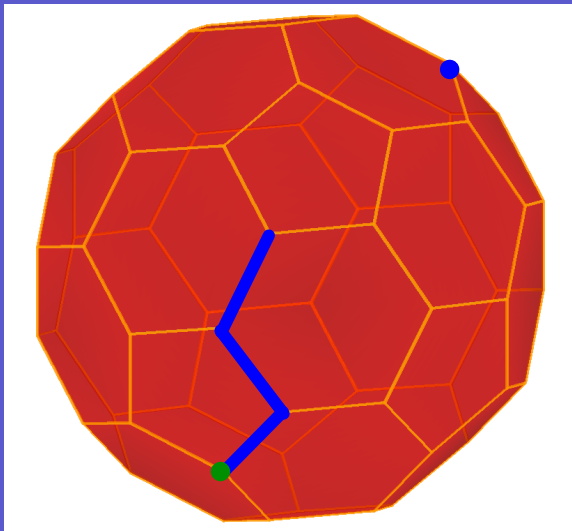
Walk along improving edge

# Geometric View on the Simplex Algorithm



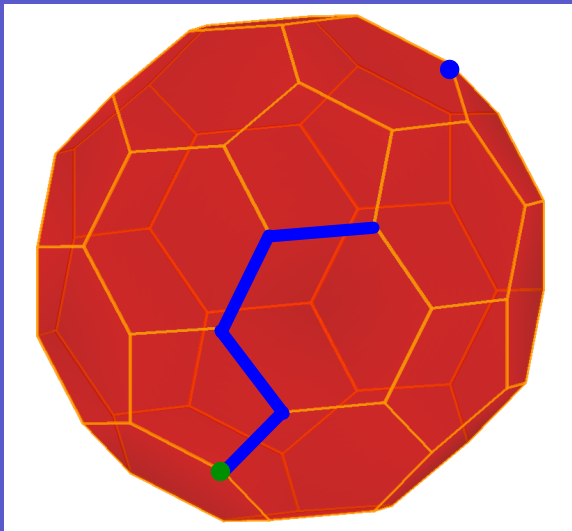
Keep walking as long as we can improve

# Geometric View on the Simplex Algorithm



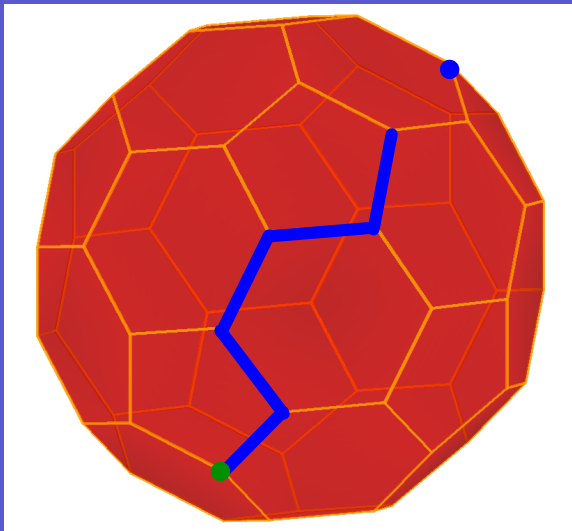
Keep walking as long as we can improve

# Geometric View on the Simplex Algorithm



Keep walking as long as we can improve

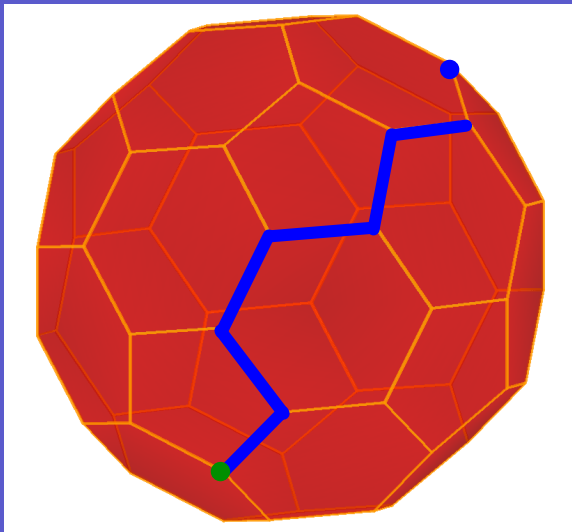
# Geometric View on the Simplex Algorithm



Keep walking as long as we can improve

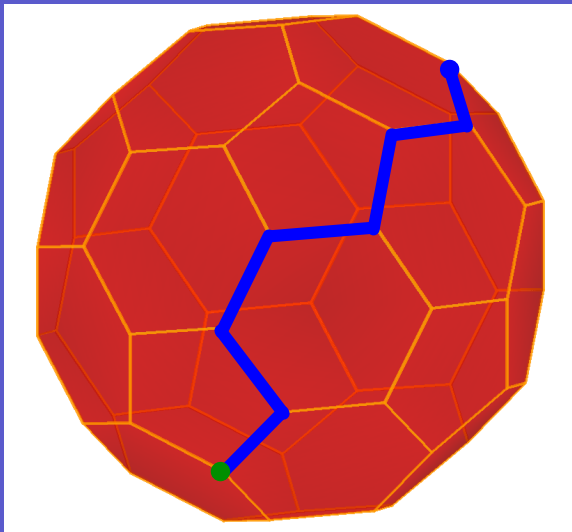


# Geometric View on the Simplex Algorithm



Keep walking as long as we can improve

# Geometric View on the Simplex Algorithm



Keep walking as long as we can improve

More examples:

- ▷ `soccerball.jvx`
- ▷ `Schrijver_KleeMinty.3.jvx`

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For most pivot rules:

Exist examples in which **simplex algorithm behaves bad**,  
i.e., takes very (exponentially) long path.

More examples:

- ▷ soccerball.jvx
- ▷ Schrijver\_KleeMinty.3.jvx

For most pivot rules:

Exist examples in which **simplex algorithm behaves bad**,  
i.e., takes very (exponentially) long path.

On the other hand:

Conjecture (Hirsch 1957)

*The shortest path between any two vertices in a  $d$ -polytope with  $m$  facets is at most  $m - d$ .*

unsolved

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Recall the main theorem:

## Theorem

*Every polytope  $P$  can be described by an **H-description**  $P = \{x : Ax \leq b\}$  or by a **V-description**  $P = \text{conv}\{v_1, \dots, v_k\}$ .*

# Conversion between Descriptions

Given,  $v_1, \dots, v_k \in \mathbb{R}^d$ , describe  $P = \text{conv}\{v_1, \dots, v_k\}$  by inequalities:

$$x = \lambda_1 v_1 + \dots + \lambda_k v_k$$

$$1 = \lambda_1 + \dots + \lambda_k$$

$$0 \leq \lambda_1, \dots, \lambda_k$$

(the variables are  $x$  and  $\lambda_1, \dots, \lambda_k$ ).

Then project out  $\lambda_1, \dots, \lambda_k$  by Fourier-Motzkin Elimination.



**Polarity** = exchange roles of points and inequalities.

## Definition

The **polar**  $P^\Delta$  of a polytope  $P$  is:

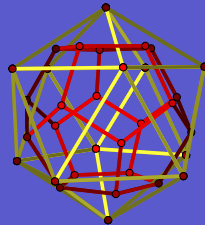
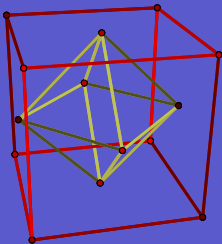
$$P^\Delta = \{c \in \mathbb{R}^d : c^T x \leq 1, \text{ for all } x \in P\}.$$

## Proposition

1. *If  $P = \text{conv}\{v_1, \dots, v_m\}$ ,  $0$  is in the interior of  $P$ , and  $V = [v_1, \dots, v_m]$ , then  $P^\Delta = \{x \in \mathbb{R}^d : V^T x \leq \mathbf{1}\}$ .*
2. *If  $P = \{x \in \mathbb{R}^d : Ax \leq \mathbf{1}\}$ , then  $P^\Delta = \text{conv}\{A_1, \dots, A_m\}$ , where  $A_i$  is the  $i$ th column of  $A$ .*

# Consequences from Polarity

- ▷ One can prove:  $P^{\Delta\Delta} = P$ , if  $0 \in P$ .
- ▷ If we can convert from  $V$  to  $H$ -description, we can convert in the other direction as well:  
**Exchange roles of inequalities and points.**
- ▷ It turns out that  $P^\Delta$  is combinatorially equivalent to “the” dual of  $P$ .



# Fourier-Motzkin Elimination

**Example:** 3-dimensional cube:

```
DIM = 11
ELIMINATION_ORDER
1 2 3 4 5 6 7 8 0 0 0
INEQUALITIES_SECTION
1 x1 -1 x2 +1 x3 +1 x4 -1 x5 +1 x6 -1 x7 -1 x8 - x9 == 0
1 x1 +1 x2 -1 x3 +1 x4 -1 x5 -1 x6 +1 x7 -1 x8 - x10 == 0
1 x1 +1 x2 +1 x3 -1 x4 +1 x5 -1 x6 -1 x7 -1 x8 - x11 == 0
x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 == 1
x1 >= 0
x2 >= 0
x3 >= 0
x4 >= 0
x5 >= 0
x6 >= 0
x7 >= 0
x8 >= 0
END
```

# Exercise 4: Fourier-Motzkin Elimination

1. Choose your favorite polytope.
2. Generate a file `filename.ieq` and insert data as above.
3. Call: `fmel filename.ieq`
4. Investigate output `filename.ieq.poi`.

- ▷ We know that the output can be exponentially large.
- ▷ Therefore hope for **output polynomial algorithm**: runs in polynomial time in the input and output.

## Open

- ▷ Exist examples in which Fourier-Motzkin elimination generates exponentially many inequalities in between.
- ▷ Exist many other algorithms, but almost all of them are known to have bad examples.
- ▷ It is in general hard to only generate vertices of an unbounded polyhedron (Khachiyan et. al [2005]).

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# Combinatorial Optimization Problems

Combinatorial Optimization problem:

Given:

- ▷ finite set  $S$  (groundset)
- ▷ set  $\mathcal{F} \subseteq 2^S$  (feasible solutions)
- ▷ weight  $w_s \in \mathbb{R}$  for each  $s \in S$

Goal:

Find  $S' \in \mathcal{F}$  with maximal weight  $w(S')$ .

$$w(S') := \sum_{s \in S'} w_s \quad \text{for } S' \subseteq S.$$

## Example: Stable Set Problem

### Given:

- ▷ graph  $G = (V, E)$
- ▷ groundset =  $V$
- ▷  $U \subseteq V$  is feasible if  
no two vertices  $u, v \in U$  are connected by an edge.  
 $U$  is called **stable** or **independent set**.
- ▷  $w_v = 1$  for all  $v \in V$ .

### Goal:

Find stable set  $U$  of maximal cardinality.



Integer programming formulation:

$$\begin{array}{ll} \max & \mathbb{1}^T x \\ \text{s.t.} & x_u + x_v \leq 1 \quad \text{for all edges } \{u, v\} \in E \\ & x \in \{0, 1\}^n. \end{array}$$

Will use LP relaxation:

$$\begin{array}{ll} \max & \mathbb{1}^T x \\ \text{s.t.} & x_u + x_v \leq 1 \quad \text{for all edges } \{u, v\} \in E \\ & Ax \leq b \quad \text{add. inequalities} \\ & x \in [0, 1]^n. \end{array}$$

# Geometry and Combinatorial Optimization

Explain general approach to combinatorial optimization problems for the stable set problem.

Study **stable set polytope**:

$$\begin{aligned} P_S &:= \text{conv}\{\chi(U) : U \text{ is a stable set}\} \\ &= \text{conv}\{x \in \{0, 1\}^n : x_u + x_v \leq 1, \text{ for all } \{u, v\} \in E\}, \end{aligned}$$

where  $\chi(U)$  is the **incidence vector** of  $U$ :

$$\chi(U)_i = \begin{cases} 1 & \text{if } i \in U \\ 0 & \text{otherwise.} \end{cases}$$

# Geometry and Combinatorial Optimization II

“Recipe” to solve the stable set problem:

1. Find classes of facets of polytope  $P_S$ .

2. For a class of facets:

Construct (fast) **separation algorithm**:

Given  $x^*$ , does there exist facet defined by  $c^T x \leq d$  such that  $c^T x^* > d$  (inequality is violated).

3. Start cutting plane algorithm:

3.1 Find solution  $x^*$  of LP relaxation.

3.2 If  $x^*$  is integer  $\rightarrow$  **optimal solution found!**

3.3 Solve separation problems for classes of facets.

3.4 Add violated inequalities to the LP relaxation.

Goto 3.1.

# Exercise 5: Finding Facets – Part I

1. Take small graph of your choice.
2. Find all stable sets.
3. Put corresponding incidence vectors into file.  
Format (filename ending with “.poi”):

```
DIM = <n>  
CONV_SECTION  
<x1> <x2> . . . <xn>  
. . .  
CONE_SECTION  
END
```

4. Compute facets: `traf <file>.poi`.
5. Analyze result. Can you explain the facets?

## Exercise 5: Finding Facets – Part II

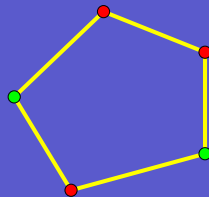
1. Write ZIMPL program, which reads any graph.  
Hint: Write graph in file: put one edge in each line  
`u v`  
Use  
`set E := read FILE as "<1n,2n>";`  
to read file. Call `zimpl -D FILE="..."`.
2. For many graphs, generate LP format output.
3. Use `lp2porta` to generate Porta output.
4. Use `vint` to generate all integer points that satisfy the given inequalities.
5. Compute facets.
6. Analyze result. Can you explain the facets?  
Can you find a separation algorithm? (May be hard!)

# “Solution”: Famous Classes

1. **Nonnegativity constraints:**  $x_u \geq 0$
2. **Odd hole inequalities:**

$$x(S) \leq \frac{|S| - 1}{2}$$

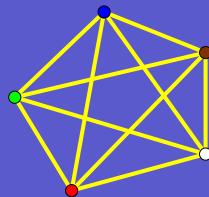
$S$  = vertices of odd cycle without chord.



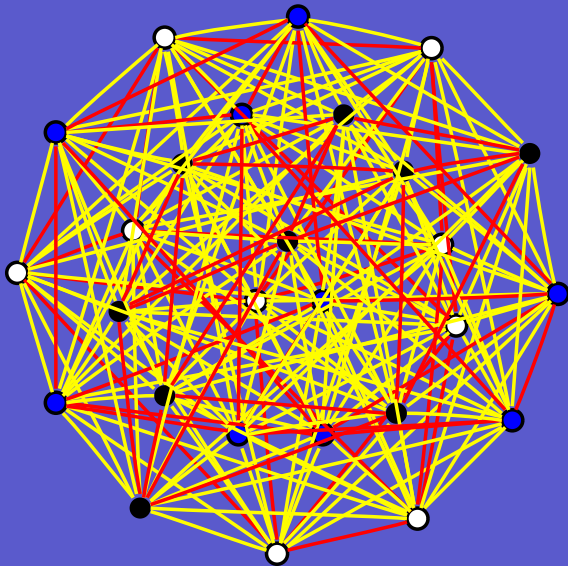
3. **Clique inequalities:**

$$x(S) \leq 1$$

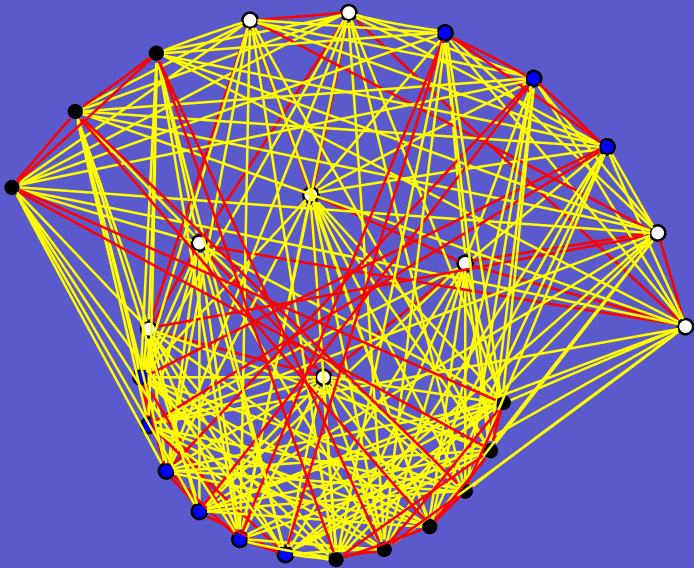
$S$  = vertices of a clique.



# Schläfli Graph

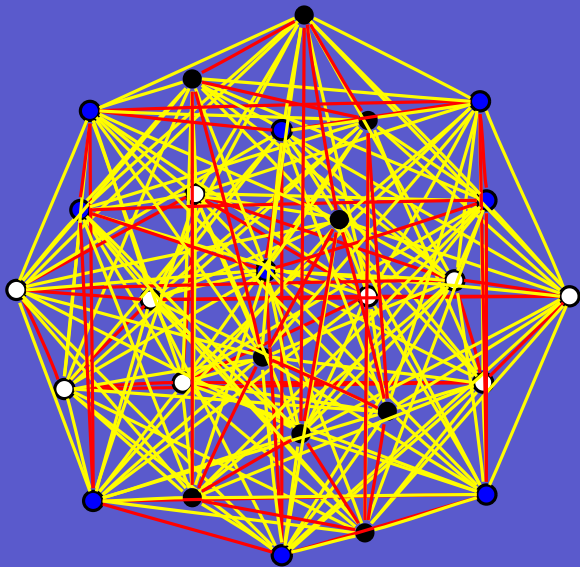


# Schläfli Graph





# Schläfli Graph



# Stable Set Polytope for the Schläfli Graph

- ▷ The Schläfli graph has 27 nodes and 216 edges.
- ▷ The Schläfli graph has 208 stable sets.
- ▷  $P_S$  has 4086 facets with 7 classes.
- ▷ Porta generated 527,962 inequalities in total.
- ▷ Maximal number of inequalities in one iteration: 14230.

1. 27 nonnegativity constraints
2. 215 5-clique inequalities, 72 6-clique inequalities  
 $x_{16} + x_{17} + x_{18} + x_{21} + x_{27} \leq 1$
3. 216 inequalities with 12 non-zero coefficients (two 2's and ten 1's) and right-hand side 2  
 $x_7 + x_8 + x_9 + x_{13} + x_{14} + x_{15} + x_{16} + x_{17} + x_{18} + 2x_{25} + x_{26} + 2x_{27} \leq 2$
4. 1080 inequalities with 14 non-zero coefficients (all 1's) and right-hand side 2  
 $x_7 + x_8 + x_9 + x_{13} + x_{14} + x_{15} + x_{16} + x_{17} + x_{18} + x_{22} + x_{24} + x_{25} + x_{26} + x_{27} \leq 2$
5. 216 inequalities with 15 non-zero coefficients (all 1's) and right-hand side 2  
 $x_7 + x_8 + x_9 + x_{13} + x_{14} + x_{15} + x_{16} + x_{17} + x_{18} + x_{19} + x_{21} + x_{22} + x_{24} + x_{25} + x_{27} \leq 2$
6. 27 inequalities with 17 non-zero coefficients (sixteen 1's and one 2) and right-hand side 2  
 $x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{13} + x_{14} + x_{15} + x_{16} + x_{17} + x_{18} + x_{21} + x_{24} + x_{25} + x_{26} + 2x_{27} \leq 2$
7. 2232 inequalities with 21 non-zero coefficients (sixteen 1's and five 2's) and right-hand side 3  
 $2x_7 + 2x_8 + 2x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14} + x_{15} + x_{16} + x_{17} + x_{18} + x_{19} + x_{21} + x_{22} + x_{24} + 2x_{25} + 2x_{27} \leq 3$

- ▷ Günter M. Ziegler  
*Lectures on Polytopes*  
Springer-Verlag, revised edition 1998.
- ▷ Branko Grünbaum  
*Convex Polytopes*  
Edited by V. Kaibel, V. Klee, G. M. Ziegler,  
Springer-Verlag, 2004
- ▷ Alexander Schrijver  
*Theory of Linear and Integer Programming*  
John Wiley & Sons, 1986
- ▷ Martin Grötschel, László Lovász, Alexander Schrijver  
*Geometric Algorithms and Combinatorial Optimization*  
Springer-Verlag, 2nd ed., 1993

# Solution Exercise 3 – Part I

$n = 8$ :



$n = 7$ :



$n = 6$ :



$n = 5$ :



the first two are equivalent.



$n = 4$ : all equivalent

# Solution to Exercise 3 – Part II

