

# Optimal Solutions of Oracle Defined Optimization Problems

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# Outline

- 1 Linear vector optimization revisited
- 2 The Separation Oracle
- 3 Vertex enumeration
- 4 Alternative approach: obtaining the vertices directly

# Vector optimization problem – our version

- The **problem space** is  $\mathbb{R}^n$ ;  $n \approx 100$ .
- The linear **constraints** are given by the  $m \times n$  matrix  $A \in \mathbb{R}^{m \times n}$ ;  $m \approx 1000$ .
- The **feasible set** of the problem (to be optimized over) is

$$\{x \in \mathbb{R}^n : Ax \geq 0\}.$$

- There are  $p$  **objectives** to be **minimized**, given by the  $p \times n$  matrix  $P \in \mathbb{R}^{p \times n}$ ;  $p \approx 10$ .
- The **objective space** is  $\mathbb{R}^p$ .
- The **optimization problem** is to find all extremal solutions of

$$\min \{ Px \in \mathbb{R}^p : x \in \mathbb{R}^n, Ax \geq 0 \}.$$

⇒

Observe:  $\{ Px : Ax \geq 0 \}$  is a **projection** of this polyhedron:

# Vector optimization problem – our version

⇒ Vector optimization problem:

$$\min \{ Px \in \mathbb{R}^p : x \in \mathbb{R}^n, Ax \geq 0 \}. \quad (*)$$

Define

$$Q = \{ Px + z \in \mathbb{R}^p : \langle x, z \rangle \in \mathbb{R}^{n+p}, z \geq 0, Ax \geq 0 \};$$

this is a  $p$ -dimensional affine projection of the  $n + p$ -dimensional polyhedron

$$S = \{ \langle x, z \rangle \in \mathbb{R}^{n+p} : (A, I)\langle x, z \rangle \geq 0 \}.$$

## Proposition\*

*Under some mild assumptions, the extremal solutions of the vector optimization problem (\*) are exactly the vertices of the  $p$ -dimensional polyhedron  $Q$ .* □

\*See also A. Löhne: *Projection of polyhedral cones*, ArXiv 1406.1708

## ~~Vector optimization~~ Vertex enumeration problem

Moving  $z$  to the problem space along with  $x$ , and increasing  $n$  by  $p$ , solving the vector optimization problem reduces to

### Vertex enumeration problem for projections

**Given** the  $m$  facets of the  $n$ -dimensional polyhedron  $\mathcal{S}$ , that is,  $\mathcal{S} = \{x \in \mathbb{R}^n : Ax \geq 0\}$  for a given  $m \times n$  matrix  $A$ ,  
**enumerate** all vertices of its  $p$ -dimensional projection  
 $\mathcal{Q} = \{Px : x \in \mathcal{S}\}$  for a given  $p \times n$  matrix  $P$ .

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Some remarks:

- The same reduction works for arbitrary ordering cone  $C$  (not only for the non-negative orthant  $C = \mathbb{R}_+^p$ ).
- $\mathcal{Q}$  is not bounded, has vertices **and** extremal rays (typically coming from the ordering cone). Both cases can be handled uniformly by using **homogeneous** coordinates.

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# An important fact on scalar linear programming

## Observation on duality

When solving a scalar LP problem, we not only get the **solution**, but also a **proof** of the correctness of the solution (dual).

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## Example:

$$Q = \{Px : x \in \mathbb{R}^n, Ax \geq 0\}.$$

- ⇒ Given points  $i \in Q$  and  $o \notin Q$ , find the intersection  $[i, o] \cap Q$ . (This problem is from the inner loop of the *outer approximation algorithm*.) To get the boundary point of  $Q$  on  $[i, o]$ ,
- ⇒ Solve the scalar LP  $\max_{\lambda, x} \{\lambda : i + \lambda(o - i) = Px, Ax \geq 0, \lambda \geq 0\}$ .
- ⇒ The solution gives the boundary point  $b = i + \hat{\lambda}(o - i) \in Q$  (primal), **and** a supporting hyperplane to  $Q$  at  $b$  proving that  $b$  is indeed optimal (coming from the dual).

# Introducing the Separation Oracle

Distilling from the example: the polyhedron  $Q \subseteq \mathbb{R}^p$  is hidden, and can only be reached through questions to the

## Facet Separation Oracle, *FSO*

Q: a point  $y \in \mathbb{R}^p$ .

A: if  $y \in Q$ , then the answer is **inside**.

if  $y \notin Q$ , then the answer is (the equation of) a **facet** of  $Q$  such that  $y$  is on the negative side of the facet.

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## Vertex enumeration problem using *FSO*

**Given** an *FSO* for the polyhedron  $Q \subseteq \mathbb{R}^p$  (and some initial data), **enumerate** all vertices and extremal rays of  $Q$ .

Getting an answer from the Oracle is typically expensive. The main complexity measure is the number of Oracle questions.

# Introducing the Separation Oracle



# Vertex enumeration using Facet Separation Oracle

Vertex enumeration for projection (restated)

**Given** the  $m \times n$  matrix  $A$  and the  $p \times n$  matrix  $P$ ,  
**enumerate** all vertices of  $Q = \{Px : x \in \mathbb{R}^n, Ax \geq 0\} \subset \mathbb{R}^p$ .

Claim

*Solving linear multiobjective optimization can be reduced to vertex enumeration using **FSO**.*

Proof.

Implement **FSO** on the query  $y \in \mathbb{R}^p$  as follows:

- 1 Pick a random inner point  $i \in Q$  (could be fixed!).
- 2 Solve the LP  $\max_{\lambda, x} \{\lambda : i + \lambda(y - i) = Px, Ax \geq 0, \lambda \geq 0\}$ .
- 3 If  $\hat{\lambda} \geq 1$  then return “ $y$  is an inner point.”
- 4 Otherwise return the supporting hyperplane to  $Q$  as the facet.

(It will be a facet!)



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# Vertex / facet enumeration

## Classical vertex enumeration problem

**Given** all facets of a polyhedron,  
**enumerate** all of its vertices.

Several vertex enumeration algorithms are known:\*

- pivot algorithms – ranked, recursive, reversed
- incremental – double description method
- primal/dual
- backtrack algorithms

All complexity questions in this area are open.

## Research problem

Which known method adapts best to our case? Other algorithms?

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\*K. Fukuda: *Vertex enumeration for polyhedra: algorithms and open problems.*

# Outer approximation using double description

## Double Description method for vertex enumeration with *FSO*

To enumerate vertices of  $Q$  generate the *approximating sequence*  $Q_j \supseteq Q_{j+1} \supseteq \dots \supseteq Q$  maintaining in each step

- ① all vertices and facets of  $Q_j$ ,
- ② for each vertex of  $Q_j$  whether it is known to be a vertex of  $Q$ .

To get  $Q_{j+1}$  from  $Q_j$  pick a vertex  $y$  of  $Q_j$  which is **not known** to be a vertex of  $Q$ . Call the *FSO* with  $y$ .

- ① If the answer is “**inner**”, mark  $y$  as a vertex of  $Q$ .
- ② Otherwise intersect  $Q_j$  with the facet of  $Q$  returned.

Stop when all vertices of  $Q_j$  are vertices of  $Q$ ; you are done.

The number of oracle calls is

number of vertices of  $Q$  + number of facets of  $Q$ .

This seems to be optimal. But is it?

# In summary

## Vector optimization problem

Find all extremal solutions of  $\min\{Px \in \mathbb{R}^p : x \in \mathbb{R}^n, Ax \geq 0\}$ .



## Vertex enumeration problem using *FSO*

**Given** an Facet Separation Oracle representation of  $Q \subseteq \mathbb{R}^p$ ,  
**enumerate** all vertices (and extremal rays) of  $Q$ .



## Double Description method for vertex enumeration with *FSO*

Use the outer approximation method, learning in each step

- either a new vertex of  $Q$ ,
- or a new facet of  $Q$ .

until the whole  $Q$  is known.

# It works!

The algorithm was used successfully for combinatorial optimization problems with  $p = 10$  (**ten!**) objectives. Some representative results:

$m$	$n$	Vertices	Facets	Time
4055	370	19	58	1:10:10
4009	370	40	103	3:24:37
3891	358	30	102	3:34:31
3963	362	167	235	9:20:19
4007	370	318	356	13:20:08
4007	370	318	356	14:34:42
4007	370	297	648	22:02:39
3913	362	779	1269	37:15:33
3987	362	4510	7966	427:43:30
3893	362	10387	13397	716:36:32

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# Vertex Separation Oracle

Suppose the projection  $Q$  can be reached by inquiring the

## Vertex Separation Oracle, $VSO$

Q: (the equation of) a closed halfspace  $H \subseteq \mathbb{R}^p$ .

A: if  $Q \subseteq H$  then the answer is **inside**.

if  $Q \not\subseteq H$ , then the answer is a **vertex** of  $Q$  not in  $H$ .

Accessing  $Q$  through a Vertex Separation Oracle  $VSO$  has some attractive properties:

- If the computation has to be aborted, we have several vertices (extremal solutions).
- The order of half-spaces submitted to the Oracle can be chosen (prioritized) so as to give a general view of the Pareto set, and refine it where necessary.

# Inner approximation using double description

## Double Description method for vertex enumeration with *VSO*

To enumerate the vertices of  $Q$  generate the *approximating sequence*  $Q_j \subseteq Q_{j+1} \subseteq \dots \subseteq Q$  maintaining in each step

- ① all vertices and facets of  $Q_j$ ,
- ② for each *facet* of  $Q_j$  whether it is known to be a facet of  $Q$ .

To get  $Q_{j+1}$  from  $Q_j$  pick a facet  $f$  of  $Q_j$  which is not known to be a facet of  $Q$ . Call the *VSO* with the half-space  $f \geq 0$ .

- ① If the answer is **inside**, mark  $f$  as a facet of  $Q$ , and continue.
- ② Otherwise let  $Q_{j+1}$  be the convex hull of  $Q_j$  and the vertex returned.

Stop when all facets of  $Q_j$  are facets of  $Q$ ; you are done.

The number of Oracle calls this algorithm makes is also  
 number of vertices of  $Q$  + number of facets of  $Q$ .

# How to implement the Vertex Separation Oracle?

Vertex enumeration for projection (restated)

**Given** the  $m \times n$  matrix  $A$  and the  $p \times n$  matrix  $P$ ,  
**enumerate** all vertices of  $Q = \{Px : x \in \mathbb{R}^n, Ax \geq 0\} \subset \mathbb{R}^p$ .

Given the closed halfplane  $\{y \in \mathbb{R}^p : \langle b, y \rangle \geq d\}$ , the naïve idea is

- 1 Solve the LP  $\hat{d} = \min_x \{\langle b, Px \rangle : Ax \geq 0\}$ .
- 2 If  $d \leq \hat{d}$ , then return “ $Q$  is inside.”
- 3 Otherwise return the point  $\hat{y} \in Q$ , where the minimum  $\hat{d} = \langle b, \hat{y} \rangle$  is taken.

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- 3 Otherwise return the point  $\hat{y} \in Q$ , where the minimum  $\hat{d} = \langle b, \hat{y} \rangle$  is taken.

The problem is that  $\hat{y}$  is a boundary point of  $Q$ , but it is not necessarily a vertex if the normal  $b$  is not in “general position.”

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- 3 Otherwise return the point  $\hat{y} \in Q$ , where the minimum  $\hat{d} = \langle b, \hat{y} \rangle$  is taken.

A correct implementation would be to take a projective image of  $Q$  first, where the ideal hyperplane is in general position (there are others). The problem and solution closely relate to the geometric duality of F. Heyde and A. Löhne.



**Thank you for your attention**