

Background To Benders' Algorithm

- Extreme points and extreme rays [DJ, p 36]

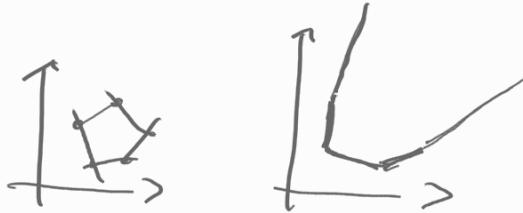
$$\min c^T x$$

$$A_1 x \leq b_1$$

$$(A_2 x \leq b_2)$$

$$x \geq 0$$

→ in general it is a polyhedron, which



can be bounded

(and hence a polytope) or unbounded

In Dantzig Wolfe dec. we rewrite the problem as

$$\min c^T x$$

$$A_1 x \leq b_1$$

$$x \in X$$

$$\text{where } X = \{x \mid A_2 x \leq b_2\}$$

If X was bounded we have seen that it can be written as a convex combination of its extreme points Φ .

$$x = l_1 \phi_1 + \dots + l_m \phi_m$$

$$X = \{x \mid x = \sum_{p \in P} \lambda_p x_p, \lambda_p \geq 0, \sum_{p \in P} \lambda_p = 1\}$$

If we want to account also for the possibility that $A_2 x \leq b_2$ is unbounded then we need the following:

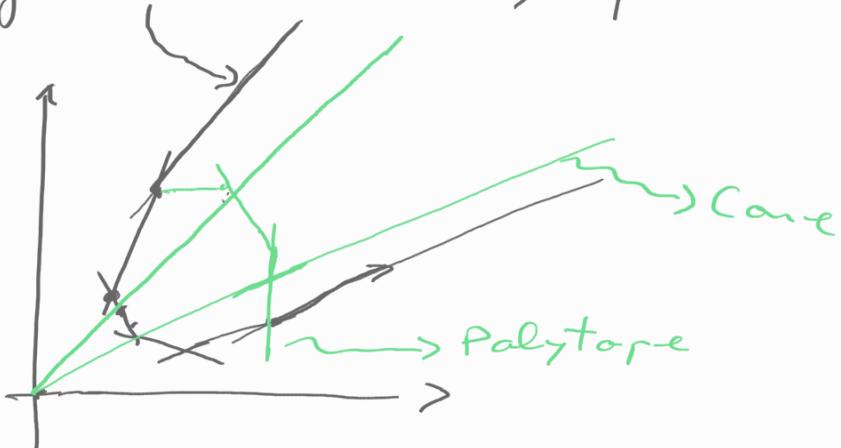
A convex cone is $\{x \mid Bx \leq 0\}$ that is the intersection of many halfspaces

A convex cone is also described by the conic combination of its extreme rays R :

$$C = \left\{ x \mid x = \sum_{r \in R} \delta_r x_r, \delta_r \geq 0 \right\}.$$


Minkowski-Weyl theorem from polyhedral analysis:

$$\text{Polyhedron} = \text{Polytope} + \text{Cone}$$



Hence, a point of a polyhedron can be described as:

$$X = \left\{ x \mid x = \sum_{p \in P} \lambda_p x_p + \sum_{r \in R} \delta_r x_r, \right. \\ \left. \begin{array}{l} \lambda_p \geq 0, \\ \sum_{p \in P} \lambda_p = 1, \\ \delta_r \geq 0 \end{array} \right\}$$

Hence in Dantzig Wolfe decomposition
the substitution would be

$$\begin{array}{ll} \min c^T x & \min c^T \sum_{p \in P} \lambda_p x_p + c^T \sum_{r \in R} \delta_r x_r \\ A_1 x \leq b_1 & A_1 \left(\sum_{p \in P} \lambda_p x_p + \sum_{r \in R} \delta_r x_r \right) \leq b_1 \\ x \in X & \sum_{p \in P} \lambda_p = 1 \\ & \lambda_p \geq 0 \quad \forall p \in P \\ & \delta_r \geq 0 \quad \forall r \in R \end{array}$$

We previously ignored the rays and can continue to do so if the pricing problem is bounded and feasible.

- How do we find the extreme rays?
From the simplex:

$$\begin{aligned} \max \quad & x_1 + x_2 \\ -2x_1 + x_2 + x_3 = 1 \end{aligned}$$

$$x_1 - x_2 + x_4 = 1$$

$$x_1, x_2, x_3, x_4 \geq 0$$

$$\left| \begin{array}{cccc|c|c} x_1 & x_2 & x_3 & x_4 & & \\ \hline -2 & 1 & 0 & 0 & 1 & \\ 1 & -1 & 0 & 1 & 0 & 1 \\ 1 & 1 & 0 & 0 & 1 & 0 \end{array} \right| \Rightarrow \left| \begin{array}{cccc|c|c} x_1 & x_2 & x_3 & x_4 & & \\ \hline -2 & 1 & 1 & 0 & 0 & 1 \\ -1 & 0 & 1 & 1 & 0 & 2 \\ 1 & 3 & 0 & -1 & 0 & 1 \\ 1 & 1 & 0 & 0 & 1 & 0 \end{array} \right|$$

↑

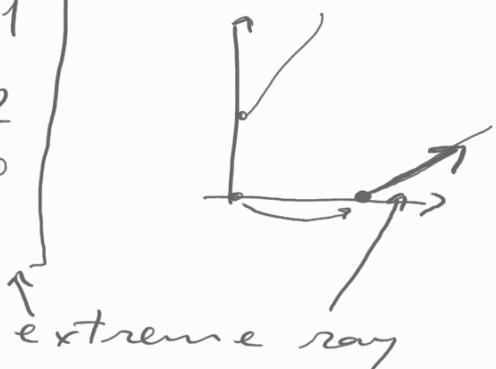
basic sol. $[0, 1, 0, 2]$. Trying to bring x_1 in basis unveils that we can increase x_1 arbitrarily

Hence:

$$x_2 = 1 + 2x_1(-x_3) \rightarrow \begin{matrix} \text{not in basis} \\ \text{stays zero} \end{matrix}$$

$$x_4 = 2 + x_1(-x_3)$$

$$x = \begin{bmatrix} u \\ 1+2u \\ 0 \\ 2+u \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 2 \end{bmatrix} + K \begin{bmatrix} 1 \\ 2 \\ 0 \\ 1 \end{bmatrix}$$



Hence, we find it with the simplex and it is returned by solvers when the problem is unbounded.

Note: if a problem is infeasible its dual is unbounded and the extreme ray is a proof of infeasibility for the primal (Farkas Lemma)

Farkas Lemma

Let $A \in \mathbb{R}^{m \times n}$ and $b \in \mathbb{R}^m$. Then:

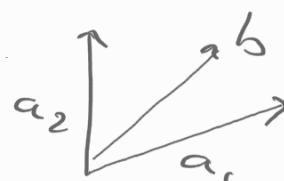
either : $\exists x \in \mathbb{R}^n$: $Ax \leq b$ and $x \geq 0$

or : $\exists u \in \mathbb{R}^m$: $u^\top A \geq 0$ and $u^\top b < 0, u \geq 0$

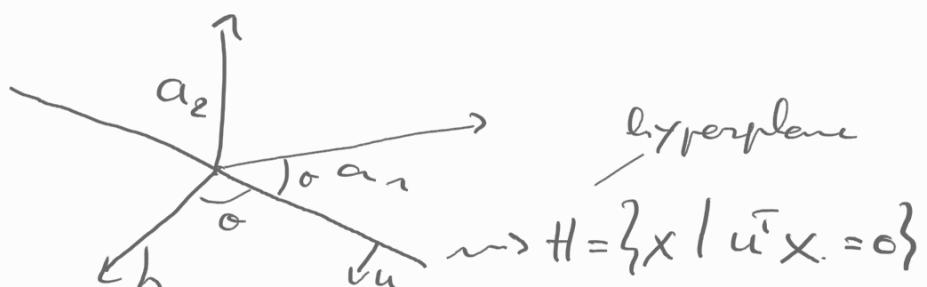
(consult in case wikipedia for variants)

Meaning:

either the system $Ax \geq b$ admits a solution $x \geq 0$, that is,
b can be attained as a linear combination of the column vectors
of A :



or the system $Ax \geq b$ does not admit a solution and there is a certificate of this fact:
an hyperplane described by $u \geq 0$
that separates b from A:



$$u^\top b = \|u\| \|b\| \cos \alpha$$

$$u^\top A = [\|u\| \|a_i\| \cos \alpha]$$

if different sign
then vector b
stays on one side
and a_i on the
other of H

Benders' reformulation

(OP)
original problem

$$\max c^T x + b^T y$$

$$F^T x + G^T y \leq d$$

$$x \in X \subseteq \mathbb{R}_+^m$$

$$y \in \mathbb{R}_+^P$$

(EF)
extended formulation

$$\max c^T x + \eta$$

$$v^T(d - F^T x) \geq 0 \quad v \in R$$

$$u^P(d - F^T x) \geq \eta \quad P \in P$$

$$x \in X$$

$$\eta \in \mathbb{R}^1$$

extreme rays of Δ_{SP}

extreme points of Δ_{SP}

Derivation of the reformulation.

Let's rewrite (OP) as:

$$Z = \max_x \{c^T x + \phi(x) : x \in X\}$$

$$\text{where } \phi(x) = \max \left\{ b^T y : G^T y \leq d - F^T x, y \in \mathbb{R}_+^P \right\}$$

Subproblem (SP)

SP is feasible \Leftrightarrow by Farkas:

$$H u \in \mathbb{R}^m : a^T u \geq 0, u \geq 0$$

$$u(d - Fx) \geq 0$$

u are the extreme rays of $u^T q \geq 0$ (cone)

Let's call them

$$v^r, r \in \mathbb{R}$$

if $\phi(x)$ exists and is finite:
by strong duality theorem of CP:

$$\phi(x) = \min_u \{ u(d - Fx) \mid u^T q \geq b, u \in \mathbb{R}_+^m \}$$

Dual Subproblem DSP

If s_P bounded and DSP bounded, then
the solution of DSP will be in one of
the extreme points by the fundamental
th. of lin. programming. Hence:

$$\phi(x) = \min_{p \in P} \{ \omega^p(d - Fx) \}$$

where $\omega^p, p \in P$ are the
extreme points of $\{u^T q \geq b, u \geq 0\}$

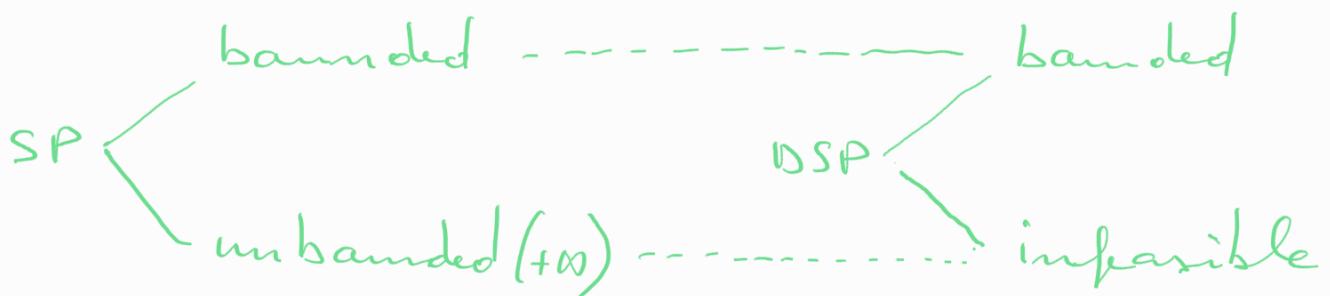
similar to $\min \{2x, 3x, 5x, 6x\}$

that can be linearized as $y \leq 2x$

$$y \leq 3x$$

$$y \leq 5x$$

$$y \leq 6x$$



After linearization this is equivalent to:

$$(EF) \quad \begin{aligned} \max_x \quad & c^T x + y \\ \text{s.t.} \quad & w^p (d - Fx) \geq y \quad \forall p \in P \\ & v^r (d - Fx) \geq 0 \quad \forall r \in R \\ & x \in X \end{aligned}$$

from def. of \bar{x} ,
ie, SP is feasible

which is the reformulation we wanted
to achieve. Hence

OP has feasible sol \Leftrightarrow EF has feasible sol

OP has unbounded sol \Leftrightarrow EF has unbounded sol

OP is infeasible \Leftrightarrow EF is infeasible

EF has too many constraints to list

then all hence, we solve it by defining it incrementally:

$$\begin{aligned}
 Z^* = \max_{\mathbf{x}} \quad & c \mathbf{x} + \eta \\
 \text{s.t.} \quad & w^r(d - F\mathbf{x}) \geq \eta \quad \forall p \in P \subseteq P \\
 (REF) \quad & v^r(d - F\mathbf{x}) \geq 0 \quad \forall r \in R \subseteq R \\
 & \mathbf{x} \in X
 \end{aligned}$$

Benders' algorithm

Solve (REF) and find (η^*, \mathbf{v}^*)

$$\begin{aligned}
 \text{Solve (DSP)} \quad \phi(\mathbf{x}^*) = \min_u \quad & u(d - F\mathbf{x}^*) \\
 & uG \geq h \\
 & u \in \mathbb{R}_+^m
 \end{aligned}$$

if unbounded then

$$\exists v^r: v^r(d - F\mathbf{x}^*) < 0 \Rightarrow v^r(d - F\mathbf{x}) \geq 0$$

must be added to make

the primal feasible (from formulation of \bar{x})

if bounded and $\phi(\mathbf{x}^*) < \eta^*$

then the solution to DSP is \mathbf{x}^*

gives a new extreme point;
whose relative const. in EF is violated

$$\phi(x^*) = \omega^P(d - Fx^*) < \eta^*$$

hence we add

$$\omega^P(d - Fx) \geq \eta$$

if bounded and $\phi(x^*) = \eta^*$

then all const. are satisfied

so the lin. prog EF is solved



REF is a relaxation of EF, hence
if the sol is feasible for EF it
is opt. for EF

Resuming:

if (REF) has no feasible sol \Rightarrow stop and
return infeasible, adding
constraints will not
remove infeas -

if (REF) has an unbounded sol $\Rightarrow \eta^* = +\infty$
guess x^* and solve (DSP)

if (REF) has a bounded sol $\Rightarrow (\eta^*, x^*)$ and solve (DSP)

if (DSP) is infeasible \Rightarrow stop the (EF) is unbounded.

if (DSP) is unbounded \Rightarrow add extreme ray const.

if (DSP) is bounded \Rightarrow add extreme point const.
or STOP because opt. found.

- if $X \subseteq \mathbb{Z}^n$ instead of \mathbb{R}^n

then Branch and Cut [Wo, p. 237]

Solve LP relaxation by Benders' alg
at each node of the enumeration
tree. Early pruning if DSP is
infeasible.

- if $y \in \mathbb{Z}^p$ instead of \mathbb{R}^p ($+ X \subseteq \mathbb{Z}^n$)

then integer subproblems

- 1) branch on (x, y, y) space
(extension of the branch and cut
alg seen above)

branch on x var except when
they are integer.

In DSP branching constraints
 $e \leq y \leq u$

yield the following changes

$$\min \{ u(d - Fx^*) - u^1 l + u^2 u : (u, u^1, u^2) \in U^* \}$$

where $U^* = \left\{ u, u^1, u^2 \in \mathbb{R}_+^m \times \mathbb{R}_+^p \times \mathbb{R}_+^p : \right.$
$$\left. uG - u^1 I_m + u^2 I_m = h \right\}$$

2) branch on (x, y)

subproblems are integer
programs \Rightarrow cuts are weak

i) $SP^T(x^*)$ infeasible

add no-good cuts

ii) $SP^T(x^*)$ has $\cup \emptyset \neq \emptyset'$

add no-good optimality cuts

iii) $\emptyset' = \phi^T(x^*) = y^*$

(x^*, y^*, γ^*) is a feasible sol to (OP)

update incumbent and

Continue the search.