# RAIRO Analyse numérique

# P. J. LAURENT C. CARASSO An algorithm of successive minimization in convex programming

*RAIRO – Analyse numérique*, tome 12, nº 4 (1978), p. 377-400. <a href="http://www.numdam.org/item?id=M2AN\_1978\_12\_4\_377\_0">http://www.numdam.org/item?id=M2AN\_1978\_12\_4\_377\_0</a>

## © AFCET, 1978, tous droits réservés.

L'accès aux archives de la revue « RAIRO – Analyse numérique » implique l'accord avec les conditions générales d'utilisation (http://www.numdam.org/ legal.php). Toute utilisation commerciale ou impression systématique est constitutive d'une infraction pénale. Toute copie ou impression de ce fichier doit contenir la présente mention de copyright.

# $\mathcal{N}$ umdam

Article numérisé dans le cadre du programme Numérisation de documents anciens mathématiques http://www.numdam.org/ RAIRO Analyse numérique/Numerical Analysis (vol 12, n° 4, 1978, p 377 à 400)

## AN ALGORITHM OF SUCCESSIVE MINIMIZATION IN CONVEX PROGRAMMING (\*)

by P. J. LAURENT (<sup>1</sup>) and C. CARASSO (<sup>2</sup>)

Abstract – A general exchange algorithm is given for the minimization of a convex function with equality and inequality constraints. It is a generalization of the Cheney-Goldstein algorithm, but following an idea given by Topfer, a finite sequence of sub-problems the dimension of which is decreasing, is considered at each iteration Given a positive number  $\varepsilon$ , under very general conditions, it is proved that the method, after a finite number of iterations, leads to an " $\varepsilon$ -solution"

In 1959, Cheney and Goldstein [6] (see also Goldstein [7]) proposed an algorithm for solving the problem of minimizing a convex function:

$$f(x) = \max_{\iota \in S} \left( \sum_{\iota=1}^{n} b_{\iota}(\iota) x_{\iota} - c(\iota) \right)$$

under the constraints:

$$\sum_{i=1}^{n} b_i(t) x_i \leq c(t) \quad \text{for all } t \in U,$$

where S and U are two disjoint compact sets and  $b_1, \ldots, b_n$ , c are continuous real functions defined on  $S \cup U$ .

At each iteration v of this algorithm, a polyhedral approximation of the problem is associated to a suitable subset  $A^v$  consisting of n+1 points of  $S \cup U$ . Using the exchange theorem (Stiefel [11, 12, 13]; see also [8, 9]) a new element  $t^v \in S \cup U$  is introduced:  $A^{v+1} = (A^v \setminus t_0^v) \cup t^v$ .

We propose here a new algorithm which is an extension of the Cheney-Goldstein algorithm for solving the same problem but under much weaker assumptions: the sets S and U are arbitrary and the mappings  $b_1, \ldots, b_n, c$  are

<sup>(\*)</sup> Recu decembre 1977

<sup>(1)</sup> Mathematiques appliquees I M A G Universite Scientifique et Médicale de Grenoble

<sup>(&</sup>lt;sup>2</sup>) U E R de Sciences, Université de Saint-Étienne, Saint-Étienne

only supposed to be bounded. Moreover, no Haar condition is introduced. At each iteration, we consider a sequence of nested minimization problems. The algorithm is based on an extension of the exchange theorem in which the exchanged quantities are not just a single point (see [3, 4]).

The idea of the algorithm is similar to the recursive method introduced by Töpfer [14], [15] (see also [3]) for problems of Tchebycheff best approximation.

In the case of a best approximation problem the algorithm becomes an extension of the Rémès algorithm (see [5]). For other applications, see [2].

#### **1. PROBLEM AND ASSUMPTIONS**

We denote by E the *n*-dimensional Euclidean space and by  $\langle x, x' \rangle$  the usual inner-product of x and x' in E.

#### 1.1. The minimization problem

We denote by L a finite set with l elements (l < n) and by S and U two arbitrary sets. Suppose that L, S and U have no common point and let  $T = S \cup U$ .

Let b and c be two bounded mappings from  $L \cup T$  into E and **R** respectively (i. e., b(T) and c(T) are bounded).

We define the functionals f and g by:

$$f(x) = \sup_{t \in S} (\langle x, b(t) \rangle - c(t)),$$

 $g(x) = \sup_{t \in U} \left( \langle x, b(t) \rangle - c(t) \right).$ 

It is easy to see that f and g are continuous convex functionals defined on E with values in **R**.

We define the affine variety W by:

$$W = \{ x \in E \mid \langle x, b(t) \rangle = c(t), t \in L \}.$$

It is convenient to suppose that the b(t),  $t \in L$  are linearly independent and that they span a *l*-dimensional subspace:

$$V = \mathscr{L}(b(t) \mid t \in L).$$

Thus, the affine variety W is parallel to  $V^{\perp}$ , the orthogonal complement of V, and has the dimension n-l.

The problem (P) consists in minimizing f(x) with x satisfying:

$$\langle x, b(t) \rangle = c(t)$$
 for  $t \in L$ 

R.A.I.R.O. Analyse numérique/Numerical Analysis

and

 $\langle x, b(t) \rangle \leq c(t)$  for  $t \in U$ .

i. e.,  $x \in W$  and  $g(x) \leq 0$ .

Put:

(P) 
$$\alpha = \inf_{\substack{x \in W \\ g(x) \leq 0}} f(x)$$

and suppose that  $\alpha$  is finite.

An element  $\overline{x} \in E$  is called a solution of (P) if:

$$\overline{x} \in W$$
,  $g(\overline{x}) \leq 0$  and  $f(\overline{x}) = \alpha$ .

An element  $\tilde{x} \in E$  will be called an  $\varepsilon$ -solution of (P) (with  $\varepsilon > 0$ ) if:

$$\widetilde{x} \in W$$
,  $g(\widetilde{x}) \leq \varepsilon$  and  $f(\widetilde{x}) \leq \alpha + \varepsilon$ .

For a given  $\varepsilon > 0$  (arbitrarily small), the algorithm that we are going to describe, will give, after a finite number of iterations (depending on  $\varepsilon$ ), an  $\varepsilon$ -solution of (P). The effective use of the method requires that for numbers  $\hat{\varepsilon}$  satisfying  $\eta \leq \hat{\varepsilon} \leq \varepsilon$  (where  $\eta$  is a positive number such that  $\eta < \varepsilon/2^{n-1}$ ) and for any  $x \in W$ , it is possible to determine

$$s \in S$$
 such that  $\langle x, b(s) \rangle - c(s) \ge f(x) - \hat{\varepsilon}$ 

and

 $u \in U$  such that  $\langle x, b(u) \rangle - c(u) \ge g(x) - \hat{\varepsilon}$ .

The exact values of f(x) and g(x) are not directly used: only an upper bound in the calculation of the supremum is necessary.

#### 1.2. Assumptions

We assume that:

(H1) there exist  $\check{x} \in W$  and  $\omega > 0$  such that  $\langle \check{x}, h(t) \rangle - c(t) \leq -\omega$ , for all  $t \in U$ , (this implies the regularity of the constraints);

(H2) the set:

$$K = \{ x \in E \mid \langle x, b(t) \rangle = 0, t \in L; \langle x, b(t) \rangle \leq 0, t \in T \}$$

is a linear subspace.

Note that the preceding set K is equal to the recession cone of all non-empty level sets:

 $S_{\lambda} = \{ x \in W | f(x) \leq \lambda, g(x) \leq 0 \}.$ 

Thus, the condition (H2) implies the existence of solutions for the probleme (P). The condition (H2) is also equivalent to:

(H2')  $0 \in \operatorname{rico}(b(T)) + V$ ,

where rico (b(T)) denotes the relative interior of the convex hull of b(T).

As a consequence of (H2), there exist  $\sigma \in \mathbf{R}$  and  $\tau \in \mathbf{R}$  such that for all  $x \in W$ :

(M) 
$$\sup_{t \in T} |\langle x, b(t) \rangle| \leq \sigma \sup_{t \in T} \langle x, b(t) \rangle + \tau.$$

#### 1.3. Application to best approximation problems

The preceding formulation includes the general problem of best approximation in a finite dimensional subspace with equality and inequality constraints. In this case, the function to minimize is:

$$f(x) = \left\| \sum_{i=1}^{n} x_i y_i - y_0 \right\|,$$

where  $y_0, y_1, \ldots, y_n$  are n+1 given elements of a normed linear space Y, the norm of which is denoted by ||y||, for  $y \in Y$ . It is possible to find a subset  $S \subset Y'$  (the topological dual of Y) such that f can be written in the following form:

$$f(x) = \sup_{y' \in S} (\langle x, b(y') \rangle - c(y')),$$

with

$$b(y') = [(y_1, y'), \dots, (y_n, y')],$$
  

$$c(y') = (y_0, y'),$$

where (y, y') represents the value at y of the continuous linear functional  $y' \in Y'$ . For example, take for S the unit sphere of Y' or the set of its extremal points.

#### 2. MINIMAL CONVEX SUPPORT (m. c. s.)

Subsequently, we will need the notion of minimal convex support of a linear subspace of E. This notion will be used not only relatively to V but also for other linear subspaces occuring in the algorithm.

te A

Let  $\mathscr{V}$  be a *d*-dimensional linear subspace of *E* spanned by the elements b(t),  $t \in D$  (not necessarely independent), where *D* is a finite subset of  $L \cup T$ .

#### 2.1. Convex support of a linear subspace

A non-empty and finite subset  $A \subset T$  will be called a *convex support* of  $\mathscr{V}$  if there exist coefficients  $\rho(t) \ge 0$ ,  $t \in A$  satisfying  $\sum \rho(t) = 1$  such that:

 $\sum_{t\in A}\rho(t) b(t)\in \mathscr{V}$ 

[i. e. if co  $(b(A)) \cap \mathscr{V} \neq \emptyset$ ].

#### 2.2. Minimal convex support of a linear subspace

A convex support A of  $\mathscr{V}$  will be called *minimal* if there does not exist a convex support of  $\mathscr{V}$  that is strictly included in A.

A subset  $A = \{t_1, \ldots, t_{k+1}\}$  consisting of k+1 points of T is a minimal convex support (m. c. s.) of  $\mathscr{V}$  if and only if:

(a) there exist positive coefficients  $\rho(t)$ ,  $t \in A$  satisfying  $\sum_{t \in A} \rho(t) = 1$  such that  $\sum_{t \in A} \rho(t) = \alpha(t) h(t) = \alpha(t)$ .

 $\sum_{t\in A}\rho(t)b(t)\in\mathscr{V};$ 

(b) the subspace  $\mathscr{L}(b(t) | t \in D \cup A)$  spanned by the b(t),  $t \in D \cup A$ , has the dimension d+k.

Every convex support contains at least a m.c.s., and using Caratheodory's theorem, one shows that a m.c.s. contains at most n-d+1 elements.

#### 2.3. Coefficients associated with a m.c.s.

One also proves that  $A \subset T$  is a m. c. s. if and only if there exist *unique* positive coefficients  $\rho_A(t)$ ,  $t \in A$ , satisfying:

$$\sum_{t \in A} \rho_A(t) b(t) \in \mathscr{V} \quad \text{and} \quad \sum_{t \in A} \rho_A(t) = 1.$$

These coefficients  $\rho_A(t)$ ,  $t \in A$ , will be called the coefficients associated with the m.c.s. A. It will also be useful to introduce coefficients  $\lambda_A(t)$ ,  $t \in D$ , such that:

$$\sum_{t \in A} \rho_A(t) b(t) + \sum_{t \in D} \lambda_A(t) b(t) = 0$$

[these  $\lambda_A(t)$  are not necessarily unique].

#### 2.4. Minimization associated with a m.c.s.

Let  $A = \{t_1, \ldots, t_{k+1}\}$  be a m. c. s. of V, consisting of k+1 elements such that  $A \cap S \neq \emptyset$ . Put:

$$f_A(x) = \max_{t \in A \cap S} (\langle x, b(t) \rangle - c(t)),$$

 $g_{A}(x) = \max_{t \in A \cap U} (\langle x, b(t) \rangle - c(t))$ 

[if  $A \cap U = \emptyset$ , then  $g_A(x) \equiv -\infty$ ] and consider the problem (P<sub>A</sub>) of minimizing  $f_A(x)$  for x belonging to W and satisfying  $g_A(x) \leq 0$ . Put:

$$(\mathbf{P}_{A}) \quad \alpha_{A} = \min_{\substack{x \in W \\ g_{A}(x) \leq 0}} f_{A}(x) \leq \alpha$$

We denote by  $W_A$  the set of solutions of  $(\mathbf{P}_A)$ , i. e. of elements  $\overline{x} \in W$  satisfying  $g_A(\overline{x}) \leq 0$ ,  $\alpha_A = f_A(\overline{x})$ .

Then we have the following result, the proof of which is simple:

THEOREM: The amount  $\alpha_A$  of  $(\mathbf{P}_A)$  is given by:

 $\alpha_{A} = \frac{1}{s_{A}} \sum_{t \in A} \rho_{A}(t) \ (\langle x_{0}, b(t) \rangle - c(t)),$ (where  $x_{0}$  is an arbitrary element of W and  $s_{A} = \sum_{t \in A \cap S} \rho_{A}(t)$ ), or by:

$$\alpha_{A} = -\frac{1}{s_{A}} \left( \sum_{t \in L} \lambda_{A}(t) c(t) + \sum_{t \in A} \rho_{A}(t) c(t) \right).$$

The set of solutions in given by:

$$W_{A} = \left\{ x \in W \middle| \langle x, b(t) \rangle - c(t) + \delta(t) \alpha_{A} = \alpha_{A}, t \in A \right\},\$$

where

$$\delta(t) = \begin{cases} 0 & \text{if } t \in S, \\ 1 & \text{if } t \in U. \end{cases}$$

Thus, the set  $W_A$  is an affine variety which is parallel to  $V_A^{\perp}$ , with:  $V_A = \mathscr{L}(b(t) | t \in L \cup A),$ 

the dimension of which is equal to l+k. Therefore, the dimension of  $W_A$  is n-(l+k) and the solution of  $(P_A)$  is unique when k=n-l.

R.A.I.R.O. Analyse numérique/Numerical Analysis

For a given  $\varepsilon > 0$ , the algorithm that we are going to describe, will build a finite sequence  $A^0, A^1, \ldots, A^{\mu}$  of m.c.s. of V and of associated solutions  $x^0, x^1, \ldots, x^{\mu}(x^{\nu} \in W_{A^{\nu}})$  such that the corresponding amounts  $\alpha^0, \alpha^1, \ldots, \alpha^{\mu}$  ( $\alpha^{\nu} = \alpha_{A^{\nu}}$ ) build a non-decreasing sequence such that:

$$x^{\mu} \in W$$
,  $g(x^{\mu}) \leq \varepsilon$  and  $f(x^{\mu}) \leq \alpha^{\mu} + \varepsilon$ .

As  $\alpha^{\mu} \leq \alpha$ , the element  $x^{\mu}$  will be an  $\varepsilon$ -solution of (P).

#### **3. A CONVERGENCE RESULT**

The convergence result that we will state in this section corresponds, for the time being, to a theoretical algorithm, the use of which is not quite specified. We will need it, on the one hand, as a guide for the definition of the effective algorithm, on the other hand as a basis for proving its convergence.

Let  $A^{\nu}$ ,  $\nu = 0, 1, ...$ , be an infinite sequence of m.c.s. of V such that  $A^{\nu} \cap S \neq \emptyset$ ,  $\nu = 0, 1, ...$ ; and let  $f^{\nu} = f_{A^{\nu}}$ ,  $g^{\nu} = g_{A^{\nu}}$  be the associated polyhedral functionals (as in § 2.4.). Put:

$$\alpha^{\mathsf{v}} = \min_{\substack{x \in W \\ g^{\mathsf{v}}(x) \leq 0}} f^{\mathsf{v}}(x) \leq \alpha$$

and denote by  $W^{\vee} = W_{A^{\vee}}$  the set of solutions. Let  $\rho^{\vee}(t) = \rho_{A^{\vee}}(t)$ ,  $t \in A^{\vee}$ , be the positive coefficients corresponding to  $A^{\vee}$  (as in § 2.3.) and let  $s^{\vee} = \sum_{t \in A^{\vee} \cap S} \rho^{\vee}(t)$ .

Consider the functional  $h^{v}$  defined by:

$$h^{\mathsf{v}}(x) = \max \left( f(x); g(x) + \alpha^{\mathsf{v}} \right)$$
$$= \sup_{t \in \mathcal{T}} \left( \langle x, b(t) \rangle - c(t) + \delta(t) \alpha^{\mathsf{v}} \right).$$

Finally, suppose that the set:

$$\hat{N} = \left\{ \nu \in \mathbf{N} / A^{\nu+1} \neq A^{\nu} \right\},\,$$

is infinite and let  $\hat{A}^{\nu} = A^{\nu+1} \setminus A^{\nu}$ , for  $\nu \in \hat{N}$ .

Then, we have the following result:

**3.1.** THEOREM: If, for all  $v \in \hat{N}$ , there exists  $x^{\nu} \in W^{\nu}$  such that:

$$h^{\mathsf{v}}(x^{\mathsf{v}}) - (\langle x^{\mathsf{v}}, b(t) \rangle - c(t) + \delta(t) \alpha^{\mathsf{v}}) \leq \hat{\varepsilon},$$

for all  $t \in \hat{A}^{\vee}$ , then we have:

$$\alpha^{\nu+1} \geq \alpha^{\nu} + \frac{1}{s^{\nu+1}} \left( \sum_{t \in \tilde{A}^{\nu}} \rho^{\nu+1}(t) \right) \left( h^{\nu}(x) - \alpha^{\nu} - \hat{\varepsilon} \right)$$

(this proves that  $\alpha^{\nu+1} > \alpha^{\nu}$ , as long as  $h^{\nu}(x^{\nu}) - \alpha^{\nu} - \hat{\epsilon} > 0$ ) and for any  $\epsilon > \hat{\epsilon}$ , there exists  $\mu \in \hat{N}$  such that:

$$h^{\mu}(x^{\mu}) - \alpha^{\mu} \leq \varepsilon$$

[what implies that  $x^{\mu}$  is an  $\varepsilon$ -solution of (P)].

*Proof: First part:* By theorem 2.4, we have:

$$\alpha^{\nu+1} = \frac{1}{s^{\nu+1}} \sum_{t \in A^{\nu+1}} \rho^{\nu+1}(t) (\langle x^{\nu}, b(t) \rangle - c(t)).$$

For  $t \in \hat{A}^{\vee}$ , we have:  $\langle x^{\vee}, b(t) \rangle - c(t) + \delta(t) \alpha^{\vee} \ge h^{\vee}(x^{\vee}) - \hat{\epsilon}$ . For  $t \in A^{\vee+1} \cap A^{\vee}$ , we have:  $\langle x^{\vee}, b(t) \rangle - c(t) + \delta(t) \alpha^{\vee} = \alpha^{\vee}$ . Hence:

$$\begin{aligned} \alpha^{\nu+1} &\ge \frac{1}{s^{\nu+1}} \sum_{t \in A^{\nu+1} \cap A^{\nu}} \rho^{\nu+1}(t) (\alpha^{\nu} - \delta(t) \alpha^{\nu}) \\ &+ \frac{1}{s^{\nu+1}} \sum_{t \in A^{\nu}} \rho^{\nu+1}(t) (h^{\nu}(x^{\nu}) - \delta(t) \alpha^{\nu} - \hat{\varepsilon}) \\ &= \frac{1}{s^{\nu+1}} \sum_{t \in A^{\nu+1}} \rho^{\nu+1}(t) (\alpha^{\nu} - \delta(t) \alpha^{\nu}) \\ &+ \frac{1}{s^{\nu+1}} \sum_{t \in A^{\nu}} \rho^{\nu+1}(t) (h^{\nu}(x^{\nu}) - \alpha^{\nu} - \hat{\varepsilon}) \end{aligned}$$

and finally:

$$\alpha^{\nu+1} \geqq \alpha^{\nu} + \frac{1}{s^{\nu+1}} \left( \sum_{t \in A^{\nu}} \rho^{\nu+1}(t) \right) \left( h^{\nu}(x^{\nu}) - \alpha^{\nu} - \hat{\varepsilon} \right).$$

Second part: The following lemma is a property of the positive coefficients that are associated with a sequence of m.c.s. We will only use the fact that  $A^{v}$ ,  $v=0,1,\ldots$ , is a sequence of m.c.s. such that  $\hat{N}$  is infinite.

LEMMA: There exist an infinite subset  $\tilde{N}$  in  $\hat{N}$  and a bipartition of  $A^{\vee}$  in  $B^{\vee}$  and  $C^{\vee} \neq \emptyset$  for  $\nu \in \hat{N}$  such that:

1°  $\lim_{v\in\tilde{N}} \left(\sum_{t\in B^{v}} \rho^{v}(t)\right) = 0;$ 

R.A.I.R.O. Analyse numérique/Numerical Analysis

2° there exists m > 0 such that  $\rho^{v}(t) \ge m$ , for all  $t \in C^{v}$  and all  $v \in \tilde{N}$ ;

3° C<sup>v</sup> is not included in  $A^{v^-}$ , for all  $v \in \tilde{N}$ , where  $v^-$  denotes the integer preceding v in  $\tilde{N}$ .

The proof of this lemma has been given in [4].

Third part: Suppose that we have  $h^{\nu}(x^{\nu}) - \alpha^{\nu} > \varepsilon$ , for all  $\nu \in \hat{N}$  (with  $\varepsilon > \hat{\varepsilon}$ ) and prove that this leads to a contradiction.

As a consequence of the first part, we then have:

$$\alpha^{\nu+1} \ge \alpha^{\nu} + \frac{1}{s^{\nu+1}} \left( \sum_{t \in \hat{A}^{\nu}} \rho^{\nu+1}(t) \left( \varepsilon - \hat{\varepsilon} \right) \quad \text{for all} \quad \nu \in \hat{N},$$

hence,  $\alpha^{v}$  is a non decreasing sequence such that  $\alpha^{v} \leq \alpha$ .

Let  $\tilde{\alpha}$  ( $\tilde{\alpha} \leq \alpha$ ) be its limit.

First we show that there exists a positive constant s such that

$$s^{\mathsf{v}} = \sum_{t \in A^{\mathsf{v}} \cap S} \rho^{\mathsf{v}}(t) \ge s > 0.$$

It is equivalent to prove that  $C^{\nu} \cap S \neq \emptyset$  for all  $\nu \in \tilde{N}$ , sufficiently large. By theorem 2.4, we have:

$$\langle x^{\mathsf{v}}, b(t) \rangle - c(t) = \alpha^{\mathsf{v}} - \delta(t) \alpha^{\mathsf{v}}$$
 for  $t \in A^{\mathsf{v}}$ 

and by (H1):

$$\langle \check{x}, b(t) \rangle - c(t) \leq -\omega$$
 (with  $\omega > 0$ ) for  $t \in A^{\vee}$ .

The mappings b and c being bounded and  $\alpha^{v}$  converging to  $\tilde{\alpha}$ , it is possible to find a constant  $\xi$  such that:

 $|\langle x^{\mathsf{v}} - \check{x}, b(t) \rangle| \leq \xi$  for all  $t \in A^{\mathsf{v}}$  and all  $\mathsf{v} \in \widetilde{N}$ .

Suppose that there exists an infinite subset  $N \subset \tilde{N}$  such that  $C^{\vee} \cap S \neq \emptyset$ , for all  $v \in \tilde{N}$  and prove that this leads to a contradiction. Consider  $\sum_{v \in V} \rho^{v}(t) \langle x^{\vee} - \tilde{x}, b(t) \rangle$  and decompose the sum according to  $B^{\vee}$  and  $C^{\vee}$ .

Letting 
$$\eta^{v} = \sum_{t \in B^{v}} \rho^{v}(t)$$
, we have:  
 $\left| \sum_{t \in B^{v}} \rho^{v}(t) \langle x^{v} - \check{x}, b(t) \rangle \right| \leq \eta^{v} m$ ,  
with  $\lim_{v \in \tilde{N}} \eta^{v} = 0$ .  
vol. 12, n° 4, 1978

On the other hand, as  $C^{\nu} \subset U$ , we have  $\langle x^{\nu}, b(t) \rangle - c(t) = 0$ , for all  $t \in C^{\nu}$ , and letting  $\theta^{\nu} = \sum_{t \in C^{\nu}} \rho^{\nu}(t)$ , we obtain:

$$\sum_{t \in C^{\mathsf{v}}} \rho^{\mathsf{v}}(t) \langle x^{\mathsf{v}} - \check{x}, b(t) \rangle \ge \theta^{\mathsf{v}} \omega \quad \text{for all } \mathsf{v} \in \overset{\approx}{N},$$

with  $\lim_{v \in \tilde{N}} \theta^{v} = 1$  and  $\omega > 0$ .

Thus, for  $v \in \widetilde{\widetilde{N}}$  sufficiently large, we would have  $\sum_{\substack{t \in A^{\vee} \\ \text{ is a } m. c. s.}} \rho^{\vee}(t) \langle x^{\vee} - \check{x}, b(t) \rangle > 0$ , what is in contradiction with the facts that  $A^{\vee}$  is a m.c.s. of V and that  $x^{\vee} - \check{x} \in V^{\perp}$ .

Using the result of the second part, we see that for all  $v \in \tilde{N}$ , the set  $C^{v}$  is not included in  $A^{v^{-}}$ . Therefore, there exist elements of  $C^{v}$  that have been introduced between the iteration  $v^{-}$  and the iteration v. Let  $\hat{v} \in \hat{N}$  be the last iteration satisfying  $v^{-} \leq \hat{v} < v$  for which at least one element of  $C^{v}$  has been introduced.

By theorem 2.4, we have:

$$\alpha^{\mathsf{v}} = \frac{1}{s^{\mathsf{v}}} \sum_{t \in A^{\mathsf{v}}} \rho^{\mathsf{v}}(t) \left( \left\langle x^{\hat{\mathsf{v}}}, b(t) \right\rangle - c(t) \right)$$

with  $s^{v} \geq s > 0$ .

We decompose again the sum according to  $B^{v}$  and  $C^{v}$ :

(a) Sum corresponding to  $B^{\vee}$ .

As a consequence of (H2), [see (M) in §.1.2.], for all  $t \in T$ , we have:

$$\left|\langle x^{\hat{\mathbf{v}}}, b(t) \rangle\right| \leq \sigma \sup_{t \in T} \langle x^{\hat{\mathbf{v}}}, b(t) \rangle + \tau.$$

As the mapping c is bounded and the sequence  $\alpha^{v}$  is also bounded, we can find a constant  $\chi$  such that:

$$\left|\langle x^{\hat{\mathbf{v}}}, b(t) \rangle - c(t)\right| \leq \sigma h^{\mathbf{v}}(x^{\hat{\mathbf{v}}}) + \chi,$$

for all  $t \in T$ . Hence, we have:

$$\left|\frac{1}{s^{\mathsf{v}}}\sum_{t\in B^{\mathsf{v}}}\rho^{\mathsf{v}}(t)\left(\langle x^{\hat{\mathsf{v}}}, b(t)\rangle - c(t)\right)\right| \leq \frac{1}{s}\eta^{\mathsf{v}}(\sigma h^{\mathsf{v}}(x^{\hat{\mathsf{v}}}) + \chi).$$

( $\beta$ ) Sum corresponding to  $C^{\vee}$ .

For  $t \in C^{\mathsf{v}} \cap \hat{A}^{\circ} \neq \emptyset$ :  $\langle x^{\circ}, b(t) \rangle - c(t) \ge h^{\circ}(x^{\circ}) - \delta(t) \alpha^{\circ} - \hat{\varepsilon}$ . For  $t \in C^{\mathsf{v}} \setminus \hat{A}^{\circ}$ :  $\langle x^{\circ}, b(t) \rangle - c(t) = \alpha^{\circ} - \delta(t) \alpha^{\circ}$ .

R.A.I.R.O. Analyse numérique/Numerical Analysis

Hence, we have:

$$\begin{split} \frac{1}{s^{\mathbf{v}}} & \sum_{t \in C^{\mathbf{v}}} \rho^{\mathbf{v}}(t) \left( \langle x^{\hat{\mathbf{v}}}, b(t) \rangle - c(t) \right) \\ & \geq \frac{1}{s^{\mathbf{v}}} \left( \sum_{t \in C^{\mathbf{v}} \cap \hat{A}^{\hat{\mathbf{v}}}} \rho^{\mathbf{v}}(t) \left( h^{\hat{\mathbf{v}}}(x^{\hat{\mathbf{v}}}) - \delta(t) \alpha^{\hat{\mathbf{v}}} - \hat{\epsilon} \right) \\ & + \sum_{t \in C^{\mathbf{v}} \setminus \hat{A}^{\hat{\mathbf{v}}}} \rho^{\mathbf{v}}(t) \left( \alpha^{\hat{\mathbf{v}}} - \delta(t) \alpha^{\hat{\mathbf{v}}} \right) \right) \\ & = \frac{1}{s^{\mathbf{v}}} \left( \sum_{t \in C^{\mathbf{v}} \cap \hat{A}^{\hat{\mathbf{v}}}} \rho^{\mathbf{v}}(t) \right) \left( h^{\hat{\mathbf{v}}}(x^{\hat{\mathbf{v}}}) - \alpha^{\hat{\mathbf{v}}} - \hat{\epsilon} \right) + \frac{1}{s^{\mathbf{v}}} \alpha^{\hat{\mathbf{v}}} \left( \sum_{t \in C^{\mathbf{v}}} \rho^{\mathbf{v}}(t) \left( 1 - \delta(t) \right) \right) \\ & \geq u^{\mathbf{v}} \alpha^{\hat{\mathbf{v}}} + m(h^{\hat{\mathbf{v}}}(x^{\hat{\mathbf{v}}}) - \alpha^{\hat{\mathbf{v}}} - \hat{\epsilon}), \end{split}$$

with  $u^{v} = (1/s^{v}) \sum_{t \in C^{v}} \rho^{v}(t) (1 - \delta(t))$  satisfying  $\lim_{v \in N} u^{v} = 1$ .

Finally, joining ( $\alpha$ ) and ( $\beta$ ) together, we obtain:

$$\begin{aligned} \alpha^{\mathsf{v}} &\geq u^{\mathsf{v}} \alpha^{\mathsf{v}} + m \left( h^{\mathsf{v}}(x^{\mathsf{v}}) - \alpha^{\mathsf{v}} - \hat{\varepsilon} \right) - \frac{\eta^{\mathsf{v}}}{s} (\sigma h^{\mathsf{v}}(x^{\mathsf{v}}) + \chi) \\ &= u^{\mathsf{v}} \alpha^{\mathsf{v}} + m \left( v^{\mathsf{v}} h^{\mathsf{v}}(x^{\mathsf{v}}) - \alpha^{\mathsf{v}} - \hat{\varepsilon} \right) - \eta^{\mathsf{v}} \frac{\chi}{s} \,, \end{aligned}$$

with  $v^{v} = 1 - (\eta^{v} \sigma/sm)$ , satisfying  $\lim_{v \in \tilde{N}} v^{v} = 1$ . As  $\alpha^{v}$  converges towards  $\tilde{\alpha}$ , we deduce that:

$$\limsup_{v \in \tilde{N}} (v^{v} h^{\hat{v}}(x^{\hat{v}}) - \alpha^{\hat{v}} - \hat{\varepsilon}) \leq 0$$

hence:

 $\limsup_{v\in\tilde{N}} v^{v}h^{\hat{v}}(x^{\hat{v}}) \leq \tilde{\alpha} + \hat{\varepsilon}.$ 

It is easy to prove that this inequality is contradictory with:

 $h^{\mathsf{v}}(x^{\mathsf{v}}) \geq \alpha^{\mathsf{v}} + \varepsilon$  for all  $\mathsf{v} \in \hat{N}$ ,

in the cas where  $\varepsilon > \hat{\epsilon}$ .

**3.2.** REMARK: If we suppose that  $A^0 \cap S \neq \emptyset$ , then, as long as we have  $h^{\nu}(x^{\nu}) - \alpha^{\nu} - \hat{\varepsilon} > 0$ , then we have  $A^{\nu+1} \cap S \neq \emptyset$ :

As a matter of fact, if we would have  $A^{\nu+1} \cap S = \emptyset$ , this would mean that  $A^{\nu+1} \cap A^{\nu} \subset U$  and that  $\hat{A}^{\nu} \subset U$ ; hence

$$\langle x^{\mathsf{v}}, b(t) \rangle - c(t) = 0$$
 for all  $t \in A^{\mathsf{v}+1} \cap A^{\mathsf{v}}$ ,  
 $\langle x^{\mathsf{v}}, b(t) \rangle - c(t) \ge h^{\mathsf{v}}(x^{\mathsf{v}}) - \alpha^{\mathsf{v}} - \hat{\varepsilon} > 0$  for all  $t \in \hat{A}^{\mathsf{v}}$ .

As we have, by (H1):

 $\langle \check{x}, b(t) \rangle - c(t) < 0$  for all  $t \in A^{\nu+1}$ ,

we would deduce that:

 $\sum_{t\in A^{\nu+1}}\langle x^{\nu}-\check{x}, b(t)\rangle > 0,$ 

in contradiction with the facts that  $A^{v+1}$  is a m. c. s. of V and that  $\check{x}^v - \check{x} \in V^{\perp}$ .

#### 4. EXCHANGE THEOREM

The preceding convergence theorem shows that the sets  $\hat{A}^{\vee}$  of new elements should be such that it is possible to exchange them with a subset  $C^{\vee}$  of  $A^{\vee}$  in such a way that  $A^{\vee+1} = (A^{\vee} \setminus C^{\vee}) \cup \hat{A}^{\vee}$  is again a m. c. s. of V.

The next theorem shows how to operate this exchange. Subsequently we will have to do this operation, not only relatively to V but also for other linear subspaces occuring in the algorithm.

Let  $\mathscr{V}$  be a d-dimensional linear subspace of E defined by:

 $\mathscr{V} = \mathscr{L}\left(b\left(t\right) \mid t \in D\right)$ 

where D is a finite subset of  $L \cup T$ .

#### 4.1. Exchange theorem

If  $A_0$  is a m. c. s. of  $\mathscr{V}$  and if  $A_1$  is a m. c. s. of

 $\mathscr{V}_{0} = \mathscr{L}(b(t) \mid t \in D \cup A_{0})$ 

then, there exists a bipartition of  $A_0$  in  $B_0$  and  $C_0 \neq \emptyset$  such that:

$$\begin{aligned} \widetilde{A}_0 &= B_0 \cup A_1 \text{ is a } m. \ c. \ s. \ of \ \mathscr{V} \\ \widetilde{A}_1 &= C_0 \text{ is a } m. \ c. \ s. \ of \\ \widetilde{\mathscr{V}}_0 &= \mathscr{L}(b(t) \mid t \in D \cup \widehat{A}_0). \end{aligned}$$

This theorem has been proved in [4]. It shows that it is possible to exchange with  $A_1$  a non-empty part  $C_0$  of  $A_0$ , in such a way that:

$$\tilde{A}_0 = (A_0 \setminus C_0) \cup A_1$$

is again a m. c. s. of  $\mathscr{V}$ .

#### 4.2. Practice of the exchange

Denote by  $\rho_0(t) > 0$ ,  $t \in A_0$  and by  $\lambda_0(t)$ ,  $t \in D$  the coefficients associated with  $A_0$ , as in paragraph 2.3:

$$(a_0) \sum_{t \in A_0} \rho_0(t) b(t) + \sum_{t \in D} \lambda_0(t) b(t) = 0,$$
$$\sum_{t \in A_0} \rho_0(t) = 1.$$

As  $A_1$  is a m. c. s. of  $\mathscr{V}_0$ , we denote by  $\rho_1(t) > 0$ ,  $t \in A_1$  and by  $\lambda_1(t)$ ,  $t \in D \cup A_0$ , the corresponding coefficients:

(a<sub>1</sub>) 
$$\sum_{t \in A_1} \rho_1(t) b(t) + \sum_{t \in D \cup A_0} \lambda_1(t) b(t) = 0,$$
  
 $\sum_{t \in A_1} \rho_1(t) b(t) = 0.$ 

Substracting r-times the relation  $(a_0)$  from the relation  $(a_1)$ , we obtain:

$$\sum_{t \in A_0} \rho_0(t) \left( \frac{\lambda_1(t)}{\rho_0(t)} - r \right) b(t) + \sum_{t \in A_1} \rho_1(t) b(t) + \sum_{t \in D} (\lambda_1(t) - r \lambda_0(t)) b(t) = 0.$$

If we choose  $r = \min_{t \in A_0} (\lambda_1(t) / \rho_0(t))$ , and we define:

$$C_0 = \left\{ t \in A_0 \left| \frac{\lambda_1(t)}{\rho_0(t)} = r \right\} \quad \text{and} \quad B_0 = A_0 \searrow C_0 \right\}$$

then the preceding relation becomes:

$$(\tilde{a}_0) \quad \sum_{t \in B_0 \cup A_1} \tilde{\rho}_0(t) b(t) + \sum_{t \in D} \tilde{\lambda}_0(t) b(t) = 0,$$

with:

$$\tilde{\rho}_{0}(t) = \begin{cases} \frac{1}{q} (\lambda_{1}(t) - r \rho_{0}(t)) & \text{if } t \in B_{0}, \\ \frac{1}{q} \rho_{1}(t) & \text{if } t \in A_{1}, \end{cases}$$

$$\begin{split} \widetilde{\lambda}_0(t) &= \frac{1}{q} \rho_1(t) \quad \text{for} \quad t \in D, \\ q &= \sum_{t \in B_0} \left( \lambda_1(t) - r \rho_0(t) \right) + \sum_{t \in A_1} \rho_1(t). \end{split}$$

The coefficients  $\tilde{\rho}_0(t)$ ,  $t \in \tilde{A}_0 = B_0 \cup A_1$  are positive with the sum equal to one.

Now the relation  $(a_0)$  can be written:

$$(\tilde{a}_1) \sum_{t \in C_0} \tilde{\rho}_1(t) b(t) + \sum_{t \in D \cup B_0 \cup A_1} \tilde{\lambda}_1(t) b(t) = 0$$

with:

$$\widetilde{\rho}_{1}(t) = \frac{1}{p} \rho_{0}(t) \quad \text{for} \quad t \in C_{0},$$

$$\widetilde{\lambda}_{1}(t) = \begin{cases} \frac{1}{p} \lambda_{0}(t) & \text{if} \quad t \in D, \\ \frac{1}{p} \rho_{0}(t) & \text{if} \quad t \in B_{0}, \\ 0 & \text{if} \quad t \in A_{1}. \end{cases}$$

$$p=\sum_{t\in C_0}\rho_0(t)$$

The coefficients  $\tilde{\rho}_1(t)$ ,  $t \in \tilde{A}_1 = C_0$  are positive with the sum equal to 1.

#### 5. STRING OF M. C. S.

#### 5.1. Successive minimization

The convergence theorem  $(\S3)$  and the exchange theorem  $(\S4)$  lead us to consider the following sub-problem:

$$(\mathbf{SP}^{\mathsf{v}}) \quad \beta^{\mathsf{v}} = \inf_{x \in W^{\mathsf{v}}} h^{\mathsf{v}}(x)$$

with

$$h^{\mathsf{v}}(x) = \max(f(x); g(x) + \alpha^{\mathsf{v}})$$
$$= \sup_{t \in T} (\langle x, b(t), \rangle - c(t) + \delta(t) \alpha^{\mathsf{v}}).$$

This sub-problem can be solved by the algorithm described in [4]:

R.A.I.R.O. Analyse numérique/Numerical Analysis

Let  $A_1^v$  be a m. c. s. of V, with  $k_1^v + 1$  elements,

 $V^{\mathsf{v}} = V_0^{\mathsf{v}} = \mathscr{L}(b(t) \mid t \in L \cup A^{\mathsf{v}}),$ 

and denote by  $h_1^{v}$  the polyhedral functional defined by:

$$h_1^{\mathsf{v}}(x) = \max_{t \in A_1^{\mathsf{v}}} (\langle x, b(t) \rangle - c(t) + \delta(t) \alpha^{\mathsf{v}}).$$

We consider the minimization of  $h_1^v(x)$  for  $x \in W^v = W_0^v$ .

Put:

 $\alpha_1^{\mathsf{v}} = \min_{x \in W_0^{\mathsf{v}}} h_1^{\mathsf{v}}(x).$ 

The set  $W_1^{\vee}$  of solutions, can be written:

$$W_1^{\mathsf{v}} = \left\{ x \in W_0^{\mathsf{v}} \middle| \langle x, b(t) \rangle - c(t) + \delta(t) \alpha^{\mathsf{v}} = \alpha_1^{\mathsf{v}}, t \in A_1^{\mathsf{v}} \right\}.$$

It is an affine variety, which is parallel to  $(V_1^{\gamma})^{\perp}$ , where

$$V_1^{\mathsf{v}} = \mathscr{L}\left(b\left(t\right) \mid t \in L \cup A_0^{\mathsf{v}} \cup A_1^{\mathsf{v}}\right)$$

is a  $l + k_0^{v} + k_1^{v}$ -dimensional linear subspace of E.

The same construction can be repeated relatively to  $V_1^v: A_2^v$  is a m. c. s. of  $V_1^v$ , with  $k_2^v + 1$  elements,  $h_2^v$  is the associated functional,  $\alpha_2^v$  the amount of its minimum on  $W_1^v$ ,  $W_2^v$  the set of solutions, and  $V_2^v = \mathcal{L}(b(t) | t \in L \cup A_0^v \cup A_1^v \cup A_2^v)$ , the dimension of which is  $l + \sum_{i=0}^2 k_i^v$ .

We continue this construction until we have  $V_{m}^{\nu} = E$ , hence  $W_{m}^{\nu}$  is reduced to a single point.

#### 5.2. String of m.c.s.

The preceding construction leads us to the notion of a string of m. c. s. (shortly "string"):

A finite sequence  $\mathscr{C} = (A_0, \ldots, A_m)$  of subsets  $A_i \subset T$  will be called a *string*, if, setting  $V_{-1} = V$ , we have:

 $A_i \text{ is a } m. c.s. \text{ of } V_{i-1},$   $V_i = \mathcal{L}(b(t) \mid t \in L; t \in A_j, j = 0, \dots, i),$   $i = 1, \dots, m,$   $V_m = E.$ vol. 12, n° 4, 1978

If the subset  $A_j$  contains  $k_j + 1$  elements (j = 0, ..., m), then the dimension of  $V_i$  is  $l + \sum_{j=0}^{i} k_j$ . Associated with each  $A_i$  of the string  $\mathscr{C}$ , we can define the coefficients  $\rho_i(t) > 0, t \in A_i$  and  $\lambda_i(t), t \in L \cup \left(\bigcup_{j=0}^{i-1} A_j\right)$  such that:  $\sum_{t \in A_i} \rho_i(t) b(t) + \sum_{t \in L} \lambda_i(t) b(t) + \sum_{j=0}^{i-1} \sum_{t \in A_j} \lambda_i(t) b(t) = 0,$  $\sum_{t \in A_i} \rho_i(t) = 1.$ 

#### 5.3. Solution associated with a string

A string  $\mathscr{C} = (A_0, \ldots, A_m)$  will be said correct if  $A_0 \cap S \neq \emptyset$ . Put:

$$\begin{cases} f_i(x) = \max_{\substack{t \in A_i \cap S \\ t \in A_i \cap U}} (\langle x, b(t) \rangle - c(t)) \\ g_i(x) = \max_{\substack{t \in A_i \cap U \\ t \in A_i \cap U}} (\langle x, b(t) \rangle - c(t)) \end{cases}$$
  $i = 0, \ldots, m.$ 

We consider the sequence of successive minimization problems associated with a correct string  $\mathscr{C}$ :

$$\alpha_0 = \min_{\substack{x \in W \\ g_0(x) \leq 0}} f_0(x)$$

the set of solutions  $W_0$  of which is an affine variety parallel to  $V_0^{\perp}$ , and

$$\alpha_i = \min_{x \in W_{i-1}} \max (f_i(x); g_i(x) + \alpha_0),$$

 $i=1,\ldots,m$ , the set of solutions  $W_i$  of which is an affine variety parallel to  $V_i^{\perp}$ .

As  $V_m = E$ , the affine variety  $W_m$  is reduced to a single point  $x = x_{\mathscr{C}}$  that we will call the solution associated with the string  $\mathscr{C}$ . As  $x = x_{\mathscr{C}}$  is a solution of the successive minimization problems, by theorem 2.4 above and theorem 2.3 of [4], it is characterized by the following conditions:

$$\langle x, b(t) \rangle = c(t), \quad t \in L \quad (l \text{ conditions})$$
  
 $\langle x, b(t) \rangle + \delta(t) \alpha_0 - \alpha_i = c(t), \quad t \in A_i \quad (k_i + 1 \text{ conditions}),$   
 $i = 0, \dots, m.$ 

R.A.I.R.O. Analyse numérique/Numerical Analysis

We can use these n+m+1 linear equations for computing the n+m+1 unknown  $x_1, \ldots, x_n, \alpha_0, \ldots, \alpha_m$ . By construction, this linear algebraic system has a unique solution.

#### 5.4. Exchange operation in a string

Let  $\mathscr{C} = (A_0, \ldots, A_m)$  be a string and  $(V_0, \ldots, V_m)$  the corresponding linear subspaces. We see that  $A_{j-1}$  is a m. c. s. of  $\mathscr{V} = V_{j-2}$  and that  $A_j$  is a m. c. s. of the linear subspace:

$$\mathscr{V}_{0} = \mathscr{L}(b(t) \mid t \in D \cup A_{j-1}) = V_{j-1},$$
  
with  $D = L \cup \bigcup_{i=0}^{j-2} A_{i}.$ 

Thus we have the same situation as in theorem 4.1.

There exists a bipartition of  $A_{i-1}$  in  $B_{i-1}$  and  $C_{i-1} \neq \emptyset$  such that, letting:

$$\tilde{A}_{j-1} = B_{j-1} \cup A_j$$
 and  $\tilde{A}_j = C_{j-1}$ ,

then  $\widetilde{\mathscr{C}} = (A_0, \ldots, \widetilde{A}_{j-1}, \widetilde{A}_j, \ldots, A_m)$  is again a string.

We will say that we have exchanged  $A_{j-1}$  and  $A_j$  in the string  $\mathscr{C}$ .

#### 5.5. Regular string

A string  $\mathscr{C} = (A_0, \ldots, A_m)$  will be said *regular* if each of the subsets  $A_i$ ,  $i=0,\ldots,m$ , has at least two elements. Thus, if  $\mathscr{C}$  is regular, the dimension of  $V_i$  is strictly greater than the dimension of  $V_{i-1}$   $(i=0,\ldots,m)$  and the integer *m* is necessarily smaller or equal to n-l-1.

If  $\mathscr{C}$  is an arbitrary string, we obtain a regular string by taking away all the m. c. s. that are reduced to a single point. If  $A_0$  is not reduced to a single point, this operation does not change the solution associated with the string as well as the amounts  $\alpha_i$  corresponding to the remaining m. c. s.  $A_i$ .

#### 6. ALGORITHM

If  $\varepsilon > 0$  is the desired accuracy, let  $\varepsilon_i$  be positive numbers satisfying:

 $(\star) \quad \varepsilon_0 = \varepsilon, \quad \varepsilon_{i+1} < \frac{\varepsilon_i}{2}, \qquad i = 0, \ldots, n_0.$ 

#### 6.1. Description of the algorithm

Suppose that, at the iteration v, we have a correct and regular string  $\mathscr{C}^{v} = \{A_{0}^{v}, \ldots, A_{m^{v}}^{v}\}$ , and denote by  $x^{v}$  the associated solution and by  $\alpha_{0}^{v}, \ldots, \alpha_{m^{v}}^{v}$  the corresponding amounts.

Determine  $t^{v} \in T$  such that

 $h^{\mathsf{v}}(x^{\mathsf{v}}) - (\langle x^{\mathsf{v}}, b(t^{\mathsf{v}}) \rangle - c(t^{\mathsf{v}}) + \delta(t^{\mathsf{v}}) \alpha_{0}^{\mathsf{v}}) \leq \varepsilon_{m^{\mathsf{v}}+1}$ 

with

$$h^{\mathsf{v}}(x) = \sup_{t \in T} \left( \langle x, b(t) \rangle - c(t) + \delta(t) \alpha_0^{\mathsf{v}} \right)$$

and put:

$$\begin{split} A^{\mathsf{v}}_{m^{\mathsf{v}}+1} &= \big\{ t^{\mathsf{v}} \big\}, \\ \alpha^{\mathsf{v}}_{m^{\mathsf{v}}+1} &= \big\langle x^{\mathsf{v}}, b(t^{\mathsf{v}}) \big\rangle - c(t^{\mathsf{v}}) + \delta(t^{\mathsf{v}}) \alpha^{\mathsf{v}}_{0}. \end{split}$$

We define the integer  $j^{v}$  by:

 $j^{\mathsf{v}} = \min(j \mid 0 \leq j \leq m^{\mathsf{v}} + 1; \alpha_{m^{\mathsf{v}}+1}^{\mathsf{v}} + \varepsilon_{m^{\mathsf{v}}+1} \leq \alpha_{j}^{\mathsf{v}} + \varepsilon_{j})$ 

[the fact that the above inequality is satisfied for a given integer j means that the corresponding sub-problem (see 5.1 and 5.3) has been sufficiently solved].

We will consider three cases according to the value of  $j^{v}$ :

First case:

 $j^{\mathbf{v}} = m^{\mathbf{v}} + 1.$ 

We introduce the new point  $t^{v}$  in the string  $\mathscr{C}^{v}$ .

Using the exchange theorem, in the string:

$$(A_0^{\mathsf{v}},\ldots,A_{m^{\mathsf{v}}}^{\mathsf{v}},A_{m^{\mathsf{v}}+1}^{\mathsf{v}} = \{t^{\mathsf{v}}\})$$

we exchange  $A_{m^{\vee}}^{\nu}$  and  $A_{m^{\vee}+1}^{\nu}$ . Then, we obtain:

either  $(A_0^{\mathsf{v}},\ldots,\widetilde{A}_{m^{\mathsf{v}}}^{\mathsf{v}},A_{m^{\mathsf{v}}+1})$ 

in which  $\widetilde{A}_{m}^{\nu}$  contains  $t^{\nu}$  but is not reduced to this single point,

or 
$$(A_{0}^{\prime}, \ldots, A_{m}^{\prime}-1, \{t^{\nu}\}, A_{m}^{\nu}),$$

what occurs in the case  $b(t^{v}) \in V_{m^{v-1}}^{v}$ .

In this latter case, we exchange again  $A_{m^{v}-1}^{v}$  and  $\{t^{v}\}$ , and so on, until we finally obtain:

either  $\widetilde{\mathscr{C}}^{\nu} = (A_0^{\nu}, \ldots, \widetilde{A}_{i'-1}^{\nu}, \widetilde{A}_{i'}^{\nu}, \ldots, A_{m^{\nu}}^{\nu}) \quad (i^{\nu} \geq 1),$ 

in which  $\tilde{A}_{t'-1}^{v}$  contains  $t^{v}$  but is not reduced to this single point,

or 
$$(\{t^{v}\}, A^{v}_{0}, \ldots, A^{v}_{m^{v}}).$$

R.A I.R.O. Analyse numérique/Numerical Analysis

But in this latter case, this means that  $x^{v}$  is an  $\varepsilon_{m^{v}+1}$ -solution of (P), with  $\varepsilon_{m^{v}-1} < \varepsilon$ : As a matter of fact,  $A_{m^{v}+1}^{v} = \{t^{v}\}$  is then a m.c.s. of V. By the remark 3.2, we will have  $t^{v} \in S$  and thus:

 $\alpha'_{m^{\prime}+1} = \langle x^{\prime}, b(t^{\nu}) \rangle - c(t^{\nu})$ will satisfy:

 $\alpha_{m^{\nu}+1}^{\nu} \leq \alpha.$ 

By the choice of  $t^{v}$ , we have:

$$h^{\mathsf{v}}(x^{\mathsf{v}}) \leq \alpha_{m^{\mathsf{v}}+1} + \varepsilon_{m^{\mathsf{v}}+1},$$

what implies that  $h^{\nu}(x^{\nu}) \leq \alpha + \varepsilon_{m^{\nu}+1}$ , i. e.  $x^{\nu}$  is a  $\varepsilon_{m^{\nu}+1}$ -solution of (P) and we stop the algorithm.

Second case:

 $1 \leq j^{\mathsf{v}} \leq m^{\mathsf{v}}$ .

Using again the exchange theorem, we then exchange  $A_{j^{\nu}-1}^{\nu}$  and  $A_{j^{\nu}}^{\nu}$  in the string  $\mathscr{C}^{\nu}$  (see § 5.4). This leads to the new string:

$$\widetilde{\mathscr{C}}^{\mathsf{v}} = (A_0^{\mathsf{v}}, \ldots, \widetilde{A}_{j^{\mathsf{v}}-1}^{\mathsf{v}}, \widetilde{A}_{j^{\mathsf{v}}}^{\mathsf{v}}, \ldots, A_{m^{\mathsf{v}}}^{\mathsf{v}}).$$

Note that  $\widetilde{A}_{j'-1}^{\nu}$  cannot be reduced to a single point, for it contains  $A_{j'}^{\nu}$  and the string  $\mathscr{C}^{\nu}$  has been supposed to be regular.

Third case:

$$j^{\prime} = 0.$$

Then, we have:

 $h^{\mathsf{v}}(x^{\mathsf{v}}) \leq \alpha_{m^{\mathsf{v}}+1}^{\mathsf{v}} + \varepsilon_{m^{\mathsf{v}}+1} \leq \alpha_{0}^{\mathsf{v}} + \varepsilon_{0}$ 

and this means that  $x^{v}$  is an  $\varepsilon_{0}$ -solution of (P) and we stop the algorithm. In short, if we put:

$$k^{\mathsf{v}} = \begin{cases} i^{\mathsf{v}} & \text{if } j^{\mathsf{v}} = m^{\mathsf{v}} + 1, \\ j^{\mathsf{v}} & \text{if } 0 \leq j^{\mathsf{v}} \leq m^{\mathsf{v}}, \end{cases}$$

we see that we stop the computation (the accuracy  $\varepsilon$  being obtained) when  $k^{\nu} = 0$ . In the other cases, the last exchange executed concerns the m. c. s. the indices of which are  $k^{\nu} - 1$  and  $k^{\nu}$ . The m. c. s.  $A_i^{\nu}$ ,  $i = 0, \ldots, k^{\nu} - 2$  are not modified.

It can happen that the new m.c.s.  $\tilde{A}_{k}^{\vee}$  is reduced to a single point. In that case, we suppress it in the string (see § 5.5). Thus, we obtain a new regular string:

$$\mathscr{C}^{v+1} = (A_0^{v+1}, \ldots, A_{m^{v+1}}^{v+1})$$

in which  $m^{v+1}$  can be equal to  $m^v - 1$ ,  $m^v$  or  $m^v + 1$ .

By the remark 3.2, the string  $\mathscr{C}^{\nu+1}$  is also correct  $(A_0^{\nu+1} \cap S \neq \emptyset)$ .

#### 6.2. Properties of the algorithm

Suppose that  $k^{v} \geq 1$ . Then we have:

(a)  $A_k^{\nu+1} = A_k^{\nu}, \alpha_k^{\nu+1} = \alpha_k^{\nu}, k = 0, \dots, k^{\nu} - 2.$ (b) For all  $t \in \hat{A}^{\nu} = A_{k^{\nu}-1}^{\nu+1} \setminus A_{k^{\nu}-1}^{\nu}$ :

 $h^{\mathsf{v}}(x^{\mathsf{v}}) - (\langle x^{\mathsf{v}}, b(t) \rangle - c(t) + \delta(t) \alpha_0^{\mathsf{v}}) \leq \varepsilon_{k^{\mathsf{v}}}.$ 

(c)  $\alpha_{k^{\nu}-1}^{\nu+1} \ge \alpha_{k^{\nu}-1}^{\nu} + \gamma^{\nu}(\varepsilon_{k^{\nu}-1} - \varepsilon_{k^{\nu}})$ , where  $\gamma^{\nu}$  is a positive number given by:

$$\gamma^{v} = \begin{cases} \frac{1}{s^{v+1}} \sum_{t \in \hat{A}^{v}} \rho_{k^{v}-1}^{v+1}(t) & \text{if } k^{v} = 1, \\ \sum_{t \in \hat{A}^{v}} \rho_{k^{v}-1}^{v+1}(t) & \text{if } k^{v} \ge 2. \end{cases}$$

*Proof:* The point (a) follows directly from the definition of the algorithm. (b) We will consider two cases:

first case:

$$j^{\mathsf{v}} = m^{\mathsf{v}} + 1; \qquad k^{\mathsf{v}} = i^{\mathsf{v}}.$$

As we exchange  $A_{k^{\nu}-1}^{\nu}$  and  $\{t^{\nu}\}$ , we have:  $A_{k^{\nu}-1}^{\nu+1} \setminus A_{k^{\nu}-1}^{\nu} = \{t^{\nu}\}$  and by the choice of  $t^{\nu}$ , we have:

$$\langle x^{\mathsf{v}}, b(t^{\mathsf{v}}) \rangle - c(t^{\mathsf{v}}) + \delta(t^{\mathsf{v}}) \alpha_0 \geq h^{\mathsf{v}}(x^{\mathsf{v}}) - \varepsilon_{m^{\mathsf{v}}+1} \geq h^{\mathsf{v}}(x^{\mathsf{v}}) - \varepsilon_{k^{\mathsf{v}}}.$$

second case:

 $1 \leq j \leq m^{\nu}; \qquad k^{\nu} = j^{\nu}.$ 

We exchange  $A_{k^{\nu}-1}^{\nu}$  and  $A_{k^{\nu}}^{\nu}$ . Thus we have  $\hat{A}^{\nu} = A_{k^{\nu}}^{\nu}$ . Now, for all  $t \in A_{k^{\nu}}^{\nu}$ , we have:

$$\langle x^{\mathsf{v}}, b(t) \rangle - c(t) + \delta(t) \alpha_0^{\mathsf{v}} = \alpha_{k^{\mathsf{v}}}^{\mathsf{v}}$$

and by definition of  $k^{v} = j^{v}$ :

$$\alpha_{k^{\nu}}^{\nu} + \varepsilon_{k^{\nu}} \ge \alpha_{m^{\nu}+1}^{\nu} + \varepsilon_{m^{\nu}+1} \ge h^{\nu}(x^{\nu}).$$

(c) The proof is similar to the first part of the proof of theorem 3.1. We will not give it here.

R.A.I.R.O. Analyse numérique/Numerical Analysis

#### 6.3. Starting the algorithm

Generally, we wish to start the algorithm with an initial string  $\mathscr{C}^0$  consisting of a single m. c. s.  $A_0^0$  of V (with exactly n-l+1 elements). The determination of  $A_0^0$  can be difficult (even impossible). In the case, it is possible to modify the problem (P) without changing its amount and one part of its solutions in such a way that the determination of  $A_0^0$  for the new problem is very easy.

Suppose we know  $x_0 \in E$  and  $r \in \mathbf{R}$  such that the problem (P) has at least a solution  $\overline{x}$  satisfying:  $\|\overline{x} - x_0\| \leq r$  and let  $\theta \in \mathbf{R}$  be constant such that  $\theta < \alpha$ . Consider then the function:

$$z(x) = \eta \| x - x_0 \| + \theta$$

where  $\eta > 0$  satisfies the condition  $\theta + \eta r \leq \alpha$ , and the new minimization problem:

$$(\tilde{P}) \quad \tilde{\alpha} = \inf_{\substack{x \in W \\ g(x) \leq 0}} \tilde{f}(x)$$

where  $\tilde{f}(x) = \max(f(x); z(x))$ .

It is easy to prove that  $\alpha = \tilde{\alpha}$  and that the set of solutions of  $(\tilde{P})$  is exactly equal to the set of solution  $\bar{x}$  of (P) satisfying the condition:

$$\|\tilde{x}-x_0\| \leq \frac{\alpha-\theta}{\eta}.$$

Note that the function z(x) can be written

$$z(x) = \sup_{x' \in S'} (\langle x, \eta x' \rangle + \theta - \eta \langle x_0, x' \rangle)$$

where S' is the unit sphere of E. Thus, the function  $\tilde{f}$  has the same form as f, replacing  $\tilde{b}$  and  $\tilde{c}$  by suitable extensions  $\tilde{b}$  and  $\tilde{c}$  to  $S \cup S'$ . It is easy to choose  $A_0^0$  in S'.

#### 7. CONVERGENCE OF THE ALGORITHM

Before proving the convergence, we need a theoretical convergence result which is very similar to theorem 3.1 but corresponds to the form of the subproblems (see § 5.1).

#### 7.1. An auxiliary convergence result

Suppose that  $\mathscr{W}$  is an affine variety which is parallel to  $\mathscr{V}^{\perp}$  (where  $\mathscr{V}$  is defined as in paragraph 2) and consider the following minimization problem:

(SP) 
$$\beta = \inf_{x \in \mathscr{W}} h_0(x),$$
  
vol. 12, n° 4, 1978

with

$$h_0(x) = \max (f(x); g(x) + \alpha_0)$$
$$= \sup_{t \in T} (\langle x, b(t) \rangle - c(t) + \delta(t) \alpha_0)$$

Let  $A^v$ ,  $v=0, 1, \ldots$ , be an infinite sequence of m. c. s. of  $\mathscr{V}$  and let  $f^v$  and  $g^v$  be the corresponding functionals (as in § 3). Put:

$$\alpha^{\mathsf{v}} = \min_{x \in \mathscr{W}} \max\left(f^{\mathsf{v}}(x); g^{\mathsf{v}}(x) + \alpha_{0}\right) \leq \beta$$

and denote by  $\mathcal{W}^{v}$  the set of solutions.

If we suppose again that the set:

$$\hat{N} = \left\{ v \in \mathbf{N} \mid A^{\nu+1} \neq A^{\nu} \right\}$$

is infinite (put  $\hat{A} = A^{\nu+1} \setminus A^{\nu}$ , for  $\nu \in \hat{N}$ ) then we have the following result:

THEOREM: If, for all  $v \in \hat{N}$ , there exists  $x^{v} \in \mathcal{W}^{v}$  such that:

$$h_0(x^{\nu}) - (\langle x^{\nu}, b(t) \rangle - c(t) + \delta(t) \alpha_0) \leq \hat{\varepsilon}$$

for all  $t \in \hat{A}^{\vee}$ , then for any  $\varepsilon > \hat{\varepsilon}$ , there exists  $\mu \in \hat{N}$  such that:

$$h_0(x^{\mu}) - \alpha^{\mu} < \varepsilon$$

[this implies that  $x^{\mu}$  is an  $\varepsilon$ -solution of (SP)].

This result is in fact a particular case of theorem 6.1 in [4].

#### 7.2. Convergence of the algorithm

THEOREM: For an arbitrary positive number  $\varepsilon$ , the algorithm described in paragraph 6, after a finite number  $\mu$  of iterations, leads to an element  $x^{\mu} \in W$  which is an  $\varepsilon$ -solution of (P).

More precisely, for a given accuracy  $\varepsilon > 0$ , there exists an integer  $\mu$  (depending on  $\varepsilon$ ) such that the element  $x^{\mu} \in W$  and the first m. c. s.  $A_0^{\mu}$  of the string  $\mathscr{C}^{\mu}$  satisfy:

$$f(x^{\mu}) \leq \alpha_0^{\mu} + \varepsilon$$
 and  $g(x^{\mu}) \leq \varepsilon$ ,

where  $\alpha_0^{\mu} = \alpha_{A_{\beta}}$  is the corresponding amount (see § 2.4). As  $\alpha_0^{\mu} \leq \alpha$ , this implies that  $x^{\mu}$  is an  $\varepsilon$ -solution of (P).

*Proof:* We only have to prove that the algorithm stops, i. e. that there exists  $\mu$  such that  $k^{\mu} = 0$ .

Suppose that we have  $k^{\nu} \ge 1$ , for all  $\nu$  and show that this leads to a contradiction:

Let  $\tilde{k} = \liminf_{v \to \infty} k^{v}$ ,  $(1 \le \tilde{k} \le n_0 + 1)$ . There exists  $v_0$  such that for all  $v \ge v_0$ , we have  $k^{v} \ge \tilde{k}$ ; and the set:

$$\tilde{N} = \left\{ v \in \mathbf{N} \, \middle| \, v \ge v_0, \, k^v = \tilde{k} \right\}$$

is infinite.

Hence, for  $v \ge v_0$ , we have  $A_k^v = A_k$  and  $V_k^v = V_k$  (independent of v) for  $k=0,\ldots, \tilde{k}-2$ .

By the definition of the algorithm, we have:

$$\begin{aligned} &\alpha_{m^{\nu}+1}^{\nu} + \varepsilon_{m^{\nu}+1} > \alpha_{\tilde{k}-1}^{\nu} + \varepsilon_{\tilde{k}-1} \quad \text{for all } \nu \in \widetilde{N}. \\ &\text{As } \widetilde{k} \leq m^{\nu} + 1 \text{, we have } \varepsilon_{m^{\nu}+1} \leq \varepsilon_{\tilde{k}} \text{, hence:} \end{aligned}$$

(i)  $\alpha_{m^{\nu}+1}^{\nu} - \alpha_{\tilde{k}-1}^{\nu} > \varepsilon_{\tilde{k}-1} - \varepsilon_{\tilde{k}}$  for all  $\nu \in \tilde{N}$ .

Put  $V_{\tilde{k}-2} = \mathscr{V}$ . Thus  $A_{\tilde{k}-1}^{\nu}$  is a m.c.s. of  $\mathscr{V}$ . For all  $\nu \ge \nu_0$ , such that  $\nu \notin \tilde{N}$ , we have  $A_{\tilde{k}-1}^{\nu+1} = A_{\tilde{k}-1}^{\nu}$  and for all  $\nu \in \tilde{N}$ , by 6.2 *b*, we have:

$$h^{\mathsf{v}}(x^{\mathsf{v}}) - [\langle x^{\mathsf{v}}, b(t) \rangle - c(t) + \delta(t) \alpha_0^{\mathsf{v}}] \leq \varepsilon_{\tilde{k}}$$

for all  $t \in \hat{A}^{\vee} = A_{\tilde{k}-1}^{\vee+1} \setminus A_{\tilde{k}-1}^{\vee}$ .

The choice of the  $\varepsilon_i$  [see condition ( $\star$ ) in § 6] implies that  $\varepsilon_{\tilde{k}-1} - \varepsilon_{\tilde{k}} > \varepsilon_{\tilde{k}}$ . Using theorem 3.1 in the case  $\tilde{k} = 1$  and theorem 7.1 in the case  $\tilde{k} > 1$  (with  $\hat{\varepsilon} = \varepsilon_{\tilde{k}}$  and  $\varepsilon = \varepsilon_{\tilde{k}-1} - \varepsilon_{\tilde{k}}$ ) there exists  $\mu \in \tilde{N}$  such that:

$$h^{\mu}(x^{\mu}) - \alpha^{\mu}_{\tilde{k}-1} \leq \varepsilon_{\tilde{k}-1} - \varepsilon_{\tilde{k}}.$$

As we have  $\alpha_{m^{\mu}+1}^{\mu} \leq h^{\mu}(x^{\mu})$ , we obtain:

(ii) 
$$\alpha_{m^{\mu}+1}^{\mu} - \alpha_{\tilde{k}-1}^{\mu} \leq \varepsilon_{\tilde{k}-1} - \varepsilon_{\tilde{k}}.$$

The two inequalities (i) and (ii) are contradictory.

#### REFERENCES

- 1. C. CARASSO, L'algorithme d'échange en optimisation convexe, Thèse, Grenoble, 1973.
- 2. C. CARASSO, Un algorithme de minimisation de fonctions convexes avec ou sans contraintes : l'algorithme d'échange. 7th I.F.I.P. Conference on Optimization Techniques, Springer-Verlag, 1975, 2, pp. 268-282.

- 3 C CARASSO and P J LAURENT, Un algorithme pour la minimisation d'une fonctionnelle convexe sur une variété affine Séminaire d'Analyse numerique, Grenoble, 18 octobre 1973
- 4 C CARASSO and P J LAURENT, Un algorithme de minimisation en chaîne en optimisation convexe Séminaire d'Analyse numérique, Grenoble, 29 janvier 1976 SIAM J Control and Optimization, vol 16, No 2, 1978 pp 209-235
- 5 C CARASSO and P J LAURENT, Un algorithme general pour l'approximation au sens de Tchebycheff de fonctions bornées sur un ensemble quelconque Approximations-Kolloquium, Bonn, June 8-12, 1976, Lecture Notes in Math, No 556, Springer-Verlag, Berlin, 1976
- 6 E W CHENEY and A A GOLDSTEIN, Newton's Method for Convex Programming and Tchebyscheff Approximation, Num Math, vol 1, 1959, pp 253-268
- 7 A A GOLDSTEIN, Constructive Real Analysis, Harper's Series in Modern Mathematics, Harper and Row, 1967
- 8 P J LAURENT, Approximation et Optimisation, Hermann Paris, 1972
- 9 P J LAURENT, Exchange Algorithm in Convex Analysis Conference on Approximation theory, The Univ of Texas, Austin, 1973, Acad Press
- 10 R T ROCKAFELLAR, Convex Analysis, Princ Univ Press 1970
- 11 E L STIEFEL, Uber diskrete und lineare Tschebyscheff-Approximationen, Num Math, vol 1, 1959, pp 1-28
- 12 E L STIEFEL Numerical Methods of Tschebycheff Approximation In On numerical Approximation, R Langer, Ed, Univ of Wisconsin, 1959, pp 217-232
- 13 E L STIEFEL, Note on Jordan Elimination, Linear Programming and Tchebycheff Approximation, Num Math, vol 2, 1960, pp 1-17
- 14 H J TOPFER, Tschebyscheff-Approximation bei nicht erfullter Haarscher Bedingung, Zeits für angew Math und Mech, vol 45, 1965, T 81-T 82
- 15 H J TOPFER, Tschebyscheff-Approximation und Austauschverfahren bei nicht erfullter Haarscher Bedingung Tagung, Oberwolfach, 1965 I S N M 7, Birkhauser Verlag 1967, pp 71-89