Convergence of the stochastic cyclic proximal algorithm Sketch of the proof

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Given a family of maximal monotone operators $\{A_i\}_{i=1}^L$ on a real Hilbert space, Passty [5] studied among other things the iterative algorithm

$$x_{n+1} = (I + \gamma_{n+1}A_L)^{-1} \cdots (I + \gamma_{n+1}A_2)^{-1} (I + \gamma_{n+1}A_1)^{-1} x_n$$

where (γ_n) is a sequence of step sizes belonging to $\ell^2 \setminus \ell^1$. When the operator $\sum_{i=1}^{L} A_i$ is maximal and when $Z(\sum_{i=1}^{L} A_i) = (\sum_{i=1}^{L} A_i)^{-1}(\{0\})$ is non empty, he showed that the sequence of weighted averages

$$\bar{x}_n = \frac{\sum_{1}^n \gamma_i x_i}{\sum_{1}^n \gamma_i}$$

converges towards an element of $Z(\sum_{i=1}^{L} A_i)$. In these notes, we replace the operators A_i with random maximal monotone operators $A_i(\xi, \cdot)$ on the Euclidean space \mathbb{R}^N , where ξ is a random variable on some probability space (Ξ, \mathscr{T}, μ) (a rigorous definition of these operators is provided in [3]), and we consider the algorithm

$$x_{n+1} = (I + \gamma_{n+1}A_L(\xi_{n+1}, \cdot))^{-1} \cdots (I + \gamma_{n+1}A_2(\xi_{n+1}, \cdot))^{-1} (I + \gamma_{n+1}A_1(\xi_{n+1}, \cdot))^{-1} x_n$$

where (ξ_n) is a sequence of independent copies of ξ . In the manuscript [6], this algorithm is referred to as the Stochastic Passty's Algorithm. For any of the operators $A_i(s, \cdot)$, let

$$\mathcal{A}_i(\cdot) = \int A_i(s, \cdot) \,\mu(ds)$$

and assume that the monotone operator $\sum A_i$ is maximal¹. Still considering that $(\gamma_n) \in \ell^2 \setminus \ell^1$ and assuming that $\mathcal{Z} = Z(\sum_{i=1}^{L} A_i)$ is nonempty, the purpose of these notes is to show that under some mild conditions, the sequence (x_n) converges almost surely to a random variable U supported by \mathcal{Z} .

In these notes, we shall confine ourselves to the case L = 2. For L > 2, the proofs are tedious without being essentially different from the case L = 2. Let us define the sets S_i for i = 1, 2 similarly as in [3] (in general, we shall reuse the notations of that paper, just adding the index *i* to specify which operator we are dealing with).

The following proposition is the analogue of [2, Prop. 1] or [3, Prop. 2].

¹The issue of the maximality of the \mathcal{A}_i was studied in [3]. The maximality of a finite sum of maximal monotone operators is a classical question in monotone operator theory, see [4] or [1].

Proposition 0.1. Assume that $\mathcal{Z} \neq \emptyset$. Assume moreover that there exists $x_{\star} \in \mathcal{Z}$ such that the set

$$\mathcal{R}_2(x_\star) = \left\{ (\varphi_1, \varphi_2) \in \mathcal{S}^2_{A_1(\cdot, x_\star)} \times \mathcal{S}^2_{A_2(\cdot, x_\star)} : \int (\varphi_1 + \varphi_2) \, d\mu = 0 \right\}$$

is not empty. Then

- 1. The sequence (x_n) is bounded almost surely and in \mathcal{L}^2 .
- 2. It holds that

$$\mathbb{E}\Big[\sum_{n} \gamma_{n}^{2} \int \Big(\|A_{1,\gamma_{n+1}}(s,x_{n})\|^{2} + \|A_{2,\gamma_{n+1}}(s,x_{n}-\gamma_{n+1}A_{1,\gamma_{n+1}}(s,x_{n}))\|^{2} \Big) \mu(ds) \Big] < \infty.$$

3. The sequence $(||x_n - x_{\star}||)_n$ converges almost surely.

Proof. Observing that

$$\begin{aligned} x_{n+1} &= J_{2,\gamma_{n+1}}(\xi_{n+1}, J_{1,\gamma_{n+1}}(\xi_{n+1}, x_n)) \\ &= J_{2,\gamma_{n+1}}(\xi_{n+1}, x_n - \gamma_{n+1}A_{1,\gamma_{n+1}}(\xi_{n+1}, x_n)) \\ &= x_n - \gamma_{n+1}A_{1,\gamma_{n+1}}(\xi_{n+1}, x_n) - \gamma_{n+1}A_{2,\gamma_{n+1}}(\xi_{n+1}, x_n - \gamma_{n+1}A_{1,\gamma_{n+1}}(\xi_{n+1}, x_n)), \end{aligned}$$

writing

$$\xi = \xi_{n+1}, \quad \gamma = \gamma_{n+1}, \quad A_{1,\gamma} = A_{1,\gamma_{n+1}}(\xi_{n+1}, x_n)$$

and
$$A_{2,\gamma} = A_{2,\gamma_{n+1}}(\xi_{n+1}, x_n - \gamma_{n+1}A_{1,\gamma_{n+1}}(\xi_{n+1}, x_n))$$

and expanding

$$||x_{n+1} - x_{\star}||^{2} = ||x_{n} - x_{\star}||^{2} + 2\langle x_{n+1} - x_{n}, x_{n} - x_{\star}\rangle + ||x_{n+1} - x_{n}||^{2},$$

we obtain

$$||x_{n+1} - x_{\star}||^{2} = ||x_{n} - x_{\star}||^{2} - 2\gamma \langle A_{1,\gamma}, x_{n} - x_{\star} \rangle - 2\gamma \langle A_{2,\gamma}, x_{n} - x_{\star} \rangle + \gamma^{2} ||A_{1,\gamma} + A_{2,\gamma}||^{2}$$

= $||x_{n} - x_{\star}||^{2} - 2\gamma X_{1} - 2\gamma X_{2} + \gamma^{2} X_{3}.$

Considering the functions φ_1 and φ_2 specified in the statement, and writing $J_{1,\gamma} = J_{1,\gamma_{n+1}}(\xi_{n+1}, x_n)$) and $J_{2,\gamma} = J_{2,\gamma_{n+1}}(\xi_{n+1}, x_n - \gamma_{n+1}A_{1,\gamma_{n+1}}(\xi_{n+1}, x_n))$, we have

$$X_1 = \langle A_{1,\gamma} - \varphi_1(\xi), J_{1,\gamma} - x_\star \rangle + \gamma \langle A_{1,\gamma} - \varphi_1(\xi), A_{1,\gamma} \rangle + \langle \varphi_1(\xi), x_n - x_\star \rangle$$

$$\geq \gamma \|A_{1,\gamma}\|^2 - \gamma \langle \varphi_1(\xi), A_{1,\gamma} \rangle + \langle \varphi_1(\xi), x_n - x_\star \rangle$$

by the monotonicity of $A_1(s, \cdot)$, and

$$\begin{split} X_2 &= \langle A_{2,\gamma} - \varphi_2(\xi), J_{2,\gamma} - x_{\star} \rangle + \gamma \langle A_{2,\gamma} - \varphi_2(\xi), A_{2,\gamma} \rangle + \gamma \langle A_{2,\gamma} - \varphi_2(\xi), A_{1,\gamma} \rangle + \langle \varphi_2(\xi), x_n - x_{\star} \rangle \\ &\geq \gamma \langle A_{2,\gamma} - \varphi_2(\xi), A_{2,\gamma} \rangle + \gamma \langle A_{2,\gamma} - \varphi_2(\xi), A_{1,\gamma} \rangle + \langle \varphi_2(\xi), x_n - x_{\star} \rangle \\ &= \gamma \|A_{2,\gamma}\|^2 - \gamma \langle \varphi_2(\xi), A_{2,\gamma} \rangle + \gamma \langle A_{2,\gamma}, A_{1,\gamma} \rangle - \gamma \langle \varphi_2(\xi), A_{1,\gamma} \rangle + \langle \varphi_2(\xi), x_n - x_{\star} \rangle \end{split}$$

by the monotonicity of $A_2(s, \cdot)$. By expanding the term X_3 , we obtain altogether

$$\begin{aligned} \|x_{n+1} - x_{\star}\|^{2} &\leq \|x_{n} - x_{\star}\|^{2} - \gamma^{2}(\|A_{1,\gamma}\|^{2} + \|A_{2,\gamma}\|^{2}) + 2\gamma^{2}\langle\varphi_{1}(\xi) + \varphi_{2}(\xi), A_{1,\gamma}\rangle + 2\gamma^{2}\langle\varphi_{2}(\xi), A_{2,\gamma}\rangle \\ &- 2\gamma\langle\varphi_{1}(\xi) + \varphi_{2}(\xi), x_{n} - x_{\star}\rangle \\ &\leq \|x_{n} - x_{\star}\|^{2} - \gamma^{2}(1 - \beta^{-1})(\|A_{1,\gamma}\|^{2} + \|A_{2,\gamma}\|^{2}) + \gamma^{2}\beta(\|\varphi_{1}(\xi)\|^{2} + \|\varphi_{1}(\xi) + \varphi_{2}(\xi)\|^{2}) \\ &- 2\gamma\langle\varphi_{1}(\xi) + \varphi_{2}(\xi), x_{n} - x_{\star}\rangle \end{aligned}$$

where we used the inequality $|\langle a, b \rangle| \leq (\beta/2) ||a||^2 + ||b||^2/(2\beta)$, with $\beta > 0$ being otherwise arbitrary.

By assumption,

$$\int (\|\varphi_1(s)\|^2 + \|\varphi_1(s) + \varphi_2(s)\|^2) \,\mu(ds) < \infty.$$

Moreover $\mathbb{E}_n \langle \varphi_1(\xi_{n+1}) + \varphi_2(\xi_{n+1}), x_n - x_\star \rangle = 0$. Thus,

$$\mathbb{E}_{n} \|x_{n+1} - x_{\star}\|^{2} \leq \|x_{n} - x_{\star}\|^{2} - \gamma_{n+1}^{2} (1 - \beta^{-1}) \int \left(\|A_{1,\gamma_{n+1}}(s, x_{n})\|^{2} + \|A_{2,\gamma_{n+1}}(s, x_{n} - \gamma_{n+1}A_{1,\gamma_{n+1}}(s, x_{n}))\|^{2} \right) \mu(ds) + C\gamma_{n+1}^{2}.$$

Choose $\beta > 1$. By the supermartingale convergence theorem along with the assumptions $(\gamma_n) \in \ell^2$, the conclusions follow.

References

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