

Quantitative information on the strong convergence of algorithms via proof mining

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Overview

Proof mining

Framework

A theorem by Yao-Noor

A theorem by Wang-Cui

Krasnosel'skiĭ-Mann iteration with Tikhonov regularization terms

Outline

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Proof mining program → analyses of mathematical proofs with the help of proof theoretic techniques, including functional interpretations, in search of concrete new information: effective bounds, algorithms, weakening of premisses, ...

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- ▶ Ensure that we are always able to extract information for the corresponding quantitative versions
- ▶ Help navigate the original proof
- ▶ Allow to avoid certain non-essential principles
- ▶ Allow to obtain explicit bounds

Proof mining with the BFI

We use the Bounded Functional Interpretation (BFI) and its characteristic principles, enriched with a new base type for elements of the space and the (universal) axioms for the Hilbert space.

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- ▶ Usually proof mining disregards precise witnesses, caring only for bounds on them
- ▶ Completely new translation of formulas
- ▶ Independence on bounded parameters is made explicit (via the interpretation itself)

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Framework

Let H be a real Hilbert space with inner product $\langle \cdot \rangle$ and norm $\|\cdot\|$ and let $T : H \rightarrow 2^H$ be an operator in H .

T is **monotone** if

$$(x, y), (x', y') \in \Gamma(T) \Rightarrow \langle x - x', y - y' \rangle \geq 0.$$

A monotone operator T is **maximal monotone** if $\Gamma(T)$ is not properly contained in the graph of any other monotone operator on H .

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We denote by $\text{zer}(T)$ the (**nonempty**) set of all **zeros** of T .

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For $c > 0$, we use J_c to denote the **resolvent** of T , i.e. the single-valued function defined by

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Notation

- ▶ $B_D(p)$ denotes the closed ball of radius D , centered at p

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- ▶ PPA is a powerful and successful algorithm in finding a solution of maximal monotone operators.
- ▶ Starting from any initial guess $z_0 \in H$, the PPA generates a sequence which approximates **weakly** the solution.

How to modify the PPA so that strong convergence is guaranteed?

Let H be a Hilbert space. The proximal point algorithm with multi-parameters is the following algorithm

$$\boxed{z_{n+1} = \lambda_n u + \gamma_n z_n + \delta_n J_{c_n}(z_n) + e_n, n \geq 0,} \quad (\text{mPPA})$$

where

- ▶ $u \in H$ is given,
- ▶ $c_n > 0$,
- ▶ $\lambda_n, \gamma_n, \delta_n \in (0, 1)$
- ▶ $\lambda_n + \gamma_n + \delta_n = 1, \forall n \geq 0$.

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A theorem by Yao-Noor

$$z_{n+1} = \lambda_n u + \gamma_n z_n + \delta_n J_{c_n}(z_n) + e_n$$

Theorem

Let (z_n) be generated by (mPPA). Assume that

1. $\lim_{n \rightarrow \infty} \lambda_n = 0$;
2. $\sum_{n=0}^{\infty} \lambda_n = \infty$;
3. $0 < \liminf_{n \rightarrow \infty} \gamma_n \leq \limsup_{n \rightarrow \infty} \gamma_n < 1$;
4. $c_n \geq c$, where c is some positive constant;
5. $c_{n+1} - c_n \rightarrow 0$;
6. $\sum_{n=1}^{\infty} \|e_n\| < \infty$.

Then (z_n) converges strongly to a point $z \in \text{zer}(T)$ which is the nearest point projection of u onto $\text{zer}(T)$.

Quantitative version of Yao-Noor's theorem I

Let (z_n) be generated by mPPA. Assume that there exist $a, b, c, d \in \mathbb{N}$ and $s \in \text{zer}(T)$ and monotone functions $\ell, L, \Delta, \Gamma, E$ such that

- (i) $\forall k \in \mathbb{N} \forall n \geq \ell(k) \left(\lambda_n \leq \frac{1}{k+1} \right)$;
- (ii) $\forall k \in \mathbb{N} \left(\sum_{i=1}^{L(k)} \lambda_i \geq k \right)$;
- (iii) $\forall n \geq a \left(\frac{1}{a+1} \leq \gamma_n \leq 1 - \frac{1}{a+1} \right)$;
- (iv) $\forall n \in \mathbb{N} \left(c_n \geq \frac{1}{c+1} \right)$;
- (v) $\forall n \in \mathbb{N} \left(\frac{1}{\Delta(n)+1} \leq \eta_n \leq 1 - \frac{1}{\Delta(n)+1} \right), \eta_n \in \{\lambda_n, \gamma_n, \delta_n\}$;
- (vi) $\forall k \in \mathbb{N} \forall n \geq \Gamma(k) \left(|c_{n+1} - c_n| \leq \frac{1}{k+1} \right)$;
- (vii) $\forall k \in \mathbb{N} \forall n \in \mathbb{N} \left(\sum_{i=E(k)+1}^{E(k)+n} \|e_i\| \leq \frac{1}{k+1} \right)$.

Quantitative version of Yao-Noor's theorem II

Theorem

Under the assumptions (i)-(vii), for every $k \in \mathbb{N}$ and every $f : \mathbb{N} \rightarrow \mathbb{N}$

$$\exists n \leq \phi(k, f) \forall i, j \in [n, n + f(n)] \left(\|z_i - z_j\| \leq \frac{1}{k+1} \right),$$

where $\phi(k, f)$ is a recursive bound explicitly given in the proof.

Remarks

- ▶ The original proof of the theorem uses explicitly the lim sup of a certain sequence. From a logical point of view this amounts to using arithmetical comprehension which requires bar recursors in order to be interpreted.

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- ▶ We were able to eliminate this dependency by using the fact that the sequence in question is bounded and using a rational approximation to the lim sup instead.
- ▶ We were also able to eliminate an argument of weak compactness.
- ▶ Moreover countable choice (in the projection argument) was eliminated due to a previous observation by Kohlenbach.
- ▶ The extracted bound is **uniform** on the parameters (depends only on $a, b, c, d, s, \ell, L, \Delta, \Gamma, E$).

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A theorem by Wang-Cui

$$z_{n+1} = \lambda_n u + \gamma_n z_n + \delta_n J_{c_n}(z_n) + e_n$$

(mPPA)

(C₁) $\lim \lambda_n = 0,$

(C₂) $\sum_{n=0}^{\infty} \lambda_n = \infty,$

(C₃) $\liminf c_n > 0,$

(C₄) $\liminf \delta_n > 0,$

(C₅) $\sum_{n=1}^{\infty} \|e_n\| < \infty$ (or $\lim \frac{\|e_n\|}{\lambda_n} = 0$).

A theorem by Wang-Cui

$$z_{n+1} = \lambda_n u + \gamma_n z_n + \delta_n J_{c_n}(z_n) + e_n$$

(mPPA)

$$(C_1) \lim \lambda_n = 0,$$

$$(C_2) \sum_{n=0}^{\infty} \lambda_n = \infty,$$

$$(C_3) \liminf c_n > 0,$$

$$(C_4) \liminf \delta_n > 0,$$

$$(C_5) \sum_{n=1}^{\infty} \|e_n\| < \infty \left(\text{or } \lim \frac{\|e_n\|}{\lambda_n} = 0 \right).$$

Theorem

Let (z_n) be generated by (mPPA). Assume that conditions $(C_1) - (C_5)$ hold. Then (z_n) converges strongly to a point $z \in \text{zer}(T)$ (the nearest point projection of u onto $\text{zer}(T)$).

Quantitative conditions

$$(Q_1) \quad \forall n \in \mathbb{N} \left(\lambda_n \geq \frac{1}{h(n)} \right)$$

$$(Q_2) \quad \forall k \in \mathbb{N} \forall n \geq \ell(k) \left(\lambda_n \leq \frac{1}{k+1} \right)$$

$$(Q_3) \quad \forall k \in \mathbb{N} \left(\sum_{i=1}^{L(k)} \lambda_i \geq k \right)$$

$$(Q_4) \quad \forall n \in \mathbb{N} \left(\min\{c_n, \delta_n^2\} \geq \frac{1}{c} \right)$$

$$(Q_{5a}) \quad \forall k \in \mathbb{N} \forall n \in \mathbb{N} \left(\sum_{i=E(k)+1}^{E(k)+n} \|e_i\| \leq \frac{1}{k+1} \right)$$

$$(Q_{5b}) \quad \forall k \in \mathbb{N} \forall n \geq E(k) \left(\frac{\|e_n\|}{\lambda_n} \leq \frac{1}{k+1} \right)$$

Quantitative version of Wang-Cui's theorem

Theorem

Let (z_n) be generated by (mPPA). Assume $(Q_1) - (Q_4)$ and either (Q_{5a}) or (Q_{5b}) . Let $D \in \mathbb{N} \setminus \{0\}$ be such that $D \geq \max\{2 \|u - p\|, \|z_0 - p\|\}$, for some $p \in \text{zer}(T)$. Then for all $k \in \mathbb{N}$ and monotone function $f : \mathbb{N} \rightarrow \mathbb{N}$

$$\exists n \leq \mu(k, f) \exists z \in B_D(p) \forall i \in [n, f(n)] \left(s_i^z \leq \frac{1}{k+1} \right)$$

with $s_n^z := \|y_n - z\|^2$ and where $\mu(k, f)$ is a recursive bound explicitly given in the proof.

Quantitative version of Wang-Cui's theorem

$$\boxed{z_{n+1} = \lambda_n u + \gamma_n z_n + \delta_n J_{C_n}(z_n) + e_n} \quad (\text{mPPA})$$

$$\boxed{y_{n+1} = \lambda_n u + \gamma_n y_n + \delta_n J_{C_n}(y_n)} \quad (\text{mPPA}_e)$$

Theorem

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Some consequences

Corollary (Metastability for (mPPA_e))

For all $k \in \mathbb{N}$ and monotone function $f : \mathbb{N} \rightarrow \mathbb{N}$,

$$\exists n \leq \mu(4(k+1)^2 - 1, f) \forall i, j \in [n, f(n)] \left(\|y_i - y_j\| \leq \frac{1}{k+1} \right).$$

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Corollary ("Metastable" version of Asymptotic regularity)

For all $k \in \mathbb{N}$ and monotone function $f : \mathbb{N} \rightarrow \mathbb{N}$, we have

(i) $\exists n \leq \tilde{\mu}(k, f) \forall i \in [n, f(n)] \left(\|J_i(y_i) - y_i\| \leq \frac{1}{k+1} \right),$

(ii) $\exists n \leq \tilde{\mu}(2k+1, f) \forall i \in [n, f(n)] \left(\|J(y_i) - y_i\| \leq \frac{1}{k+1} \right),$

where

$$\tilde{\mu}(k, f) := \max\{\mu(16c^2(k+1)^2 - 1, \check{f} + 1), \ell(4cD(k+1) - 1)\}$$

$$\text{and } \check{f}(m) := f(\max\{m, \ell(4cD(k+1) - 1)\}).$$

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- ▶ We only required a weaker version of the metric projection principle, replacing the projection point by suitable approximations
- ▶ We were able to bypass the sequential weak compactness argument
- ▶ The original proof has a discussion by cases which, in our case, imposes a discussion by cases for each approximation to the projection point
- ▶ The conditions allow for $\gamma_n \equiv 0$, in which case (mPPA) reduces to (HPPA) and we can obtain a quantitative version (under the weakest conditions but with slightly more complicated bounds)

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The T-KM iteration

$$\boxed{x_{n+1} = \beta_n x_n + \lambda_n (T(\beta_n x_n) - \beta_n x_n)} \quad (\text{T-KM})$$

where

- ▶ $x_0 \in H$
- ▶ $(\beta_n), (\lambda_n) \subset (0, 1]$
- ▶ $T : H \rightarrow H$ is a nonexpansive mapping

Boţ, Csetnek and Meier obtained a **strong convergence** result for (T-KM) and for similar modifications of the forward-backward and the Douglas-Rachford algorithms.

Boţ, Csetnek and Meier's Theorem

$$x_{n+1} = \beta_n x_n + \lambda_n (T(\beta_n x_n) - \beta_n x_n)$$

(T-KM)

Theorem

Let $(\beta_n), (\lambda_n) \subset (0, 1]$ be real sequences satisfying:

- (i) $\lim \beta_n = 1$,
- (ii) $\sum_{n \geq 0} (1 - \beta_n) = \infty$,
- (iii) $\sum_{n \geq 1} |\beta_n - \beta_{n-1}| < \infty$,
- (iv) $\liminf \lambda_n > 0$,
- (v) $\sum_{n \geq 1} |\lambda_n - \lambda_{n-1}| < \infty$.

Consider the iterative scheme (T-KM) with an arbitrary starting point $x_0 \in H$ and a nonexpansive mapping $T : H \rightarrow H$ such that $\text{Fix} T \neq \emptyset$. Then (x_n) converges strongly to $\text{proj}_{\text{Fix} T}(0)$.

Quantitative conditions

$$(K_1) \quad \forall n \in \mathbb{N} (\beta_n \geq \frac{1}{h(n)})$$

$$(K_2) \quad \forall k \in \mathbb{N} \forall n \geq b(k) (|1 - \beta_n| \leq \frac{1}{k+1})$$

$$(K_3) \quad \forall k \in \mathbb{N} \left(\sum_{i=1}^{D(k)} (1 - \beta_i) \geq k \right)$$

$$(K_4) \quad \forall k \in \mathbb{N} \forall n \in \mathbb{N} \left(\sum_{i=B(k)+1}^{B(k)+n} |\beta_i - \beta_{i-1}| \leq \frac{1}{k+1} \right)$$

$$(K_5) \quad \forall n \in \mathbb{N} (\lambda_n \geq \frac{1}{\ell})$$

$$(K_6) \quad \forall k \in \mathbb{N} \forall n \in \mathbb{N} \left(\sum_{i=L(k)+1}^{L(k)+n} |\lambda_i - \lambda_{i-1}| \leq \frac{1}{k+1} \right)$$

Boţ, Csetnek and Meier's Theorem (Quantitative version)

$$\boxed{x_{n+1} = \beta_n x_n + \lambda_n (T(\beta_n x_n) - \beta_n x_n)} \quad (\text{T-KM})$$

Theorem

Let (z_n) be generated by (T-KM). Assume that the conditions $(K_1) - (K_6)$ hold. Then for all $k \in \mathbb{N}$ and monotone function $f : \mathbb{N} \rightarrow \mathbb{N}$

$$\exists n \leq \mu(k, f) \forall i, j \in [n, f(n)] \left(\|x_i - x_j\| \leq \frac{1}{k+1} \right)$$

Remarks

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- ▶ The proof of metastability does not require the projection argument nor sequential weak compactness
- ▶ In the presence of the projection point, one can follow the arguments of the proof and show that our theorems are indeed quantitative versions of the original theorems

References I

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Thank you!