

# Convergence of the iterates and rates for an inertial forward-backward descent method

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# Outline

- ▶ Setting;
- ▶ Inertial Forward-Backward splitting: a variation on a theme of Alvarez-Attouch, 99 (Oliny-Moudafi, 03, Pock-Lorenz, 14);
- ▶ Conditions for global convergence;
- ▶ Rates and convergence for accelerated descent algorithms.

# Setting

Minimize  $F(x) = f(x) + g(x)$ ,  $x \in X$  Hilbert space, with  $\nabla f$   $L$ -Lipschitz and  $g$  “simple”, meaning that

$$\text{prox}_{\tau g}(x) := \arg \min_x g(y) + \frac{1}{2\tau} \|x - y\|^2$$

can be evaluated. Method based on Forward-Backward splitting

$$x \mapsto T_x := (I + \tau \partial g)^{-1} (I - \tau \nabla f)(x)$$

for  $\tau \leq 1/L$ . (Or  $2/L$ .)

# Inertial Method

- ▶  $x_0 = y_0$ ;
- ▶  $x_n = T(y_{n-1}), n \geq 1$ ;
- ▶  $y_n = x_n + \alpha_n(x_n - x_{n-1}), n \geq 1$ ,

where  $\alpha_n \in [0, 1]$  will be specified.

# Convergence analysis

- This is a bit technical, we sketch the arguments (from Alvarez-Attouch, originally)
- First, if  $\Phi_n = \frac{1}{2}\|x_n - x^*\|^2$ , where  $x^*$  is a solution, then a computation shows that

$$\begin{aligned}\Phi_n - \Phi_{n+1} &= \frac{1}{2}\|x_n - x_{n+1}\|^2 + \langle y_n - x_{n+1}, x_{n+1} - x^* \rangle \\ &\quad - \alpha_n \langle x_n - x_{n-1}, x_{n+1} - x^* \rangle \quad (1)\end{aligned}$$

Then, using that  $-\nabla f(x^*) \in \partial g(x^*)$  and  $y_n - x_{n+1} - \tau \nabla f(y_n) \in \tau \partial g(x_{n+1})$ , and the monotonicity of  $\partial g$ , we find

$$\langle y_n - x_{n+1}, x_{n+1} - x^* \rangle + \tau \langle \nabla f(x^*) - \nabla f(y_n), x_{n+1} - x^* \rangle \geq 0$$

which is combined with the previous to give

$$\begin{aligned} \Phi_n - \Phi_{n+1} &\geq \delta_{n+1} + \tau \langle \nabla f(y_n) - \nabla f(x^*), x_{n+1} - x^* \rangle \\ &\quad - \alpha_n \langle x_n - x_{n-1}, x_{n+1} - x^* \rangle. \end{aligned}$$

where  $\delta_{n+1} := \|x_n - x_{n+1}\|^2/2$ .

Then, the co-coercivity of  $\nabla f$  yields

$$\begin{aligned} & \langle \nabla f(y_n) - \nabla f(x^*), x_{n+1} - x^* \rangle \\ & \geq \frac{1}{L} \|\nabla f(y_n) - \nabla f(x^*)\|^2 + \langle \nabla f(y_n) - \nabla f(x^*), x_{n+1} - y_n \rangle \\ & \geq -\frac{L}{4} \|x_{n+1} - y_n\|^2. \end{aligned}$$

and it follows

$$\Phi_n - \Phi_{n+1} \geq \delta_{n+1} - \frac{\tau L}{4} \|x_{n+1} - y_n\|^2 - \alpha_n \langle x_n - x_{n-1}, x_{n+1} - x^* \rangle.$$

Then, simple computations which only use  $0 \leq \alpha_n \leq 1$  eventually lead to

$$\Phi_{n+1} - \Phi_n - \alpha_n(\Phi_n - \Phi_{n-1}) \leq - \left(1 - \frac{\tau L}{2}\right) \frac{\|x_{n+1} - y_n\|^2}{2} + 2\alpha_n \delta_n$$

so that if  $\tau \leq 2/L$ ,

$$\Phi_{n+1} - \Phi_n - \alpha_n(\Phi_n - \Phi_{n-1}) \leq 2\alpha_n \delta_n.$$

By induction, we can deduce that

$$\Phi_{n+1} - \Phi_n \leq 2 \sum_{j=2}^n \left( \prod_{l=j}^n \alpha_l \right) \delta_n$$



If we denote  $\beta_{j,n} = \prod_{l=j}^n \alpha_l$ , we deduce

$$\Phi_{n+1} - \Phi_n \leq 2 \sum_{j=2}^n \beta_{j,n} \delta_j$$

Hence (summing from 2 to  $+\infty$ )

$$\lim_{n \rightarrow \infty} \Phi_n - \Phi_2 \leq \sum_{n=1}^{+\infty} \sum_{j=2}^n \beta_{j,n} \delta_j = 2 \sum_{j=2}^{+\infty} \delta_j \sum_{n=j}^{+\infty} \beta_{j,n}$$

(where we recall  $\Phi_n = \|x_n - x^*\|^2/2$ ). In fact, what we really can show is that  $\Phi_n$  converges as soon as

$$\sum_{j=2}^{+\infty} \delta_j \sum_{n=j}^{+\infty} \beta_{j,n} < +\infty.$$

## Convergence of $(x_n)_n$

One can show (cf Opial's thm) that if  $\Phi_n$  converges, then also  $x_n$  converges weakly to some  $x^*$  solution of the problem (because then subsequences converge to fixed points).

A classical assumption is  $\alpha_n \leq \alpha < 1$ , in which case  $\beta_{j,n} \leq \alpha^{n-j}$ ,  $\sum_{n \geq j} \beta_{j,n} \leq 1/(1 - \alpha)$ , and the condition for convergence boils down to  $\sum_{j \geq 2} \delta_j < +\infty$ , which needs to be shown separately (for instance, it is true if  $\alpha \leq 1/3$ , cf Lorenz-Pock 2014).

We want however to consider more general cases with  $\alpha_n \rightarrow 1$ .  
Assume now that

$$\alpha_n \leq \frac{n-1}{n+2}.$$

Then (if  $k - j \geq 2$ )

$$\beta_{j,k} \leq \prod_{n=j}^k \frac{n-1}{n+2} \leq \left( \frac{j+1}{k} \right)^3$$

and one can show

$$\sum_{k \geq j} \beta_{j,k} \leq \frac{j+5}{2}.$$

## Condition for convergence

In this case, we obtain the convergence of the iterates whenever

$$\sum_{j=2}^{+\infty} \delta_j \frac{j+5}{2} < +\infty,$$

that is, if

$$\sum_{j \geq 2} j \|x_{j+1} - x_j\|^2 < +\infty.$$

**Remark:** up to now, we only used the monotonicity of  $\partial g$  and co-coercivity of  $\nabla f$ , as well as  $\tau \leq 2/L$ .

## Descent rule for FB descent

Now, we use the fact that we are minimizing and recall the descent rule for the FB descent (cf Tseng'08, Beck-Teboulle'09, Nesterov'05). The point  $T\bar{x}$  is by definition a minimizer of the strongly convex function

$$f(\bar{x}) + \langle \nabla f(\bar{x}), x - \bar{x} \rangle + g(x) + \frac{\|x - \bar{x}\|^2}{2\tau}$$

Hence it follows that for all  $x \in X$  (using also the convexity of  $f$ ),

$$\begin{aligned} g(x) + f(x) + \frac{\|x - \bar{x}\|^2}{2\tau} &\geq \\ &g(x) + f(\bar{x}) + \langle \nabla f(\bar{x}), x - \bar{x} \rangle + \frac{\|x - \bar{x}\|^2}{2\tau} \geq \\ &g(T\bar{x}) + f(\bar{x}) + \langle \nabla f(\bar{x}), T\bar{x} - \bar{x} \rangle + \frac{\|T\bar{x} - \bar{x}\|^2}{2\tau} + \frac{\|x - T\bar{x}\|^2}{2\tau} \\ &\geq g(T\bar{x}) + f(T\bar{x}) + \left(\frac{1}{\tau} - L\right) \frac{\|T\bar{x} - \bar{x}\|^2}{2} + \frac{\|x - T\bar{x}\|^2}{2\tau}. \end{aligned}$$

## FB Descent rule

Hence if  $\tau \leq 1/L$  (here we loose something),

$$g(x) + f(x) + \frac{\|x - \bar{x}\|^2}{2\tau} \geq g(T\bar{x}) + f(T\bar{x}) + \frac{\|x - T\bar{x}\|^2}{2\tau}. \quad (*)$$

for any  $x, \bar{x}$ .

# Inertial gradient descent

Choose in this inequality  $\bar{x} = y_n$ ,  $T\bar{x} = Ty_n = x_{n+1}$ , and  $x = ((t_{n+1} - 1)x_n + x^*)/t_{n+1}$  where  $\alpha_n$  and  $t_n$  are related:  $\alpha_n = (t_n - 1)/t_{n+1}$ . Then calculations show that

$$\begin{aligned} & (t_{n+1}^2 - t_{n+1})(g(x_n) + f(x_n) - (g(x^*) + f(x^*))) \\ & \quad + \frac{1}{2\tau} \|-t_n x_n + x^* + (t_n - 1)x_{n-1}\|^2 \\ & \geq t_{n+1}^2(g(x_{n+1}) + f(x_{n+1}) - (g(x^*) + f(x^*))) \\ & \quad + \frac{1}{2\tau} \|(t_{n+1} - 1)x_n + x^* - t_{n+1}x_{n+1}\|^2 \end{aligned}$$

So that if  $t_{n+1}^2 - t_{n+1} \leq t_n^2$ , one deduces the recursion:



## Recursion for inertial gradient descent

$$\begin{aligned} t_n^2(F(x_n) - F(x^*)) - (t_n^2 - t_{n+1}^2 + t_{n+1})(F(x_n) - F(x^*)) \\ + \frac{1}{2\tau} \|x_{n-1} - x^* + t_n(x_n - x_{n-1})\|^2 \\ \geq t_{n+1}^2(F(x_{n+1}) - F(x^*)) \\ + \frac{1}{2\tau} \|x_n - x^* + t_{n+1}(x_{n+1} - x_n)\|^2 \end{aligned}$$

Hence (choosing  $\tau = 1/L$ ):

$$F(x_n) - F(x^*) \leq \frac{L}{2t_n^2} \|x_0 - x^*\|^2$$

## FISTA acceleration (Beck and Teboulle, 2008/09)

It consists in cancelling the red terms:  $t_{n+1} = \sqrt{1 + 4t_n^2}/2$ . Similar behaviour (a bit less “optimal”) with  $t_n = (n + 1)/2$ :

$$F(x_n) - F(x^*) \leq \frac{2L}{(n+1)^2} \|x_0 - x^*\|^2$$

In this case, observe that  $\alpha_n = (n - 1)/(n + 2)$ , so that as mentioned earlier, we obtain weak convergence of the iterates as soon as

$$\sum_{j \geq 2} j \|x_{j+1} - x_j\|^2 < +\infty. \quad (**)$$

However, we do not know if (\*\*) holds.

## A variant of FISTA

Assume that  $t_n$  increases a little slower, and  $\alpha_n$  goes slower to 1:  
For instance, choose  $t_n = 1 + (n - 1)/a$ , with  $a > 2$ , so that  
 $\alpha_n = (n - 1)/(n + a) < (n - 1)/(n + 2)$ . Then one still has weak  
convergence as soon as (\*\*). Observe that now

$$t_n^2 - t_{n+1}^2 + t_{n+1} = \frac{(a - 2)n + (a - 1)^2}{a^2}$$

so that from the recursion, we get in addition (from the red terms  
in the recursion) that

$$\sum_{n \geq 0} n(F(x_n) - F(x^*)) < +\infty$$

(which is still not (\*\*)).

Then, consider the inequality (\*) again, with as before  $\bar{x} = y_n$ ,  $T\bar{x} = Ty_n = x_{n+1}$ , but now  $x = x_n$ . It follows (with still  $\tau = 1/L$ )

$$F(x_n) + L \frac{\|x_n - y_n\|^2}{2} = \\ F(x_n) + L\alpha_n^2 \frac{\|x_n - x_{n-1}\|^2}{2} \geq F(x_{n+1}) + L \frac{\|x_n - x_{n+1}\|^2}{2},$$

hence, if  $\alpha_n = (n-1)/(n+a)$  and recalling  $\delta_n = \|x_n - x_{n-1}\|^2/2$ ,

$$(n+a)^2\delta_{n+1} - (n-1)^2\delta_n \\ \leq \frac{1}{L}(n+a)^2(F(x_n) - F(x^*) - (F(x_{n+1}) - F(x^*))).$$



# Conclusion

- ▶ We still don't know if the iterates of the “standard” FISTA algorithm converge.
- ▶ An “infinitesimal” modification yields a convergent algorithm with roughly the same estimate:

$$F(x_n) - F(x^*) \leq \frac{La^2}{2(n+a-1)^2} \|x_0 - x^*\|^2$$

- ▶ (In practice we observe a very similar behaviour.) (Including for the iterates!)

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Thank you for your attention