

# Convergence of FISTA (Beck & Teboulle 2009)

---

## Andersen Ang

ECS, Uni. Southampton, UK  
andersen.ang@soton.ac.uk  
Homepage [angms.science](http://angms.science)

Version: July 17, 2023  
First draft: Dec 21, 2019

## Content

Problem setup:  $\min f(x) + g(x)$   
FISTA / Proximal gradient method + Nesterov's acceleration  
Prerequisite  
A key lemma  
Convergence rate of FISTA  $\mathcal{O}(\frac{1}{k^2})$

Reference: Amir Beck and Marc Teboulle. "A fast iterative shrinkage-thresholding algorithm for linear inverse problems." SIAM journal on imaging sciences, 2009.

# Setup

## Problem setup

$$(\mathcal{P}) : \min_{\mathbf{x}} F(\mathbf{x}) = f(\mathbf{x}) + g(\mathbf{x}).$$

- ▶  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  smooth convex
  - ▶  $f$  is smooth
    - ▶  $f$  is continuously differentiable  $f \in \mathcal{C}^1$   
 $\iff \nabla f(\mathbf{x})$  exists for all  $\mathbf{x}$
    - ▶  $\nabla f$  is globally  $L_f$ -Lipschitz  
 $L_f > 0, \|\nabla f(\mathbf{x}) - \nabla f(\mathbf{y})\| \leq L_f \|\mathbf{x} - \mathbf{y}\|$
  - ▶  $f$  is convex  $f(\mathbf{x}) \geq f(\mathbf{x}_0) + \langle \nabla f(\mathbf{x}_0), \mathbf{x} - \mathbf{x}_0 \rangle$
- ▶  $g : \mathbb{R}^n \rightarrow \bar{\mathbb{R}}$  is convex and possibly non-smooth
- ▶  $(\mathcal{P})$  is assumed solvable  $\mathcal{X}^* := \underset{\mathbf{x}}{\operatorname{argmin}} F(\mathbf{x}) \neq \emptyset$
- ▶ Notation:  $\mathbf{x}^* \in \mathcal{X}^*$  and  $F^* = F(\mathbf{x}^*)$
- ▶ See [here](#) for details on solving this problem using proximal gradient method.
- ▶ This document: convergence of FISTA (an accelerated ISTA)
- ▶ The proof is purely algebraic tricks, not special, not inspiring

---

## Algorithm 1: FISTA

---

- 1 Initialize  $\mathbf{y}_1 = \mathbf{x}_0 \in \mathbb{R}^n$
- 2 Initialize stepsize  $t_1 = 1$
- 3 **while not converge do**
- 4 
$$\mathbf{x}_k = p_L(\mathbf{y}_k) \tag{1}$$
$$t_{k+1} = \frac{1 + \sqrt{1 + 4t_k^2}}{2} \tag{2}$$
$$\mathbf{y}_{k+1} = \mathbf{x}_k + \frac{t_k - 1}{t_{k+1}}(\mathbf{x}_k - \mathbf{x}_{k-1}) \tag{3}$$

- 
- ▶ Assume known  $L \geq L_f$
  - ▶ FISTA = Nesterov's acceleration applied on ISTA
  - ▶ FISTA  $\in$  accelerated proximal gradient algorithms

## Prerequisite

$$(\mathcal{P}) : \min_{\mathbf{x}} F(\mathbf{x}) = f(\mathbf{x}) + g(\mathbf{x}), \quad f \text{ is convex, } L_f\text{-smooth, } g \text{ is convex}$$

- ▶ **Local quadratic over-estimator:**  $f$  is  $L_f$ -smooth, so for any  $L \geq L_f$ ,

$$Q(\mathbf{x}; \mathbf{y}) := f(\mathbf{y}) + \langle \nabla f(\mathbf{y}), \mathbf{x} - \mathbf{y} \rangle + \frac{L}{2} \|\mathbf{x} - \mathbf{y}\|_2^2 + g(\mathbf{x})$$

- ▶ Let  $L \geq L_f$ , the **proximal gradient update operator** on  $g$ , denoted as  $p_L : \mathbb{R}^n \rightarrow \mathbb{R}^n$

$$p_L(\mathbf{y}) = \operatorname{argmin}_{\mathbf{x}} \left\{ \frac{L}{2} \left\| \mathbf{x} - \left( \mathbf{y} - \frac{1}{L} \nabla f(\mathbf{y}) \right) \right\|_2^2 + g(\mathbf{x}) \right\},$$

- ▶ **A Lemma associated to the update operator**  $\forall \mathbf{y}, L \geq L_f > 0$  s.t.  $F(p_L(\mathbf{y})) \leq Q(p_L(\mathbf{y}); \mathbf{y})$ , then  $\forall \mathbf{x}$

$$F(\mathbf{x}) - F(p_L(\mathbf{y})) \geq \frac{L}{2} \|p_L(\mathbf{y}) - \mathbf{y}\|_2^2 + L \langle \mathbf{y} - \mathbf{x}, p_L(\mathbf{y}) - \mathbf{y} \rangle. \quad (4)$$

See [here](#) for details and proof.

- ▶ **Special Pythagoras Theorem** A key theorem in the analysis

$$\text{Scalar case} \quad (b - a)^2 + 2(a - c)(b - a) = (b - c)^2 - (a - c)^2$$

$$\text{Vector case} \quad \|\mathbf{b} - \mathbf{a}\|_2^2 + 2\langle \mathbf{a} - \mathbf{c}, \mathbf{b} - \mathbf{a} \rangle = \|\mathbf{b} - \mathbf{c}\|_2^2 - \|\mathbf{a} - \mathbf{c}\|_2^2 \quad (5)$$

- ▶ Familiar with [the proof of Nesterov's Accelerated Gradient Descent](#)

## A key lemma

$$\mathbf{x}_k = p_L(\mathbf{y}_k) \quad (1)$$

$$F(\mathbf{x}) - F(p_L(\mathbf{y})) \geq \frac{L}{2} \|p_L(\mathbf{y}) - \mathbf{y}\|_2^2 + L \langle \mathbf{y} - \mathbf{x}, p_L(\mathbf{y}) - \mathbf{y} \rangle \quad (4)$$

**Lemma** For the sequence  $\{\mathbf{x}_k, \mathbf{y}_k\}_{k \in \mathbb{N}}$  produced by FISTA with constant stepsize  $\frac{1}{L}$  with  $L \geq L_f$ , then

$$\frac{2}{L} (t_k^2 \delta_k - t_{k+1}^2 \delta_{k+1}) \geq \|\mathbf{u}_{k+1}\|_2^2 - \|\mathbf{u}_k\|_2^2, \quad \forall k$$

where  $\delta_k := F(\mathbf{x}_k) - F^*$ ,  $\mathbf{u}_k := t_k \mathbf{x}_k - (t_k - 1) \mathbf{x}_{k-1} - \mathbf{x}^*$ .

## Proof

$$F(\mathbf{x}_k) - F(p_L(\mathbf{y}_{k+1})) \geq \frac{L}{2} \|p_L(\mathbf{y}_{k+1}) - \mathbf{y}_{k+1}\|_2^2 + L \langle \mathbf{y}_{k+1} - \mathbf{x}_k, p_L(\mathbf{y}_{k+1}) - \mathbf{y}_{k+1} \rangle \quad \text{let } \mathbf{x} = \mathbf{x}_k, \mathbf{y} = \mathbf{y}_{k+1} \text{ in (4)}$$

$$F(\mathbf{x}_k) - F(\mathbf{x}_{k+1}) \stackrel{(1)}{\geq} \frac{L}{2} \|\mathbf{x}_{k+1} - \mathbf{y}_{k+1}\|_2^2 + L \langle \mathbf{y}_{k+1} - \mathbf{x}_k, \mathbf{x}_{k+1} - \mathbf{y}_{k+1} \rangle$$

$$\delta_k - \delta_{k+1} \stackrel{\delta}{\geq} \frac{L}{2} \|\mathbf{x}_{k+1} - \mathbf{y}_{k+1}\|_2^2 + L \langle \mathbf{y}_{k+1} - \mathbf{x}_k, \mathbf{x}_{k+1} - \mathbf{y}_{k+1} \rangle \quad (6)$$

$$F^* - F(p_L(\mathbf{y}_{k+1})) \geq \frac{L}{2} \|p_L(\mathbf{y}_{k+1}) - \mathbf{y}_{k+1}\|_2^2 + L \langle \mathbf{y}_{k+1} - \mathbf{x}^*, p_L(\mathbf{y}_{k+1}) - \mathbf{y}_{k+1} \rangle \quad \text{let } \mathbf{x} = \mathbf{x}^*, \mathbf{y} = \mathbf{y}_{k+1} \text{ in (4)}$$

$$F^* - F(\mathbf{x}_{k+1}) \stackrel{(1)}{\geq} \frac{L}{2} \|\mathbf{x}_{k+1} - \mathbf{y}_{k+1}\|_2^2 + L \langle \mathbf{y}_{k+1} - \mathbf{x}^*, \mathbf{x}_{k+1} - \mathbf{y}_{k+1} \rangle.$$

$$-\delta_{k+1} \stackrel{\delta}{\geq} \frac{L}{2} \|\mathbf{x}_{k+1} - \mathbf{y}_{k+1}\|_2^2 + L \langle \mathbf{y}_{k+1} - \mathbf{x}^*, \mathbf{x}_{k+1} - \mathbf{y}_{k+1} \rangle. \quad (7)$$

$$\delta_k - \delta_{k+1} \geq \frac{L}{2} \|\mathbf{x}_{k+1} - \mathbf{y}_{k+1}\|_2^2 + L \langle \mathbf{y}_{k+1} - \mathbf{x}_k, \mathbf{x}_{k+1} - \mathbf{y}_{k+1} \rangle \quad (6)$$

$$-\delta_{k+1} \geq \frac{L}{2} \|\mathbf{x}_{k+1} - \mathbf{y}_{k+1}\|_2^2 + L \langle \mathbf{y}_{k+1} - \mathbf{x}^*, \mathbf{x}_{k+1} - \mathbf{y}_{k+1} \rangle \quad (7)$$

Re-arrange (6) and (7)

$$\frac{2}{L} (\delta_k - \delta_{k+1}) \geq \|\mathbf{x}_{k+1} - \mathbf{y}_{k+1}\|_2^2 + 2 \langle \mathbf{y}_{k+1} - \mathbf{x}_k, \mathbf{x}_{k+1} - \mathbf{y}_{k+1} \rangle. \quad (6)$$

$$-\frac{2}{L} \delta_{k+1} \geq \|\mathbf{x}_{k+1} - \mathbf{y}_{k+1}\|_2^2 + 2 \langle \mathbf{y}_{k+1} - \mathbf{x}^*, \mathbf{x}_{k+1} - \mathbf{y}_{k+1} \rangle. \quad (7)$$

Tricky step 1:  $(t_{k+1} - 1)(6) + (7)$  gives

$$\frac{2}{L} \left( (t_{k+1} - 1) \delta_k - t_{k+1} \delta_{k+1} \right) \geq t_{k+1} \|\mathbf{x}_{k+1} - \mathbf{y}_{k+1}\|_2^2 + 2 \langle t_{k+1} \mathbf{y}_{k+1} - \mathbf{x}^* - (t_{k+1} - 1) \mathbf{x}_k, \mathbf{x}_{k+1} - \mathbf{y}_{k+1} \rangle$$

Tricky step 2: multiply  $t_{k+1}$  to the inequality, and apply  $t_k^2 \stackrel{(2)}{=} t_{k+1}^2 - t_{k+1}$

$$\frac{2}{L} \left( t_k^2 \delta_k - t_{k+1}^2 \delta_{k+1} \right) \geq \|t_{k+1} \mathbf{x}_{k+1} - t_{k+1} \mathbf{y}_{k+1}\|_2^2 + 2 \langle t_{k+1} \mathbf{y}_{k+1} - \mathbf{x}^* - (t_{k+1} - 1) \mathbf{x}_k, t_{k+1} \mathbf{x}_{k+1} - t_{k+1} \mathbf{y}_{k+1} \rangle$$

Tricky step 3: let  $\mathbf{a} = t_{k+1} \mathbf{y}_{k+1}$ ,  $\mathbf{b} = t_{k+1} \mathbf{x}_{k+1}$ ,  $\mathbf{c} = (t_{k+1} - 1) \mathbf{x}_k + \mathbf{x}^*$

$$\frac{2}{L} \left( t_k^2 \delta_k - t_{k+1}^2 \delta_{k+1} \right) \geq \|\mathbf{b} - \mathbf{a}\|_2^2 + 2 \langle \mathbf{a} - \mathbf{c}, \mathbf{b} - \mathbf{a} \rangle \stackrel{(5)}{=} \|\mathbf{b} - \mathbf{c}\|_2^2 - \|\mathbf{a} - \mathbf{c}\|_2^2.$$

By (3)  $\mathbf{a} = t_{k+1} \mathbf{y}_{k+1} \stackrel{(3)}{=} t_{k+1} \mathbf{x}_k + (t_k - 1)(\mathbf{x}_k - \mathbf{x}_{k-1})$ , and by definition  $\mathbf{u}_k = t_k \mathbf{x}_k - (t_k - 1) \mathbf{x}_{k-1} - \mathbf{x}^*$  we have

$$\frac{2}{L} \left( t_k^2 \delta_k - t_{k+1}^2 \delta_{k+1} \right) \geq \underbrace{\|t_{k+1} \mathbf{x}_{k+1} - (t_{k+1} - 1) \mathbf{x}_k - \mathbf{x}^*\|_2^2}_{\|\mathbf{u}_{k+1}\|_2^2} - \underbrace{\|t_{k+1} \mathbf{x}_k + (t_k - 1)(\mathbf{x}_k - \mathbf{x}_{k-1}) - (t_{k+1} - 1) \mathbf{x}_k - \mathbf{x}^*\|_2^2}_{\|\mathbf{u}_k\|_2^2} \quad \square$$

### Theorem. Convergence of FISTA

The sequence  $\{\mathbf{x}, \mathbf{y}_k\}_{k \in \mathbb{N}}$  generated by FISTA with constant stepsize  $\frac{1}{L}$  with  $L > L_f$  fulfills

$$F(\mathbf{x}_k) - F^* \leq \frac{2L\|\mathbf{x}_0 - \mathbf{x}^*\|_2^2}{(k+1)^2} \quad \forall k.$$

To prove this, we need the key lemma and

- ▶ **Lemma 1** Let  $\{a_k, b_k\}$  be positive sequence that  $a_1 + b_1 \leq c, c > 0$  and  $a_k - a_{k+1} \geq b_{k+1} - b_k \quad \forall k$ . Then  $a_k \leq c$  for all  $k$
- ▶ **Lemma 2**  $t_k$  generated by FISTA with  $t_1 = 1$  satisfies  $t_k \geq \frac{k+1}{2}$  for all  $k$ .

Note that by the definition of  $\delta_k$ , the theorem can be expressed as

$$\delta_k \leq \frac{2LR_0^2}{(k+1)^2}$$

and this give the “hint” that we should start from the key lemma.

► **Proof** By the key lemma:

$$\frac{2}{L}(t_k^2 \delta_k - t_{k+1}^2 \delta_{k+1}) \geq \|\mathbf{u}_{k+1}\|_2^2 - \|\mathbf{u}_k\|_2^2, \quad \forall k.$$

► Let  $a_k := \frac{2}{L} t_k^2 \delta_k$ ,  $b_k := \|\mathbf{u}_k\|_2^2$ ,  
 $c := \|\mathbf{y}_1 - \mathbf{x}^*\|_2^2 = \|\mathbf{x}_0 - \mathbf{x}^*\|_2^2 = R_0^2$ , then

$$a_k - a_{k+1} \leq b_{k+1} - b_k,$$

by Lemma 1, assuming  $a_1 + b_1 \leq c$  then

$a_k = \frac{2}{L} t_k^2 \delta_k \leq R_0^2 = c$ . Inequality  $\frac{2}{L} t_k^2 \delta_k \leq R_0^2$  gives

$$F(\mathbf{x}_k) - F^* =: \delta_k \leq \frac{LR_0^2}{2t_k^2}.$$

Now apply lemma 2:

$$t_k \geq \frac{k+1}{2} \implies \frac{1}{t_k} \leq \frac{2}{k+1} \implies \frac{1}{t_k^2} \leq \frac{4}{(k+1)^2}$$

$$F(\mathbf{x}_k) - F^* \leq \frac{2LR_0^2}{(k+1)^2}. \quad \square$$

► The last thing to do is to show  $a_1 + b_1 \leq c$

$$a_1 = \frac{2}{L} t_1^2 v_1 = \frac{2}{L} v_1$$

$$b_1 = \|\mathbf{u}_1\|_2^2 = \|\mathbf{x}_1 - \mathbf{x}^*\|_2^2$$

$$c = \|\mathbf{x}_0 - \mathbf{x}^*\|_2^2 = R_0^2$$

Apply Lemma (4) with  $\mathbf{x} = \mathbf{x}^*$ ,  $\mathbf{y} = \mathbf{y}_1$ ,  $p_L(\mathbf{y}_1) = \mathbf{x}_1$

$$F^* - F(\mathbf{x}_1) \geq \frac{L}{2} \|\mathbf{x}_1 - \mathbf{y}_1\|_2^2 + L \langle \mathbf{y}_1 - \mathbf{x}^*, \mathbf{x}_1 - \mathbf{y}_1 \rangle$$

$\iff$

$$\frac{2}{L} v_1 \geq \|\mathbf{x}_1 - \mathbf{y}_1\|_2^2 + 2 \langle \mathbf{y}_1 - \mathbf{x}^*, \mathbf{x}_1 - \mathbf{y}_1 \rangle$$

$$\stackrel{(5)}{=} \|\mathbf{y}_1 - \mathbf{x}^*\|_2^2 - \|\mathbf{x}_1 - \mathbf{x}^*\|_2^2$$

$$= R_0^2 - \|\mathbf{x}_1 - \mathbf{x}^*\|_2^2$$

Hence  $a_1 + b_1 \leq c$ .