# Gradient methods

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## **Contents**

1 Class of methods

1

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**Lemma 1.** Let  $q_i(x) = q_i^* + \frac{\mu}{2} \|x - \nu_i\|^2$ . Then for  $\lambda \in [0,1]$  we have for  $q_\lambda := (1-\lambda)q_0 + \lambda q_1$  the expression

$$(1 - \lambda)q_0(x) + \lambda q_1(x) = q_{\lambda}^* + \frac{\mu}{2} \|x - \nu_{\lambda}\|^2$$
 (1)

with

$$v_{\lambda} = (1 - \lambda)v_0 + \lambda v_1, \quad q_{\lambda}^* = (1 - \lambda)q_0^* + \lambda q_1^* + \frac{\lambda(1 - \lambda)\mu}{2} \|v_1 - v_0\|^2.$$
 (2)

*Furthermore* 

$$\operatorname{argmin}\left\{q_{\lambda}^{*} \mid 0 \leqslant \lambda \leqslant 1\right\} = P_{[0,1]}\left(\frac{1}{2} + \frac{q_{1}^{*} - q_{0}^{*}}{\mu \left\|\nu_{1} - \nu_{0}\right\|^{2}}\right). \tag{3}$$

*Proof.* The two functions in (1) are quadratic un x and  $\lambda$ . Since they coincide for  $\lambda = 0$  and  $\lambda = 1$ , it is sufficient to check that the second derivative with respect to  $\lambda$  of the right-hand side vanishes:

$$\frac{\partial^{2}}{\partial \lambda^{2}}\left\|x-\nu_{\lambda}\right\|^{2}=2\left\|\nu_{1}-\nu_{0}\right\|^{2}=-\frac{\partial^{2}\lambda(1-\lambda)}{\partial \lambda^{2}}\left\|\nu_{1}-\nu_{0}\right\|^{2}$$

For  $f \in \mathcal{F}_{\mu,L}(X)$  we have

$$f(x) \geqslant f(y) + \langle \nabla f(y), x - y \rangle + \frac{\mu}{2} \|x - y\|^2 = f(y) - \frac{1}{2\mu} \|\nabla f(y)\|^2 + \frac{\mu}{2} \left\| x - y + \frac{1}{\mu} \nabla f(y) \right\|^2$$

#### Algorithm 1: GM fixed step size

Input:  $x_0 \in X$ ,  $\varepsilon > 0$ .

Set  $v_0 = x_0$ ,  $q_0 = -\infty$  and k = 0.

- (1) If  $f(x_k) q_k \leqslant \epsilon$ : STOP
- (5)  $x_{k+1} = x_k \frac{1}{L}\nabla f(x_k)$

(3) 
$$w_k = x_k - \frac{1}{u} \nabla f(x_k), p_k = f(x_k) - \frac{\|\nabla f(x_k)\|^2}{2u}$$

$$Q_k = \frac{\mu}{2} \| v_k - w_k \|^2$$
,  $\lambda_k = P_{[0,1]} \left( \frac{1}{2} + \frac{p_k - q_k}{2Q_k} \right)$ 

$$q_{k+1} = (1 - \lambda_k)q_k + \lambda_k p_k + \lambda_k (1 - \lambda_k)Q_k$$

- (4)  $v_{k+1} = (1 \lambda_k)v_k + \lambda_k w_k$
- (6) Increment k and go to (1)

#### Algorithm 2: AGM fixed step size

Input:  $x_0 \in X$ ,  $\varepsilon > 0$ .

Set 
$$v_0 = x_0 - \frac{1}{\mu} \nabla f(x_0)$$
,  $q_0 = f(x_0) - \frac{\|\nabla f(x_0)\|^2}{2\mu}$  and  $k = 0$ .

- (1) If  $f(x_k) q_k \leqslant \epsilon$ : STOP
- (2)  $y_k = (1 \theta_k)v_k + \theta_k x_k$

(3) 
$$Q_k = \frac{\mu}{2} \left\| v_k - y_k + \frac{1}{\mu} \nabla f(y_k) \right\|^2$$

$$q_{k+1} = (1-\lambda_k)q_k + \lambda_k \left(f(y_k) - \frac{\|\nabla f(y_k)\|^2}{2\mu}\right) + \lambda_k (1-\lambda_k)Q_k$$

(4) 
$$v_{k+1} = (1 - \lambda_k)v_k + \lambda_k \left(y_k - \frac{1}{\mu}\nabla f(y_k)\right)$$

(5) 
$$x_{k+1} = y_k - \frac{1}{L} \nabla f(y_k)$$

(6) Increment k and go to (1)

For  $\theta_k < 1$  we have

$$v_k - y_k = \frac{\theta_k}{1 - \theta_k} (y_k - x_k)$$

such that

$$Q_k = \frac{\theta_k^2 \mu}{2(1-\theta_k)^2} \left\| x_k - y_k \right\|^2 + \frac{1}{2\mu} \left\| \nabla f(y_k) \right\|^2 + \frac{\theta_k}{1-\theta_k} \langle \nabla f(y_k), y_k - x_k \rangle.$$

Let  $e_k := f(x_k) - q_k$  Then we have

$$\begin{aligned} e_{k+1} \leqslant & f(y_k) - \frac{1}{2L} \|\nabla f(y_k)\|^2 - q_{k+1} \\ = & (1 - \lambda_k)(f(y_k) - q_k) - \frac{1}{2L} \|\nabla f(y_k)\|^2 + \lambda_k \frac{\|\nabla f(y_k)\|^2}{2\mu} - \lambda_k (1 - \lambda_k) Q_k \\ = & (1 - \lambda_k)e_k + \underbrace{(1 - \lambda_k)\langle\nabla f(y_k), y_k - x_k\rangle + \left(\frac{\lambda_k}{2\mu} - \frac{1}{2L}\right) \|\nabla f(y_k)\|^2 - \lambda_k (1 - \lambda_k) Q_k}_{Y} \end{aligned}$$

$$\begin{split} X = & (1 - \lambda_k) \left( 1 - \frac{\lambda_k \theta_k}{1 - \theta_k} \right) \left\langle \nabla f(y_k), y_k - x_k \right\rangle + \left( \frac{\lambda_k}{2\mu} - \frac{1}{2L} - \frac{\lambda_k (1 - \lambda_k)}{2\mu} \right) \left\| \nabla f(y_k) \right\|^2 \\ & - \frac{\lambda_k (1 - \lambda_k) \theta_k^2 \mu}{2(1 - \theta_k)^2} \left\| x_k - y_k \right\|^2 \\ = & \frac{(1 - \lambda_k) (1 - (1 + \lambda_k) \theta_k)}{1 - \theta_k} \left\langle \nabla f(y_k), y_k - x_k \right\rangle + \left( \frac{\lambda_k^2}{2\mu} - \frac{1}{2L} \right) \left\| \nabla f(y_k) \right\|^2 \\ & - \frac{\lambda_k (1 - \lambda_k) \theta_k^2 \mu}{2(1 - \theta_k)^2} \left\| x_k - y_k \right\|^2 \end{split}$$

Under the condition  $(1 + \lambda_k)\theta_k < 1$  we get

$$X \leqslant \left(\frac{2(1-\theta_k)^2}{4\lambda_k(1-\lambda_k)\theta_k^2\mu}\frac{(1-\theta_k)^2}{(1-\lambda_k)^2(1-(1+\lambda_k)\theta_k)^2} + \frac{\lambda_k^2}{2\mu} - \frac{1}{2L}\right)\left\|\nabla f(y_k)\right\|^2$$