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## Gradient projection method for convex functions and strongly convex sets

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#### Abstract

Let $\mathbb{H}$ be the real Hilbert space. $\langle p, x\rangle$ is scalar product for vectors $p, x \in \mathbb{H}$. Let $B_{R}(x)=\{y \in \mathbb{H}:\|y-x\| \leq R\}$. We denote the boundary of the set $A \subset \mathbb{H}$ by $\partial A$. The metric projection of the point $x \in \mathbb{H}$ on the set $A \subset \mathbb{H}$ is defined as follows: $P_{A}(x)=\left\{y \in A \mid\|x-y\|=\inf _{a \in A}\|x-a\|\right\}$. We denote the normal cone to the closed convex set $A$ at the point $a \in A$ by $N(A ; a)$, i.e. $N(A ; a)=\left\{p \in \mathbb{H}:\langle p, a\rangle=\sup _{x \in A}\langle p, x\rangle\right\}$.


Definition. [1, Definition 4.3.1] A nonempty set $A \subset \mathbb{H}$ is called strongly convex of radius $R$, if it can be represented as the intersection of closed balls of radius $R>0$.

Consider the minimization problem: $\min _{x \in A} f(x)$. We consider the standard gradient projection algorithm: $x_{1} \in \partial A, x_{k+1}=P_{A}\left(x_{k}-\alpha_{k} f^{\prime}\left(x_{k}\right)\right), \quad \alpha_{k}>0$.

Suppose that:
(i) $A \subset \mathbb{H}$ is strongly convex with radius $R$,
(ii) $f: \mathbb{H} \rightarrow \mathbb{R}$ is convex, differentiable and the gradient $f^{\prime}(x)$ satisfies the Lipschitz condition with constant $M>0$,
(iii) for any $k \in \mathbb{N}$ there exists a unit vector $n\left(x_{k}\right) \in N\left(A ; x_{k}\right)$ such that $\left\langle n\left(x_{k}\right), f^{\prime}\left(x_{k}\right)\right\rangle \leq 0$,
(iv) the problem (1) has a unique solution $x_{*} \in \partial A$.

Theorem 1. Suppose that conditions (i)-(iv) hold. Then for $\alpha_{k}=\alpha \in\left(0, \frac{2}{M}\right]$ we have the following estimate of the convergence rate of the algorithm:

$$
\left\|x_{k+1}-x_{*}\right\| \leq \sqrt[4]{\frac{R^{2}}{R^{2}+\alpha^{2}\left\|f^{\prime}\left(x_{k}\right)\right\|^{2}}}\left\|x_{k}-x_{*}\right\|
$$

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1. E. S. Polovinkin, M. V. Balashov. Elements of convex and strongly convex analysis. Fizmatlit, Moscow, 2007 (in Russian).
