

Subgradient method with Polyak's step in transformed space

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Outline

- 1 Polyak's subgradient method ($m = 1$)
- 2 Method A ($m \geq 1$)
- 3 Method B ($m \geq 1$, matrix B)
- 4 Comparison of the methods A and B

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Formulation of the problem

We consider the following problem

$$\text{to find } x^* = \arg \min_{x \in R^n} f(x) \quad \text{if } f^* \text{ is known,} \quad (1)$$

where $f(x)$ is a convex function and $f^* = f(x^*) = \min_{x \in R^n} f(x)$.

Main inequality for the subgradient

For convex function $f(x)$ the following inequality is valid

$$(x - x^*, g_f(x)) \geq f(x) - f^*, \quad \forall x \in R^n, \quad (2)$$

where $g_f(x)$ is a subgradient (gradient) of the function $f(x)$.

Inequality (2) is used for the calculation of the step in subgradient method, which B.T. Polyak offered in 1969.



Polyak B.T. Minimization of unsmooth functionals // *USSR Comput. Math. Math. Phys.* 1969. Vol. 9, No. 3, pp. 14-29.

Polyak's subgradient method

Polyak's subgradient method has the iterative form

$$x_{k+1} = x_k - h_k \frac{g_f(x_k)}{\|g_f(x_k)\|}, \quad h_k = \frac{f(x_k) - f^*}{\|g_f(x_k)\|}, \quad k=0, 1, \dots \quad (3)$$

Step h_k is called Polyak's step.

Decrease of the distance to the minimum point

Theorem 1 (Polyak, 1969)

The sequence $\{x_k\}_{k=0}^{\infty}$, generated by the method (3), satisfies the inequalities

$$\|x_{k+1} - x^*\|^2 \leq \|x_k - x^*\|^2 - \frac{(f(x_k) - f^*)^2}{\|g_f(x_k)\|^2}, \quad k = 0, 1, 2, \dots$$

Remark. Theorem 1 guarantees that in Polyak's method the distance to the minimum point decreases monotonically.

m -inequality for the subgradient

We consider methods A and B for convex functions $f(x)$, subgradients $g_f(x)$ of which satisfy the following condition:

$$(x - x^*, g_f(x)) \geq m(f(x) - f^*), \quad \forall x \in R^n, \quad (4)$$

where parameter $m \geq 1$.

The method A uses the step h_k in **original space**, the method B uses h_k in **transformed space** of variables.

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Method A

Initialization. f^* , $m \geq 1$, $x_0 \in R^n$, $\varepsilon > 0$.

A1. Calculate $f(x_k)$ and $g_f(x_k)$.

If $f(x_k) - f^* \leq \varepsilon$, then STOP ($k^* = k$, $x_\varepsilon^* = x_k$).

A2. Calculate the next point

$$x_{k+1} = x_k - h_k \frac{g_f(x_k)}{\|g_f(x_k)\|}, \quad h_k = \frac{m(f(x_k) - f^*)}{\|g_f(x_k)\|}.$$

A3. Go to the $(k + 1)$ -th iteration with x_{k+1} .

Decrease of the distance to the minimum point

Theorem 2 (Stetsyuk, 2012)

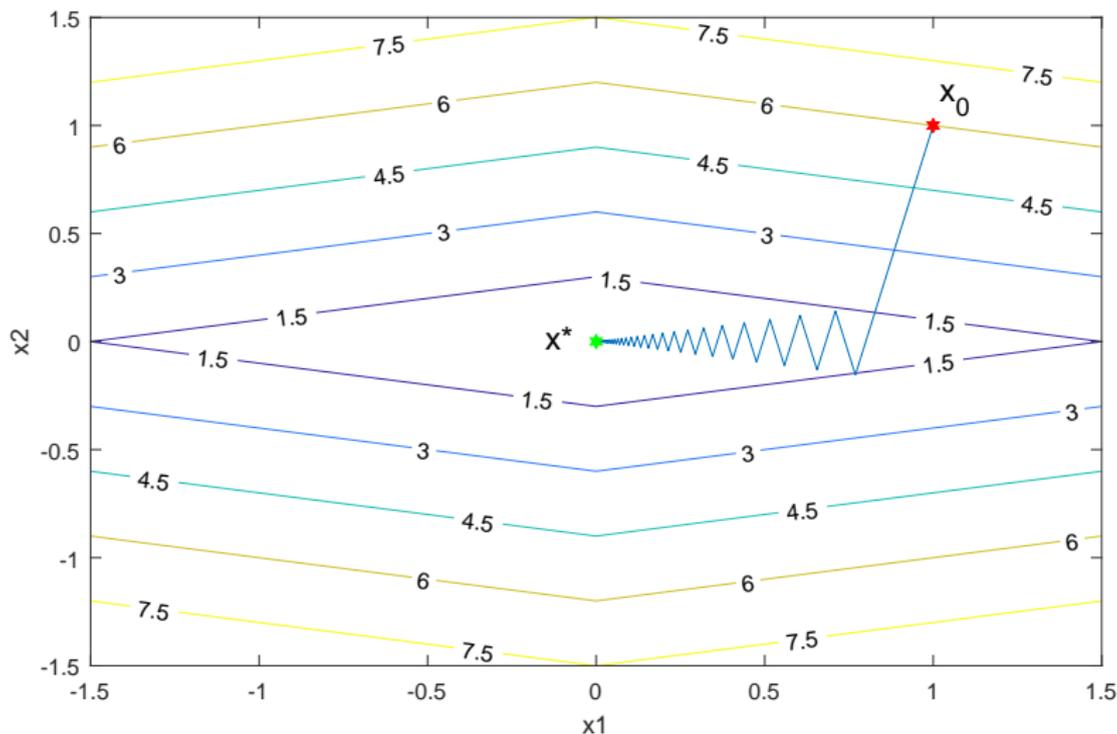
The sequence $\{x_k\}_{k=0}^{k^-1}$, generated by the method A, satisfies the inequalities*

$$\|x_{k+1} - x^*\|^2 \leq \|x_k - x^*\|^2 - \frac{m^2(f(x_k) - f^*)^2}{\|g_f(x_k)\|^2}, \quad k = 0, 1, 2, \dots \quad (5)$$

Remark. Theorem 2 guarantees that in Polyak's subgradient method the distance to the minimum point decreases monotonically if the inequality (4) is used.

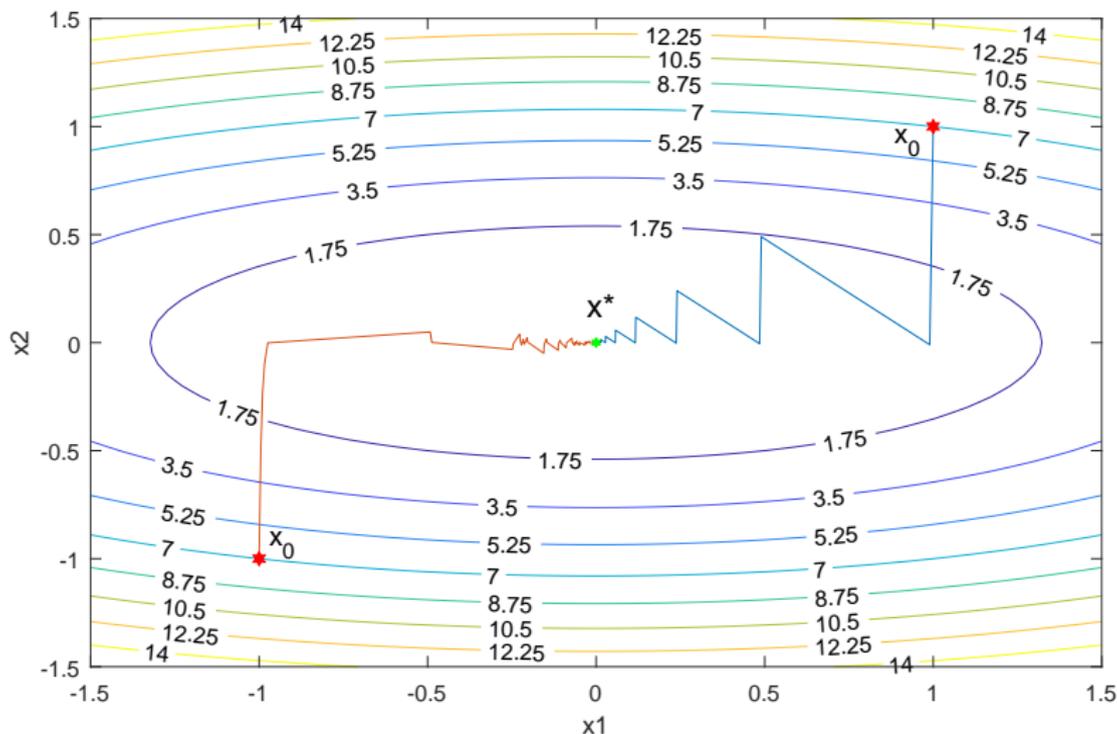
Method A for $f_1(x_1, x_2) = |x_1| + 5|x_2|$

$$m = 1, f^* = 0, x_0 = (1, 1)^T$$



Method A for $f_2(x_1, x_2) = x_1^2 + 6x_2^2$

Blue - $m = 2$, $x_0 = (1, 1)^T$ Red - $m = 1$, $x_0 = (-1, -1)^T$



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Method B

Initialization. f^* , $m \geq 1$, $x_0 \in R^n$, $n \times n$ matrix B , $\varepsilon > 0$.

A1. Calculate $f(x_k)$ and $g_f(x_k)$.

If $f(x_k) - f^* \leq \varepsilon$, then STOP ($k^* = k$, $x_{\varepsilon}^* = x_k$).

A2. Calculate the next point

$$x_{k+1} = x_k - h_k B \frac{B^T g_f(x_k)}{\|B^T g_f(x_k)\|}, \quad h_k = \frac{m(f(x_k) - f^*)}{\|B^T g_f(x_k)\|}.$$

A3. Go to the $(k + 1)$ -th iteration with x_{k+1} .

Here h_k is the Polyak's step in the transformed space of variables.

If B is an identity matrix, the method B turns into the method A.

Decrease of the distance to the minimum point

Theorem 3 (Stetsyuk, Stovba, Chernousova, 2018)

The sequence $\{x_k\}_{k=0}^{k^*-1}$, generated by the method B, satisfies the inequalities

$$\|A(x_{k+1} - x^*)\|^2 \leq \|A(x_k - x^*)\|^2 - \frac{m^2(f(x_k) - f^*)^2}{\|B^T g_f(x_k)\|^2}, \quad k = 0, 1, \dots \quad (6)$$

Remark. Theorem 3 guarantees that in the method B the distance to the minimum point decreases monotonically in the transformed space if the inequality (4) is used.

Method B for $f_1(x_1, x_2) = |x_1| + 10|x_2|$

$$B = \begin{pmatrix} 1 & 0 \\ 0 & 1/\alpha \end{pmatrix}, x_0 = (1, 1)^T$$

ε_f	$\alpha = 1$	$\alpha = 1.5$	$\alpha = 2$	$\alpha = 3$	$\alpha = 4$	$\alpha = 5$
1.0e-02	262	114	62	19	17	13
1.0e-04	492	216	119	44	31	22
1.0e-06	722	319	177	70	45	31
1.0e-08	952	421	234	95	60	40
1.0e-10	1183	523	292	121	74	49

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1. Piecewise quadratic function

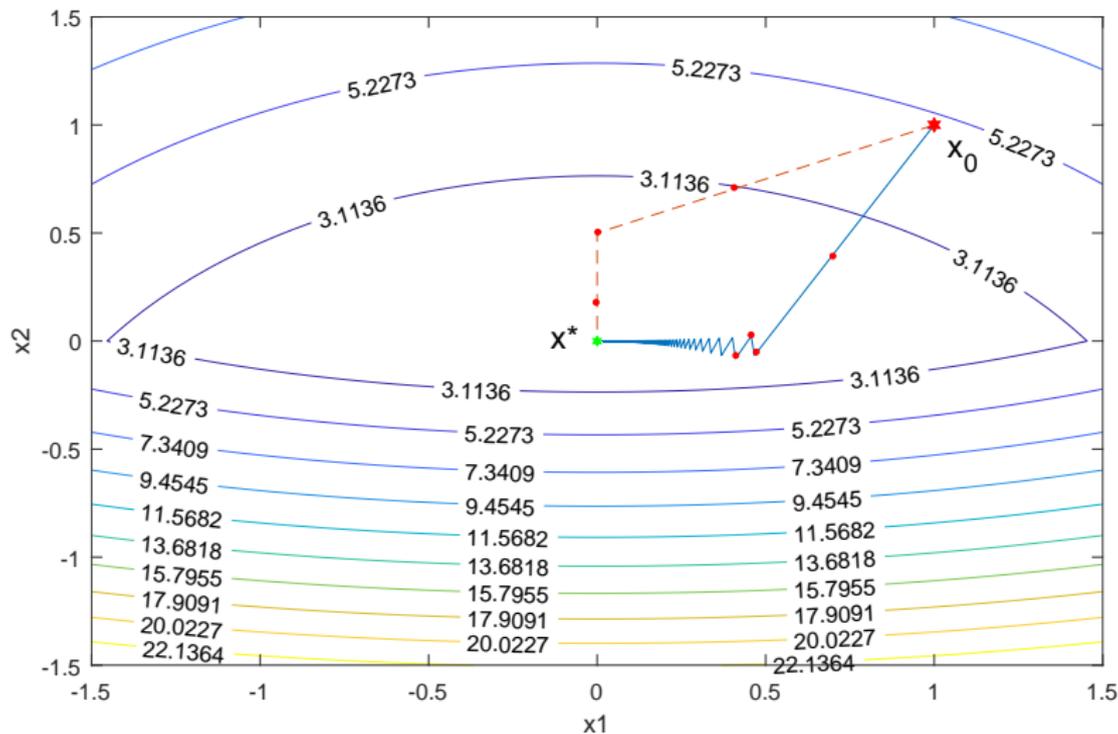
$$f_3(x_1, x_2) = \max \left\{ x_1^2 + (2x_2 - 2)^2 - 3, x_1^2 + (x_2 + 1)^2 \right\},$$
$$x^* = (0, 0), f^* = 1.$$

2. Quadratic function

$$f_4(x) = \|Ax - b\|^2, x^* = (1, 1, \dots, 1)^T, f^* = 0.$$

Methods A and B for piecewise quadratic function

$x_0 = (1, 1)^T$, $B = \text{diag}(1; 0.5)$, method A - blue, method B - red



Methods A and B for function $f_4(x) = \|Ax - b\|^2$

$$A = \|a_{ij}\|_{i,j=1}^{l,n} = \begin{pmatrix} 100 & 0 & 0 \dots 0 \\ 0 & 100 & 0 \dots 0 \\ & & A_1 \end{pmatrix}, \quad b_i = \sum_{j=1}^n a_{ij}, \quad i = \overline{1, l},$$

A_1 - 498×100 random matrix, $m = 2$.

ε	Method A		Method B	
	itnA	$\ x_\varepsilon^* - x^*\ $	itnB	$\ x_\varepsilon^* - x^*\ $
1.0e-04	695	8.0612e-04	86	1.2175e-04
1.0e-06	1085	8.2861e-05	108	1.1773e-05
1.0e-08	1491	8.3470e-06	130	1.1353e-06
1.0e-10	1901	8.3881e-07	150	1.3526e-07
1.0e-12	2313	8.3994e-08	172	1.3018e-08
1.0e-14	2725	8.4440e-09	194	1.2523e-09

Conclusions

1. Using parameter $m \geq 1$ in Polyak's subgradient method gives a possibility to minimize special classes of convex functions more effectively.
2. Application of space transformation can accelerate convergence of subgradient method with Polyak's step in the transformed space of variables.

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THANK YOU
FOR YOUR ATTENTION

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