

# Report on “Smooth saddle-point representation of a non-smooth function”

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May 26, 2024

## 1 SP-MD

The Section 5.2 describes the saddle point mirror descent algorithm as applying mirror descent to a product space  $Z = X \times Y$  using a specific mirror map  $\Phi(z) = a\Phi_X(x) + b\Phi_Y(y)$ . This approach utilizes a mirror map defined on the combined domain  $D = D_X \times D_Y$ , with coefficients  $a$  and  $b$  being positive real number that will be determined based on the problem’s specific requirements. The vector field  $g : Z \rightarrow \mathbb{R}^n \times \mathbb{R}^m$  is defined based on the gradients with respect on the function  $\phi$  at  $(x, y)$  positions.

The update formula for the algorithm is:

$$z_{t+1} \in \arg \min_{z \in Z \cap D} (\eta \langle g_t, z \rangle + D_\Phi(z, z_t)),$$

where  $g_t$  includes subgradients  $g_{X,t}$  and  $g_{Y,t}$  corresponding to the respective parts of the function  $\phi$ . The implementation of the SP-MD algorithm is available at: <https://github.com/WenbinWang024/Intern/blob/main/SP-MD.py>. This algorithm applies the SP-MD method by transforming the problem into an iterative optimization over the product space  $Z = X \times Y$  using a mirror map  $\Phi$ . Here,  $\Phi$  combines individual mirror maps on  $X$  and  $Y$  along with adjusted coefficients  $a$  and  $b$ , to accommodate the Lipschitz continuity properties of the two subspaces. The update step uses the Bregman divergence to measure distances between points, guiding the search direction for the next step.

## 2 SP-MP

The saddle point mirror prox is an advanced algorithm designed for scenarios where the function  $\phi$  is smooth. The algorithm is detailed under the assumption that  $\phi$  possesses certain smoothness properties defined by constants  $\beta_{11}, \beta_{12}, \beta_{22}$  and  $\beta_{21}$ . These properties specify the smoothness of the partial derivatives of  $\phi$  with respect to  $x$  and  $y$  over the domain  $X$  and  $Y$ . This method relies on smoothness to guarantee that the vector field  $g$  used in the algorithm maintains Lipschitz continuity under a suitably defined norm. The coefficients  $a$  and  $b$  in the mirror map  $\Phi(z) = a\Phi_X(x) + b\Phi_Y(y)$  are chosen based on the relationship to the Lipschitz constants and some intrinsic measures of the sets  $X$  and  $Y$ , like their radius. The implementation of the SP-MP algorithm is available at: <https://github.com/WenbinWang024/Intern/blob/main/SP-MP.py>. This algorithm enhances the mirror descent method through a two-stage optimization process suitable for scenarios where the function  $\varphi$  is smooth. It begins by computing an intermediate point  $w_{t+1}$  which introduces an additional predictive step. Then, the main iteration point  $z_{t+1}$  is updated in the principal step. Each step involves solving an optimization problem to minimize a modified objective function that includes the gradients of  $\varphi$  and the Bregman divergence  $D_\Phi$ . This method is designed to leverage the smoothness of  $\varphi$  to enhance the stability and convergence speed of the algorithm.

## 3 Application

The implementations of the applications are available at: <https://github.com/WenbinWang024/Intern/tree/main>.