# Gradient Sliding for SP with one composite

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## Establishing goals

We consider following saddle point problem with one composite:

$$\min_{x} \max_{y} F(x, y) + f(x)$$

where functions f(x) and F(x,y) are correspondingly  $L_f$  and  $L_F$  Lipshitz continuous

We also consider the sum of the two functions to be  $\mu_x$  and  $\mu_y$  strongly convex

# **Gradient Sliding**

- ► The first step is to use the gradient sliding algorithm (Kovalev et al. [2022]).
- ▶ The oracle of  $\nabla f(x)$  will be called  $\tilde{\mathcal{O}}\left(\sqrt{\frac{L_f}{\mu_x}}\right)$

**Problem:** we need to solve a suboptimization task.

### **FOAM**

The gradient sliding produces the following suboptimization task:

$$\min_{x} \max_{y} \left( \langle \nabla f(x_g^k), x - x_g^k \rangle + \|x - x_g^k\|^2 + F(x, y) \right)$$

► The complexity of this task via FOAM (Kovalev and Gasnikov [2022]) is  $\tilde{\mathcal{O}}\left(\frac{L_F}{\sqrt{(L_F + \mu_X) \cdot \mu_y}}\right)$ 

## Final complexity

The final oracle complexity for  $\nabla F$  can now be calculated:

$$\tilde{\mathcal{O}}\left(\sqrt{\frac{L_f}{\mu_x}}\right)\cdot\tilde{\mathcal{O}}\left(\frac{L_F}{\sqrt{(L_f+\mu_x)\cdot\mu_y}}\right)=\tilde{\mathcal{O}}\left(\frac{L_f}{\mu_x\mu_y}\right)$$

### The Experiment

The test is performed on a Bilinear Quadratic SP.

$$\min_{x} \max_{y} \left( \underbrace{p(x)}_{y} + \langle y, Ax \rangle - g(y) \right)$$

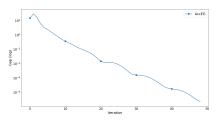


Figure: Visualized results

#### Conclusion

- ► A comparison with other methods is necessary
- Run method on more SP tasks
- ▶ The theory needs to be finished

### References

Dmitry Kovalev, Aleksandr Beznosikov, Ekaterina Borodich, Alexander Gasnikov, and Gesualdo Scutari. Optimal gradient sliding and its application to distributed optimization under similarity. 2022. doi: 10.48550/ARXIV.2205.15136. URL https://arxiv.org/abs/2205.15136.

Dmitry Kovalev and Alexander Gasnikov. The first optimal algorithm for smooth and strongly-convex-strongly-concave minimax optimization. 2022. doi: 10.48550/ARXIV.2205.05653. URL https://arxiv.org/abs/2205.05653.

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