Sign operator for (L_0, L_1) -smooth Optimization

Optimization with sparsity constraints is challenging yet crucial for modern applications.

This work introduces a novel approach to tackle these challenges efficiently.

The problem

to investigate optimization under (L_0, L_1) -smoothness assumption with heavy-tailed noise

Sign-
$$(L_0, L_1)$$
 Method

leverages the sign function for robust gradient updates

The solution

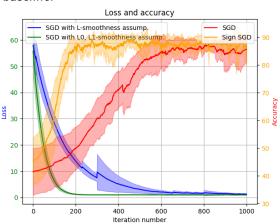
our results promise faster convergence and contribute to sparse solutions

- 1) set the problem with (L_0, L_1) and HT-noise conditions,
- 2) put the sign function to handle noisy gradients,
- 3) get improved bounds for convergence.

Graphical Highlights of Sign- (L_0, L_1)

Convergence rates improve significantly with Sign- (L_0, L_1) with HT-noise compared to baseline.

$$\begin{split} \|\nabla^2 f(x)\|_2 &\leq \\ L_0 + L_1 \|\nabla f(x)\|, \\ \text{Subjects:} \\ \text{Sign-based} \\ \text{methods, } (L_0, L_1) \\ \text{smoothness,} \\ \text{high-probability} \\ \text{convergence,} \\ \text{heavy-tailed noise.} \end{split}$$



Methods can be applied to problems with HT-noise and high communication costs.