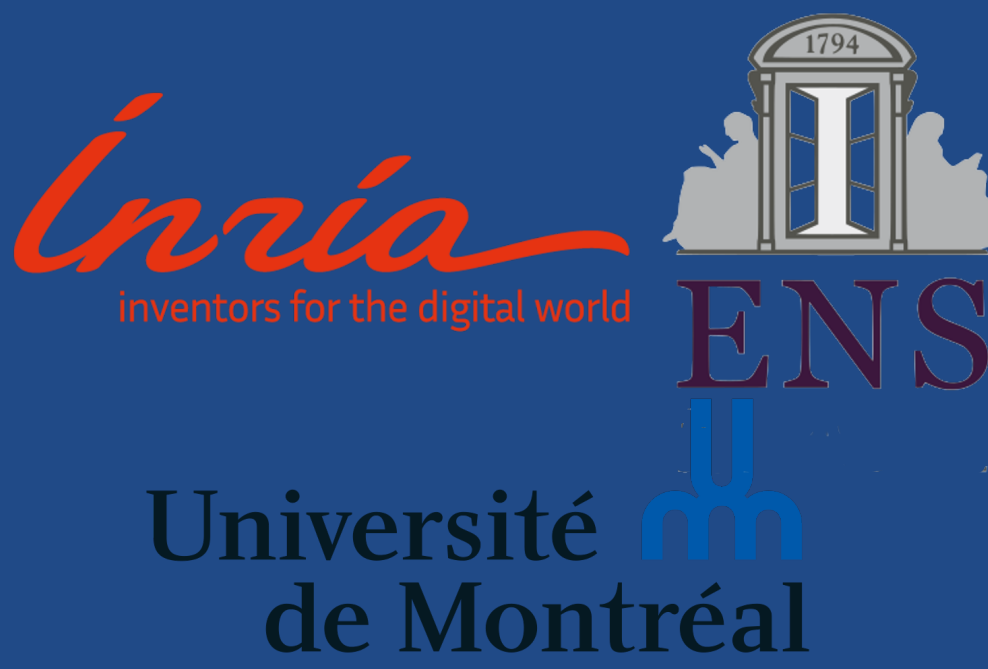


# Breaking the Nonsmooth Barrier: A Scalable Parallel Method for Composite Optimization

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## Summary

Optimization methods need to be adapted to the **parallel** setting to leverage modern computer architectures.

Highly efficient variants of stochastic gradient descent have been recently proposed, such as Hogwild [1], Kromagnon [2], ASAGA [3].

They assume that the objective function is smooth, so are inapplicable to problems such as Lasso, optimization with constraints, etc.

### Contributions:

1. **Sparse Proximal SAGA**, a sparse variant of the linearly-convergent proximal SAGA algorithm.
2. **ProxASAGA**, the first parallel asynchronous variance-reduced method that supports composite objective functions.

## Problem setting

We consider optimization problems of the form

$$\underset{\boldsymbol{x} \in \mathbb{R}^p}{\text{minimize}} \, f(\boldsymbol{x}) + h(\boldsymbol{x}), \quad \text{with } f(x) \stackrel{\text{def}}{=} \frac{1}{n} \sum_{i=1}^n f_i(\boldsymbol{x})$$

where  $f$  is differentiable with  $L$ -Lipschitz gradient. We also assume that  $h$  is block-separable and “simple”, in the sense that we have access to its proximal operator, defined as

$$\text{prox}_h \stackrel{\text{def}}{=} \arg \min_{\boldsymbol{x}} h(\boldsymbol{x}) + \frac{1}{2} \|\boldsymbol{x} - \boldsymbol{z}\|^2.$$

## A new algorithm: Sparse Proximal SAGA

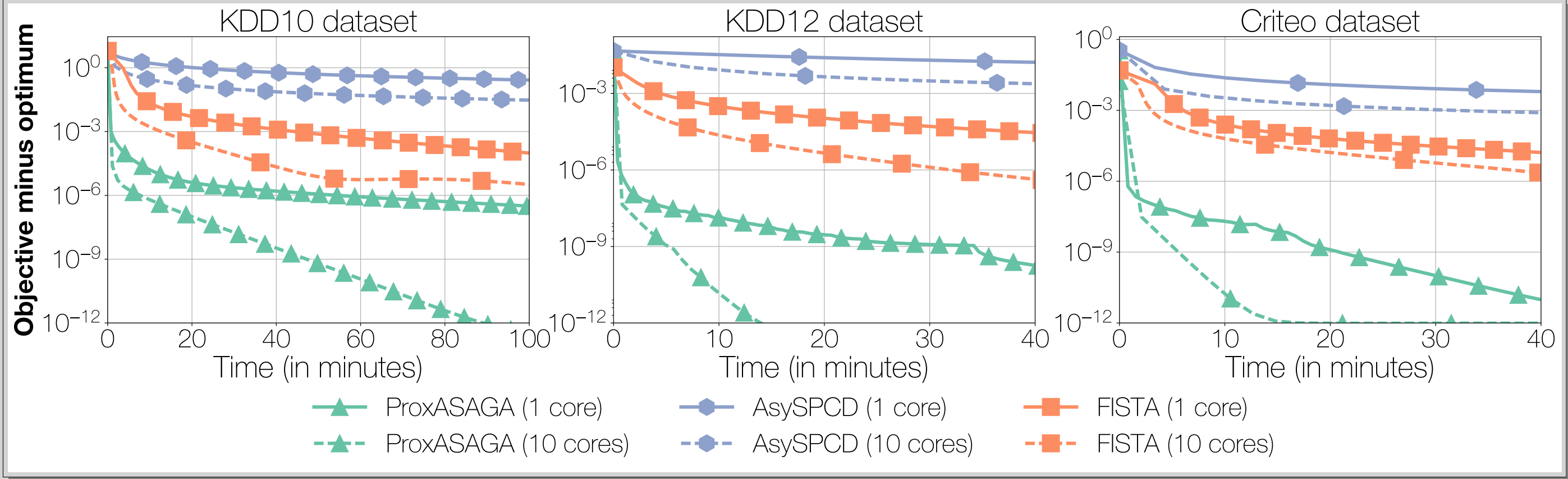
$$\boldsymbol{v}_i = \nabla f_i(\boldsymbol{x}) - \boldsymbol{\alpha}_i + \boldsymbol{D}_i, \quad \boldsymbol{x}^+ = \text{prox}_{\gamma \varphi_i}(\boldsymbol{x} - \gamma \boldsymbol{v}_i)$$

## Proximal Asynchronous SAGA (ProxASAGA)

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## Experimental results

### Comparison on 3 large-scale datasets



## References

1. Niu, F., Recht, B., Re, C. & Wright, S. *Hogwild: A lock-free approach to parallelizing stochastic gradient descent.* in *NIPS* (2011).
2. Mania, H. *et al.* Perturbed iterate analysis for asynchronous stochastic optimization. *SIAM Journal on Optimization* (2017).
3. Leblond, R., Pedregosa, F. & Lacoste-Julien, S. ASAGA: asynchronous parallel SAGA. *AISTATS* (2017).