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▶ To cite this version:

Théo Guyard, Gilles Monnoyer, Cédric Herzet, Clément Elvira. Safe peeling for L0-penalized problems. PGMO DAYS 2023, Nov 2023, Paris Saclay, France. hal-04313015

HAL Id: hal-04313015

https://hal.science/hal-04313015

Submitted on 28 Nov 2023

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Safe Peeling for L0-Penalized Problems

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Keywords: Sparse problems, ℓ_0 -norm, mixed-integer programming, bound tightening.

In many application fields such as machine learning, signal processing or inverse problems, it is of paramount interest to solve optimization problems of the form

$$\min_{\mathbf{x} \in \mathbb{R}^n} f(\mathbf{x}) + \lambda \|\mathbf{x}\|_0 \tag{1}$$

where $f(\cdot)$ is some loss function, $\|\cdot\|_0$ is the so-called ℓ_0 -norm that promotes sparsity in the optimizers by counting the number of non-zeros in its argument and $\lambda > 0$ is an hyperparameter [1]. This problem can be solved effectively via Branch-and-Bound (BnB) algorithms by adding a bound constraint $\mathbf{l} \leq \mathbf{x} \leq \mathbf{u}$ to reformulate the ℓ_0 -norm using a binary variable through Big-M relations [2]. In practice, the choice of the values in the bounds \mathbf{l} and \mathbf{u} must respect two conflicting imperatives. On the one hand, the additional constraint must not modify the solutions of the initial problem. Hence, the interval $[\mathbf{l}, \mathbf{u}] \subseteq \mathbb{R}^n$ should be chosen wide enough to contain all the optimizers of (1). On the other hand, the quality of the relaxations constructed across the BnB tree degrades with the spread of $[\mathbf{l}, \mathbf{u}]$. Hence, setting a too-wide interval is likely to increase the number of nodes explored during the BnB algorithm and penalize the solution time.

We propose a new strategy, dubbed *peeling*, to address this issue. By leveraging a duality property linking consecutive nodes, we locally tighten the interval $[\mathbf{l}, \mathbf{u}]$ which results in a strengthening of the relaxations constructed. This procedure is carried out in a *safe* manner, *i.e.*, we guarantee the validity of the BnB procedure and the optimality of the solutions obtained. Peeling can be seen as a generalization of the *screening* methodology recently introduced for ℓ_0 -penalized problems [3]. Through different numerical experiments, we show that peeling allows significant gains in the number of nodes explored and in the solution time.

References

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