

Heuristic Optimality Checks for Noise-Aware Sparse Recovery by ℓ_1 -Minimization

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Primal Problem

We consider the general class

$$\min_{x \in \mathbb{R}^n} \|x\|_1 \quad \text{s.t.} \quad \|Ax - b\| \leq \delta \quad (\text{P1})$$

of ℓ_1 -minimization problems. This class includes some of the most popular approaches for sparse reconstruction from incomplete linear measurements. Therein, $A \in \mathbb{R}^{m \times n}$ has full rank m , $\delta \geq 0$ is an estimate of measurement noise, $\|\cdot\|$ is an arbitrary norm and $b \in \mathbb{R}^m$ satisfies $\|b\| > \delta$.

Dual Problem

With the dual norm $\|\cdot\|_*$ of $\|\cdot\|$, the dual problem of (P1) is

$$\max_{y \in \mathbb{R}^m} -b^\top y - \delta \|y\|_* \quad \text{s.t.} \quad \|A^\top y\|_\infty \leq 1. \quad (\text{D1})$$

Optimality Conditions

A solution \hat{x} is optimal for (P1) if and only if there exists $\hat{y} \in \mathbb{R}^m$ with

$$-A^\top \hat{y} \in \text{Sign}(\hat{x}) \quad \text{and} \quad A\hat{x} \in b + \delta \partial \|\hat{y}\|_*.$$

The tuple (\hat{x}, \hat{y}) is called a primal-dual optimal pair. A primal-dual optimal pair can also be characterized by the conditions

$$\|\hat{x}\|_1 + b^\top \hat{y} + \delta \|\hat{y}\|_* = 0 \quad , \quad \|A\hat{x} - b\| = \delta \quad \text{and} \quad \|A^\top \hat{y}\|_\infty = 1.$$

Motivation

- A typical feature of iterative solvers for (P1) is that convergence becomes slow towards the end.
- We propose a heuristic optimality check (HOC) that often allows to “jump” from an iterate to an approximately optimal solution.
- Given a point $x \in \mathbb{R}^n$ with approximate support S , the idea of HOC is to construct an optimal point \hat{x} with support $\hat{S} \subseteq S$ exploiting optimality conditions and duality.
- The HOC scheme can be integrated into any solver for (P1).

Algorithm: HOC for (P1)

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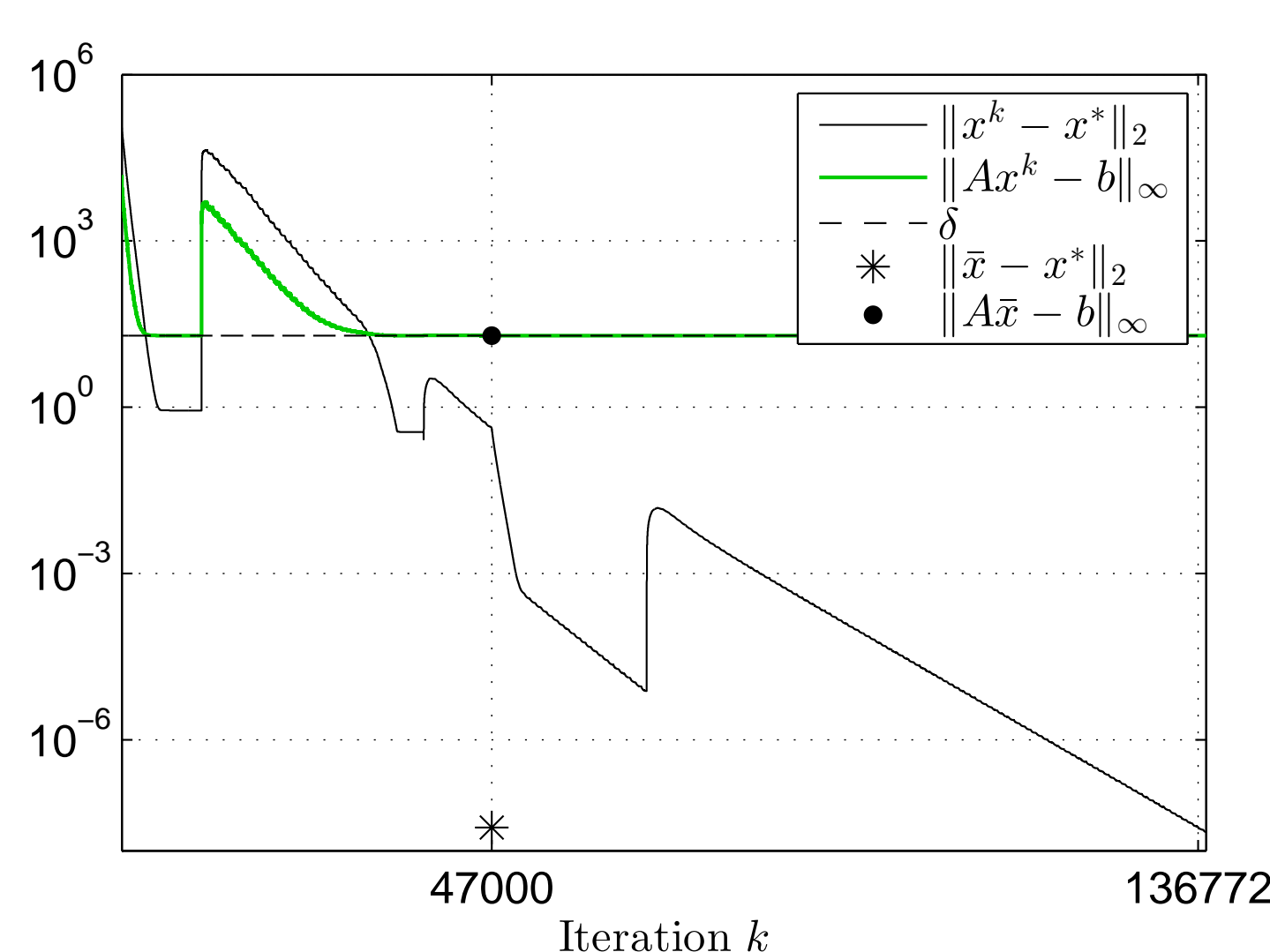
input:  $A \in \mathbb{R}^{m \times n}, b \in \mathbb{R}^m, \delta \geq 0, x \in \mathbb{R}^n$ 
1  $S \leftarrow$  approximate support deduced from  $x$ 
2  $\hat{y} \leftarrow$  approximate solution to  $-A_S^\top w = \text{sign}(x_S)$ 
3 if  $\|A^\top \hat{y}\|_\infty \approx 1$  then
4    $\hat{x}_S \leftarrow$  approximate solution to  $A_S z \in b + \delta \partial \|\hat{y}\|_*$ 
5    $\hat{x}_{S^c} \leftarrow 0$ 
6   if  $\|A\hat{x} - b\| \approx \delta$  then
7     if  $(\|\hat{x}\|_1 + \delta \|\hat{y}\|_* + b^\top \hat{y}) / \|\hat{x}\|_1 \approx 0$  then
8       return approximate primal-dual optimal pair  $(\hat{x}, \hat{y})$ 

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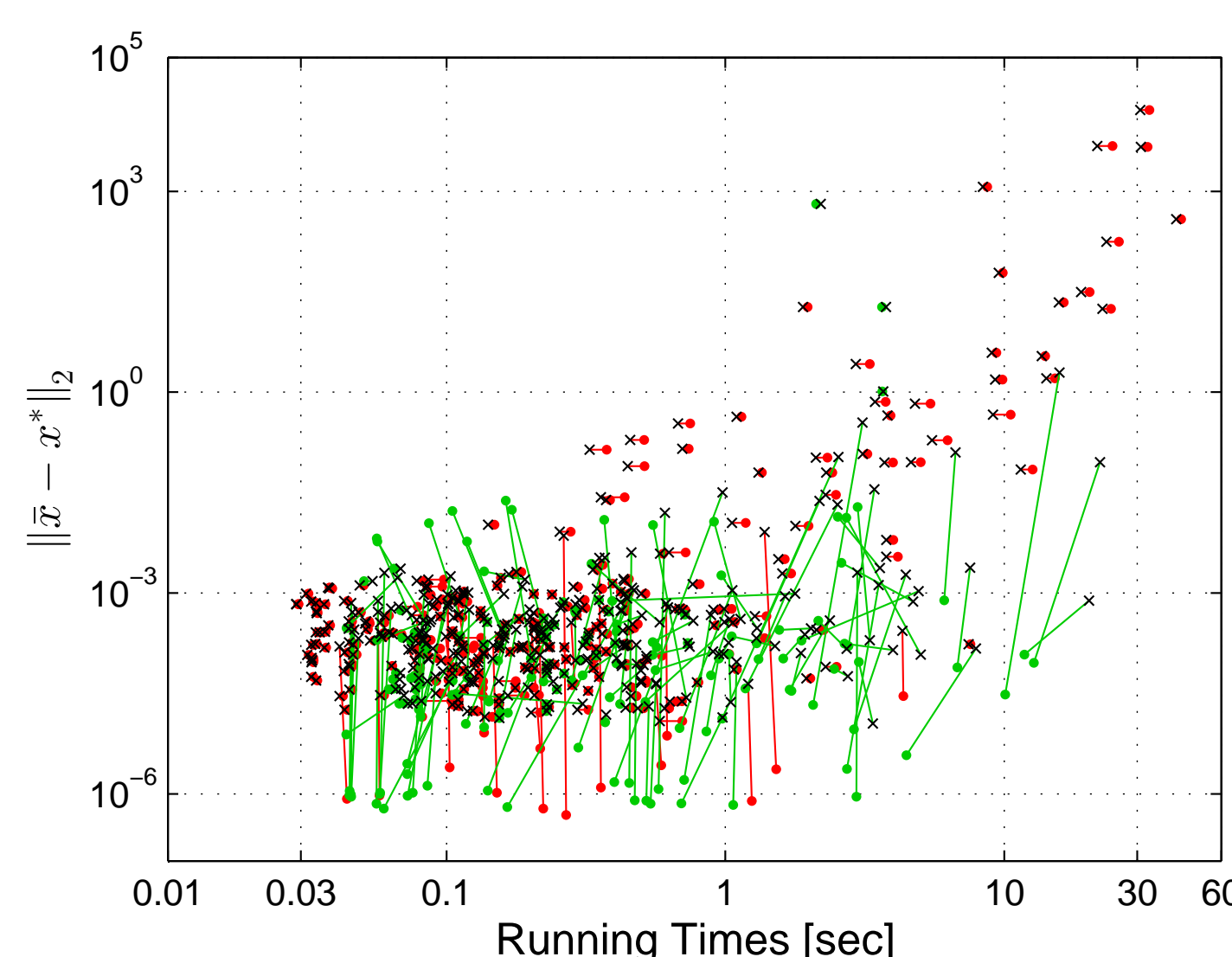
The choice of $\|\cdot\|$ affects Step 4 crucially. We display two examples:

- $\|\cdot\| = \|\cdot\|_\infty \implies \partial \|\hat{y}\|_* = \partial \|\hat{y}\|_1 = \text{Sign}(\hat{y})$ (multivalued)
- $\|\cdot\| = \|\cdot\|_2 \implies \partial \|\hat{y}\|_* = \partial \|\hat{y}\|_2 = \hat{y} / \|\hat{y}\|_2$ (singlevalued)

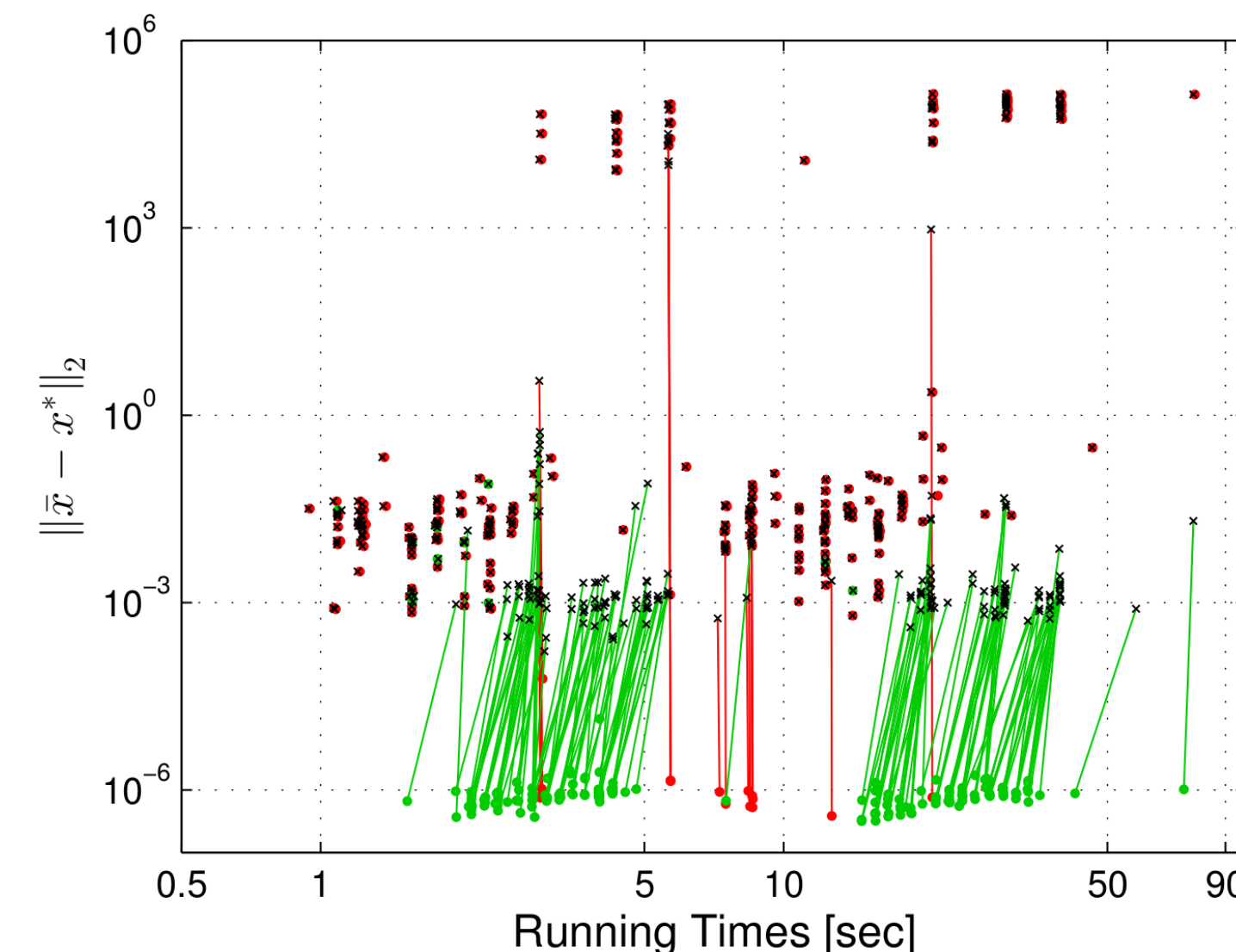
Numerical Results



Example for HOC efficiency in the incremental subgradient method, applied to (P1) with ℓ_∞ -constraints.



HOC impact within SPGL1 applied to (P1) with ℓ_2 -constraints. Average $\delta \approx 0.1$.



HOC impact within SolveBP/PDCO applied to $\min \lambda \|x\|_1 + \frac{1}{2} \|Ax - b\|_2^2$. Average $\lambda = 10$.

Middle and right: 444 test instances with varying solution sparsities and dimensions. Crossmarks represent results without HOC, dots those with HOC. Results for the same instance are connected by lines (green/red: faster/slower with HOC).

Conclusion

The HOC scheme is empirically demonstrated to allow for early termination and to improve solution speed and/or accuracy for different methods and a large number of instances of (P1).

Extensions and Future Work

- HOC schemes for the problems $\min \lambda \|x\|_1 + \frac{1}{2} \|Ax - b\|_2^2$ and $\min \|D^\top x\|_1$ s.t. $\|Ax - b\| \leq \delta$ exist as well.
- Our results can be used to generate test instances for (P1).

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