Exact Recovery of Partially Sparse Vectors

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Problem

We consider the general problem

 $\min_{x \in \mathbb{R}^n} f(x_1, x_2) \quad \text{s.t.} \quad A_1 x_1 + A_2 x_2 = b$

 x_1 is assumed to be sparse, x_2 is expected to be dense

Recovery Conditions

Theorem (weighted ℓ_1 **-** ℓ_2 **-norm case):** A point x^* with $x_2^* \neq 0$ is a solution of

 $\min_{x \in \mathbb{R}^n} \|x\|_{M,\alpha} \quad \text{s.t.} \quad Ax = Ax^*$

if and only if there exists $w^* \in \mathbb{R}^m$ such that

Mixed Norms

As objective function, we consider • the weighted ℓ_1 - ℓ_2 -norm

 $\|x\|_{M,\alpha} \coloneqq \|x_1\|_1 + \alpha \|x_2\|_2$

the Luxemburg norm

$$\begin{aligned} \|x\|_{L,\beta} &\coloneqq \inf\left\{\lambda > 0 \,:\, \|\frac{x_1}{\lambda}\|_1 + \beta \|\frac{x_2}{\lambda}\|_2^2 \le 1\right\} \\ &= \frac{1}{2} \|x_1\|_1 + \sqrt{\frac{1}{4}} \|x_1\|_1^2 + \beta \|x_2\|_2^2 \end{aligned}$$

Relationship

- $||x||_{M,\alpha} = ||x||_{L,\beta}$ if $x_1 = 0$ and $\beta = \alpha^2$ (or if $x_2 = 0$)
- $\beta = \alpha^2$ minimizes ratio of largest and smallest values of $\|\cdot\|_{L,\beta}$ on the unit sphere w.r.t. $\|\cdot\|_{M,\alpha}$

 $(A_1)^{\top} w^* \in \partial \|x_1^*\|_1$ and $(A_2)^{\top} w^* = \frac{\alpha}{\|x_2^*\|_2} \cdot x_2^*.$

(1)(2)

Theorem (Luxemburg norm case): A point $x^* \neq 0$ is a solution of

 $\min_{x \in \mathbb{R}^n} \|x\|_{L,\beta} \quad \text{s.t.} \quad Ax = Ax^*$

if and only if there exists $w^* \in \mathbb{R}^m$ such that

$$(A_{1})^{\top}w^{*} \in \left(\frac{1}{2} + \frac{\|x_{1}^{*}\|_{1}}{4\sqrt{\frac{1}{4}}\|x_{1}^{*}\|_{1}^{2} + \beta\|x_{2}^{*}\|_{2}^{2}}\right) \partial \|x_{1}^{*}\|_{1}$$
(3)
and $(A_{2})^{\top}w^{*} = \left(\frac{\beta}{\sqrt{\frac{1}{4}}\|x_{1}^{*}\|_{1}^{2} + \beta\|x_{2}^{*}\|_{2}^{2}}\right) \cdot x_{2}^{*}.$ (4)

Comparison of Unit Balls





3D-unit ball of $\|\cdot\|_{M,1}$ ($x_2 \in \mathbb{R}^2$)



3D-unit ball of $\|\cdot\|_{L,1}$ ($x_2 \in \mathbb{R}^2$)

Computational Experiments





Reconstruction from $b := A_1 x_1 + A_2 t$ with Gaussian $A_1 \in \mathbb{R}^{60 \times 200}$, $A_2 \in \mathbb{R}^{60}$ vector of all ones, some sparse x_1^* (left plot), and t = 2. Middle: recovered x_1 by the Luxemburg norm $\|\cdot\|_{L,\alpha^2}$, right: recovered x_1 by the mixed norm $\|\cdot\|_{M,\alpha}$, each with $\alpha = 4$.

Recovery error in the same setting as in leftmost figure. Left: error $||x_1^* - x_1||_2$ recovered with $||\cdot||_{L,16}$ (blue circles) and $||\cdot||_{M,4}$ (red stars) as a function of t^* . Right: error $|t^* - t|$ as a function of t^* . We conclude that the Luxemburg norms recovers x_1^* and t^* exactly for t^* large enough, the mixed norm for $t^* = 0$ only.



