Basis Pursuit Part II

Review of the null space property and stability

The basis pursuit the convex relaxation of the ℓ^0 -minimization problem, which is NP-hard in general. The null space property (NSP) of order s is a necessary and sufficient condition for the exact reconstruction of every s-sparse vector via the basis pursuit. Here, "exact reconstruction" means that every vector $x \in \mathbb{C}^N$ with at most s non-zero entries is the unique minimizer of the following optimization problem:

$$\min \{ ||z||_1 \, | \, Az = Ax \} \ .$$

Furthermore, one can see that if this is the case, then x is also the unique minimizer of the corresponding ℓ^0 -minimization problem.

The basis pursuit is stable under a sparsity defect in the vector x, if the measurement matrix satisfies the stable null space property (SNSP).

4.3 Robustness

The goal of this section is to take the noise in the measurements into account, additinally to the sparsity defect in the vector x. More precisely, we want to investigate the following optimization problem, for a given noise level η :

$$\min \left\{ \|z\|_1 \, | \, z \in \mathbb{C}^N, \ \|Az - (Ax + e)\| \le \eta \right\}. \tag{P_{1,\eta}}$$

For that we need to strengthen the assumptions on the measurement matrix A to satisfy the robust null space property (RNSP), depending on the ℓ^q -norm, in which the error bound is desired. We start with q = 1:

Definition: Robust null space property

A matrix $A \in \mathbb{C}^{m \times N}$ is said to satisfy the robust null space property with respect to $\|\cdot\|$ with the constants $\rho \in (0,1)$ and $\tau > 0$ relative to a set $S \subseteq [N]$ iff

$$\forall v \in \mathbb{C}^{N} : \|v_{S}\|_{1} \leq \rho \|v_{\overline{S}}\|_{1} + \tau \|Av\|.$$
 (RNSP(\|\cdot\|, \rho, \tau, S))

A satisfies the robust null space property of order s with respect to $\|\cdot\|$ with the constants $\rho \in (0,1)$ and $\tau > 0$ RNSP($\|\cdot\|, \rho, \tau, s$) iff A satisfies RNSP($\|\cdot\|, \rho, \tau, S$) for all sets $S \subseteq [N]$ with $|S| \le s$.

We note that the RNSP directly implies the SNSP. The main result is the following theorem.

Theorem: (Theorem 4.20 in the book)

A matrix $A \in \mathbb{C}^{m \times N}$ satisfies $\text{RNSP}(\|\cdot\|, \rho, \tau, S)$ if and only if

$$\forall x, z \in \mathbb{C}^N :$$

$$||z - x||_1 \le \frac{1 + \rho}{1 - \rho} (||z||_1 - ||x||_1 + 2||x_{\overline{S}}||_1) + \frac{2\tau}{1 - \rho} ||A(x - z)||.$$

This is a generalization of the previously discussed theorem for the SNSP (cf. Theorem 4.14 in the book). The following corollary, an ℓ^1 -bound for the error, is a key result:

Corollary: (Theorem 4.19 in the book)

Assume that $A \in \mathbb{C}^{m \times N}$ satisfies $\text{RNSP}(\|\cdot\|, \rho, \tau, s)$ with $0 < \rho < 1$ and $\tau > 0$, $e \in \mathbb{C}^m$ with $\|e\| \le \eta$ and let $x \in \mathbb{C}^N$. Then, if $\mathcal{L}_{x,\eta}$ is the set of all minimizers of the problem

$$\min\left\{\|z\|_1\,\big|\,z\in\mathbb{C}^N,\ \|Az-(Ax+e)\|\leq\eta\right\}\,,$$

then

$$\sup_{x^{\#} \in \mathcal{L}_x} \|x - x^{\#}\|_1 \le \frac{2(1+\rho)}{1-\rho} \sigma_s(x)_1 + \frac{4\tau}{1-\rho} \eta,$$

i.e. the solution set \mathcal{L}_x is contained in a ball of radius $\frac{2(1+\rho)}{1-\rho}\sigma_s(x)_1 + \frac{4\tau}{1-\rho}\eta$ around x in the ℓ^1 -norm, which is small for small η and small sparsity defect $\sigma_s(x)_1$.

Now, let us move on to establish an error bound in ℓ^q for $q \in [1, \infty)$. We need:

$egin{aligned} Definition: \ell^q ext{-robust null space property} \end{aligned}$

Let $q \geq 1$. A matrix $A \in \mathbb{C}^{m \times N}$ satisfies the ℓ^q -robust null space property of order $s \in \mathbb{N}$ (with respect to $\|\cdot\|$) with the constants $\rho \in (0,1)$ and $\tau \geq 0$ iff

$$\forall S \subseteq [N], \ |S| \le s \ \forall v \in \mathbb{C}^N: \ \|v_S\|_q \le \frac{\rho}{s^{1-\frac{1}{q}}} \|v_{\overline{S}}\|_1 + \tau \|Av\|.$$

Remark

• For $v \in \mathbb{C}^N$, $S \subset [S]$ with $|S| \leq s$ and $1 \leq p \leq q$, we have

$$||v_S||_p \leq s^{\frac{1}{p} - \frac{1}{q}} ||v_S||_q$$
.

• Hence,

$$\ell^q$$
-RNSP $(\|\cdot\|, \rho, \tau, s) \implies \ell^p$ -RNSP $(s^{\frac{1}{p} - \frac{1}{q}} \|\cdot\|, \rho, \tau, s)$

for $1 \le p \le q$.

• In particular, the previously known ℓ^1 -RNSP is implied by the ℓ^q -RNSP for $1 \le q < \infty$.

We have the following result:

Theorem: (Theorem 4.25 in the book)

If a matrix $A \in \mathbb{C}^{m \times N}$ satisfies ℓ^q -RNSP($\|\cdot\|, \rho, \tau, s$) for some $1 \leq q < \infty, \rho \in (0, 1), \tau > 0$, then

$$\forall x, z \in \mathbb{C}^N, \ \forall p \in [1, q] :$$

$$\|z - x\|_p \le \frac{C}{s^{1 - \frac{1}{p}}} (\|z\|_1 - \|x\|_1 + 2\sigma_s(x)_1) + Ds^{\frac{1}{p} - \frac{1}{q}} \|A(x - z)\|,$$

where
$$C = \frac{(1+\rho)^2}{1-\rho}$$
 and $D := (3+\rho)\tau/(1-\rho)$.

The main result is the following corollary of this theorem:

Corollary 0.1: Robustness of the quadratically constrained basis pursuit

Assume $A \in \mathbb{C}^{m \times N}$ satisfies ℓ^2 -RNSP($\|\cdot\|_2, \rho, \tau, s$) and let $x \in \mathbb{C}^N$, $e \in \mathbb{C}^m$ with $\|e\| \leq \eta$. Then, if $\mathcal{L}_{x,\eta}$ is the set of all minimizers of the problem

$$\min \left\{ \|z\|_1 \, \big| \, z \in \mathbb{C}^N, \ \|Az - (Ax + e)\|_2 \leq \eta \right\} \, ,$$

then

$$\sup_{x^{\#} \in \mathcal{L}_{x,\eta}} \|x - x^{\#}\|_{p} \le \frac{C}{s^{1 - \frac{1}{p}}} \sigma_{s}(x)_{1} + Ds^{\frac{1}{p} - \frac{1}{2}} \eta$$

for $p \in [1, 2]$ and for some constants C, D > 0 that only depend on ρ, τ .

4.4 Recovery of individual vectors