

- Last time:
- Mutual coherence & sparsity
 - Uniqueness via coherence & μ_1 -coherence
 - Analysis of OMP via coherence & μ_1 -coherence

- Today:
- Analysis of BP via coherence & μ_1 -coherence
 - Analysis of Thresholding algo.

Analysis of Basis Pursuit

Any conditions guaranteeing success of all ℓ_1 -approx CS using OMP with $|S|$ iterations also guarantee success of BP for all sets supported on S .

Thus it is:

$$\max_{j \notin S} |(A^H x)_j| > \max_{k \in S} |(A^H x)_k| \Leftrightarrow \begin{cases} \mu_1 < 1 \\ \mu_1 < \frac{1}{|S|} \end{cases}$$

implies the NOMP.
Given $v \in \mathcal{N}(A) \setminus \{0\}$, we have

$$\|v\|_2 = \|A^H A v\|_2 = \|A^H A v\|_2 \leq \|A^H\|_2 \|v\|_2 \leq \|A\|_2 \|v\|_2$$

$$\|v\|_2 \leq \|A\|_2 \|v\|_2 \Rightarrow \|v\|_2 (1 - \|A\|_2) \leq 0$$

Thus, we get the following result:

Thm 5.15 $A \in \mathbb{C}^{m \times n}$, ℓ_1 normalized cols.
If $\mu_1(A) + \mu_1(A^{(-)}) < 1$, then every n -sparse $x \in \mathbb{C}^n$ is exactly recovered from $y = Ax$ via BP.

Proof: [Thm 4.5]: rec. & suff. t.p.t. A satisfies $\mathcal{N}(A) = \{0\}$.

If $\mu_1 < 1$, $\mu_1 < \frac{1}{|S|}$ and $|S| \geq 1$.

Let a_1, \dots, a_n denote the cols of A .
 $v \in \mathcal{N}(A) \Rightarrow \sum_{j=1}^n v_j a_j = 0$.

Taking the inner product with a_i , i.e. S ,
 $v_i = \sum_{j \in S} v_j \langle a_i, a_j \rangle = - \sum_{j \notin S} v_j \langle a_i, a_j \rangle$

$$|v_i| \leq \sum_{j \in S} |v_j| |\langle a_i, a_j \rangle| + \sum_{j \notin S} |v_j| |\langle a_i, a_j \rangle|$$

Sum over all $i \in S$, interchange summation:
 $\|v\|_1 \leq \sum_{i \in S} |v_i| \leq \sum_{i \in S} |v_i| \sum_{j \in S} |\langle a_i, a_j \rangle| + \sum_{j \notin S} |v_j| \sum_{i \in S} |\langle a_i, a_j \rangle|$

$$\leq \sum_{i \in S} |v_i| \mu_1 + \sum_{j \notin S} |v_j| \mu_1(A^{(-)})$$

$$= \mu_1 \|v\|_1 + \mu_1(A^{(-)}) \|v\|_1$$

$\Rightarrow (1 - \mu_1(A^{(-)})) \|v\|_1 \leq \mu_1 \|v\|_1$
Thus, if $1 - \mu_1(A^{(-)}) > \mu_1(A)$,
 $\|v\|_1 < \|v\|_1$ is satisfied. \square

We have seen that $\mu_1(A) \geq \sqrt{\frac{m-n}{m+n}}$, $A \in \mathbb{R}^{m \times n}$
So if $\mu_1(A) + \mu_1(A^{(-)}) < 1$ with $\frac{m-n}{m+n} \leq \mu_1(A)$ and $A \in \mathbb{R}^{m \times n}$, and if $m \geq 2n$, we must have $1 > \mu_1(A) + \mu_1(A^{(-)}) \geq (2n-1) \sqrt{\frac{m-n}{m+n}}$
 $\Rightarrow \sqrt{\frac{m-n}{m+n}} > \frac{1}{2n-1}$ a contradiction.
 $m \geq c n^2$ is a fundamental threshold with coherence based bounds. $(c \geq \frac{1}{(2n-1)^2})$

Analysis of thresholding based algorithms

Basic thresholding

$$S^0 = L_2(A^H y)$$

$$x^0 = \arg \min_{z \in \mathbb{C}^n} \{ \|y - Az\|_2, \text{supp}(z) \leq s \}$$

Prop. 5.7: $x \in \mathbb{C}^n$ is recoverable via BT if $\min_{j \in S} |(A^H y)_j| > \max_{j \notin S} |(A^H y)_j|$.

Thm 5.16 $A \in \mathbb{C}^{m \times n}$, ℓ_2 normalized cols.
 $x \in \mathbb{C}^n$, $\text{supp}(x) = S$, $|S| = s$.
If $\mu_1(A) + \mu_1(A^{(-)}) < \frac{\min_{i \in S} |x_i|}{\max_{j \notin S} |x_j|}$

then x is exactly recovered via BT.

Proof: a_1, \dots, a_n : cols of A , $\|a_i\|_2 = 1$.
Using Prop. 5.7, we need t.p.t. for any $j \in S, k \notin S$,
 $|\langle Ax, a_j \rangle| > |\langle Ax, a_k \rangle|$

$$|\langle Ax, a_j \rangle| = |\sum_{i \in S} x_i \langle a_i, a_j \rangle| \leq \left(\max_{i \in S} |x_i| \right) \mu_1(S)$$

$$|\langle Ax, a_k \rangle| = \left| \sum_{i \in S} x_i \langle a_i, a_k \rangle \right|$$

$$\geq |x_j| \left(1 - \sum_{i \in S, i \neq j} |x_i| |\langle a_i, a_k \rangle| \right)$$

$$\geq \min_{i \in S} |x_i| \left(\min_{i \in S} |x_i| \mu_1(S^{(-)}) \right)$$

Thus, when $\mu_1(A) + \mu_1(A^{(-)}) < \frac{\min_{i \in S} |x_i|}{\max_{j \notin S} |x_j|}$,

$$\left(\min_{i \in S} |x_i| \right) \left(\min_{i \in S} |x_i| \mu_1(S^{(-)}) \right) > \left(\max_{j \notin S} |x_j| \right) \mu_1(S)$$

Analysis of iterative hard thresholding: see ex 5.10

Analysis of hard thresholding pursuit

HTP: Init: $x^0 = 0, S^0 = \emptyset$

Iterate: $S^{m+1} = L_2(x^m + A^H(y - Az^m))$

[HTP]: $x^{m+1} = \arg \min_{z \in \mathbb{C}^n} \{ \|y - Az\|_2, \text{supp}(z) \leq s^m \}$

Until STOP

Thm 5.17 $A \in \mathbb{C}^{m \times n}$, ℓ_2 normalized cols.

If $2\mu_1(A) + \mu_1(A^{(-)}) < 1$, every n -sparse $x \in \mathbb{C}^n$ is exactly recovered from $y = Ax$ after at most k iterations of hard thresholding pursuit.

Proof: Consider indices j_1, j_2, \dots, j_n s.t.

$$|x_{j_1}| \geq |x_{j_2}| \geq \dots \geq |x_{j_n}| = 0$$

Let $v = \sum_{i=1}^n x_{j_i} a_{j_i}$, the vector v is n -sparse and $y = Av$ as x is n -sparse.

The largest abs. entries of v are $x_{j_1}, x_{j_2}, \dots, x_{j_n}$.

$$x_{j_1}^2 \leq x_{j_1}^2, A^H(y - Ax) = x_{j_1}^2(A^H(Ax) - Ax) = 0$$

(The rest of $S^k \setminus S$ is empty) and hence $x^k = x$ by (HTP)

For $k > 1$, it is sufficient to prove

$$\min_{i \in S^k} |x_i| > \max_{j \notin S^k} |x_j| \quad \square$$

Note that, $\# j \in \mathbb{C} \cup \mathbb{R}$.

$$z_j^{(n)} = \frac{1}{n} \sum_{k=1}^n (x_k - x_j) \langle x_k, x_j \rangle$$

$$= \frac{1}{n} \sum_{k=1}^n (x_k - x_j) \langle x_k, x_j \rangle \quad (\because \langle x_k, x_k \rangle = 1)$$

$$\therefore \|z_j^{(n)} - x_j\| \leq \frac{1}{n} \sum_{k=1}^n \|x_k - x_j\| + \sum_{k=1}^n |\langle x_k, x_j \rangle| \quad \text{--- (1)}$$

Now, for $\{k \in \mathbb{N}\}$ and $\ell \in \mathbb{S}$

$$\|z_k^{(n)}\| \geq \|z_k\| - \mu_k(x) \|x_k - x_j\| - \mu_k(x) \|x_k - x_j\| \rightarrow 0$$

$$\|z_k^{(n)}\| \leq \mu_k(x) \|x_k - x_j\| + \mu_k(x) \|x_k - x_j\| \rightarrow 0$$

In particular, for $n \geq \frac{1}{\mu_k(x) \|x_k - x_j\|} + \frac{1}{\mu_k(x) \|x_k - x_j\|} \Rightarrow \|z_k^{(n)}\| \rightarrow 0$

$\therefore \mu_k(x) < 1$. Therefore, the above case of the induction hypothesis in (1) holds for $n \geq 1$. Assume that the hypothesis holds for $n-1$, with $n \geq 1$.

$\Rightarrow \{z_1, \dots, z_{n-1}\} \in \mathbb{S}^*$

Now, $\{z_1, \dots, z_n\}$ with n replaced by $(n-1)$ gives (using Lemma 2.1)

$$(A^{n-1} (z - z^*))_{\mathbb{S}^*} = 0$$

Hence, for any $j \in \mathbb{S}^*$, define \mathbb{O} of \mathbb{S}^{n-1} as $\{z_1, \dots, z_{n-1}\}$

$$\text{Rank } \mathbb{O} \Rightarrow$$

$$\|z_j^{(n)} - x_j\| \leq \underbrace{\mu_k(x) \|x_k - x_j\|}_{\substack{\text{The last term in } \mathbb{S} \\ \text{and by the induction hypothesis} \\ \text{at step (1) above in the second form of (1)}}} + \underbrace{\mu_k(x) \|z_{n-1}\|}_{\substack{\text{The last term in } \mathbb{S} \\ \text{and by the induction hypothesis} \\ \text{at step (1) above in the second form of (1)}}}$$

Taking max over $j \in \mathbb{S}^*$ and rearranging \Rightarrow

$$\|(z - z^*)_{\mathbb{S}^*}\| \leq \frac{\mu_k(x)}{(1 - \mu_k(x))} \|z_{n-1}\|$$

Substituting (1) \Rightarrow for $\{k \in \mathbb{N}\}$ and $\ell \in \mathbb{S}$

$$\|z_k^{(n)}\| \geq \left(1 - \frac{\mu_k(x)}{\mu_k(x)}\right) \|z_{n-1}\| \quad \text{--- (2)}$$

$$\|z_k^{(n)}\| \leq \frac{\mu_k(x)}{1 - \mu_k(x)} \|z_{n-1}\| \quad \text{--- (3)}$$

Since $\frac{\mu_k(x)}{1 - \mu_k(x)} < \frac{1}{2}$, this \Rightarrow (2) holds for $n \geq 1$. This completes the proof by induction. \square

(Note: $\|z_k^{(n)}\| > \mu_k(x) \|z_k^{(n)}\|$)