

Compressed Sensing

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Setup:

$X_0 \in \mathbb{R}^n$ unknown signal

$\|X_0\|_0 = S \ll n$, so X_0 is sparse

measurement $a_i \cdot X_0 = b_i$ or $a_i \cdot X_0 = b_i + \epsilon_i$ noise

Given a_i & b_i , find X .

Want:

$$\min \|X\|_0 \text{ such that } Ax = b \quad \forall A = \begin{pmatrix} -a_1 \\ -a_2 \\ \vdots \\ -a_m \end{pmatrix}$$

Can't minimize $\|X\|_0$ b/c NP-hard (combinatorially expensive)

Instead

$$\min \|X\|_1 \text{ such that } Ax = b$$

¶

Plaus? If

When do these problems succeed

at finding X_0 ? Stable to noise?

Stable to not-quite sparse signals?

Key ideas for recovery conditions

spark

incoherence

null space property

restricted isometry property

Spark

The spark of $A \in \mathbb{R}^{m \times n}$ is \wedge^{smallest} $\#$ linly dependent columns

Theorem:

[For any $y \in \mathbb{R}^m$, there exists at most one k -sparse signal X st $y = Ax$] if and only if $\text{spark}(A) > 2k$.

Proof:

\Rightarrow : Assuming at most one k -sparse X st $y = Ax$.

Suppose $\text{spark}(A) \leq 2k$. $\exists h \in N(A)$ st $\|h\|_0 \leq 2k$

while $h = x - x'$ w/ $\|x\|_0 \leq k$ & $\|x'\|_0 \leq k$, $Ax = Ax'$.

Contradiction.

\Leftarrow : Suppose $\text{spark}(A) > 2k$. Suppose for some $y \in \mathbb{R}^m$ st $\|y\|_0 \leq k$ st $y = Ax = Ax'$. Then $x - x' \in N(A)$. $\|x - x'\|_0 \leq 2k$. $\|x\|_0 \leq k$ & $\|x'\|_0 \leq k$.
so $Ah = 0$ for $h = x - x'$. $\|h\|_0 \leq 2k \Rightarrow h = 0 \Rightarrow x = x'$. ~~contradiction~~ ~~so $\text{spark}(A) \geq 2k$~~

1) Setup: Let $x_0 \in \mathbb{R}^n$, $\|x_0\|_0 = s$.
 Let $A \in \mathbb{R}^{m \times n}$, $m < n$.
 Let $b = Ax_0$.

Find x_0 by solving

$$\min \|x\|_0 \text{ st. } Ax = b. \quad (*)$$

Claim: ^{Fix A.} If $\text{spark}(A) > 2s$, then ^{$\forall x_0$} , x_0 is unique soln to $(*)$.

^{Fix A} If $\text{spark}(A) \leq 2s$, $\exists x_0$ such that x_0 not unique soln to $(*)$

Gist: If x_0 & x_1 are s -sparse solns of $Ax = b$
 $x_0 - x_1 \in N(A)$ & $\|x_0 - x_1\|_0 \leq 2s$.

Q: Why isn't the following a counter example?

$$\underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \end{pmatrix}}_A \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad (**)$$

\rightarrow A has spark ~~3~~ ^A (take $k=2$) ≤ 2 (but $k=1$)
 but there is exactly one soln to $(**)$ with sparsity ≤ 1 :

$$x = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

A: Spark condition is equivalent to $(*)$ working for all x_0

If spark is too low, $(*)$ may work for some x_0 but will necessarily fail for some other x_0 .

Cohherence

$$\text{Let } A = \begin{pmatrix} a_1 & a_2 & \cdots & a_n \end{pmatrix}$$

$$\mu(A) = \max_{i \neq j} \frac{|\langle a_i, a_j \rangle|}{\|a_i\|_2 \|a_j\|_2} \quad \text{is coherence.}$$

Low values of coherence are good.

Fact: $\mu(A) \in \left[\sqrt{\frac{n-m}{m(n-1)}}, 1 \right]$

↑
Welch
bound

4) A is $m \times n$

$$\begin{bmatrix} | & | & & | \\ a_1 & a_2 & \cdots & a_n \\ | & | & & | \end{bmatrix}$$

$a_i \in \mathbb{R}^m$ is column of A .

Low coherence:

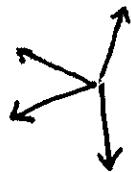
$$\begin{array}{c} \uparrow \\ \rightarrow \end{array} \text{ in } \mathbb{R}^2$$

lowest coherence with 3 vectors in \mathbb{R}^2



~~lowest coherence with 4 vect~~

lowest coherence with 4 vectors in \mathbb{R}^2 ?



? No coherence = 1 in this case



Large coherence: parallel vectors

5) $A \in \mathbb{R}^{m \times n}$

$$\begin{bmatrix} 1 & \cdots & d_n \\ a_1 & \cdots & 1 \end{bmatrix}$$

$$\mu(A) \geq \sqrt{\frac{n-m}{m(n-1)}} \quad \text{Welch bound}$$

If $n \gg m$, this bound becomes $\mu(A) \geq \frac{1}{\sqrt{m}}$

This bound appears to depend ??!!
on wrong variable \nearrow

The ambient dimension

If n is larger, vectors must start "clumping"

So minimal incoherence should be ~~decreasing~~ growing

in n

Resolution: Equality in Welch bound can only occur when $n \leq \binom{m+1}{2}$

~~So Welch bound is tight in case $n \leq \binom{m+1}{2}$~~

In that case $\mu(A) \geq \frac{1}{\sqrt{m}}$ — ambient dimension controls best incoherence up to a certain n . Then n controls best incoherence

Lemma: For any $A \in \mathbb{R}^{m \times n}$, $\text{Spark}(A) \geq 1 + \frac{1}{\mu(A)}$

Proof (by Gershgorin's Circle Theorem):

WLOG, assume A has unit ^{magnitude} columns.

Let $\Lambda \subseteq \{1, \dots, n\}$ be s.t. $|\Lambda| = p$.

Let $G = A_{\Lambda}^T A_{\Lambda}$. Then $g_{ii} = 1 + i$ & $|g_{ij}| \leq \mu(A)$ $\forall i \neq j$.

By Gershgorin's Circle Thm: if $\sum_{j \neq i} |g_{ij}| < |g_{ii}|$, then

G is PSD. The cols of A_{Λ} are linearly independent

~~We have $p < 1 + \frac{1}{\mu(A)}$ $\nabla p < \text{spark}(A)$~~

Hence if $(p-1)\mu < 1$ then $\text{spark}(A) > p$

If $p < 1 + \frac{1}{\mu}$ then $\text{spark}(A) > p$

So $\text{spark}(A) \geq 1 + \frac{1}{\mu}$

Thm: If $k < \frac{1}{2} \left(1 + \frac{1}{\mu(A)}\right)$ then

$\forall y \in \mathbb{R}^m$ there is at most one x s.t. $y = Ax$ & $\|x\|_0 \leq k$.

With bound $\mu(A) \geq \frac{1}{\sqrt{m}}$ that means
can only guarantee recovery for $k \leq \sqrt{m}$.