
The Sparsity Gap



Joel A. Tropp

Computing & Mathematical Sciences
California Institute of Technology
`jtropp@acm.caltech.edu`

Introduction

Systems of Linear Equations

We consider linear systems of the form

$$m \left\{ \underbrace{\begin{bmatrix} \Phi \end{bmatrix}}_N \right\} \begin{bmatrix} x \end{bmatrix} = \begin{bmatrix} b \end{bmatrix}$$

Assume that

- Φ has dimensions $m \times N$ with $N > m$
- Φ has full row rank
- The columns of Φ have unit ℓ_2 norm

The Trichotomy Theorem

Theorem 1. *For a linear system $\Phi x = b$, exactly one of the following situations occurs.*

- 1. No solution exists.*
- 2. The equation has a unique solution.*
- 3. The solutions form a linear subspace of positive dimension.*

Regularization via Sparsity

A principled approach to underdetermined systems:

$$\min \|\mathbf{x}\|_0 \quad \text{subject to} \quad \Phi \mathbf{x} = \mathbf{b} \quad (\text{P0})$$

where $\|\mathbf{x}\|_0 = \#\text{supp}(\mathbf{x}) = \#\{j : x_j \neq 0\}$

- When $\|\mathbf{x}\|_0 \leq s$, then \mathbf{x} is called *s-sparse*
- If $\Phi \mathbf{x} = \mathbf{b}$ and \mathbf{x} is *s-sparse*, then \mathbf{x} is an *s-sparse representation* of \mathbf{b}
- Since Φ has full row-rank, every \mathbf{b} has an *m-sparse* representation
- **Question:** What can we say about sparser representations?

Geometry

Key Insight

Sparse representations are well behaved
when the matrix Φ is sufficiently nice

(Column submatrices should not be singular and
individual columns should not look like sparse signals)

Quantifying Niceness I

🐼 We call Φ a *tight frame* when

$$\Phi\Phi^* = \frac{N}{m} \cdot \mathbf{I}$$

🐼 Equivalently, the rows of Φ form an orthonormal family (up to scaling)

🐼 Observe that N/m is the *redundancy* of the frame

🐼 Tight frames have minimal spectral norm among conformal matrices

Quantifying Niceness II

• The *coherence* of Φ is the quantity

$$\mu = \max_{j \neq k} |\langle \varphi_j, \varphi_k \rangle|$$

• Measures the angle between columns

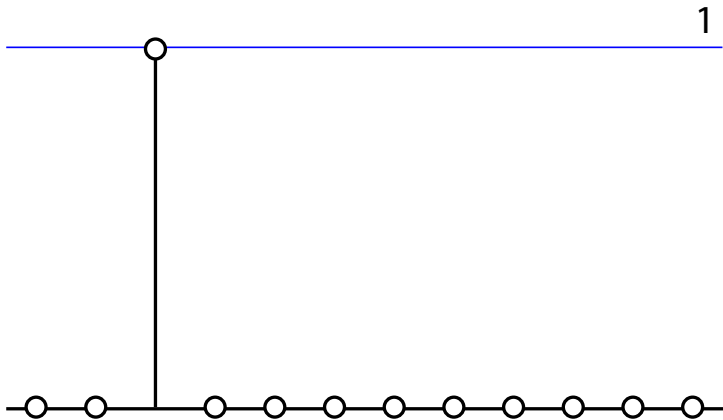
• When $N \geq 2m$, the coherence satisfies

$$\mu \gtrsim \frac{1}{\sqrt{m}}$$

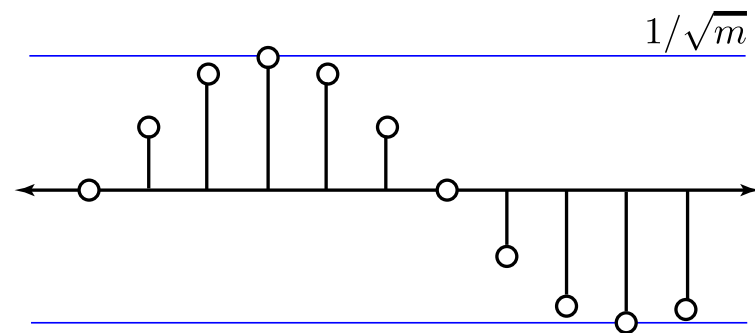
• Incoherent matrices appear often in signal processing applications

References: [Welch 1974, Mallat–Zhang 1993, Donoho–Huo 2001, Gribonval–Nielsen 2003, Strohmer–Heath 2003]

Example: Identity + Fourier



Impulses



Complex Exponentials

A very incoherent tight frame

Uniqueness

Uncertainty implies Uniqueness

Theorem 2. *Suppose that a vector \mathbf{b} has two representations: $\Phi\mathbf{x} = \mathbf{b} = \Phi\mathbf{y}$. Then*

$$\|\mathbf{x}\|_0 + \|\mathbf{y}\|_0 > \mu^{-1}.$$

Corollary 3. *Suppose that $\mathbf{b} = \Phi\mathbf{x}$ where*

$$\|\mathbf{x}\|_0 < \frac{1}{2} \cdot \mu^{-1}.$$

Then \mathbf{x} is the unique solution to (P0).

☛ Very strict requirement since $\mu^{-1} \lesssim \sqrt{m}$

References: [Donoho–Stark 1989, Donoho–Huo 2001, Gribonval–Nielsen 2003, Donoho–Elad 2003]

The Square-Root Threshold

🦉 Sparse representations are not necessarily unique past the \sqrt{m} threshold

Example: The Dirac Comb

🦉 Consider the Identity + Fourier matrix with $m = p^2$

🦉 There is a vector \mathbf{b} that can be written as either p spikes or p sines

🦉 By the Poisson summation formula,

$$\mathbf{b}(t) = \sum_{j=0}^{p-1} \delta_{pj}(t) = \frac{1}{\sqrt{m}} \sum_{j=0}^{p-1} e^{-2\pi i p j t / m} \quad \text{for } t = 0, 1, \dots, m$$

References: [Donoho–Stark 1989]

Enter Probability

Insight:
The bad vectors are atypical

An Uncertainty Principle for Generic Signals

Theorem 4. [T 2010] *Suppose that $\mathbf{b} = \Phi \mathbf{x}$ where the nonzero components of \mathbf{x} have a continuous distribution. With probability one, the vector \mathbf{b} has no representation $\mathbf{b} = \Phi \mathbf{y}$ where $\text{supp}(\mathbf{x}) \cap \text{supp}(\mathbf{y}) = \emptyset$ unless*

$$\|\mathbf{x}\|_0 + \|\mathbf{y}\|_0 > \mu^{-1} \|\mathbf{x}\|_0^{1/2}.$$

Corollary 5. *When $\mu \leq m^{-1/2}$, condition becomes*

$$\|\mathbf{y}\|_0 > \|\mathbf{x}\|_0 \left[\sqrt{\frac{m}{\|\mathbf{x}\|_0}} - 1 \right]$$

- 🐼 Even with refinements, this approach does not yield uniqueness!
- 🐼 **Problem:** Some *supports* could be bad

References: [*The Sparsity Gap*]

Enter More Probability

Insight:
The bad supports are atypical

A Simple Model for Random Sparse Vectors

Model (M0) for $\mathbf{b} = \Phi \mathbf{x}$

The matrix Φ is a unit-norm tight frame of size $m \times N$
with coherence $\mu \leq c / \log N$.

The support of \mathbf{x} has cardinality $s \leq cm / \log N$ and
is uniformly random.

The nonzero entries of \mathbf{x} have a continuous distribution.

Random Submatrices & Sparse Representation

Theorem 6. [T 2006, 2008] *Assume all parameters satisfy Model (M0). Draw a uniformly random set S of s columns from Φ , and define the random column submatrix $\mathbf{A} = \Phi_S$. Then*

$$\text{Prob} \left\{ \|\mathbf{A}^* \mathbf{A} - \mathbf{I}\| < \frac{1}{2} \right\} \geq 99.72\%$$

and

$$\text{Prob} \left\{ \max_{n \notin S} \|\mathbf{A}^* \varphi_n\|_2 < \frac{1}{2} \right\} \geq 99.72\%.$$

References: [*Random Subdictionaries, Random Submatrices*]

The Sparsity Gap

Theorem 7. [T 2008, 2010] *Let $\mathbf{b} = \Phi \mathbf{x}$ be a vector drawn from Model (M0). With probability at least 99.44%, the following statements hold.*

- 1. The vector \mathbf{x} is the unique solution to (P0).*
- 2. Furthermore, there is no disjoint representation*

$$\mathbf{b} = \Phi \mathbf{y} \quad \text{where} \quad \text{supp}(\mathbf{x}) \cap \text{supp}(\mathbf{y}) = \emptyset$$

unless

$$\|\mathbf{y}\|_0 > \|\mathbf{x}\|_0 \left(1 + 2 \cdot \frac{m}{N}\right).$$

References: [Candès–Romberg 2006, *Random Subdictionaries, Random Submatrices, The Sparsity Gap*]

To learn more...

Web: <http://www.acm.caltech.edu/~jtropp>

E-mail: jtropp@acm.caltech.edu

Relevant papers:

- 📄 “Conditioning of random subdictionaries,” *ACHA*, 2008
- 📄 “Norms of random submatrices,” *CRAS*, 2008
- 📄 “Spikes and sines,” *JFAA*, 2008
- 📄 “The sparsity gap,” *Proc. CISS*, 2010