

# Compressed Sensing: Day 2

## Proof structure

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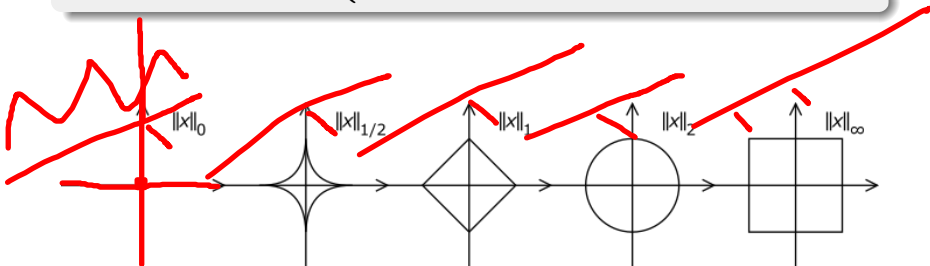
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# Norm minimization

## Definition (p-norm minimization)


$$(P_p) \begin{cases} \min_{g \in \mathbb{C}^N} & \|g\|_p \\ \text{s.t.} & \hat{g} = \hat{f} \text{ in } \Omega. \end{cases} \quad |\Omega| \geq 2 |\Gamma|$$



# When do minimizers coincide?

Definition ( $(P_1)$  minimizer)

$$f^\# \in \operatorname{argmin}(P_1).$$

- When is  $f^\# = f$ ?
- What if  $\Omega$  is random? 

ltau

# Reconstruction with high probability

Fix  $f \in \mathbb{C}^N$ ,  $\tau \in (0, 1)$  and  $M > 0$ .

Theorem ([CRT06, Theorem 1.3])

Denote  $T := \text{supp } f$ . If

$$|\Omega| \geq 2|T|$$

$$\tau N \geq |T| \log N \frac{1}{\alpha(M)},$$

then,  $f = f^\#$  with probability at least  $1 - O(N^{-M})$ .

$|\Omega|$  is “at least” to  $\tau N$

Recall that  $\Omega$ , which defines  $(P_1)$ , is a random set such that  $\mathbb{E}[|\Omega|] = \tau N$ . Moreover,

Lemma (Concentration bound)

$$\mathbb{P}(|\Omega| > (1 - \varepsilon_M)\tau N) \geq \underline{1 - N^{-M}},$$

where  $\varepsilon_M := \sqrt{\frac{2M \log N}{|\tau N|}}$ .

# Proof

## Lemma (Concentration bound)

$$\mathbb{P}(|\Omega| > (1 - \varepsilon_M)\tau N) \geq 1 - N^{-M},$$

## Proof.

Since  $|\Omega| \sim \text{Bin}(N, \tau)$ ,

$$\begin{aligned} \mathbb{P}(|\Omega| \leq \tau N - \varepsilon_M \tau N) &\leq \exp\left[\frac{-(\varepsilon_M \tau N)^2}{2\tau N}\right] && ; \text{Chernoff bound} \\ &= \exp[-M \log N] && ; \varepsilon_M := \sqrt{\frac{2M \log N}{|\tau N|}} \\ &= N^{-M}. \end{aligned}$$



# Reconstruction with high probability

Fix  $f \in \mathbb{C}^N$ ,  $\tau \in (0, 1)$  and  $M > 0$ .

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then,  $f = f^\sharp$  with probability at least  $1 - O(N^{-M})$ .

Fourier data

# Sensing basis

Let  $f \in \mathbb{C}^N$ . We sense

$$\{\hat{f}(\omega) : \omega \in \Omega\} = \{\langle \mathcal{F}_{N \rightarrow \{\omega\}}, f \rangle : \omega \in \Omega\}.$$

- Could we sense something else?
- Can we use other device to sense the signal  $f$ ?
- What if we sense  $\{\langle \varphi, f \rangle : \varphi \in \Omega\}$  with  $\Omega \subseteq \Phi$  random?



## Description basis

Let  $f \in \mathbb{C}^N$ . Recall that

Definition (Support)

$$\text{supp } f := \{t \in [N] : f(t) \neq 0\}.$$

Again,

$$\{t \in [N] : f(t) \neq 0\} = \{t \in [N] : \langle e_t, f \rangle \neq 0\}.$$

- Could we describe differently?
- Can we use other device to express the signal  $f$ ?
- What if we care about sparsity in  $\Psi$ :  $\{\psi \in \Psi : \langle \psi, f \rangle \neq 0\}$ ?

# How similar are two basis?

## Definition (Coherence)

Let  $\Phi, \Psi$  be two basis of  $\mathbb{C}^N$ .

$$\mu(\Phi, \Psi) := \sqrt{N} \max_{\phi \in \Phi, \psi \in \Psi} |\langle \phi, \psi \rangle|.$$

The coherence measures the largest correlation between any two elements of  $\Phi$  and  $\Psi$ .

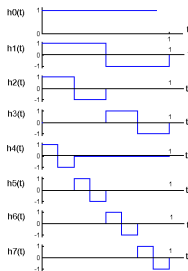
## Exercise (Bounds for coherence)

$$\mu(\Phi, \Psi) \in [1, \sqrt{N}].$$

# Examples

Possible basis are the following.

- Spikes:  $\{e_j = (\delta_j(t))_{t \in [N]} : j \in [N]\}$ .
- Fourier:  $\left\{ \left( \frac{1}{\sqrt{N}} e^{-i\omega_j t} \right)_{t \in [N]} : \omega_j = \frac{2\pi j}{N}, j \in [N] \right\}$ .
- Haar,  $N = 2^m$ :  $\{\gamma_{j,k} : 2^j + k \in [N]\}$ , e.g.

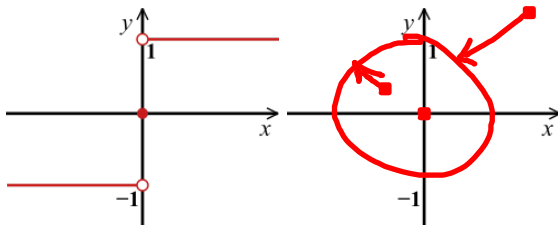


# More notation

## Definition (Sign)

Let  $x \in \mathbb{C}$ , define

$$\text{sign } x := \begin{cases} \frac{x}{|x|} & ; x \neq 0 \\ 0 & ; \sim \end{cases} .$$



# Minimizer for given basis

Definition (1-norm minimization with custom basis)

$$(P_1(\Psi, \Phi)) \begin{cases} \min_{g \in \mathbb{C}^N} & \sum_{\psi \in \Psi} |\langle \psi, g \rangle| = \|g\|_1 \\ \text{s.t.} & \langle \phi, g \rangle = \langle \phi, f \rangle \quad \forall \phi \in \Omega \subseteq \Phi. \\ & \hat{g} = \hat{f} \end{cases}$$

Definition

$$\underline{f_{\Psi, \Phi}^{\#}} \in \operatorname{argmin}(P_1(\Psi, \Phi)).$$

# General result

Random sensing  $\Omega$ , fixed signal  $f$



$\Phi = \Psi$   
Coherence =  $\sqrt{N}$

Theorem ([CR07, Theorem 1.1])

Denote  $T := \{\psi : \langle \psi, f \rangle \neq 0\}$ . If

$$\tau N \geq \underline{\mu^2(\Phi, \Psi)} |T| \log N \frac{1}{\alpha(M)},$$

then,  $f = \underline{f_{\Psi, \Phi}^\#}$  with probability at least  $1 - O(N^{-M})$ .

## General result: Deterministic set, random signal

Let  $C_0, C_1$  be fixed numerical constants. Fix  $\Phi, \Psi$  basis of  $\mathbb{C}^N$ ,  
 $T \subseteq \Psi, \Omega \subseteq \Phi$ .

Theorem ([CR07, Theorem 1.1])

Fix  $\delta > 0$ . If

$$|\Omega| \geq \max \left\{ C_0 \mu^2(\Phi, \Psi) |T| \log \left( \frac{N}{\delta} \right), C_1 \log^2 \left( \frac{N}{\delta} \right) \right\},$$

then,  $f = f_{\Psi, \Phi}^\#$  with probability at least  $1 - \delta$ .

Probability on what?  $f$  is random in its sign: choose the sign  $\langle \psi, f \rangle$  randomly for each  $\psi \in T$ . Then, the reconstruction holds independently of the magnitude of  $\langle \psi, f \rangle$ .

# Reconstruction with high probability

Fix  $f \in \mathbb{C}^N$ ,  $\tau \in (0, 1)$  and  $M > 0$ .

Theorem ([CRT06, Theorem 1.3])

Denote  $T := \text{supp } f$ . If

$$\tau N \geq |T| \log N \frac{1}{\alpha(M)},$$

then,  $f = \underline{f^\#}$  with probability at least  $1 - O(N^{-M})$ .



# Reduction to $(P_1)$

Let  $f \in \mathbb{C}^N$  and  $\underline{\Omega}$  non-empty. Denote  $T := \text{supp } f$ .

Lemma (Sufficient conditions)

If  $\mathcal{F}_{T \rightarrow \Omega} : \mathbb{C}^T \rightarrow \mathbb{C}^\Omega$  is injective and there exists  $P \in \mathbb{C}^N$  such that

$$\begin{cases} \hat{P}(\omega) = 0 & ; \omega \in \Omega^c \\ P(t) = \text{sign } f(t) & ; t \in T \\ |P(t)| < 1 & ; t \in T^c, \end{cases}$$

then,  $\underline{f} = \underline{f}^\#$  is the unique minimizer of  $(P_1)$ .

# Proof: $f$ is a minimizer

Proof.

Let  $g \in \mathbb{C}^N$  such that  $\hat{g}|_{\Omega} = \hat{f}|_{\Omega}$ . Define  $h := g - f$ . We will show that  $\|g\|_1 \geq \|f\|_1$ .

Let  $t \in T$ . Then,

$$\begin{aligned} |g(t)| &= |f(t) + h(t)| \\ &= ||f(t)| + h(t)\overline{\text{sign } f(t)}| && ; \cdot \overline{\text{sign } f(t)} \\ &\geq |f(t)| + \text{Re}(h(t)\overline{\text{sign } f(t)}) && ; |\cdot| \geq \text{Re}(\cdot) \\ &= |f(t)| + \text{Re}(h(t)\overline{P(t)}) && ; P(t) = \text{sign } f(t). \end{aligned}$$

Therefore, for  $t \in T$ ,  $|g(t)| \geq |f(t)| + \text{Re}(h(t)\overline{P(t)})$ .  
(...)



Proof:  $f$  is a minimizer

## Proof.

We will show that  $\|g\|_1 \geq \|f\|_1$ .

Let  $t \in T^c$ . Then,

$$\begin{aligned} |g(t)| &= |h(t)| \\ &\geq |h(t)\overline{P(t)}| && ; |P(t)| < 1 \\ &\geq \operatorname{Re}(h(t)\overline{P(t)}) && ; |\cdot| \geq \operatorname{Re}(\cdot). \end{aligned}$$

Therefore, for  $t \in T^c$ ,  $|g(t)| \geq |f(t)| + \operatorname{Re}(h(t)\overline{P(t)})$ .

(...)



Proof:  $f$  is a minimizer

Proof.

Then,

$$\|g\|_1 \geq \|f\|_1 + \sum_{t \in [M]} \operatorname{Re}(h(t)\overline{P(t)}).$$

Note that  $\hat{h}|_{\Omega} = \hat{g}|_{\Omega} - \hat{f}|_{\Omega} = 0$ . Moreover, by Parseval's formula,

$$\sum_{t \in [M]} \operatorname{Re}(h(t)\overline{P(t)}) = \frac{1}{N} \sum_{\omega} \operatorname{Re}(\hat{h}(\omega)\overline{\hat{P}(\omega)}) = 0,$$

since either  $\hat{h}$  or  $\hat{P}$  vanishes for each  $\omega$ .

Thus,

$$\|g\|_1 \geq \|f\|_1.$$



# Proof: $f$ is the unique minimizer

Proof.

Assume  $\|g\|_1 = \|f\|_1$ . By the previous inequalities that  $|h(t)| = |h(t)P(t)|$  and so, for  $t \in T^C$ ,

$$\underline{h(t)} = 0.$$

Since  $\hat{h}|_{\Omega} = 0$  and  $\mathcal{F}_{T \rightarrow \Omega}$  is injective (and linear),

$$h = g - f \equiv 0.$$



# Reduction to $(P_1)$

Let  $f \in \mathbb{C}^N$  and  $\Omega$  non-empty. Denote  $T := \text{supp } f$ .

Lemma (Sufficient conditions)

If  $\mathcal{F}_{T \rightarrow \Omega} : \mathbb{C}^T \rightarrow \mathbb{C}^\Omega$  is injective and there exists  $P \in \mathbb{C}^N$  such that

$$\begin{cases} \hat{P}(\omega) = 0 & ; \omega \in \Omega^c \\ P(t) = \text{sign } f(t) & ; t \in T \\ |P(t)| < 1 & ; t \in T^c, \end{cases}$$

then,  $f = f^\#$  is the unique minimizer of  $(P_1)$ .

## Clever construction

## Definition (Extrapolation)

Denote the extrapolation by zero by the operator  $\iota: \mathbb{C}^T \rightarrow \mathbb{C}^N$ .

Its adjoint  $\iota^*: \mathbb{C}^N \rightarrow \mathbb{C}^T$  is the restriction map. Note that  $\iota^*\iota: \mathbb{C}^T \rightarrow \mathbb{C}^T$  is the identity map.

We will prove that there is a suitable operator  $\underline{H}: \mathbb{C}^T \rightarrow \mathbb{C}^N$  such that

$$P := \left( \iota - \frac{1}{|\Omega|} H \right) \left( \underset{\text{Id}}{\iota^* \iota} - \frac{1}{|\Omega|} \iota^* H \right) \overset{-1}{\iota^* \text{sign } f}$$

works for our purposes.

# Why can we make it work?

Recalling the sufficient conditions, we would have to prove that

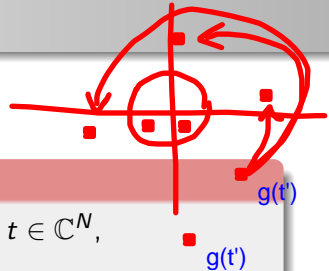
$$\begin{cases} \hat{P}(\omega) = 0 & ; \omega \in \Omega^c \\ P(t) = \text{sign } f(t) & ; t \in T \\ |P(t)| < 1 & ; t \in T^c, \end{cases}$$

(and that  $\mathcal{F}_{T \rightarrow \Omega}$  is injective) occurs with high probability.

We will prove that if  $\left( \iota^* \iota - \frac{1}{|\Omega|} \iota^* H \right)$  is invertible, then  $\mathcal{F}_{T \rightarrow \Omega}$  is injective, so we can concentrate on  $P$ .



## Definition



## Definition (White noise operator)

Define the operator  $H: \mathbb{C}^T \rightarrow \mathbb{C}^N$  by, for all  $t \in \mathbb{C}^N$ ,

$$Hg(t) := - \sum_{\omega \in \Omega} \sum_{\substack{t' \in T \\ t' \neq t}} e^{i\omega(t-t')} g(t').$$

## Exercise

Prove that  $\iota^* H: \mathbb{C}^T \rightarrow \mathbb{C}^T$  is self-adjoint.

# What will we prove?

Lemma (Constrained Fourier transform)  $\hat{P}$  is supported in  $\Omega$

For all  $g \in \mathbb{C}^T$ , the Fourier transform of  $(\iota - \frac{1}{|\Omega|} H) g$  (which is an element of  $\mathbb{C}^N$ ) is supported in  $\Omega$ .

Lemma (sign on  $T$ )

For all  $t \in T$ , we have that  $P(t) = \text{sign } f(t)$ .

Lemma (Small magnitude)

With high probability  $|P(t)| < 1$ , for all  $t \in T^c$ .

Lemma (Well defined and injectivity)

The operator  $(\iota^* \iota - \frac{1}{|\Omega|} \iota^* H)$  is invertible with high probability, and its invertibility implies that  $\mathcal{F}_{T \rightarrow \Omega}$  is injective.

## Easy proof I

## Lemma (Constrained Fourier transform)

For all  $g \in \mathbb{C}^T$ , the Fourier transform of  $\left(\iota - \frac{1}{|\Omega|} H\right) g$  (which is an element of  $\mathbb{C}^N$ ) is supported in  $\Omega$ .

## Constrained Fourier transform.

Fix  $g \in \mathbb{C}^T$ . Note that, for all  $t \in [N]$ ,

$$\left(\iota - \frac{1}{|\Omega|} H\right) g(t) = \frac{1}{|\Omega|} \sum_{\omega \in \Omega} \sum_{t' \in T} e^{i\omega(t-t')} g(t') = \frac{1}{|\Omega|} \sum_{\omega \in \Omega} e^{i\omega t} \hat{g}(\omega).$$

Then, the Fourier transform of  $\left(\iota - \frac{1}{|\Omega|} H\right) g$  is supported in  $\Omega$ .  $\square$

## Easy proof II

Lemma (sign on  $T$ )

For all  $t \in T$ , we have that  $P(t) = \text{sign } f(t)$ .

sign on  $T$ .Fix  $t \in T$ . Note that,

$$\begin{aligned} P(t) &= \iota^* P(t) && ; t \in T \\ &= \iota^* \left( \cancel{\iota - \frac{1}{|\Omega|} H} \right) \left( \cancel{\iota^* \iota - \frac{1}{|\Omega|} \iota^* H} \right)^{-1} \iota^* \text{sign } f(t) && ; \text{def } P \\ &= \iota^* \text{sign } f(t) \\ &= \text{sign } f(t) && ; t \in T. \end{aligned}$$



## More complex proof

Lemma (Small magnitude)

*With high probability  $|P(t)| < 1$ , for all  $t \in T^c$ .*

Proof.

Next week. □

## Easy proof III

## Lemma (Well defined and injectivity)

The operator  $\left(\iota^* \iota - \frac{1}{|\Omega|} \iota^* H\right)$  is invertible with high probability, and its invertibility implies that  $\mathcal{F}_{T \rightarrow \Omega}$  is injective.

## Proof.

Invertibility with high probability will be proven next week. For now, let us prove the implication. Since

$\left(\iota - \frac{1}{|\Omega|} H\right) g(t) = \frac{1}{|\Omega|} \sum_{\omega \in \Omega} e^{i\omega t} \hat{g}(\omega)$ , we have that

$$\iota^* \iota - \frac{1}{|\Omega|} \iota^* H = \frac{1}{|\Omega|} \underbrace{(\mathcal{F}_{T \rightarrow \Omega})^*}_{\text{red underline}} \underbrace{\mathcal{F}_{T \rightarrow \Omega}}_{\text{red underline}}.$$

Then, the invertibility of  $\iota^* \iota - \frac{1}{|\Omega|} \iota^* H$  implies that  $\mathcal{F}_{T \rightarrow \Omega}$  is injective. □

# Summary

## Lemma (Small magnitude (next week))

*For all  $t \in T^c$ , we have that  $|P(t)| < 1$  occurs with high probability.*

## Lemma (Invertibility (next week))

*The operator  $\iota^* \iota - \frac{1}{|\Omega|} \iota^* H$  is invertible with high probability.*

## What will we prove?

### Lemma (Constrained Fourier transform)

For all  $g \in \mathbb{C}^T$ , the Fourier transform of  $\left(\iota - \frac{1}{|\Omega|} H\right) g$  (which is an element of  $\mathbb{C}^N$ ) is supported in  $\Omega$ .

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## Reduction to $(P_1)$

Let  $f \in \mathbb{C}^N$  and  $\Omega$  non-empty. Denote  $T := \text{supp } f$ .

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then,  $f = f^\#$  is the unique minimizer of  $(P_1)$ .

# Reconstruction with high probability

Fix  $f \in \mathbb{C}^N$ ,  $\tau \in (0, 1)$  and  $M > 0$ .



Theorem ([CRT06, Theorem 1.3])

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## References I

-  Emmanuel Candès and Justin Romberg.  
Sparsity and incoherence in compressive sampling.  
*Inverse Problems*, 23(3):969–985, apr 2007.
-  E.J. Candes, J. Romberg, and T. Tao.  
Robust uncertainty principles: exact signal reconstruction from highly incomplete frequency information.  
*IEEE Transactions on Information Theory*, 52(2):489–509, February 2006.