Algorithms for phase retrieval problems Part I

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Inverse problems

Inverse problems are problems of the form :

Reconstruct an "unknown object" from some set of "measurements" ?

In this mini-course, we will discuss a family of inverse problems, called phase retrieval problems.

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Outline

This morning : introduction to phase retrieval problems.

- Definition, motivations
- ▶ When is there hope to solve a phase retrieval problem ?
 - First case : "generic" problems
 - Second case : "non-generic" problems

This afternoon : a first family of reconstruction algorithms.

Tomorrow : a second family of reconstruction algorithms.

Phase retrieval problems : definition

Let

- ► V be a fixed complex vector space;
- x be an unknown element of V;
- $(L_s)_{s\in S}$ be a (known) family of linear forms on V.

The corresponding linear inverse problem would be :

Reconstruct
$$x \in V$$
 from $(L_s(x))_{s \in S}$?

Typically "easy" to solve.



Phase retrieval problems : definition

Here, we assume that we do not have access to the phase of $L_s(x)$.

 \rightarrow We have only its modulus.

General form of a phase retrieval problem :

Reconstruct $x \in V$ from $(|L_s(x)|)_{s \in S}$?

(Called "phase retrieval" because, if you can retrieve the phases, you can easily reconstruct x by solving the linear inverse problem.)

Global phase ambiguity

Reconstruct $x \in V$ from $(|L_s(x)|)_{s \in S}$?

Let $u \in \mathbb{C}$ have modulus 1. For any $s \in S$,

$$|L_s(ux)| = |uL_s(x)|$$
$$= |L_s(x)|.$$

 \Rightarrow It is not possible to distinguish x from ux.

We only aim at recovering x up to multiplication by a unitary complex number.

We call it reconstruction up to a global phase.

First motivation : X-ray imaging

Screen



[Schechtman, Eldar, Cohen, Chapman, Miao, and Segev, 2015]

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First motivation : X-ray imaging

We aim at recovering the support function of the object :

$$\begin{array}{rccc} f & : & \mathbb{R}^2 & \to & \mathbb{C} \\ & & (x,y) & \to & 1 & \text{if point } (x,y) \text{ of the plate is} \\ & & & & \text{masked by the object,} \\ & & & (x,y) & \to & 0 & \text{otherwise.} \end{array}$$

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The diffracted wave, when it arrives to the screen, can be described by a function $F_{screen} : \mathbb{R}^2 \to \mathbb{C}$.

This F_{screen} is (approximately) the Fourier transform of f:

$$F_{screen} = \hat{f}$$

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Recall that $F_{screen} = \hat{f}$.

Reconstruct a compactly supported ffrom $(|\hat{f}(x, y)|)_{(x,y) \in \mathbb{R}^2}$?

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Reconstruct a compactly supported ffrom $(|\hat{f}(x, y)|)_{(x,y) \in \mathbb{R}^2}$?

This is a phase retrieval problem.

(The most important one, historically and in applications.)

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Second motivation : audio processing

Goal of audio processing : develop algorithms to solve tasks involving sound recordings, that would be too difficult or tedious for human persons to perform.

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Main difficulty : an audio signal is a complicated object.

It can be represented by a function

 $f: \mathbb{R} \to \mathbb{R},$ usually with no obvious "structure".



Second motivation : audio processing

Idea : use an intermediate representation, easier to analyze.

$$(f:\mathbb{R}\to\mathbb{R}) \longrightarrow (|Wf|:\mathbb{R}^+\times\mathbb{R}\to\mathbb{R}^+).$$

|Wf| : modulus of the "short-time Fourier transform" or "wavelet transform" of f.





$|Wf|: \mathbb{R}^+ \times \mathbb{R} \to \mathbb{R}^+$



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Typical pipeline of an algorithm



When the expected result is a new audio signal,



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Typical pipeline of an algorithm



When the expected result is a new audio signal,



Reconstruct
$$x \in V$$
 from $(|L_s(x)|)_{s \in S}$?

When can we hope to solve this problem?

At least two possible issues :

No uniqueness of the reconstruction :

There exist $x \neq y$ such that $\forall s, |L_s(x)| = |L_s(y)|$.

No stability

There exist $x \not\approx y$ such that $\forall s, |L_s(x)| \approx |L_s(y)|$.

If one of these issues arises, we say the the problem is ill-posed.

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For a given a family of linear forms $(L_s)_{s \in S}$, can we determine whether there is uniqueness and stability?

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 \rightarrow It is difficult.

Two main situations

- 1. The L_s are "generic" or random, chosen according to a simple probability law.
 - Analysis is doable.
 - Uniqueness and stability tend to hold.
- 2. The L_s are deterministic, imposed by a specific application.
 - Analysis is usually difficult or impossible.
 - Uniqueness and stability may fail to hold.

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Generic / random case, in finite dimension

Reconstruct
$$x \in V$$
 from $(|L_s(x)|)_{s \in S}$?

To simplify, we assume $\dim(V)$ and $\operatorname{Card}(S)$ to be finite. We can assume :

•
$$V = \mathbb{C}^n$$
;

•
$$S = \{1, ..., m\};$$

∀s, L_s = ⟨., f_s⟩, for some f_s ∈ Cⁿ.
 (⟨.,.⟩ is the usual hermitian product.)

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V = Cⁿ;
S = {1,...,m}; → "measurement vectors"
∀s, L_s = ⟨..(f_s), for some f_s ∈ Cⁿ. (⟨.., ⟩ is the usual hermitian product.)

Generic / random case, in finite dimension

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Reconstruct $x \in \mathbb{C}^n$ from $(|\langle x, f_k \rangle|)_{1 \le k \le m}$?

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Theorem (Uniqueness for generic measurements)

If $m \ge 4n - 4$, then, for generic vectors f_1, \ldots, f_m , uniqueness holds :

$$\forall x, y \in \mathbb{C}^n, \qquad (\forall k, |\langle x, f_k \rangle| = |\langle y, f_k \rangle|) \quad \Rightarrow \quad (x = y).$$

"Generic" means that the set of *m*-tuples (f_1, \ldots, f_m) for which the property holds is open and dense in $(\mathbb{C}^n)^m$.

[Balan, Casazza, and Edidin, 2006] [Conca, Edidin, Hering, and Vinzant, 2015]

Reconstruct
$$x \in \mathbb{C}^n$$
 from $(|\langle x, f_k \rangle|)_{1 \le k \le m}$?

Generic uniqueness : idea of proof

Principle : dimension counting.

(Not all x are uniquely determined by $(|\langle x, f_k \rangle|)_{1 \le k \le m}$.)

$$\iff \begin{pmatrix} \text{There exist } x, y \in \mathbb{C}^n \text{ such that} \\ \forall k \leq m, |\langle x, f_k \rangle| = |\langle y, f_k \rangle|. \end{pmatrix}$$
$$\iff \begin{pmatrix} \text{There exist } x, y \in \mathbb{C}^n \text{ and } \phi_1, \dots, \phi_m \in \mathbb{R} \text{ such} \\ \text{that} \\ \forall k \leq m, \langle x, f_k \rangle = e^{i\phi_k} \langle y, f_k \rangle. \end{pmatrix}$$

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 $\begin{pmatrix} \text{There exist } x, y \in \mathbb{C}^n \text{ and } \phi_1, \dots, \phi_m \in \mathbb{R} \text{ such that} \\ \forall k \leq m, \quad \langle x - e^{i\phi_k}y, f_k \rangle = 0. \end{cases}$

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$$\begin{pmatrix} \text{There exist } x, y \in \mathbb{C}^n \text{ and } \phi_1, \dots, \phi_m \in \mathbb{R} \text{ such that} \\ \forall k \leq m, \quad \langle x - e^{i\phi_k}y, f_k \rangle = 0. \end{pmatrix}$$
$$\iff \left((f_1, \dots, f_m) \in \bigcup_{\substack{x, y \in \mathbb{C}^n \\ \phi_1, \dots, \phi_m \in \mathbb{R}}} \{x - e^{i\phi_1}y\}^{\perp} \times \dots \times \{x - e^{i\phi_m}y\}^{\perp} \right)$$

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$$\begin{pmatrix} \text{There exist } x, y \in \mathbb{C}^n \text{ and } \phi_1, \dots, \phi_m \in \mathbb{R} \text{ such that} \\ \forall k \leq m, \quad \langle x - e^{i\phi_k}y, f_k \rangle = 0. \end{pmatrix}$$
$$\iff \begin{pmatrix} (f_1, \dots, f_m) \in \bigcup_{\substack{x, y \in \mathbb{C}^n \\ \phi_1, \dots, \phi_m \in \mathbb{R}}} \{x - e^{i\phi_1}y\}^{\perp} \times \dots \times \{x - e^{i\phi_m}y\}^{\perp} \\ \dim_{\mathbb{R}} = m(2n-2) \end{pmatrix}$$
$$\dim_{\mathbb{R}} = 2n + 2n + m$$

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 $\dim_{\mathbb{R}} = 2n + 2n + m$

$$\begin{pmatrix} \text{There exist } x, y \in \mathbb{C}^n \text{ and } \phi_1, \dots, \phi_m \in \mathbb{R} \text{ such that} \\ \forall k \leq m, \quad \langle x - e^{i\phi_k}y, f_k \rangle = 0. \end{pmatrix}$$
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$$\dim_{\mathbb{R}} = 2n + 2n + m$$

The set of all (f_1, \ldots, f_m) for which uniqueness does not hold is a (union of) manifold(s), with dimension at most 2n + 2n + m + m(2n - 2) = 2mn + 4n - m.

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$$\begin{pmatrix} \text{There exist } x, y \in \mathbb{C}^n \text{ and } \phi_1, \dots, \phi_m \in \mathbb{R} \text{ such that} \\ \forall k \leq m, \quad \langle x - e^{i\phi_k}y, f_k \rangle = 0. \end{pmatrix}$$
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The set of all (f_1, \ldots, f_m) for which uniqueness does not hold is a (union of) manifold(s), with dimension at most 2n + 2n + m + m(2n - 2) = 2mn + 4n - m.

 $2mn + 4n - m < 2mn = \dim_{\mathbb{R}}(\mathbb{C}^n)^m$ when $m \ge 4n + 1$.

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Reconstruct
$$x \in \mathbb{C}^n$$
 from $(|\langle x, f_k \rangle|)_{1 \le k \le m}$?

When $m \ge 4n - 4$, generic measurements \Rightarrow uniqueness.

Can we have uniqueness with less measurements than 4n - 4?

- If m < 4n − O(log(n)), uniqueness never holds. [Heinosaari, Mazzarella, and Wolf, 2013]
- ▶ For some n, when m < 4n − 4, uniqueness never holds. [Conca, Edidin, Hering, and Vinzant, 2015]
- ▶ For some n, uniqueness can be reached with m < 4n 4. [Vinzant, 2015]



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Summary

- $m \sim 4n$ measurements generically ensure uniqueness.
- This is essentially optimal.



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We are never given the exact value of $(|\langle x, f_k \rangle|)_{1 \le k \le m}$. In this case, can we do approximate reconstruction?

Summary

- $m \sim 4n$ measurements generically ensure uniqueness.
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What about stability?

We are never given the exact value of $(|\langle x, f_k \rangle|)_{1 \le k \le m}$. In this case, can we do approximate reconstruction?

For example, is reconstruction Lipschitz?

$$\forall x, y \in \mathbb{C}^{n}, \quad ||x-y||_{2} \leq C \left| \left| \begin{pmatrix} |\langle x, f_{1} \rangle| \\ \vdots \\ |\langle x, f_{m} \rangle| \end{pmatrix} - \begin{pmatrix} |\langle y, f_{1} \rangle| \\ \vdots \\ |\langle y, f_{m} \rangle| \end{pmatrix} \right| \right|_{2}?$$

(for some C > 0)

Stability

Theorem

When uniqueness holds, reconstruction is always Lipschitz.

[Balan and Zou, 2016]



Stability

Theorem

When uniqueness holds, reconstruction is always Lipschitz.

[Balan and Zou, 2016]

Problem : the Lipschitz constant can be terrible. In particular, it can depend on n and m.

 \rightarrow Stronger assumptions on f_1, \ldots, f_m ?

We assume that the f_k are chosen at random, independently, according to normal distributions :

$$f_k \overset{i.i.d.}{\sim} \mathcal{N}_{\mathbb{C}}(0, \mathrm{Id}_n), \quad k = 1, \ldots, m.$$

Theorem

There exist constants $\alpha, \gamma, C > 0$ such that, if

 $m \geq \alpha n$,

reconstruction is C-Lipschitz with proba at least $1 - O(e^{-\gamma m})$.

[Candès and Li, 2014]

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Summary

For generic/random f_k 's, the problem is well-posed.

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- Uniqueness holds "generically".
- Stability holds with high probability.

Summary

For generic/random f_k 's, the problem is well-posed.

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- Uniqueness holds "generically".
- Stability holds with high probability.

But in applications, measurements are neither generic nor random.

 \rightarrow Is the phase retrieval problem still well-posed ?

Non-generic/random phase retrieval

Is the phase retrieval problem still well-posed?

Answering this question is difficult. Only a handful of particular cases can be precisely studied.

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We will discuss the main one : the Fourier transform. (\rightarrow X-ray imaging)

Phase retrieval for the Fourier transform

Let $x \in \mathbb{C}^n$ be a vector. Its (semi)discrete Fourier transform is

$$\hat{x}: \omega \in [0; 2\pi[\rightarrow \hat{x}(\omega) = \sum_{k=0}^{n-1} x_k e^{-ik\omega}]$$

The corresponding phase retrieval problem is

Reconstruct $x \in \mathbb{C}^n$ from $(|\hat{x}(\omega)|)_{\omega \in [0;2\pi[}?$

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Reconstruct
$$x \in \mathbb{C}^n$$
 from $(|\hat{x}(\omega)|)_{\omega \in [0;2\pi[}?$

For all
$$\omega$$
, $|\hat{x}(\omega)|^2 = \left|\sum_{k=0}^{n-1} x_k e^{-ik\omega}\right|^2$
$$= \left(\sum_{k=0}^{n-1} x_k e^{-ik\omega}\right) \left(\sum_{k=0}^{n-1} \overline{x_k} e^{+ik\omega}\right)$$
$$= P_x(e^{-i\omega})\overline{P_x}\left(\frac{1}{e^{-i\omega}}\right),$$

where

$$P_x(X) = \sum_{k=0}^{n-1} x_k X^k.$$

Reconstruct
$$x \in \mathbb{C}^n$$
 from $(|\hat{x}(\omega)|)_{\omega \in [0;2\pi[}?)$

$$ig(\mathsf{Knowing}\;(|\hat{x}(\omega)|)_{\omega\in[0;2\pi[}ig) \ \iff ig(\mathsf{Knowing}\;P_x(X)\overline{P}_x(1/X)ig)$$

Let us write $P_x(X) = c \prod_{s=1}^{n-1} (X - z_s)$. Then $P_x(X)\overline{P}_x(1/X) = \tilde{c}X^{-(n-1)} \prod_{s=1}^{n-1} (X - z_s)(X - \frac{1}{z_s})$.

> From $|\hat{x}|$, we can determine z_1, \ldots, z_{n-1} up to the transformation $z \to \frac{1}{z}$.

Phase retrieval for the Fourier transform Summary

From $(|\hat{x}(\omega)|)_{\omega \in [0;2\pi[}$, we can recover P_x (and hence x) up to a "flipping" of each root.

Since there are n-1 roots, for most vectors x,

there are 2^{n-1} vectors y such that $|\hat{x}| = |\hat{y}|$.

 \Rightarrow No uniqueness.



And in dimension $d \ge 2$?

We can define a "dimension d" Fourier transform, on vectors $x \in \mathbb{C}^{n^d}$, and look at the same phase retrieval problem.

The picture is different :

"Almost all" x are uniquely determined from $|\hat{x}|$.

But reconstruction is not Lipschitz.

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Other problems than the Fourier transform?

Other measurement systems can be analyzed, but they need to have special properties.

("Special properties" : typically, a relation with harmonic analysis that allows to use similar tools as for Fourier.)

For these other measurement systems, the problem is

- possibly "less ill-posed" than for Fourier;
- ► typically "more ill-posed" than for random measurements.

Other problems than the Fourier transform?

Other measurement systems can be analyzed, but they need to have special properties.

("Special properties" : typically, a relation with harmonic analysis that allows to use similar tools as for Fourier.)

For these other measurement systems, the problem is

- possibly "less ill-posed" than for Fourier;
- ► typically "more ill-posed" than for random measurements.
- \rightarrow We will briefly discuss the wavelet transform case.

Phase retrieval for the wavelet transform

$$egin{array}{rcl} W:& L^2(\mathbb{R})&
ightarrow&(L^2(\mathbb{R}))^{\mathbb{Z}}\ &f&
ightarrow&Wf=(f\star\psi_j)_{j\in\mathbb{Z}}, \end{array}$$

where $(\psi_j)_{j \in \mathbb{Z}}$ is a (specific) family of band-pass filters.



Phase retrieval for the wavelet transform

To be able to analyze this problem, we need an assumption on the filters : we need them to be Cauchy wavelets.

Reason :

If the ψ_j 's are Cauchy wavelets, then, for each j, $f \star \psi_j$ is the restriction onto a line of some specific holomorphic function.

Theorem (Mallat and Waldspurger [2015])

We assume the filters to be Cauchy wavelets.

- Uniqueness holds.
- Reconstruction is not Lipschitz, but "locally stable".

Phase retrieval for the short-time Fourier transform

$$W: L^2(\mathbb{R}) \to (L^2(\mathbb{R}))^{\mathbb{Z}}$$

 $f \to Wf = (\widehat{w_n f})_{n \in \mathbb{Z}},$

with w a compactly-supported window, w_n its translation by n.

Reconstruct *f* from
$$|Wf| = (|\widehat{w_n f}|)_{n \in \mathbb{Z}}$$
?

Phase retrieval for the short-time Fourier transform

$$egin{array}{rcl} W: & L^2(\mathbb{R}) & o & (L^2(\mathbb{R}))^\mathbb{Z} \ & f & o & Wf = (\widehat{w_n f})_{n\in\mathbb{Z}}, \end{array}$$

with w a compactly-supported window, w_n its translation by n.

Reconstruct *f* from
$$|Wf| = (|\widehat{w_n f}|)_{n \in \mathbb{Z}}$$
?

For Gaussian windows, similar results as for wavelets :

Uniqueness holds.

Reconstruction is not Lipschitz, but "locally stable".
 [Alaifari, Daubechies, Grohs, and Yin, 2016]
 [Grohs and Rathmair, 2017]

Phase retrieval for the short-time Fourier transform

$$egin{array}{rcl} W: & L^2(\mathbb{R}) & o & (L^2(\mathbb{R}))^\mathbb{Z} \ & f & o & Wf = (\widehat{w_n f})_{n\in\mathbb{Z}}, \end{array}$$

with w a compactly-supported window, w_n its translation by n.

Reconstruct *f* from
$$|Wf| = (|\widehat{w_n f}|)_{n \in \mathbb{Z}}$$
?

For general windows,

- Almost all signals are uniquely determined.
- Stability unknown.

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[Jaganathan, Eldar, and Hassibi, 2016]
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Summary

- We have introduced a family of inverse problems.
- We have discussed two applications.
- ► We have discussed the well-posedness of these problems.
 - ► In a generic/random setting, well-posed.
 - In a non-generic setting, possibly ill-posed; difficult to study.

This afternoon

- ► Focus on the generic/random setting.
- Discuss reconstruction algorithms.