

Phase Retrieval with Random Illumination

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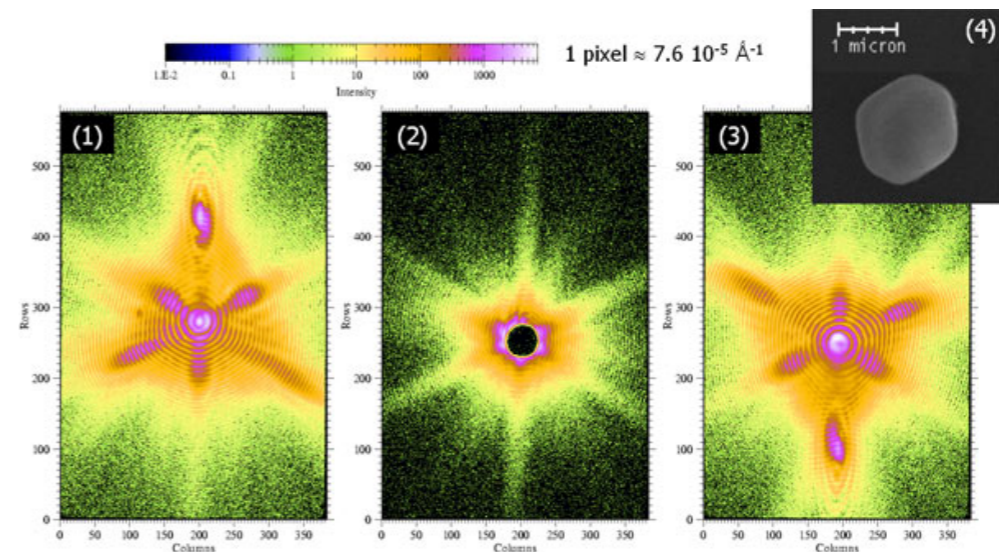
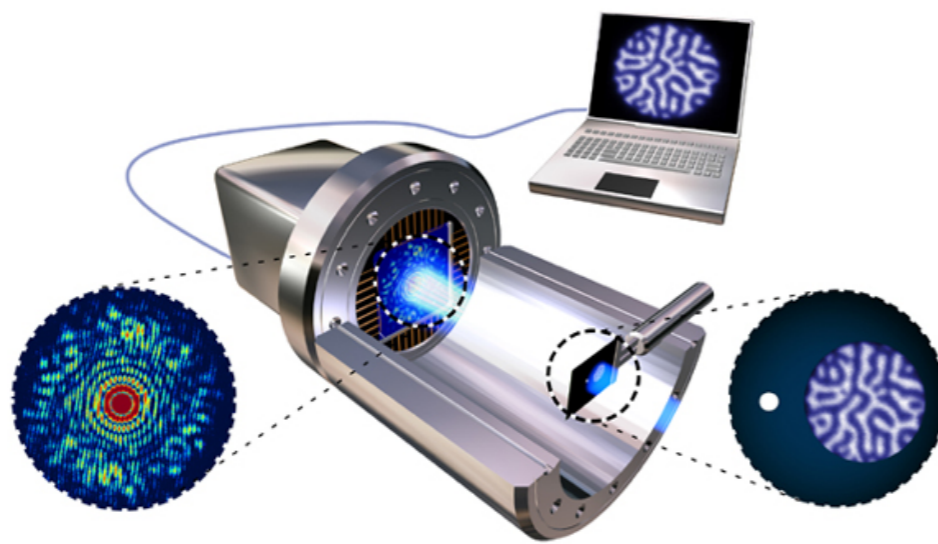
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Phase retrieval

Problem

Reconstruct the object f from the Fourier magnitude $|\Phi f|$.

Why do we care?



X-ray crystallography, single-molecule imaging, astronomy etc.

1985 Nobel Prize in Chemistry for Hauptman and Karle: partial solution of phasing problem.

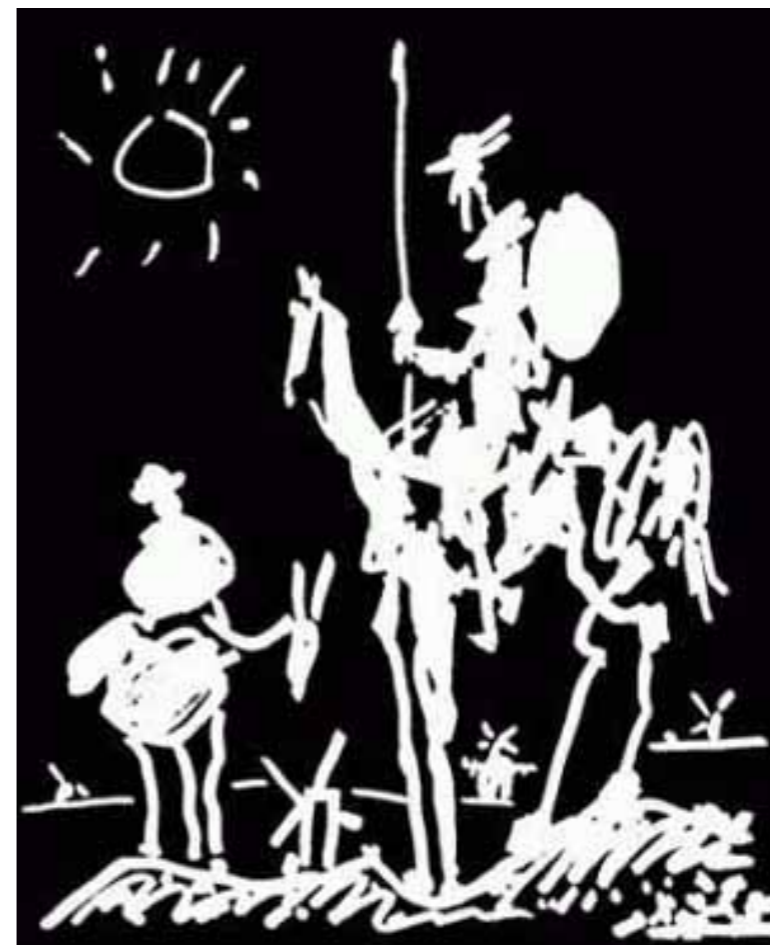
Phasing problem formulation

Discrete finite objects

Let $\mathbf{n} = (n_1, n_2) \in \mathbb{Z}^2$ and $\mathbf{z} = (z_1, z_2) \in \mathbb{C}^2$.

$$\text{multi-index: } \mathbf{z}^{\mathbf{n}} = z_1^{n_1} z_2^{n_2}$$

Let the *object* be represented by $f(\mathbf{n}), \mathbf{n} \leq \mathbf{N} = (N, N)$.



Binary objects: white = 1, black = 0.



$$f_L = \text{Lena}$$

$$F_L(\mathbf{w}) = |F_L(\mathbf{w})|e^{i\theta_L(\mathbf{w})}$$

$$F_1(\mathbf{w}) = |F_B(\mathbf{w})|e^{i\theta_L(\mathbf{w})}$$

$$f_1 = |\Phi^* F_1|$$

$$f_B = \text{Barbara}$$

$$F_B(\mathbf{w}) = |F_B(\mathbf{w})|e^{i\theta_B(\mathbf{w})}$$

$$F_2(\mathbf{w}) = |F_L(\mathbf{w})|e^{i\theta_B(\mathbf{w})}$$

$$f_2 = |\Phi^* F_2|$$



$$F_1(\mathbf{w}) = |F_B(\mathbf{w})| e^{i\theta_L(\mathbf{w})}$$
$$f_1 = |\Phi^* F_1|$$

$$F_2(\mathbf{w}) = |F_L(\mathbf{w})| e^{i\theta_B(\mathbf{w})}$$
$$f_2 = |\Phi^* F_2|$$

Phase retrieval

Discrete finite objects

Let $\mathbf{n} = (n_1, n_2) \in \mathbb{Z}^2$ and $\mathbf{z} = (z_1, z_2) \in \mathbb{C}^2$.

multi-index : $\mathbf{z}^{\mathbf{n}} = z_1^{n_1} z_2^{n_2}$

Let the object be represented by $f(\mathbf{n}), \mathbf{n} \leq \mathbf{N} = (N, N)$

Fourier transform describes **wave propagation**

$$F(e^{i2\pi w_1}, e^{i2\pi w_2}) = \sum_{\mathbf{n}} f(\mathbf{n}) e^{-i2\pi \mathbf{n} \cdot \mathbf{w}}$$

Analytic continuation \implies z -transform

$$F(\mathbf{z}) = \sum_{\mathbf{n}} f(\mathbf{n}) \mathbf{z}^{-\mathbf{n}}.$$

Discrete phase retrieval problem:

Determine $f(\mathbf{n})$ from Fourier magnitude data

$$|F(\mathbf{w})|, \quad \forall \mathbf{w} = (e^{i2\pi w_1}, e^{i2\pi w_2}) \in [0, 1]^2$$

Fourier magnitude data:

$$\begin{aligned} |F(\mathbf{w})|^2 &= \sum_{\mathbf{n}=-N}^N \sum_{\mathbf{m}} f(\mathbf{m} + \mathbf{n}) f^*(\mathbf{m}) e^{-i2\pi\mathbf{n}\cdot\mathbf{w}} \\ &= \sum_{\mathbf{n}=-N}^N C_f(\mathbf{n}) e^{-i2\pi\mathbf{n}\cdot\mathbf{w}} \end{aligned}$$

where

$$C_f(\mathbf{n}) = \sum_{\mathbf{m}} f(\mathbf{m} + \mathbf{n}) f^*(\mathbf{m})$$

is the **autocorrelation** function of f .

Fourier magnitude data contain complete information about autocorrelation function.

Sampling Theorem:

$\text{supp}(C_f) \subset [-N, N]^2 \implies [0, 1]^2$ is reduced to the Nyquist grid

$$\mathcal{M} = \left\{ (k_1, k_2) : k_j = 0, \frac{1}{2N+1}, \frac{2}{2N+1}, \dots, \frac{2N}{2N+1} \right\}$$

Oversampling ratio

$$\sigma = \frac{\text{Fourier magnitude data number}}{\text{unknown image pixel number}}$$

Standard ratio: $\sigma = 2^d$

Compressed sensing: $\sigma < 2^d$

Trivial ambiguities

Autocorrelation:

$$C_f(\mathbf{n}) = \sum_{\mathbf{m} + \mathbf{n} \in \mathcal{N}} f(\mathbf{m} + \mathbf{n}) f^*(\mathbf{m})$$

Invariant under:

(i) global phase,

$$f(\mathbf{n}) \longrightarrow e^{i\theta} f(\mathbf{n}), \quad \text{for some } \theta \in [0, 2\pi],$$

(ii) spatial translation

$$f(\mathbf{n}) \longrightarrow f(\mathbf{n} \oplus \mathbf{m}), \quad \mathbf{n} \oplus \mathbf{m} = \mathbf{n} + \mathbf{m} \bmod(N), \quad \text{some } \mathbf{m} \in \mathbb{Z}^2$$

(iii) conjugate inversion (twin image)

$$f(\mathbf{n}) \longrightarrow f^*(\mathbf{N} - \mathbf{n}).$$

Nontrivial ambiguity

THEOREM (Hayes 82, Pitts-Greenleaf 03)

Let the z -transform $F(z)$ of a finite complex-valued sequence $\{f(n)\}$ be given by

$$F(z) = \alpha z^{-m} \prod_{k=1}^p F_k(z), \quad m \in \mathbb{N}^2, \alpha \in \mathbb{C}$$

where $F_k, k = 1, \dots, p$ are nontrivial irreducible polynomials. Let $G(z)$ be the z -transform of another finite sequence $g(n)$. Suppose $|F(w)| = |G(w)|, \forall w \in [0, 1]^2$. Then $G(z)$ must have the form

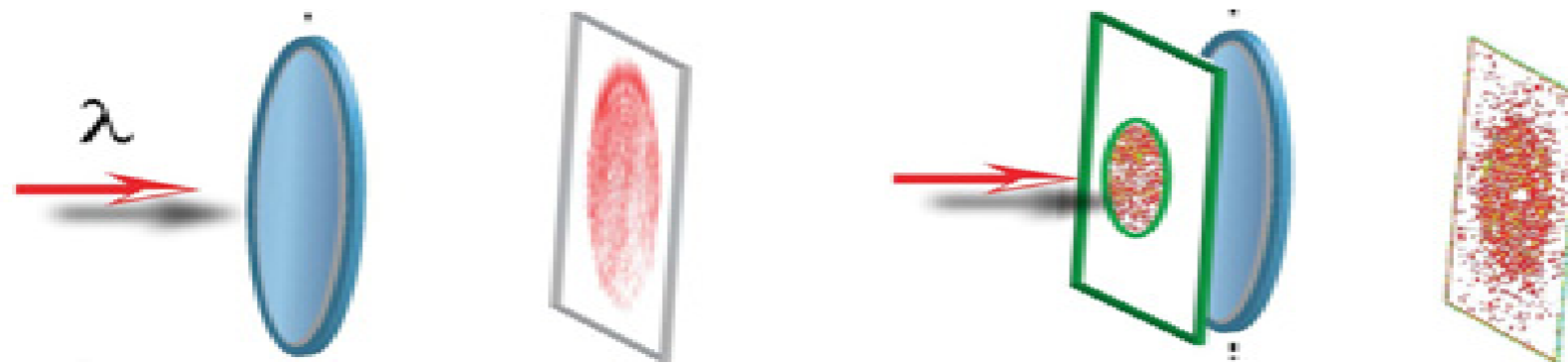
$$G(z) = |\alpha| e^{i\theta} z^{-p} \left(\prod_{k \in I} F_k(z) \right) \left(\prod_{k \in I^c} F_k^*(1/z^*) \right), \quad p \in \mathbb{N}^2, \theta \in \mathbb{R}$$

where I is a subset of $\{1, 2, \dots, p\}$.

Nontrivial ambiguity: Partial conjugate inversion on factors.

Random illumination

Coded aperture imaging



Diffuser generated speckle pattern: Garcia-Zalevsky-Fixler 05

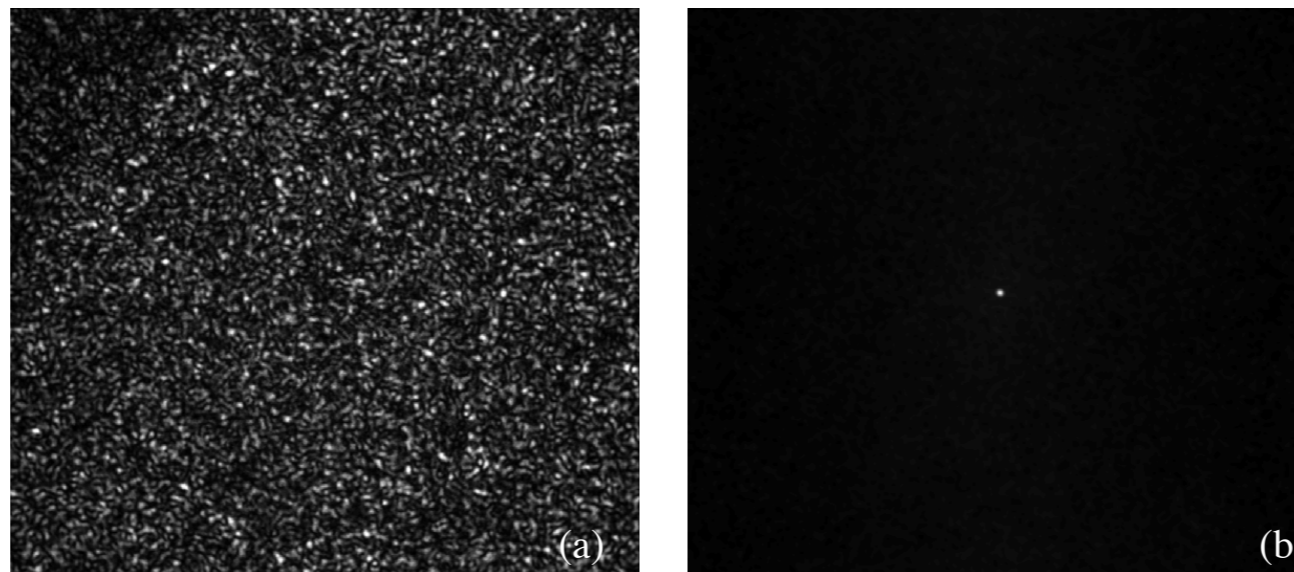
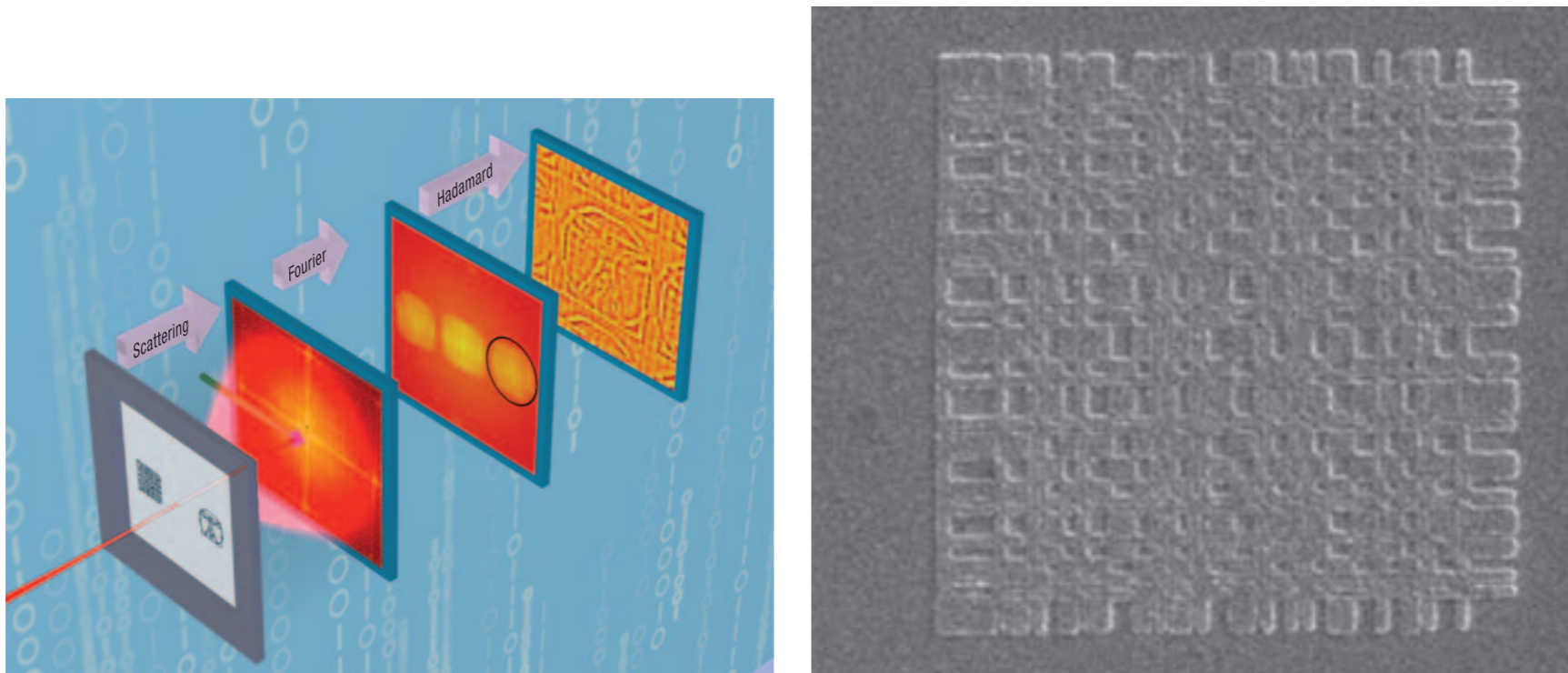


Fig. 2. (a) Encoding speckle pattern. (b) Autocorrelation of the encoding pattern.



X-ray holography with URA (Marchesini *et al.* 08).

Coherent X-ray imaging will be a key technique for developing nanoscience and nanotechnology.

One shot imaging: Bright and ultrashort X-ray pulse vaporizes the sample right after the pulse passes the sample.

Random illumination amounts to replacing the original object $f(\mathbf{n})$ by

$$\tilde{f}(\mathbf{n}) = f(\mathbf{n})\lambda(\mathbf{n}) \quad (\text{illuminated object})$$

Random illumination

$$\tilde{f}(n) = f(n)\lambda(n) \quad (\text{illuminated object})$$

$\lambda(n)$, representing the illumination field, is a **known** sequence of samples of random variables.

Let $\lambda(n)$ be continuous random variables with respect to the Lebesgue measure on \mathbb{S}^1 (the unit circle), \mathbb{R} or \mathbb{C} .

Case of \mathbb{S}^1 can be facilitated by a **random phase modulator** with

$$\lambda(n) = e^{i\phi(n)}$$

where $\phi(n)$ are continuous random variables on $[0, 2\pi]$.

Case of \mathbb{R} : **random amplitude modulator**.

Case of \mathbb{C} : both phase and amplitude modulations.

Irreducibility

THEOREM. Suppose that the support of the object $\{f(\mathbf{n})\}$ has **rank ≥ 2** . Then the the z -transform of the illuminated object $f(\mathbf{n})\lambda(\mathbf{n})$ is irreducible with probability one.

False for **rank 1** objects: fundamental thm of algebra

$$\begin{aligned}\sigma &= \frac{\text{Fourier magnitude data number}}{\text{unknown image pixel number}} \\ &= 2^d\end{aligned}$$

Absolute uniqueness

Positivity

THEOREM If $f(n)$ is **real and nonnegative** for every n then, with probability one, f is determined **absolutely** uniquely by the Fourier magnitude measurement on the lattice \mathcal{L} .

Sector constraint

THEOREM Suppose the phases of the object belong to $[a, b] \subset [0, 2\pi]$. Then the solution to the Fourier phasing problem has a unique solution with probability exponentially close to unity (depending on the **sparsity** and the phase range $|b - a|$.)

Complex objects w/o constraint

THEOREM. Suppose that $\{\lambda_1(n)\}$ are i.i.d. continuous random variables with respect to the Lebesgue measure on S^1 , \mathbb{R} or \mathbb{C} and in addition either one of the following conditions is true.

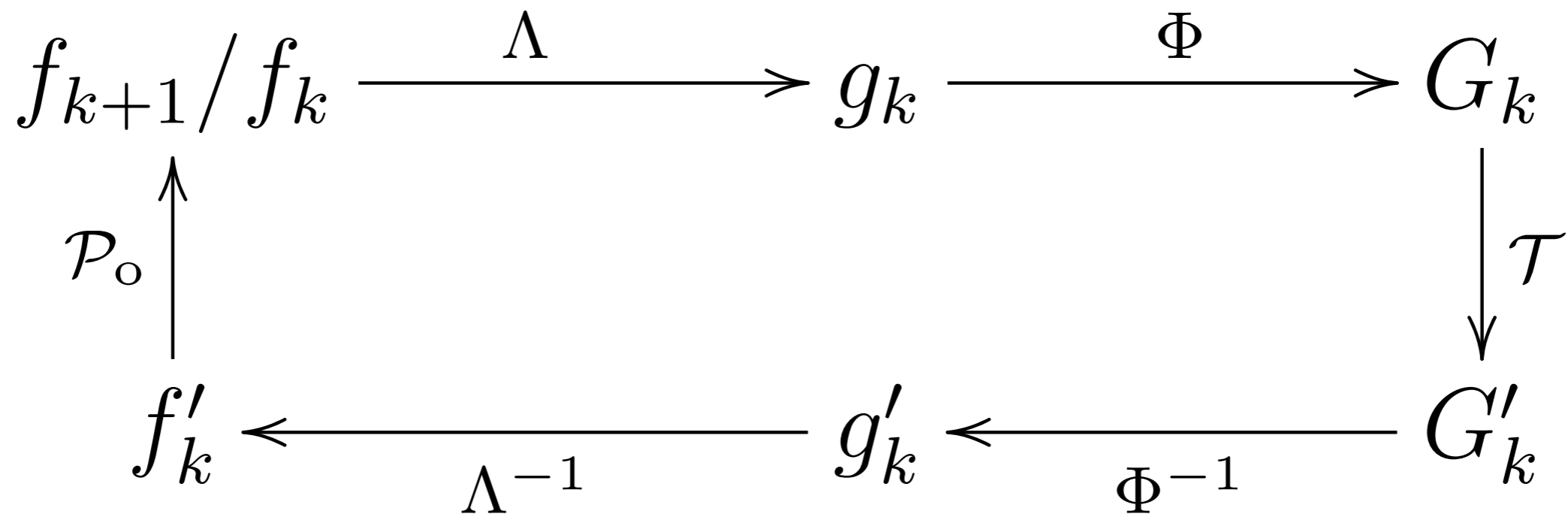
(i) $\{\lambda_2(n)\}$ are i.i.d. continuous random variables with respect to the Lebesgue measure on S^1 , \mathbb{R} or \mathbb{C} and $\{\lambda_2(n)\}$ are **independent** of $\{\lambda_1(n)\}$.

(ii) $\{\lambda_2(n)\}$ are **deterministic**.

Then with probability one $f(n)$ is uniquely determined, **up to a constant phase factor**, by the Fourier magnitude measurements with two illuminations λ_1 and λ_2 .

Alternated projections

Gerchberg-Saxton; Error Reduction (Fienup)

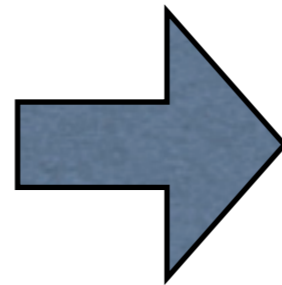


Oversampling method (Miao et al.)

$$\sigma = \frac{\text{image pixel number} + \text{zero-padding pixel number}}{\text{image pixel number}}.$$

Zero-padding

cameraman



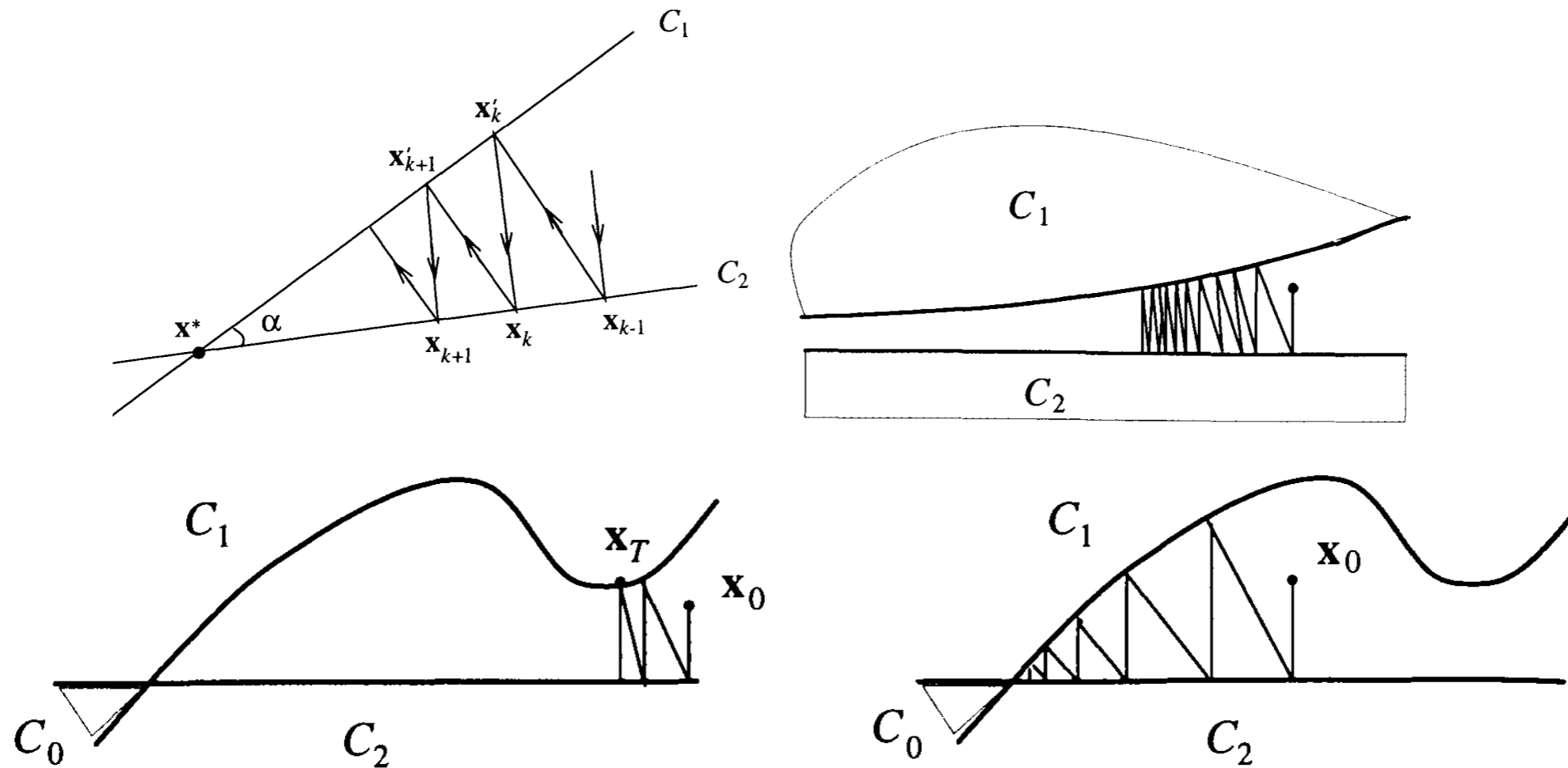
Multiple illuminations

$$\mathcal{P}_1 = \Lambda_1^{-1} \Phi^{-1} \mathcal{T}_1 \Phi \Lambda_1$$

$$\mathcal{P}_2 = \Lambda_2^{-1} \Phi^{-1} \mathcal{T}_2 \Phi \Lambda_2.$$

$$f_{k+1} = \mathcal{P}_0 \mathcal{P}_2 \mathcal{P}_1 f_k.$$

Error Reduction (Gerchberg-Saxton)



Bregman 65: **convex** constraints \implies convergence to **a feasible solution**.

Fourier magnitude data are a non-convex constraint!

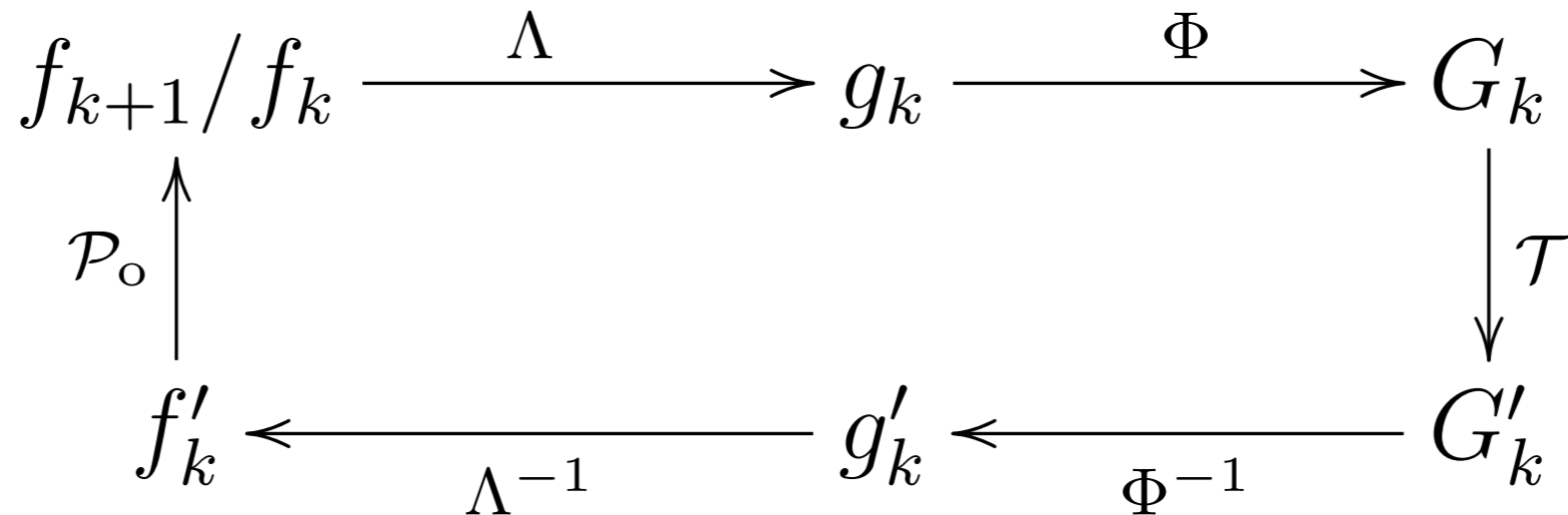
Nonconvexity or nonuniqueness ?

Convergence

Theorem 4. *Let $f \in \mathcal{C}(\mathcal{N})$ be an array with $f(\mathbf{0}) \neq 0$ and $\text{rank} \geq 2$. Let $\lambda(\mathbf{n})$ be i.i.d. continuous random variables on \mathbb{S}^1 . Let the Fourier magnitude be sampled on \mathcal{L} . Let h be a fixed point of $\mathcal{P}_o \mathcal{P}_f^\theta$ such that $\mathcal{P}_f^\theta h$ satisfies the zero-padding condition.*

- (a) *If f is real-valued, $h = \pm f$ with probability one,*
- (b) *If f satisfies the sector condition of Theorem 2, then $h = e^{i\nu} f$, for some ν , and satisfies the same sector constraint with probability at least $1 - |\mathcal{N}|(\beta - \alpha)^{\llbracket S/2 \rrbracket} (2\pi)^{-\llbracket S/2 \rrbracket}$.*

Hybrid-Input-Output (HIO)



$$\Re(f_{k+1}(\mathbf{n})) = \Re(f'_k(\mathbf{n}))$$

$$\Im(f_{k+1}(\mathbf{n})) = \Im(f_k(\mathbf{n})) - \beta \cdot \Im(f'_k(\mathbf{n})),$$

Real-valued objects

Error metrics

Relative error

$$e(\hat{f}) = \begin{cases} \|f - \hat{f}\| / \|f\| \\ \min_{\nu \in [0, 2\pi)} \|f - e^{i\nu} \hat{f}\| / \|f\| \end{cases}$$

Relative residual

$$r(\hat{f}) = \frac{\|Y - |\Phi \Lambda \mathcal{P}_o\{\hat{f}\}| \|}{\|Y\|}$$



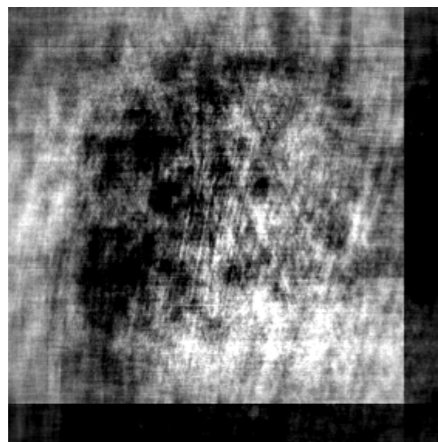
(a)

269 x 269

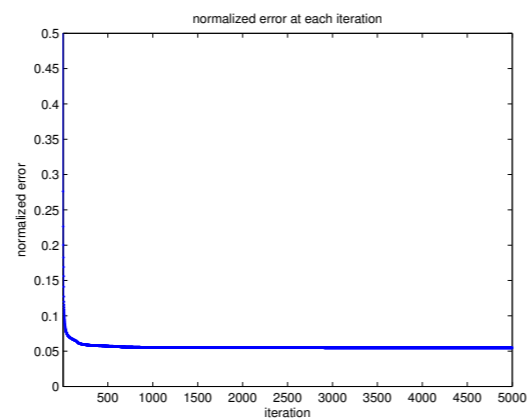


(b)

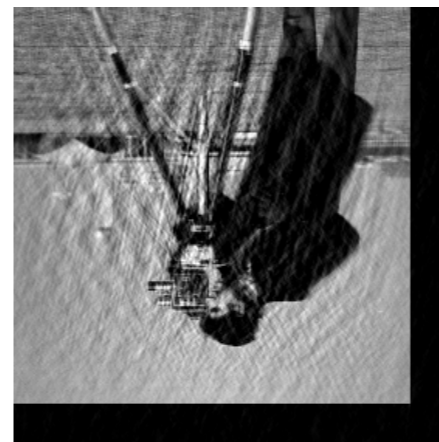
200 x 200



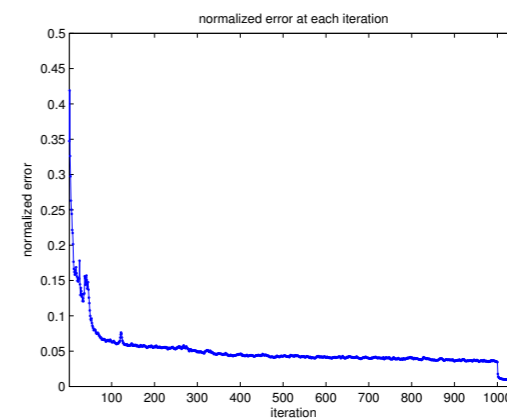
(a)



(b)



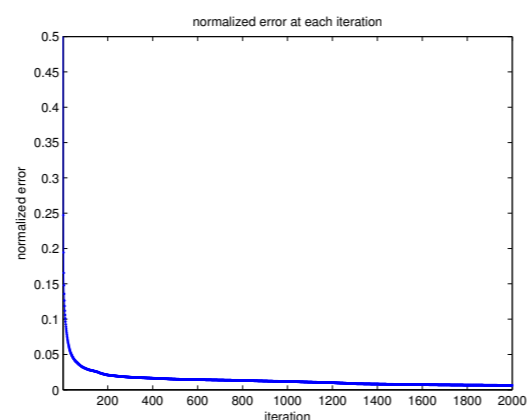
(c)



(d)



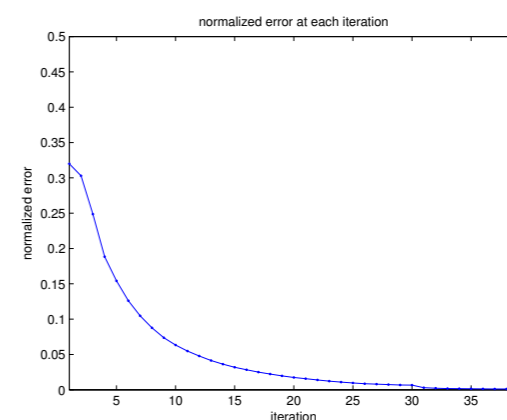
(e)



(f)



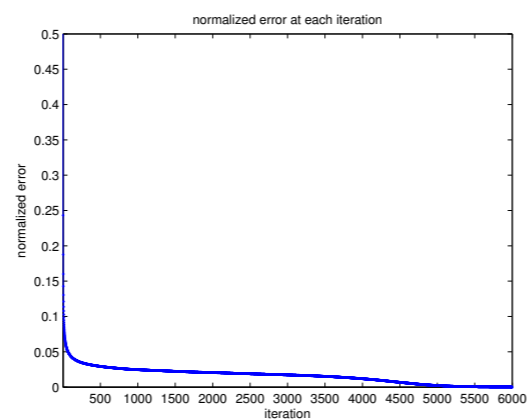
(g)



(h)



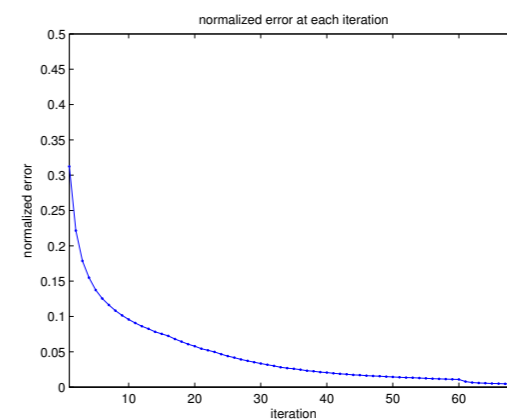
(i)



(j)

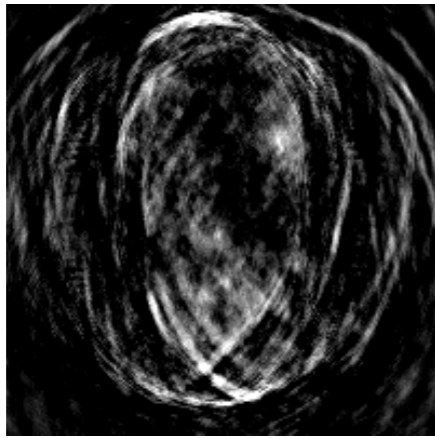


(k)

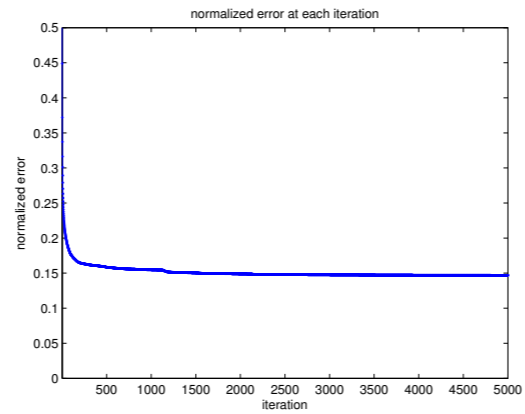


(l)

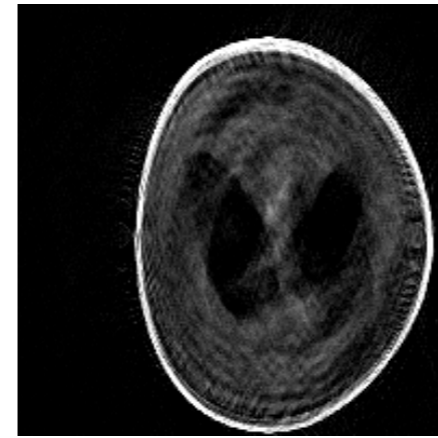
(e)-(h) Low resolution 40 x 40 block illumination with OR=2
(i)-(l) High resolution illumination with OR=1



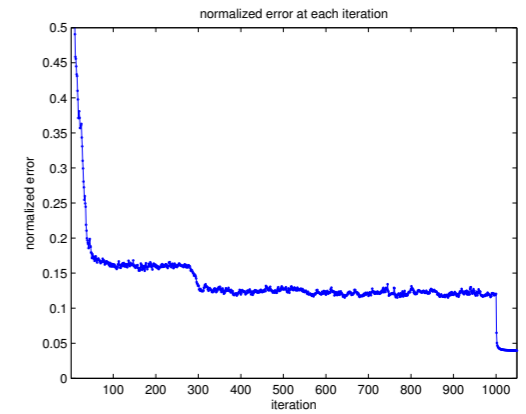
(a)



(b)



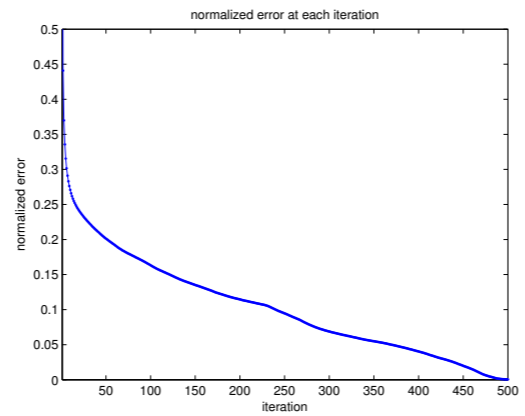
(c)



(d)



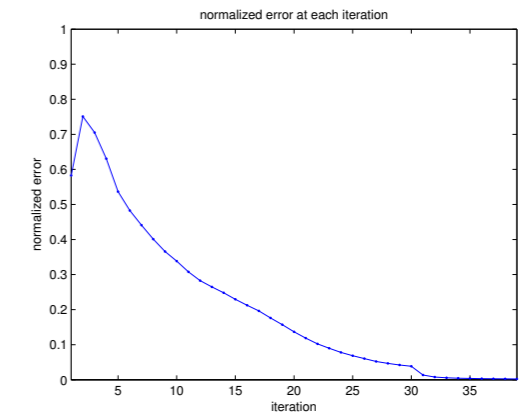
(e)



(f)



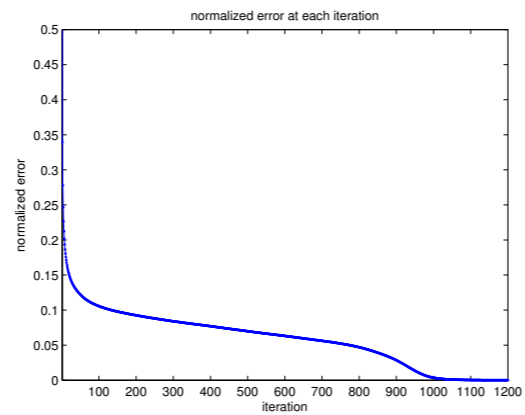
(g)



(h)



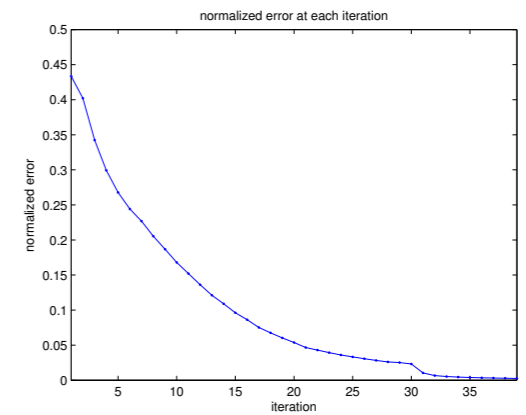
(i)



(j)



(k)

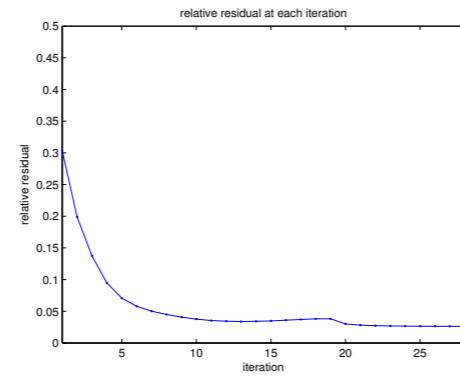


(l)

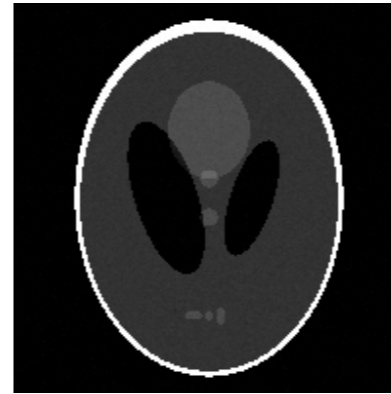
(e) - (h) Low resolution 40 x 40 block illumination with OR=2
(i) - (l) High resolution illumination with OR=1



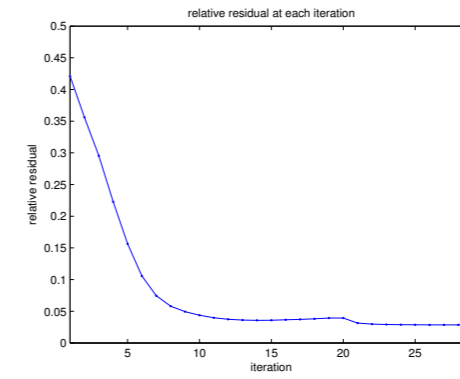
(a)



(b)



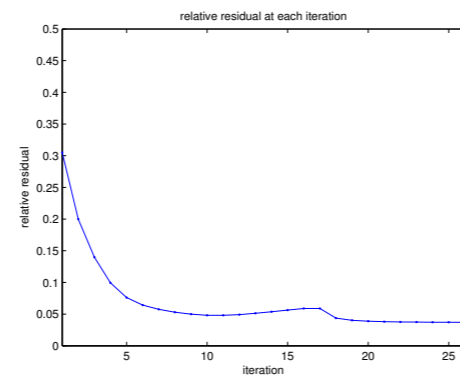
(c)



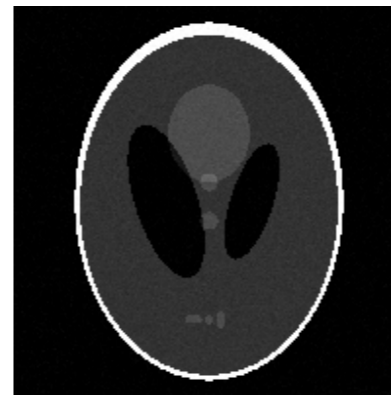
(d)



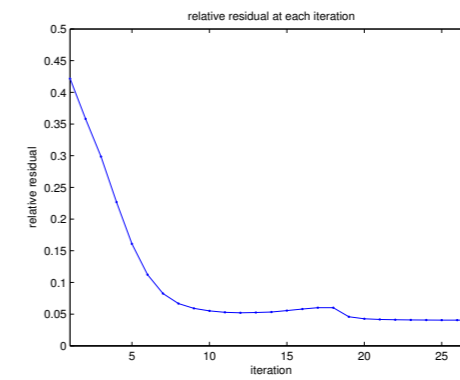
(e)



(f)



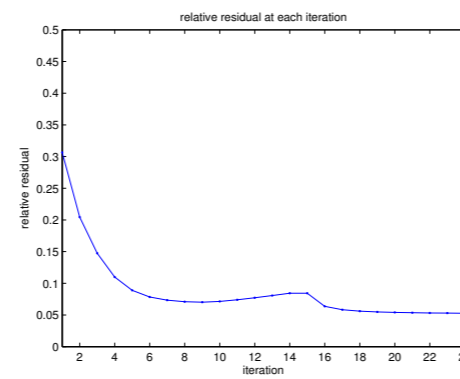
(g)



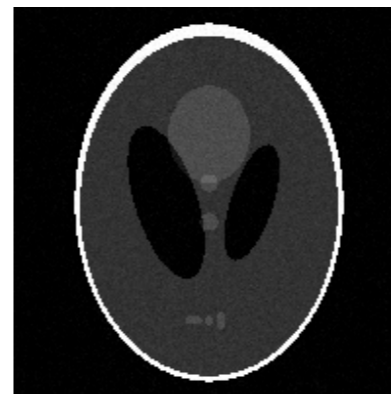
(h)



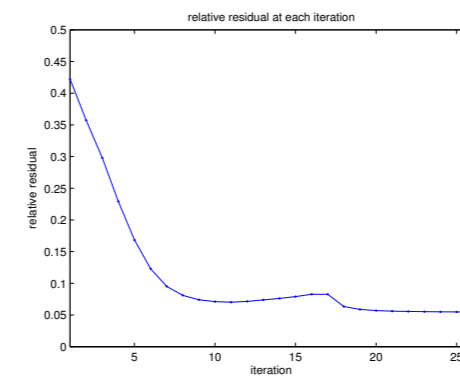
(i)



(j)



(k)

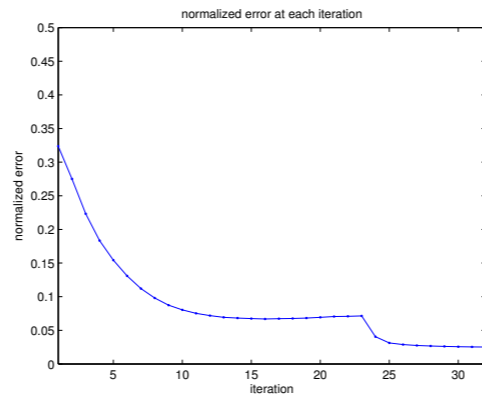


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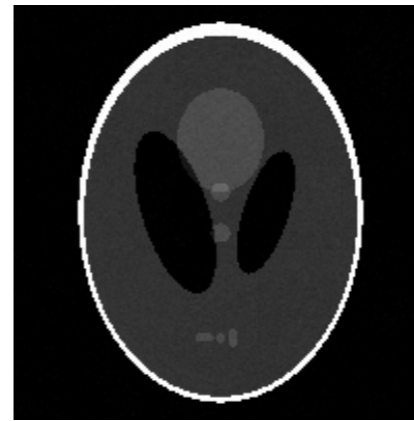
High resolution illumination with 5% Gaussian, Poisson and illuminator errors



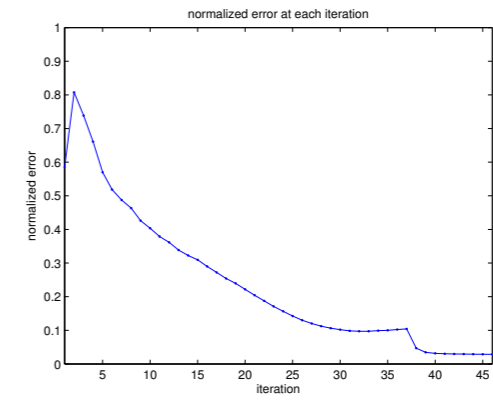
(a)



(b)



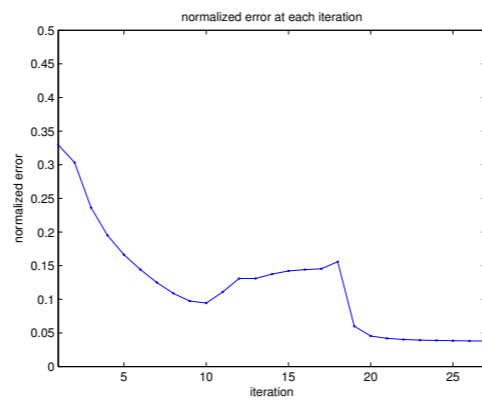
(c)



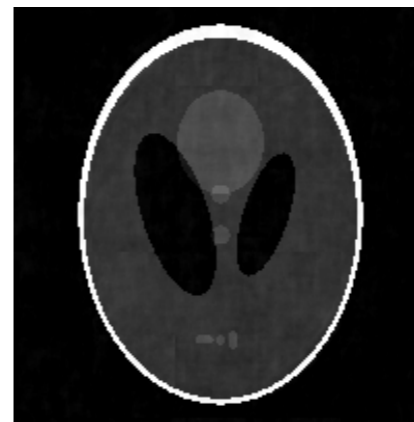
(d)



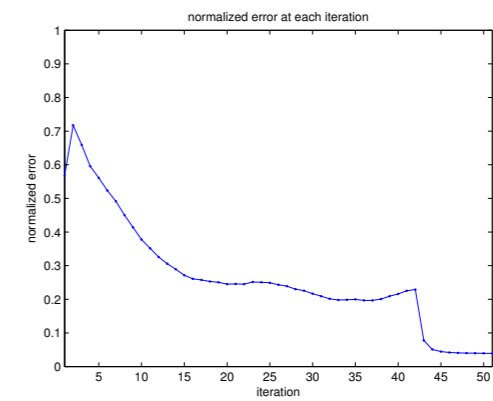
(e)



(f)



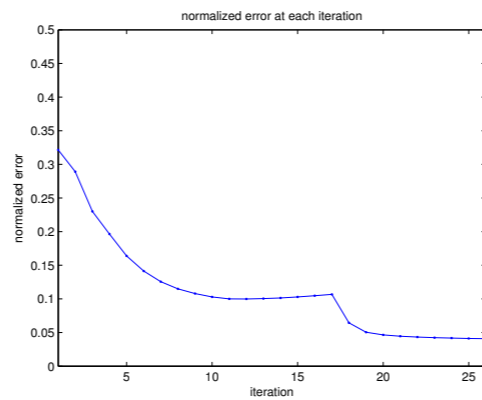
(g)



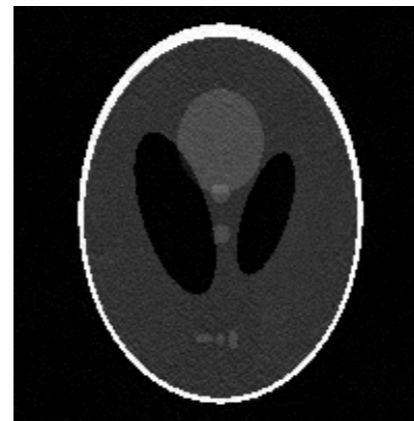
(h)



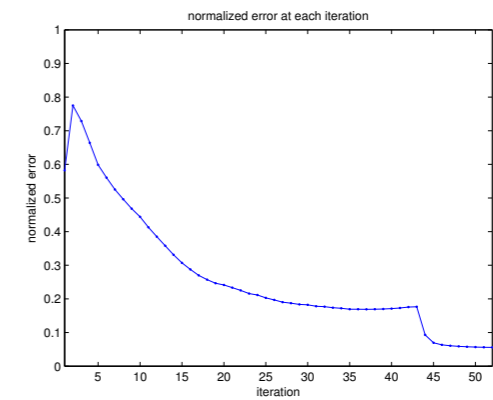
(i)



(j)



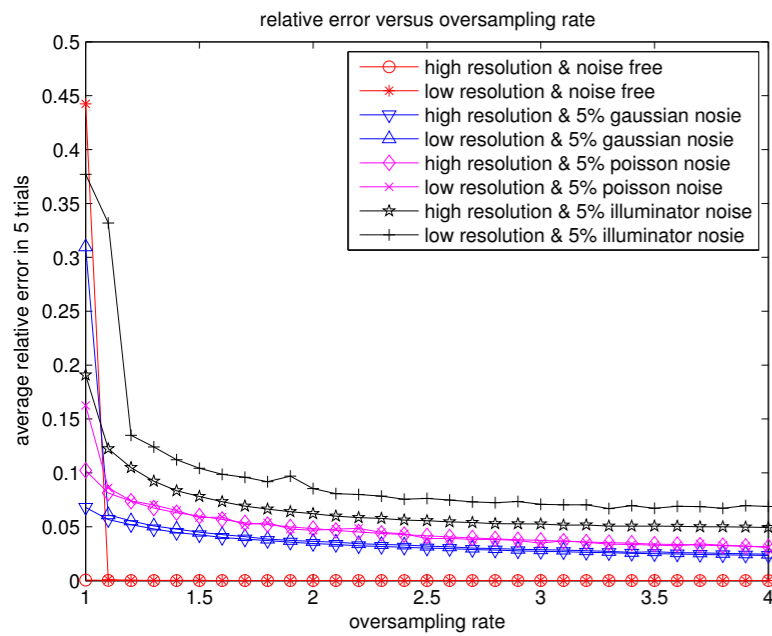
(k)



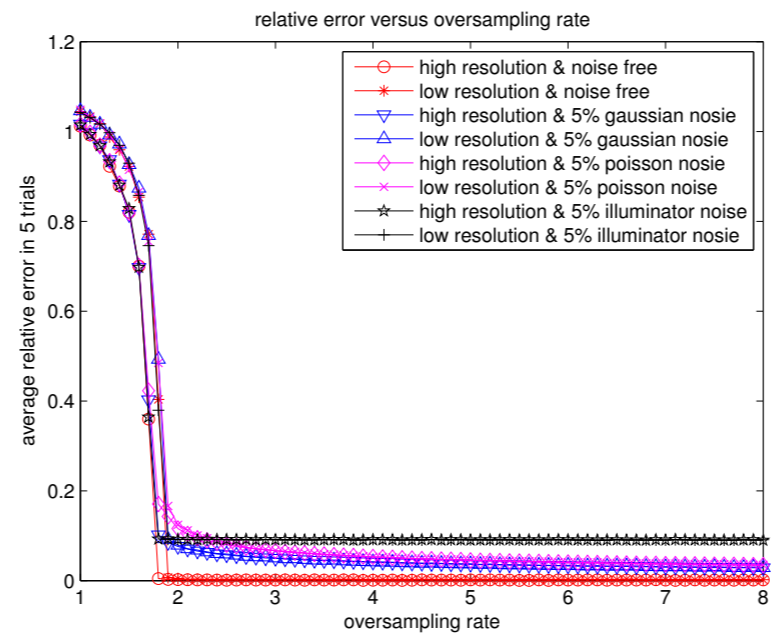
(l)

Low resolution illumination with 5% Gaussian, Poisson and illuminator errors

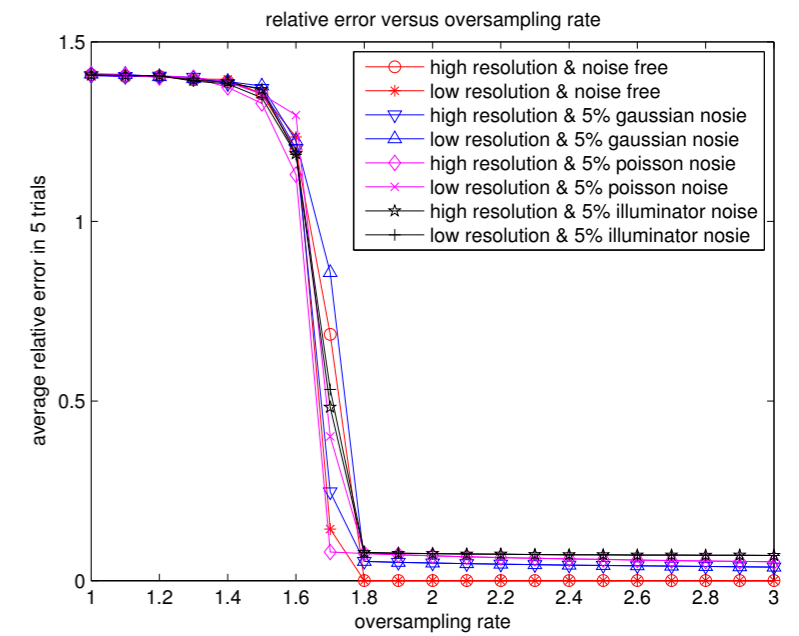
Compressed measurement



(a)



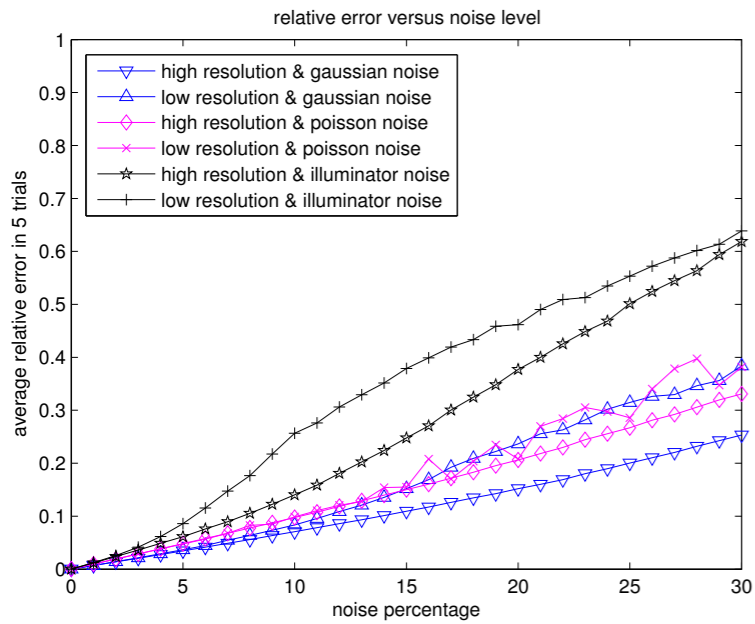
(b)



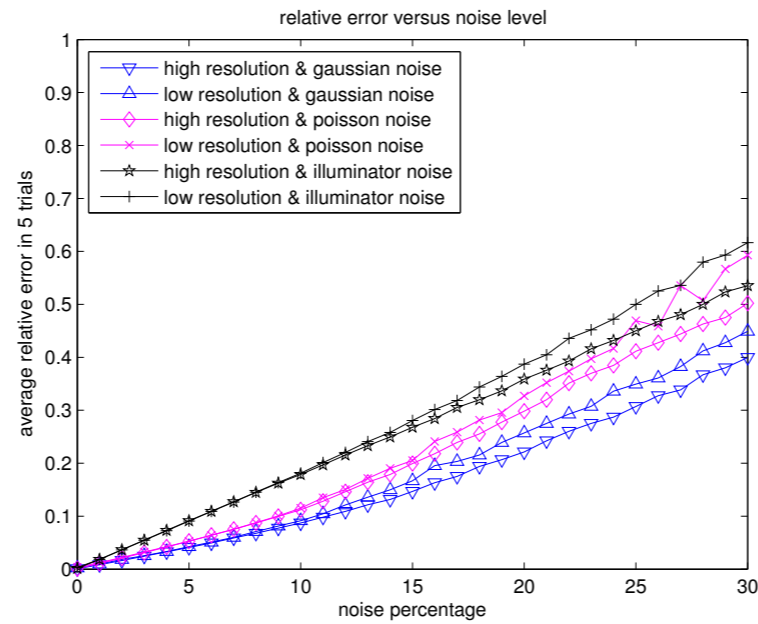
(c)

- (a) real-valued
- (b) positive real & imaginary parts
- (c) no constraint

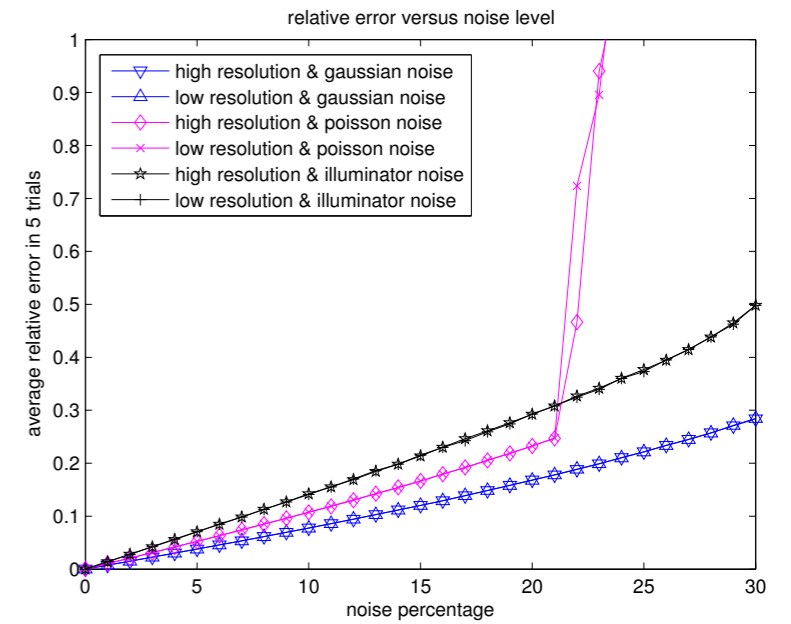
Noise stability



(a)



(b)



(c)

- (a) real-valued objects
- (b) positive real & imaginary parts
- (c) no constraint

Conclusions

- Random illumination as enabling tool for phase retrieval.
- Absolute uniqueness
- Fast convergence
- $OR = 1$ (real) or 2 (complex)
- Proof of convergence: HIO?
- References:
 - A. Fannjiang [Absolute uniqueness in phase retrieval with random illumination](#) Inverse Problems 2012 (arXiv:1110.5097)
 - A. Fannjiang and W. Liao [Phase retrieval with random phase illumination](#) arXiv:1206.1001