

Uniqueness and reconstruction formulas for phase retrieval via reference objects with applications in coherent diffractive imaging

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SFB 755 - Nanoscale Photonic Imaging

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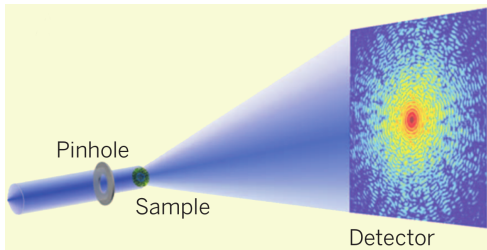


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Motivation

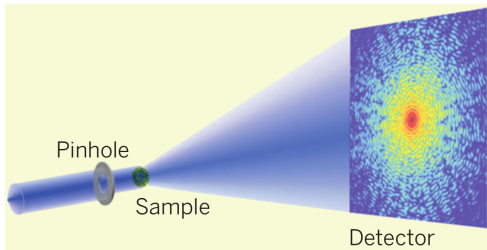
Motivation: Coherent Diffractive Imaging (CDI)



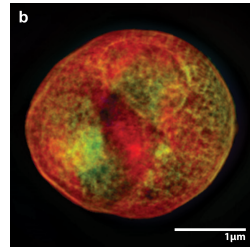
(a) Sketch of a CDI setup [Miao et al., 2015]

- ▶ Microscopic sample is illuminated by coherent X-Rays
- ▶ Measured far-field intensity pattern encodes object structure

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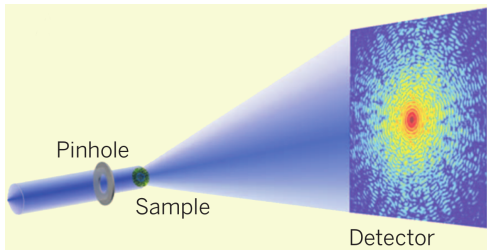
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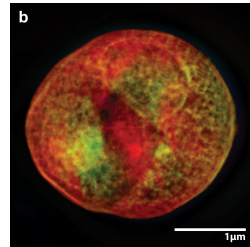
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Forward map: $F : f = \underbrace{-i\phi - \mu}_{\text{Phase shifts (+absorption)}} \mapsto I = \left| \underbrace{\mathcal{F}}_{\text{Fourier transform}} \underbrace{(p \cdot \exp(-i\phi - \mu))}_{=f \text{ (exit wave field)}} \right|^2$

The Phase Retrieval Problem

Inverse Problem 1 (Phase Retrieval)

Recover $f \in A \subset \{h \in L_c^1(\mathbb{R}^m) : \text{supp}(h) \subset \Omega\}$ from $\hat{g} := |\mathcal{F}(f)|^2$ for some (compact) support domain $\Omega \subset \mathbb{R}^m$.

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Difficulties:

► *Non-uniqueness:*

- In $m = 2D$: discrete problem is “almost always” unique [Barakat and Newsam, 1984]
- *Counter-example:* for $f(x, y) = \mathbf{1}_{[0,1]^2}(x, y) \cdot \exp(x)$, IP1 has infinitely many real-valued solutions $\tilde{f} \in L^2([0, 1]^2)$.

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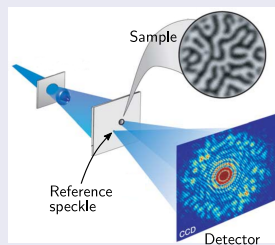
This talk: What if a part of the signal $f = p + h$ is known *a priori*?

Holography: Phase Retrieval with Reference Signals

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In the setting of Inverse Problem 1, let $f = p + h$ with $p = \delta_0$ (Dirac-delta) and $\text{supp}(h) \subset \Omega_h \subset \mathbb{R}^m$ with $\text{dist}(\Omega_h, 0) > \text{diam}(\Omega_h)$. Then

$$h = \mathcal{F}^{-1} (|\mathcal{F}(p + h)|^2) |_{\Omega_h}$$



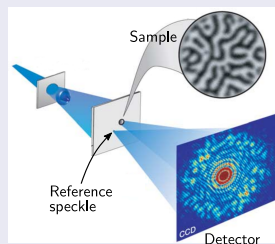
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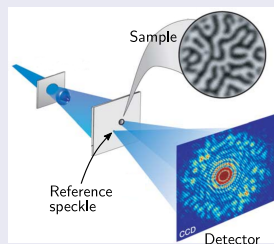
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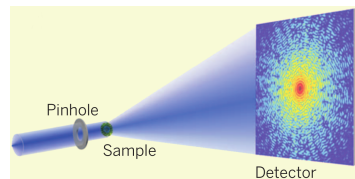
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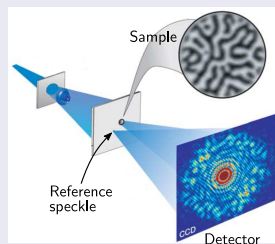
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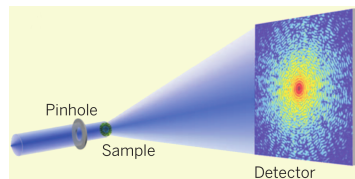
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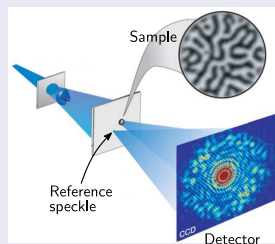
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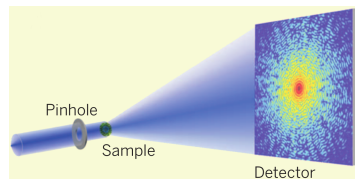
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↪ Simplest achievable setting: $p = p_0 \cdot \mathbf{1}_{\text{pinhole}}$



Basic Notions and Concepts

Abstract Setting

Definitions/Notation:

- *Convolution:* $f * g : x \mapsto \int_{\mathbb{R}^m} f(x - y)g(y) dy$
- *Reflection:* $f^\cdot : x \mapsto \overline{f(-x)}$
- *Hermitean and anti-hermitean parts:* $f^h := \frac{1}{2}(f + f^\cdot)$, $f^a := \frac{1}{2}(f - f^\cdot)$
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Observation: Extra information $2p^h * h^h + 2p^a * h^a$ if terms are separable!

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Theorem 2 (Support of convolutions (e.g. [Hörmander, 2003]))

Let $u_1, u_2 \in L_c^1(\mathbb{R}^m)$ be compactly supported. Then

$$\text{conv}(\text{supp}(u_1 * u_2)) = \text{conv}(\text{supp}(u_1)) + \text{conv}(\text{supp}(u_2)),$$

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- ▶ $(p * \Delta h)|_M \neq 0 \rightsquigarrow$ contradiction! □

Uniqueness Results and Reconstruction Formulas for Hermitean Reference Objects

Reconstruction of Complex-valued Objects

Theorem 4 (Reconstruction of small inclusions)

Let $p \in L_c^1(\mathbb{R}^m) \setminus \{0\}$ be known and hermitean with $\Omega_p := \text{conv}(\text{supp}(p))$. Let

$$A := \left\{ h \in L_c^1(\mathbb{R}^m) : \text{supp}(h) \subset 1/2 \cdot \Omega_p \right\}. \quad (3.1)$$

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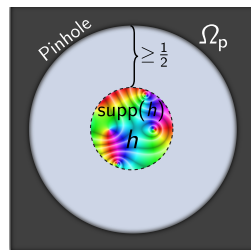
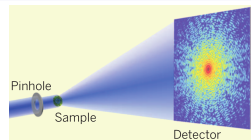


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$$(2p * h^h) |_{\mathbb{R}^m \setminus \Omega_p} = \mathcal{F}^{-1}(\Delta I) |_{\mathbb{R}^m \setminus \Omega_p} \quad (3.2a)$$

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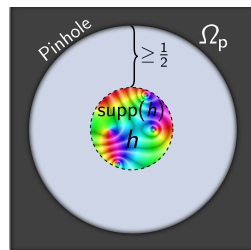
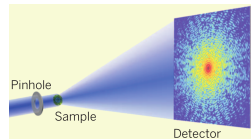


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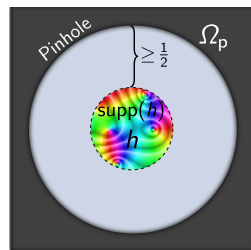
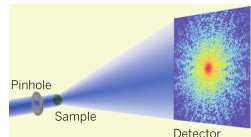


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Proof:

- Since $p^a = 0$, $p = p^h$: $\mathcal{F}^{-1}(\Delta I) = 2p * h^h + \underbrace{h^h * h^h - h^a * h^a}_{\text{supp}(\cdot) \subset 2 \cdot \text{conv}(\text{supp}(h)) \subset \Omega_p}$.

Reconstruction of Complex-valued Objects

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Let $p \in L_c^1(\mathbb{R}^m) \setminus \{0\}$ be known and hermitean with $\Omega_p := \text{conv}(\text{supp}(p))$. Let

$$A := \left\{ h \in L_c^1(\mathbb{R}^m) : \text{supp}(h) \subset 1/2 \cdot \Omega_p \right\}. \quad (3.1)$$

Then any $h \in A$ is uniquely determined by the data $\Delta I = |\mathcal{F}(p+h)|^2 - |\mathcal{F}(p)|^2$ up to reflection $h \mapsto h^\cdot$. Moreover, h^h and $\pm h^a$ can be reconstructed via

$$(2p * h^h) |_{\mathbb{R}^m \setminus \Omega_p} = \mathcal{F}^{-1}(\Delta I) |_{\mathbb{R}^m \setminus \Omega_p} \quad (3.2a)$$

$$-\mathcal{F}(h^a)^2 = \Delta I - \mathcal{F}(2p * h^h + h^h * h^h) \quad (3.2b)$$

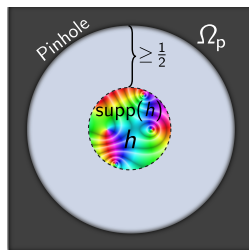
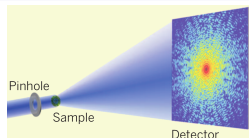


Fig.: Setting of Theorem 4

Proof:

- ▶ Since $p^a = 0$, $p = p^h$: $\mathcal{F}^{-1}(\Delta I) = 2p * h^h + \underbrace{h^h * h^h - h^a * h^a}_{\text{supp}(\cdot) \subset 2 \cdot \text{conv}(\text{supp}(h)) \subset \Omega_p}$.
- ▶ Corollary 3: $\mathcal{F}^{-1}(\Delta I) |_{\mathbb{R}^m \setminus \Omega_p} = (2p * h^h) |_{\mathbb{R}^m \setminus \Omega_p}$ uniquely determines h^h

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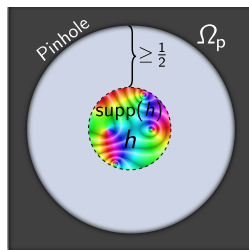
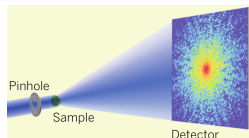


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- ▶ The *sign retrieval problem* (3.2b) is uniquely solvable up to “ \pm ” □

Reconstruction of Weak Phase Objects

Theorem 5 (Reconstruction of weak phase objects)

Let $p \in L_c^1(\mathbb{R}^m) \setminus \{0\}$ be known, hermitean and real-valued with $\Omega_p := \text{conv}(\text{supp}(p))$. Let

$$A := \left\{ i\phi : \phi \in L_c^1(\mathbb{R}^m) \text{ real-valued, } \text{supp}(\phi) \subset \Omega_p \right\}. \quad (3.3)$$

Then any $h \in A$ is uniquely determined by the data $\Delta I = |\mathcal{F}(p+h)|^2 - |\mathcal{F}(p)|^2$ up to reflection $h \mapsto h^-$. Moreover, h^h and $\pm h^a$ can be reconstructed via

$$2p * h^h = i\mathfrak{I}(\mathcal{F}^{-1}(\Delta I)) \quad (3.4a)$$

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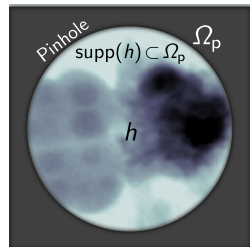
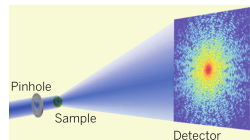


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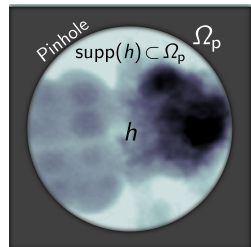
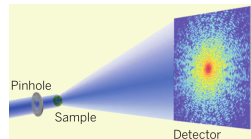


Fig.: Setting of Theorem 5

Proof:

- ▶ We have $\mathcal{F}^{-1}(\Delta I) = 2p * h^h + h^h * h^h - h^a * h^a$
- ▶ $2p * h^h$ is purely imaginary and $h^h * h^h - h^a * h^a$ is real-valued □

Simple Objects

X-ray CDI-contrast: $h \propto \exp\left(-i \underbrace{\phi}_{\text{phase shifts}} - \underbrace{\mu}_{\text{attenuation}}\right) - 1.$

Definition 6 (Simple Objects)

We call a class of functions $A \subset L_c^1(\mathbb{R}^m)$ *simple* if for any $h \in A$ and any $\Omega \subset \mathbb{R}^m$, the anti-Hermitian part $h^a|_{\Omega}$ is uniquely determined by $h^h|_{\Omega}$.

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Classes of Simple Objects:

- *Weak homogeneous objects*: For $a, b \in \mathbb{R} \setminus \{0\}$ fixed, let $A = \{(a + ib)\phi : \phi \in L_c^1(\mathbb{R}^m) \text{ real-valued}\}$. Then

$$h^a(x) = \frac{a}{b} \Im(h^h(x)) + \frac{ib}{a} \Re(h^h(x)) \quad \text{a.e. for all } h \in A$$

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- **Pure phase objects:** For fixed sign “ \pm ”, let $A = \{\exp(\pm i\phi) - 1 : \phi \in L_c^1(\mathbb{R}^m) \text{ real-valued}, 0 \leq \phi < \frac{\pi}{2} \text{ a.e.}\}$. Then

$$h^a(x) = \pm i \frac{h^h(x) + 1}{|h^h(x) + 1|} \cdot (1 - |h^h(x) + 1|^2)^{\frac{1}{2}} \quad \text{a.e. for all } h \in A$$

Reconstruction of Simple Objects

Theorem 7 (Reconstruction of simple objects)

Let $p \in L_c^1(\mathbb{R}^m) \setminus \{0\}$ be known and hermitean with $\Omega_p := \text{conv}(\text{supp}(p))$. For some $\varepsilon > 0$, let

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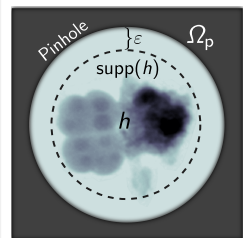
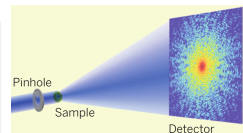


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Initialize: $k = 0, h_0 = 0$

Iterate: $g_k = (\mathcal{F}^{-1}(\Delta I) - h_k * h_k^-) |_{M_k}$

Solve $(2p * h_{k+1}^h) |_{M_k} = g_k$

$h_{k+1} = (h_{k+1}^h + h^a(h_{k+1}^h)) |_{\Omega_k}$

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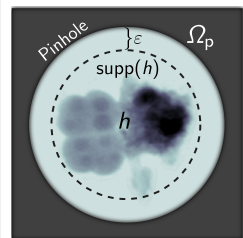
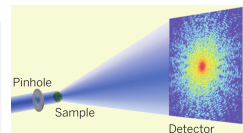


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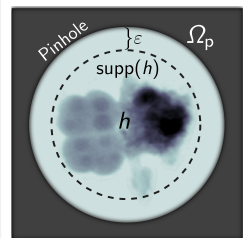
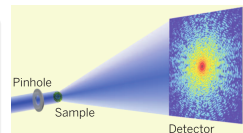


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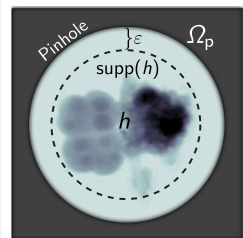
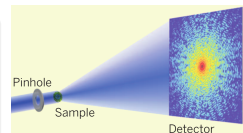


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Remark:

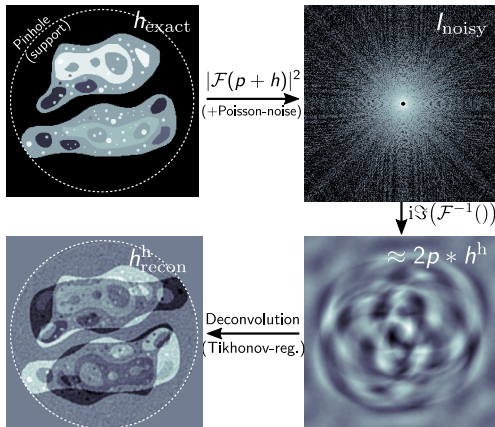
Linear inversions +
nonlinear data updates
 \rightsquigarrow Newton algorithm

Numerical Examples and Conclusions

Proof-of-principle: Simulation of Weak Phase Object

Simulation:

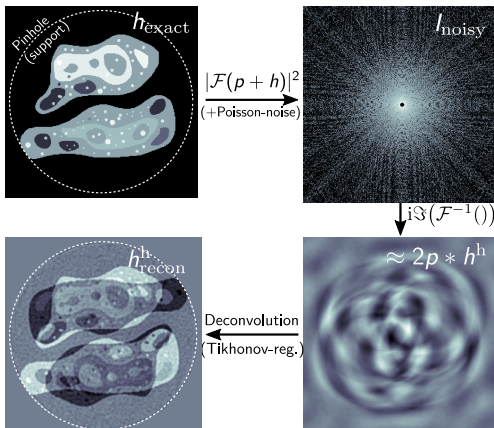
- ▶ Circular probe $p = \mathbf{1}_{\text{circle}}$
(= reference signal)
- ▶ Weak phase object $h = p \cdot (\exp(-i\phi) - 1)$
with $\phi \in [0; 0.3]$
- ▶ $I_{\text{noisy}} = |\mathcal{F}(p + h)|^2 +$
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- ▶ Reconstruct $h^h \approx -i\phi^a$
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(cf. Theorem 5)



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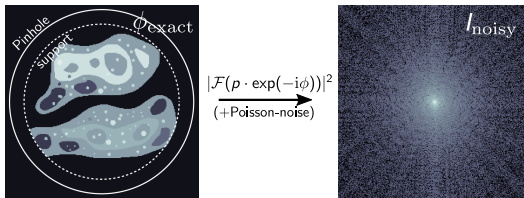
Observations:

- Stable and linear reconstruction of hermitean part h^h
- Remaining *nonlinear* reconstruction of h^a still challenging

Proof-of-principle: Simulation of Strong Phase Object

Simulation:

- ▶ Circular probe $p = \mathbf{1}_{\text{circle}}$ (= reference signal)
- ▶ Strong phase object $h = p \cdot (\exp(-i\phi) - 1)$ with $\phi \in [0; 1]$ with smaller support than p
- ▶ $I_{\text{noisy}} = F(\phi) := |\mathcal{F}(p \cdot \exp(-i\phi))|^2 + \text{Poisson noise}$

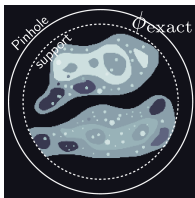


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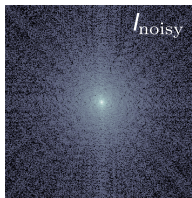
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- ▶ Reconstruction by regularized Newton-method (\rightsquigarrow Theorem 7)

$$\phi_{k+1} = \underset{\phi}{\operatorname{argmin}} \|F(\phi_k) + F[\phi_k](\phi - \phi_k) - I_{\text{noisy}}\|^2 + \beta \|\phi - \phi_k\|_{L^2}^2 + \alpha \|\phi\|_{L^2}^2$$



$$\begin{array}{c} |\mathcal{F}(p \cdot \exp(-i\phi))|^2 \\ \xrightarrow{(+\text{Poisson-noise})} \end{array}$$

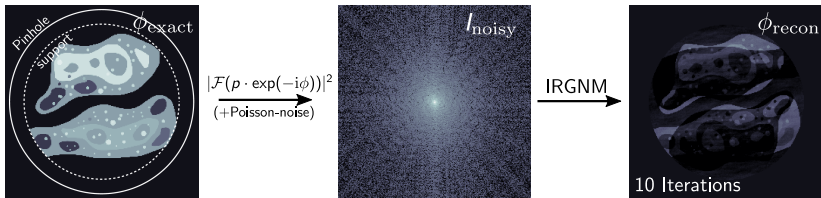


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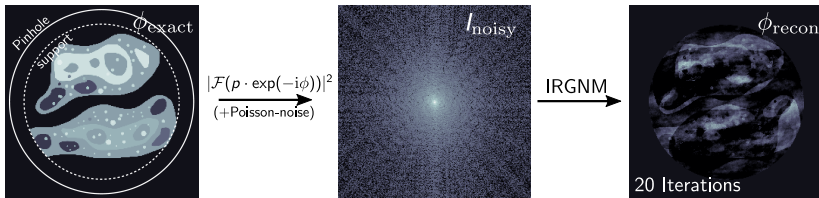


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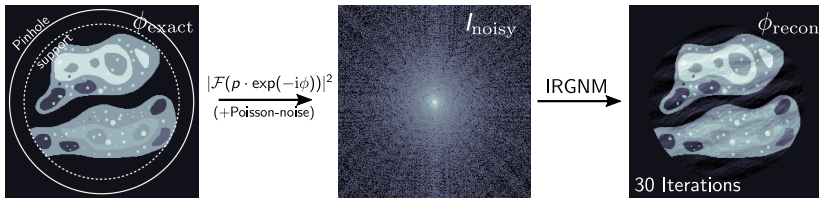


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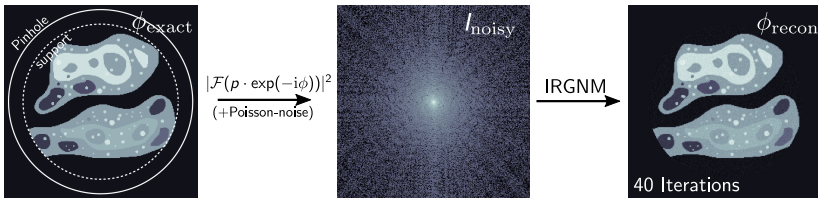


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Observation: Feasible reconstruction as found in [Hohage and Werner, 2013]

Conclusions

Summary:

- CDI enables nanoscale X-ray imaging but requires phase retrieval
- Image reconstruction suffers from non-uniqueness + non-convexity
- ✓ Proved uniqueness and reconstruction formulas in the presence of a suitable reference signal \rightsquigarrow *holographic approach*
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Future Work:

- Extend to more general (*non-Hermitian*) reference signals
- Apply holographic approach to other experimental settings

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