Blind Ptychography: Theory & Algorithms

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Outline

• Introduction and motivation: Phase retrieval vs. ptychography.

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- Mathematical set-up
- Inherent ambiguities
- Local uniqueness of the exit waves
- Raster grid pathology
- Global uniqueness for mixing schemes.
- Reconstruction
- Conclusion

Phase retrieval



• Mask/probe μ + propagation \mathcal{F} + intensity measurement:

data = diffraction pattern = $|\mathcal{F}(f \otimes \mu)|^2$, \mathcal{F} = Fourier transform.

Ptychography: extended objects

Hoppe (1969): electron microscopy.



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• Inverse problem with windowed Fourier intensities.

Fourier ptychography

Zheng et al. (2013): Convolution in the Fourier domain.



Fourier ptychography

Zheng et al. (2014)



Mask/probe retrieval

Thibault et al. (08/09) - Lensless coherent diffractive imaging



- Relative residual reduces (from 32% to 18%) after mask recovery routine is turned on.
- Simultaneous recovery of the mask and the object?

Maiden et al. 2017



• A randomly masked aperture of approximately the same size as the true probe was used as an initial probe estimate and free-space was used as the initial object estimate.

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• Overlap ratio 70 - 80%.

Twin image ambiguity: Chen & F (2017)

Fresnel mask $\mu^{0}(\mathbf{k}) := \exp\left\{i\pi\rho|\mathbf{k}|^{2}/m\right\}$



No uniqueness for certain ρ even if the mask is known!

Notation & set-up

- $\mathcal{T}:$ all the shifts $t\in\mathbb{Z}^2$ involved in the measurement.
- μ^{0} the initial mask; $\mu^{\mathbf{t}}$ the $\mathbf{t}\text{-shifted}$ mask
- $\mathcal{M}^0 = \mathbb{Z}^2_m$; \mathcal{M}^t the domain of μ^t .
- $\mathcal{M} := \cup_{t \in \mathcal{T}} \mathcal{M}^t$
- f^t : the object restricted to \mathcal{M}^t
- $\operatorname{Twin}(f^{\mathsf{t}})$: 180°-rotation of $\overline{f^{\mathsf{t}}}$ around the center of \mathcal{M}^{t}
- $f = \lor_{\mathbf{t}} f^{\mathbf{t}} \subseteq \mathcal{M}$ and refer to each $f^{\mathbf{t}}$ as a part of f.

The original object is broken up into a set of overlapping object parts, each of which produces a coded diffraction pattern (coded by μ^{t}).

Linear phase ambiguity

Consider the probe and object estimates

$$\begin{split} \nu^0(\mathbf{n}) &= \mu^0(\mathbf{n}) \exp(-\mathrm{i} a - \mathrm{i} \mathbf{w} \cdot \mathbf{n}), \quad \mathbf{n} \in \mathcal{M}^0\\ g(\mathbf{n}) &= f(\mathbf{n}) \exp(\mathrm{i} b + \mathrm{i} \mathbf{w} \cdot \mathbf{n}), \quad \mathbf{n} \in \mathbb{Z}_n^2 \end{split}$$

for any $a, b \in \mathbb{R}$ and $\mathbf{w} \in \mathbb{R}^2$. For any \mathbf{t} , we have the following calculation

$$\nu^{\mathbf{t}}(\mathbf{n}) = \nu^{0}(\mathbf{n} - \mathbf{t})$$

= $\mu^{0}(\mathbf{n} - \mathbf{t}) \exp(-i\mathbf{w} \cdot (\mathbf{n} - \mathbf{t})) \exp(-ia)$
= $\mu^{\mathbf{t}}(\mathbf{n}) \exp(-i\mathbf{w} \cdot (\mathbf{n} - \mathbf{t})) \exp(-ia)$

and hence for all $n \in \mathcal{M}^t, t \in \mathcal{T}$

$$\nu^{\mathbf{t}}(\mathbf{n})g^{\mathbf{t}}(\mathbf{n}) = \mu^{\mathbf{t}}(\mathbf{n})f^{\mathbf{t}}(\mathbf{n})\exp(\mathrm{i}(b-a))\exp(\mathrm{i}\mathbf{w}\cdot\mathbf{t})$$

implying g and ν^0 produce the same ptychographic data as f and μ^0 since for each t, $\nu^t \odot g^t$ is a constant phase factor times $\mu^t \odot f^t_* \odot f^t_* \odot g^t_* \odot g^t_*$

Measurement scheme

- $\bullet \ \mathcal{M}^t = \mathsf{nodes}$
- Two nodes are s-connected if $|\mathcal{M}^{\mathbf{t}} \cap \mathcal{M}^{\mathbf{t}'} \cap \operatorname{supp}(f)| \ge s$.





Block phase

Theorem (F & Chen 2018)

Let the scheme be s-connected and each f^t is a non-line object. Suppose that some f^t has a tight support in \mathcal{M}^t and that $\mu^0 \neq 0$ has independently distributed random phases over at least the range of length π . Suppose that ν^0 with

(MPC)
$$\Re\left[\overline{\nu^0}(\mathbf{n})\mu^0(\mathbf{n})\right] > 0, \quad \forall \mathbf{n} \in \mathcal{M}^0,$$

and an arbitrary object $g = \bigcup_k g^k$ produce the same ptychographic data as f and μ^0 . Then with probability at least $1 - c^s$, c < 1,

$$\nu^{\mathbf{t}} \odot g^{\mathbf{t}} = e^{\mathrm{i}\theta_{\mathbf{t}}} \mu^{\mathbf{t}} \odot f^{\mathbf{t}}, \quad \forall \mathbf{t} \in \mathcal{T},$$

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where c depends on the mask phase distribution.

 $\theta_{\mathbf{t}} =$ block phases depending on \mathbf{t} but not \mathbf{n} .

Object support constraint (OSC)

 f^{t} has a tight support in \mathcal{M}^{t} : \mathcal{M}^{t} is the smallest rectangle containing f^{t} .



OSC is a relaxation of the tight support condition.

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OSC counter-example

Let
$$m = 2n/3$$
, $\mathbf{t} = (m/2, 0)$ $f^0 = [0, f_1^0]$ and $f^{\mathbf{t}} = [f_0^1, 0]$ with $f_1^0 = f_0^1$.
Likewise, $\mu^0 = [\mu_0^0, \mu_1^0], \mu^{\mathbf{t}} = [\mu_0^1, \mu_1^1]$.
Let $\nu^0 = \mu^0, \nu^{\mathbf{t}} = \mu^{\mathbf{t}}$ and $g^0 = [g_0^0, 0], g^{\mathbf{t}} = [0, g_1^1]$ where

$$\begin{array}{lll} g^0(\mathbf{n}) &=& \bar{f}^0(\mathbf{N}-\mathbf{n})\bar{\mu}^0(\mathbf{N}-\mathbf{n})/\mu^0(\mathbf{n}), & \forall \mathbf{n} \in \mathcal{M}^0\\ g^{\mathbf{t}}(\mathbf{n}) &=& \bar{f}^{\mathbf{t}}(\mathbf{N}+2\mathbf{t}-\mathbf{n})\bar{\mu}^{\mathbf{t}}(\mathbf{N}+2\mathbf{t}-\mathbf{n})/\mu^{\mathbf{t}}(\mathbf{n}), & \forall \mathbf{n} \in \mathcal{M}^{\mathbf{t}}. \end{array}$$

Hence $g^0 \odot \mu^0$ and $g^t \odot \mu^t$ produce the same diffraction patterns as $f^0 \odot \mu^0$ and $f^t \odot \mu^t$ but

$$\begin{array}{rcl} g^{0} \odot \mu^{0} & \neq & e^{\mathrm{i}\theta_{0}}f^{0} \odot \mu^{0} \\ g^{\mathrm{t}} \odot \mu^{\mathrm{t}} & \neq & e^{\mathrm{i}\theta_{\mathrm{t}}}f^{\mathrm{t}} \odot \mu^{\mathrm{t}} \end{array}$$

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even when the mask estimate is perfectly accurate.

Mask phase constraint (MPC)

- Let μ⁰ be a nonvanishing random mask with phase at each pixel continuously and independently distributed according to a nonvanishing probability density function p_γ on (−γπ, γπ] with a constant γ ≤ 1.
- Suppose our mask estimate is ν^0 . Write the relative mask error as

$$\alpha(\mathbf{n}) \exp[\mathrm{i}\phi(\mathbf{n})] = \nu^0(\mathbf{n})/\mu^0(\mathbf{n}), \quad \alpha(\mathbf{n}) > 0.$$

• F. & Chen (2018): ν^0 satisfies MPC if

 $|\phi(\mathbf{n}) - \phi_0| < \min\{\gamma, 1/2\} \mod 2\pi,$

for a constant ϕ_0 , i.e. $\nu^0(\mathbf{n})$ is pointing in the half plane in \mathbb{C} with the normal vector $\mu^0(\mathbf{n})$ for all \mathbf{n} .

Counter-examples exist!

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Raster scan

Raster scan: $\mathbf{t}_{kl} = \tau(k, l), k, l \in \mathbb{Z}$ where τ is the step size. $\mathcal{M} = \mathbb{Z}_n^2, \ \mathcal{M}^0 = \mathbb{Z}_m^2, n > m$, with the periodic boundary condition.





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Block phase

Theorem (F 2018)

Let $\mathcal{T} = \{\mathbf{t}_k\}$ be a **v**-generated cyclic group of order q and M^k the \mathbf{t}_k -shifted mask domain. Suppose that

$$u^k({\mathsf{n}}) {\mathsf{g}}^k({\mathsf{n}}) = e^{{\mathrm{i}} heta_k} \mu^k({\mathsf{n}}) f^k({\mathsf{n}}), \quad \textit{for all } {\mathsf{n}} \in \mathcal{M}^k \textit{ and } {\mathsf{t}}_k \in \mathcal{T}.$$

lf

$$\mathcal{M}^k \cap \mathcal{M}^{k+1} \cap supp(f) \cap (supp(f) \oplus \mathbf{v}) \neq \emptyset, \quad \forall k$$

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then $\{\theta_0, \theta_1, \dots, \theta_{q-1}\}$ form an arithmetic progression.

Non-APA ambiguity

For
$$q = 3, \tau = m/2$$
, let

$$f = \begin{bmatrix} f_{00} & f_{10} & f_{20} \\ f_{01} & f_{11} & f_{21} \\ f_{02} & f_{12} & f_{22} \end{bmatrix}, \quad g = \begin{bmatrix} f_{00} & e^{i2\pi/3}f_{10} & e^{i4\pi/3}f_{20} \\ e^{i2\pi/3}f_{01} & e^{i4\pi/3}f_{11} & f_{21} \\ e^{i4\pi/3}f_{02} & f_{12} & e^{i2\pi/3}f_{22} \end{bmatrix}$$

be the object and its reconstruction, respectively, where $f_{ij}, g_{ij} \in \mathbb{C}^{n/3 \times n/3}$. Let

$$\mu^{kl} = \begin{bmatrix} \mu_{00}^{kl} & \mu_{10}^{kl} \\ \mu_{01}^{kl} & \mu_{11}^{kl} \end{bmatrix}, \quad \nu^{kl} = \begin{bmatrix} \mu_{00}^{kl} & e^{-i2\pi/3}\mu_{10}^{kl} \\ e^{-i2\pi/3}\mu_{01}^{kl} & e^{-i4\pi/3}\mu_{11}^{kl} \end{bmatrix}, \quad k, l = 0, 1, 2,$$

be the probe and its estimate, respectively, where $\mu_{ij}^{kl}, \nu_{ij}^{kl} \in \mathbb{C}^{n/3 \times n/3}$. It is verified straightforwardly that $\nu^{ij} \odot g^{ij} = e^{i(i+j)2\pi/3}\mu^{ij} \odot f^{ij}$.

Periodic ambiguity (raster grid pathology)

 $(au=m/2)~{f t}_{kl}$ -shifted probes μ^{kl} and u^{kl} can be written as

$$\mu^{kl} = \begin{bmatrix} \mu_{00}^{kl} & \mu_{10}^{kl} \\ \mu_{01}^{kl} & \mu_{11}^{kl} \end{bmatrix}, \quad \nu^{kl} = \begin{bmatrix} \epsilon \odot \mu_{ij}^{kl} \end{bmatrix}$$

Let

$$\epsilon = [\alpha(\mathbf{n}) \exp(\mathrm{i}\phi(\mathbf{n}))], \quad \epsilon^{-1} = [\alpha^{-1}(\mathbf{n}) \exp(-\mathrm{i}\phi(\mathbf{n}))] \in \mathbb{C}^{\tau \times \tau}.$$

Consider the two objects

$$f = \begin{bmatrix} f_{00} & \dots & f_{q-1,0} \\ \vdots & \vdots & \vdots \\ f_{0,q-1} & \dots & f_{q-1,q-1} \end{bmatrix}, \quad g = \begin{bmatrix} \epsilon^{-1} \odot f_{ij} \end{bmatrix}$$

Two exit waves $\mu^{kl} \odot f^{kl}$ and $\nu^{kl} \odot g^{kl}$ are identical. But the estimates are far off.

Mixing schemes

- Rank-one perturbation $\mathbf{t}_{kl} = \tau(k, l) + (\delta_k^1, \delta_l^2).$
- Full-rank perturbation $\mathbf{t}_{kl} = \tau(k, l) + (\delta_{kl}^1, \delta_{kl}^2).$





Global uniqueness

Theorem

Suppose f does not vanish in \mathbb{Z}_n^2 . Let $a_j^i = 2\delta_{j+1}^i - \delta_j^i - \delta_{j+2}^i$ and let $\{\delta_{j_k}^i\}$ be the subset of perturbations satisfying $\operatorname{gcd}_{j_k}\{|a_{j_k}^i|\} = 1$, i = 1, 2, and

$$\begin{aligned} \tau &\geq \max_{i=1,2} \{ |a_{j_k}^i| + \delta_{j_k+1}^i - \delta_{j_k}^i \} \\ 2\tau &\leq m - \max_{i=1,2} \{ \delta_{j_k+2}^i - \delta_{j_k}^i \}, \quad (> 50\% \text{ overlap}) \\ m - \tau &\geq 1 + \max_{k'} \max_{i=1,2} \{ |a_{j_k}^i| + \delta_{k'+1}^i - \delta_{k'}^i \}. \end{aligned}$$

Then APA and SF are the only ambiguities, i.e. for some explicit r

$$g(\mathbf{n})/f(\mathbf{n}) = \alpha^{-1}(0) \exp(i\mathbf{n} \cdot \mathbf{r}),$$

$$\nu^{0}(\mathbf{n})/\mu^{0}(\mathbf{n}) = \alpha(0) \exp(i\phi(0) - i\mathbf{n} \cdot \mathbf{r}),$$

$$\theta_{kl} = \theta_{00} + \mathbf{t}_{kl} \cdot \mathbf{r}.$$

Theorem (F & Chen 2018)

If ${\mathcal T}$ satisfies the mixing property, then

$$g(\mathbf{n})/f(\mathbf{n}) = \alpha^{-1}(0) \exp(i\mathbf{n} \cdot \mathbf{r}),$$

$$\nu^{0}(\mathbf{n})/\mu^{0}(\mathbf{n}) = \alpha(0) \exp(i\phi(0) - i\mathbf{n} \cdot \mathbf{r}),$$

$$\theta_{\mathbf{t}} = \theta_{0} + \mathbf{t} \cdot \mathbf{r}.$$

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Alternating minimization

 $|\mathcal{F}(\mu, f)| = b$: the ptychographic data. Define $A_k h := \mathcal{F}(\mu_k, h)$, $B_k \eta := \mathcal{F}(\eta, f_{k+1})$. We have $A_k f_{j+1} = B_j \mu_k$.

- Initial guess μ_1 .
- **2** Update the object estimate $f_{k+1} = \underset{g \in \mathbb{C}^{n \times n}}{\operatorname{argmin}} \mathcal{L}(A_k^*g)$
- **③** Update the probe estimate $\mu_{k+1} = \underset{\nu \in \mathbb{C}^{m \times m}}{\operatorname{argmin}} \mathcal{L}(B_k^* \nu)$
- Iterminate when ||B^{*}_kµ_{k+1}| − b|| is less than tolerance or stagnates. If not, go back to step 2 with k → k + 1.

Two non-convex log-likelihood functions:

Poisson:
$$\mathcal{L}(y) = \sum_{i} |y[i]|^2 - b^2[i] \ln |y[i]|^2$$

Gaussian: $\mathcal{L}(y) = \frac{1}{2} |||y| - b||^2$.

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The Gaussian is non-differentiable and the hight SNR and near critical limit of the Poisson.

Initialization

• Mask/probe initialization

 $\mu_1(\mathbf{n}) = \mu^0(\mathbf{n}) \exp{[i\phi(\mathbf{n})]},$

where $\phi(\mathbf{n})$ i.i.d. uniform on $(-\pi/2,\pi/2) \Longrightarrow$

$$\Re\Big[\overline{\mu_1}(\mathbf{n})\mu^0(\mathbf{n})\Big]>0,\quad \forall\mathbf{n}\in\mathcal{M}^0,$$



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Relative error of the mask estimate

$$\sqrt{\frac{1}{\pi} \int_{-\pi/2}^{\pi/2} |e^{i\phi} - 1|^2 d\phi} = \sqrt{2(1 - \frac{2}{\pi})} \approx 0.8525$$

• Object initialization: $f_1 = \text{constant}$ or random phase object.

Douglas-Rachford splitting (DRS)

Each minimization can be cast into the form:

$$\arg\min_{g\in\mathbb{C}^{n\times n}}\mathcal{L}(A_k^*g) \Longrightarrow \arg\min_u K(u) + \mathcal{L}(u)$$

where

K =Indicator function of $\{A_k^* x : x \in \mathbb{C}^{n \times n}\}.$

DRS is defined by the following iteration for $l = 1, 2, 3 \cdots$

$$y^{l+1} = \operatorname{prox}_{K/\rho}(u^{l});$$

$$z^{l+1} = \operatorname{prox}_{\mathcal{L}/\rho}(2y^{l+1} - u^{l})$$

$$u^{l+1} = u^{l} + z^{l+1} - y^{l+1}$$

where

$$\operatorname{prox}_{\mathcal{K}/\rho}(u) = A_k^* (A_k^*)^{\dagger} u$$

$$\operatorname{prox}_{\mathcal{L}/\rho}(u) = \operatorname{argmin}_{x} \mathcal{L}(x) + \frac{\rho}{2} ||x - u||^2.$$

Fixed point algorithm with Gaussian log-likelihood

- $\rho = 1$
- Reflectors: $R_k = 2P_k I$, $S_k = 2Q_k I$.
- Gaussian:

$$u_k^{l+1} = \frac{1}{2}u_k^l + \frac{1}{2}b \odot \operatorname{sgn}(R_k u_k^l)$$
$$v_k^{l+1} = \frac{1}{2}v_k^l + \frac{1}{2}b \odot \operatorname{sgn}(S_k v_k^l).$$

Poisson:

$$\begin{aligned} u_k^{l+1} &= \ \frac{1}{2} u_k^l - \frac{1}{3} R_k u_k^l + \frac{1}{6} \sqrt{|R_k u_k^l|^2 + 24b^2} \odot \operatorname{sgn} \left(R_k u_k^l \right) \\ v_k^{l+1} &= \ \frac{1}{2} v_k^l - \frac{1}{3} S_k v_k^l + \frac{1}{6} \sqrt{|S_k v_k^l|^2 + 24b^2} \odot \operatorname{sgn} \left(S_k v_k^l \right). \end{aligned}$$

Proposition

Let $\rho \ge 1$. Let (u, v) be a linearly stable fixed point. Then $f_{\infty} := A_{\infty}^{\dagger} u$ and $\mu_{\infty} := B_{\infty}^{\dagger} v$ are a solution to blind ptychography, i.e. $|\mathcal{F}(\mu_{\infty}, f_{\infty})| = b$.

Proposition

Let (u, v) be the true solution. Then $||J_A(\eta)||_2 \le ||\eta||_2$, $||J_B(\xi)||_2 \le ||\xi||_2$ for all $\eta, \xi \in \mathbb{C}^N$ and the equality holds in the direction $\pm ib/||b||$ (and possibly elsewhere on the unit sphere).

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Test objects and error metric



 $\mathsf{RE}(k) = \min_{\alpha \in \mathbb{C}, \mathbf{k} \in \mathbb{R}^2} \frac{\|f(\mathbf{r}) - \alpha e^{-i\frac{2\pi}{n}\mathbf{k} \cdot \mathbf{r}} f_k(\mathbf{r})\|_2}{\|f\|_2}$



correlation length c = 0, 0.4m, 0.7m, 1m

Scanning schemes

- Rank-one perturbation $\mathbf{t}_{kl} = 30(k, l) + (\delta_k^1, \delta_l^2)$ where δ_k^1 and δ_l^2 are randomly selected integers in [-4, 4].
- Full-rank perturbation $\mathbf{t}_{kl} = 30(k, l) + (\delta_{kl}^1, \delta_{kl}^2)$ where δ_{kl}^1 and δ_{kl}^2 are randomly selected integers in [-4, 4].
- The adjacent probes overlap by roughly 50%.
- Boundary conditions:

Periodic BC Dark-field (enforced or not) Bright-field (enforced or not)



Dark-field vs. periodic BC



Figure: (Left) Reconstructed moduli with dark-field BC & 30 inner iterations; (right) Reconstructed phase error with periodic BC & 80 inner iterations

Experiment: Rank-one vs. full-rank



Experiment: Independent vs. correlated mask



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Noise robustness



Boundary conditions



(a) CiB with PPC(0, 0, 0.5)

(b) RPP with PPC(0, 0, 0.4)

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Full-rank scheme

Boundary conditions



(c) CiB with PPC(0, 0, 0.5)

(d) RPP with PPC(0, 0, 0.4)

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Rank-1 scheme

Conclusion

- Theory: blind ptychography can recover simultaneously the object and the probe/mask up to an affine phase factor and a constant amplitude offset.
 - \rightarrow Mixing schemes
 - $\rightarrow~\mathsf{Raster}$ scan pathology
- Algorithm: MPC Initialization + AMDRS (Convergence proof?)

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Osition uncertainty ?



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Thank you!

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