Recovery of Compactly Supported Functions from Spectrogram Measurements via Lifting

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Joint work with...



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Continuous Phase Retrieval

Joint work with...



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Background

Motivation

Applications: The phase-retrieval problem arises whenever the detectors can only capture *intensity* measurements. For example,

X-ray crystallography Diffraction imaging Ptychographic Imaging

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Our goals: Approaching realistic measurement designs compatible with, e.g., ptychography, coupled with computationally efficient and robust recovery algorithms.

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Background

Motivating Application



Diffraction Pattern

Ptychographic Image

Figure: Ptychographic Imaging

Algorithms for Discrete Phase Retrieval

- There has been a good deal of work on signal recovery from phaseless STFT measurements in the *discrete setting*:
 - \blacktriangleright First f and g are modeled as vectors ab initio,
 - ► Then recovered from discrete STFT magnitude measurements.
- Recovery techniques include
 - Iterative methods (Alt. Proj. for STFT) along the lines of Griffin and Lim [8, 12],
 - Alternating Projections [7],
 - Graph theoretic methods for Gabor frames based on polarization [11, 9],
 - Semidefinite relaxation-based methods [5], and others [2, 1, 4, 3].

Signal Recovery from STFT Measurements

- In 1-D ptychography [10, 7], a compactly supported specimen
 f: ℝ → ℂ, is scanned by a focused beam g: ℝ → ℂ which translates across the specimen in fixed overlapping shifts l₁,..., l_K ∈ ℝ.
- At each such shift a phaseless diffraction image is sampled by a detector.
- The measurements are modeled as STFT magnitude measurements:

$$b_{k,j} := \left| \int_{-\infty}^{\infty} f(t) g(t - l_k) e^{-2\pi i \omega_j t} dt \right|^2, \ 1 \le k \le K, \ 1 \le j \le N.$$
 (1)

• We aim to approximate f (up to a global phase) using these $b_{k,j}$ measurements.

• Given stacked spectrogram samples from (1),

$$\vec{b} = (b_{1,1}, \dots, b_{1,N}, b_{2,1}, \dots, b_{K,N})^T \in [0,\infty)^{NK},$$
 (2)

approximately recover a piecewise smooth and compactly supported function $f: \mathbb{R} \to \mathbb{C}$ up to a global phase.

- WLOG assume that the support of f is contained in [-1,1].
- Motivated by ptychography, we primarily consider the beam function g to also be (effectively) compactly supported within $[-a,a] \subsetneq [-1,1]$.
- Assume also that g is smooth enough that its Fourier transform decays relatively rapidly in frequency space compared to \hat{f} . Examples of such g include both Gaussians, as well as compactly supported C^{∞} bump functions [6].

- Using techniques from [4, 3] on discrete PR adapted to continuous PR, recover samples of \hat{f} at frequencies in $\Omega = \{\omega_1, \ldots, \omega_N\}$, giving $\vec{f} \in \mathbb{C}^N$ with $f_j = \hat{f}(\omega_j)$:
 - First, a truncated lifted linear system is inverted in order to learn a portion of the rank-one matrix $\vec{f}\vec{f}^*$.
 - Then, angular synchronization is used to recover \vec{f} from the portion of $\vec{f}\vec{f}^*$ above.
- Reconstruct \widehat{f} via standard sampling theorems.
- Invert this approximation in order to learn f.
- This linear system is banded and Toeplitz, with band size determined by the decay of \hat{g} : if g is effectively bandlimited to $[-\delta, \delta]$ the computational cost is $\mathcal{O}\left(\delta N(\log N + \delta^2)\right)$ **essentially FFT-time** in N for small δ .

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Lifted Formulation

Lemma (Lifting Lemma)

Suppose $f : \mathbb{R} \to \mathbb{C}$ is piecewise smooth and compactly supported in [-1,1]. Let $g \in L^2([-a,a])$ be supported in $[-a,a] \subset [-1,1]$ for some a < 1, with $||g||_{L^2} = 1$. Then for all $\omega \in \mathbb{R}$,

$$\left|\mathcal{F}\left[f \cdot S_{l}g\right](\omega)\right| = \frac{1}{2} \left|\sum_{m \in \mathbb{Z}} e^{-\pi i lm} \hat{f}\left(\frac{m}{2}\right) \hat{g}\left(\frac{m}{2} - \omega\right)\right|$$

for all shifts $l \in [a-1, 1-a]$.

Proof.

Plancherel's Theorem implies that

$$\left|\mathcal{F}\left[f \cdot S_{l}g\right](\omega)\right|^{2} = \left|\int_{-\infty}^{\infty} \hat{f}(\omega - \eta)\,\hat{g}(-\eta)\,e^{-2\pi i l\eta}d\eta\right|^{2}.$$

• Applying Shannon's Sampling theorem to \hat{f} and recalling that $\mathcal{F}[f\star g]=\hat{f}\hat{g}$ yield

$$\begin{aligned} |\mathcal{F}[f \cdot S_l g](\omega)|^2 &= \left| \sum_{m \in \mathbb{Z}} \hat{f}\left(\frac{m}{2}\right) \left[\hat{g}\left(\cdot\right) e^{-2\pi i l\left(\cdot\right)} \star \operatorname{sinc}\pi\left(m+2\left(\cdot\right)\right) \right] \left(-\omega\right) \right|^2 \\ &= \frac{1}{4} \left| \sum_{m \in \mathbb{Z}} \hat{f}\left(\frac{m}{2}\right) e^{-\pi i l\left(m-2\omega\right)} \int_{-l-1}^{-l+1} g\left(u\right) e^{-2\pi i u\left(\frac{m}{2}-\omega\right)} du \right|^2 . \end{aligned}$$

• The result follows by noting the support of g and the Fourier type integral in the last equality.

Lifted Form

We write our measurements in a lifted form

$$|\mathcal{F}[f \cdot S_l g](\omega)|^2 \approx \frac{1}{4} \vec{X_l^*} \vec{Y_\omega} \vec{Y_\omega^*} \vec{X_l}$$

where $\vec{X_l} \in \mathbb{C}^{4\delta+1}$ and $\vec{Y_\omega} \in \mathbb{C}^{4\delta+1}$ are the vectors

$$\vec{X}_{l} = \begin{pmatrix} e^{\pi i l(2\delta)} \hat{g}(-\delta) \\ e^{\pi i l(2\delta-1)} \hat{g}\left(\frac{1}{2}-\delta\right) \\ \vdots \\ e^{\pi i l \cdot 0} \hat{g}(0) \\ \vdots \\ e^{\pi i l(1-2\delta)} \hat{g}\left(\delta-\frac{1}{2}\right) \\ e^{\pi i l(-2\delta)} \hat{g}(\delta) \end{pmatrix}, \vec{Y}_{\omega} = \begin{pmatrix} \frac{\hat{f}(\omega-\delta)}{\hat{f}\left(\omega-\delta+\frac{1}{2}\right)} \\ \frac{\vdots}{\hat{f}(\omega)} \\ \vdots \\ \frac{\hat{f}(\omega)}{\hat{f}\left(\omega+\delta-\frac{1}{2}\right)} \\ \hat{f}(\omega+\delta) \end{pmatrix}$$

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Lifted Form

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$$|\mathcal{F}[f \cdot S_l g](\omega)|^2 \approx \frac{1}{4} \vec{X}_l^* \vec{Y}_\omega \vec{Y}_\omega^* \vec{X}_l$$

Here, $\vec{Y}_{\omega}\vec{Y}_{\omega}^{*}$ is the rank-one matrix

$$\begin{bmatrix} \left| \hat{f}(\omega-\delta) \right|^2 & \cdots & \overline{\hat{f}(\omega-\delta)} \hat{f}(\omega) & \cdots & \overline{\hat{f}(\omega-\delta)} \hat{f}(\omega+\delta) \\ \vdots & \ddots & \vdots & \vdots & \vdots \\ \overline{\hat{f}(\omega)} \hat{f}(\omega-\delta) & \cdots & \left| \hat{f}(\omega) \right|^2 & \cdots & \overline{\hat{f}(\omega)} \hat{f}(\omega+\delta) \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \overline{\hat{f}(\omega+\delta)} \hat{f}(\omega-\delta) & \cdots & \overline{\hat{f}(\omega+\delta)} \hat{f}(\omega) & \cdots & \left| \hat{f}(\omega+\delta) \right|^2 \end{bmatrix}$$

Note the occurrence of the magnitudes of \hat{f} on the diagonal, and the relative phase terms elsewhere.

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• Let $\mathbf{F} \in \mathbb{C}^{N \times N}$ be defined as

$$\mathbf{F}_{i,j} = \begin{cases} \overline{\hat{f}\left(\frac{i-2n-1}{2}\right)} \hat{f}\left(\frac{j-2n-1}{2}\right), & \text{if } |i-j| \le 2\delta, \\ 0, & \text{otherwise,} \end{cases}, \text{ where } n = \frac{N-1}{4} \end{cases}$$

- **F** is composed of overlapping segments of matrices $\vec{Y}_{\omega}\vec{Y}_{\omega}^*$ for $\omega \in \{-n, \dots, n\}$.
- Thus, our spectrogram measurements can be written as

$$\vec{b} \approx \text{diag}(\mathbf{GFG}^*),$$
 (3)

where $\mathbf{G} \in \mathbb{C}^{NK \times N}$ is a block Toeplitz matrix encoding the $\vec{X_l}$'s.

• We consistently vectorize (3) to obtain a linear system which can be inverted to learn \vec{F} , a vectorized version of \mathbf{F} .

In particular, we have

$$\vec{b} \approx \mathbf{M}\vec{F},$$
 (4)

where $\mathbf{M} \in \mathbb{C}^{NK \times N^2}$ is computed by passing the canonical basis of $\mathbb{C}^{N \times N}$ through (3).

- We solve the linear system (4) as a least squares problem.
- Experiments have shown that M is of rank NK.
- The process of recovering the Fourier samples of f from \vec{F} is known as angular synchronization.

Angular Synchronization

• Angular synchronization is the process recovering d angles $\phi_1, \phi_2, \ldots, \phi_d \in [0, 2\pi)$ given noisy and possibly incomplete difference measurements of the form

$$\phi_{ij} := \phi_i - \phi_j, \quad (i,j) \in \{1,2,\dots,d\} \times \{1,2,\dots,d\}.$$

• We are interested in angular synchronization problems that arise when performing *phase retrieval* from *local correlation measurements* [4, 3].



Leading Eigenvector \leftrightarrow Phase Vector for $\delta = 1/2$ $\begin{bmatrix} |\vec{f_1}|^2 \ \vec{f_1}\vec{f_2}^* \ \vec{f_2}\vec{f_1}^* \ |\vec{f_2}|^2 \ \vec{f_2}\vec{f_3}^* \ \vec{f_3}\vec{f_2}^* \ |\vec{f_3}|^2 \ \vec{f_3}\vec{f_4}^* \ \vec{f_4}\vec{f_3}^* \ |\vec{f_4}|^2 \end{bmatrix}^T \\ \downarrow (\text{re-arrange})$ $\begin{bmatrix} |\vec{f_1}|^2 & \vec{f_1}\vec{f_2}^* & 0 & 0\\ \vec{f_2}\vec{f_1}^* & |\vec{f_2}|^2 & \vec{f_2}\vec{f_3}^* & 0\\ 0 & \vec{f_3}\vec{f_2}^* & |\vec{f_3}|^2 & \vec{f_3}\vec{f_4}^*\\ 0 & 0 & \vec{f_4}\vec{f_3}^* & |\vec{f_4}|^2 \end{bmatrix}$ (**F**, 4 δ + 1 entries in band) Continuous Phase Retrieval SampTA 2017

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Leading Eigenvector \leftrightarrow Phase Vector for $\delta = 1/2$ $\begin{bmatrix} |\vec{f_1}|^2 \ \vec{f_1}\vec{f_2}^* \ \vec{f_2}\vec{f_1}^* \ |\vec{f_2}|^2 \ \vec{f_2}\vec{f_3}^* \ \vec{f_3}\vec{f_2}^* \ |\vec{f_3}|^2 \ \vec{f_3}\vec{f_4}^* \ \vec{f_4}\vec{f_3}^* \ |\vec{f_4}|^2 \end{bmatrix}^T \\ | \text{ (re-arrange)}$ $\begin{bmatrix} |\vec{f_1}|^2 & \vec{f_1}\vec{f_2}^* & 0 & 0\\ \vec{f_2}\vec{f_1}^* & |\vec{f_2}|^2 & \vec{f_2}\vec{f_3}^* & 0\\ 0 & \vec{f_3}\vec{f_2}^* & |\vec{f_3}|^2 & \vec{f_3}\vec{f_4} \\ 0 & 0 & \vec{f_4}\vec{f_3}^* & |\vec{f_4}|^2 \end{bmatrix} (\mathbf{F}, 4\delta + 1 \text{ entries in band})$ (normalize entrywise) $\begin{bmatrix} 1 & e^{\mathrm{i}(\phi_1 - \phi_2)} & 0 & 0 \\ e^{\mathrm{i}(\phi_2 - \phi_1)} & 1 & e^{\mathrm{i}(\phi_2 - \phi_3)} & 0 \\ 0 & e^{\mathrm{i}(\phi_3 - \phi_2)} & 1 & e^{\mathrm{i}(\phi_3 - \phi_4)} \\ 0 & 0 & e^{\mathrm{i}(\phi_4 - \phi_3)} & 1 \end{bmatrix}$ Continuous Phase Retrieval SampTA 2017 15 / 24 Leading Eigenvector \leftrightarrow Phase Vector for $\delta = 1/2$ $\begin{bmatrix} |\vec{f_1}|^2 \ \vec{f_1}\vec{f_2}^* \ \vec{f_2}\vec{f_1}^* \ |\vec{f_2}|^2 \ \vec{f_2}\vec{f_3}^* \ \vec{f_3}\vec{f_2}^* \ |\vec{f_3}|^2 \ \vec{f_3}\vec{f_4}^* \ \vec{f_4}\vec{f_3}^* \ |\vec{f_4}|^2 \end{bmatrix}^T \\ | \text{ (re-arrange)}$ $\begin{bmatrix} |\vec{f_1}|^2 & \vec{f_1}\vec{f_2}^* & 0 & 0\\ \vec{f_2}\vec{f_1}^* & |\vec{f_2}|^2 & \vec{f_2}\vec{f_3}^* & 0\\ 0 & \vec{f_3}\vec{f_2}^* & |\vec{f_3}|^2 & \vec{f_3}\vec{f_4}^*\\ 0 & 0 & \vec{f_4}\vec{f_3}^* & |\vec{f_4}|^2 \end{bmatrix} (\mathbf{F}, 4\delta + 1 \text{ entries in band})$ (normalize entrywise) $\begin{bmatrix} 1 & e^{\mathrm{i}(\phi_1 - \phi_2)} & 0 & 0 \\ e^{\mathrm{i}(\phi_2 - \phi_1)} & 1 & e^{\mathrm{i}(\phi_2 - \phi_3)} & 0 \\ 0 & e^{\mathrm{i}(\phi_3 - \phi_2)} & 1 & e^{\mathrm{i}(\phi_3 - \phi_4)} \\ 0 & 0 & e^{\mathrm{i}(\phi_4 - \phi_3)} & 1 \end{bmatrix}$ (top eigenvector) $\approx [\mathbf{e}^{\mathbf{i}\phi_1} \mathbf{e}^{\mathbf{i}\phi_2} \mathbf{e}^{\mathbf{i}\phi_3} \mathbf{e}^{\mathbf{i}\phi_4}]^T$ Continuous Phase Retrieval SampTA 2017 15 / 24 Leading Eigenvector \leftrightarrow Phase Vector for $\delta = 1/2$ $\begin{bmatrix} |\vec{f_1}|^2 \ \vec{f_1}\vec{f_2}^* \ \vec{f_2}\vec{f_1}^* \ |\vec{f_2}|^2 \ \vec{f_2}\vec{f_3}^* \ \vec{f_3}\vec{f_2}^* \ |\vec{f_3}|^2 \ \vec{f_3}\vec{f_4}^* \ \vec{f_4}\vec{f_3} \ |\vec{f_4}|^2 \end{bmatrix}^T \\ | \text{ (re-arrange)}$ $\begin{bmatrix} |\vec{f_1}|^2 & \vec{f_1}\vec{f_2}^* & 0 & 0\\ \vec{f_2}\vec{f_1}^* & |\vec{f_2}|^2 & \vec{f_2}\vec{f_3}^* & 0\\ 0 & \vec{f_3}\vec{f_2}^* & |\vec{f_3}|^2 & \vec{f_3}\vec{f_4}^*\\ 0 & 0 & \vec{f_4}\vec{f_3}^* & |\vec{f_4}|^2 \end{bmatrix} (\mathbf{F}, 4\delta + 1 \text{ entries in band})$ (normalize entrywise) $\begin{bmatrix} 1 & e^{\mathrm{i}(\phi_1 - \phi_2)} & 0 & 0 \\ e^{\mathrm{i}(\phi_2 - \phi_1)} & 1 & e^{\mathrm{i}(\phi_2 - \phi_3)} & 0 \\ 0 & e^{\mathrm{i}(\phi_3 - \phi_2)} & 1 & e^{\mathrm{i}(\phi_3 - \phi_4)} \\ 0 & 0 & e^{\mathrm{i}(\phi_4 - \phi_3)} & 1 \end{bmatrix}$ (top eigenvector) $\approx [\mathbf{e}^{\mathbf{i}\phi_1} \mathbf{e}^{\mathbf{i}\phi_2} \mathbf{e}^{\mathbf{i}\phi_3} \mathbf{e}^{\mathbf{i}\phi_4}]^T$ (Reconstruction of \hat{f} samples) $\left[|\vec{f_1}| e^{i\phi_1} \quad |\vec{f_2}| e^{i\phi_2} \quad |\vec{f_3}| e^{i\phi_3} \quad |\vec{f_4}| e^{i\phi_4} \right]^T$ Continuous Phase Retrieval SampTA 2017 15 / 24

- Consider the Oscillatory Gaussian $f(x) = 2^{\frac{1}{4}}e^{-8\pi x^2}\cos(24x)\chi_{[-1,1]}$.
- Take as window the Gaussian $g(x) = c \cdot e^{-16\pi x^2} \chi_{\left[-\frac{1}{2}, \frac{1}{2}\right]}.$



• We use a total of 11 shifts of g, and choose 61 values of ω from [-15, 15] sampled in half-steps, and set $\delta = 7$.

The reconstruction in physical space is shown at selected grid points in the figure below.

The relative ℓ^2 error in physical space is 1.872×10^{-2} .



- Consider the Characteristic Function $f(x) = \chi_{\left[-\frac{1}{5}, \frac{1}{5}\right]}$.
- Take as window the Gaussian $g(x) = c \cdot e^{-32\pi x^2} \chi_{\left[-\frac{1}{2}, \frac{1}{2}\right]}$.



• We use a total of 21 shifts of g, and choose 293 values of ω from [-73,73] sampled in half-steps, and set $\delta = 10$.

The reconstruction in physical space is shown in the figure below.

The relative ℓ^2 error in physical space is 1.509×10^{-1} .



- Consider the Peicewise Smooth Function $f(x) = \frac{1}{2}\chi_{\left[-\frac{3}{20},0\right]} + \left(\frac{-10}{3}x + 1\right)\chi_{\left[0,\frac{3}{20}\right]}.$
- Take as window the Gaussian $g(x) = c \cdot e^{-32\pi x^2/5} \chi_{\left[-\frac{1}{2}, \frac{1}{2}\right]}$.



• We use a total of 41 shifts of g, and choose 281 values of ω from [-70, 70] sampled in half-steps, and set $\delta = 10$.

The reconstruction in physical space is shown in the figure below.

The relative ℓ^2 error in physical space is 1.162×10^{-1} .



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