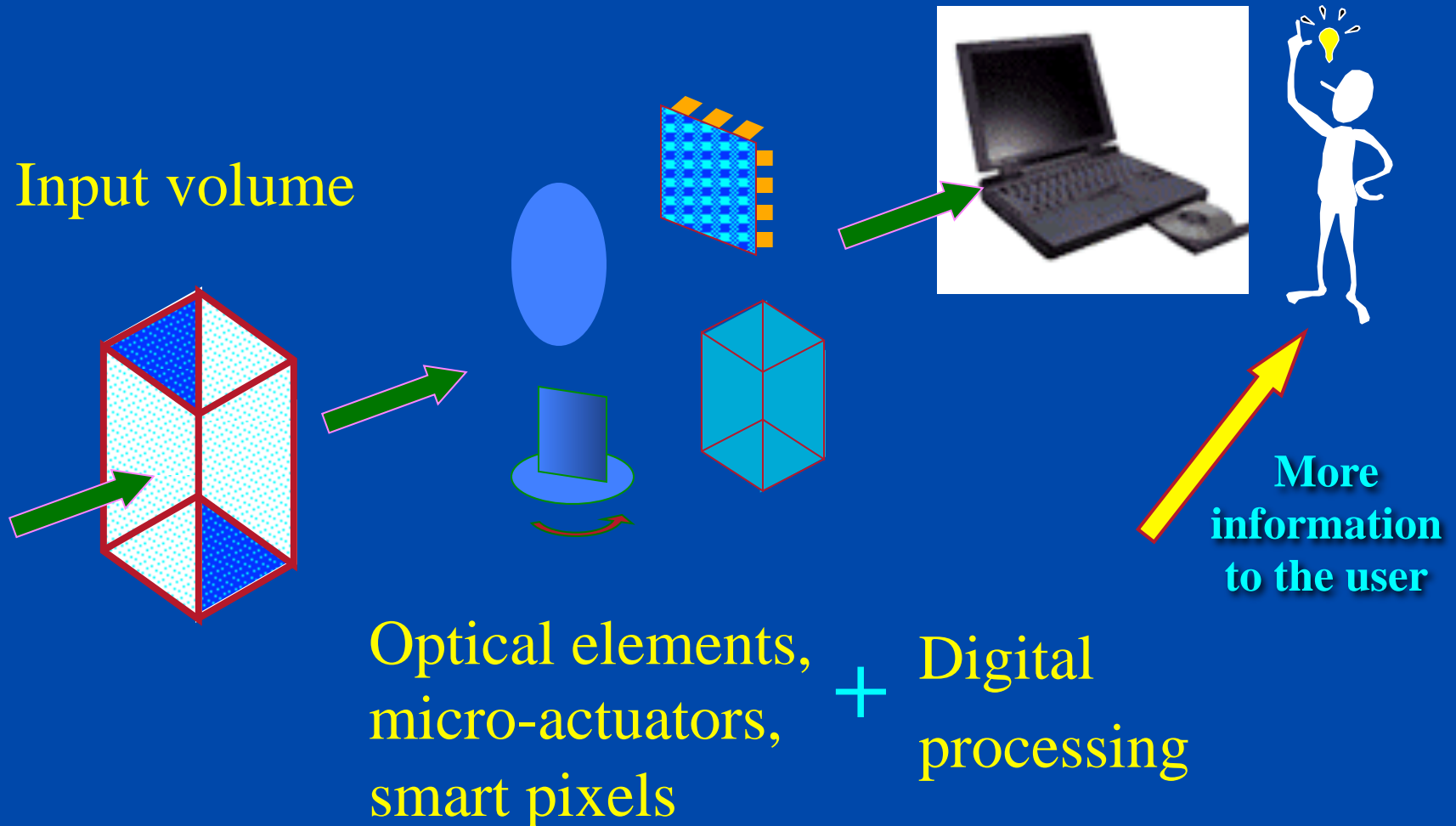
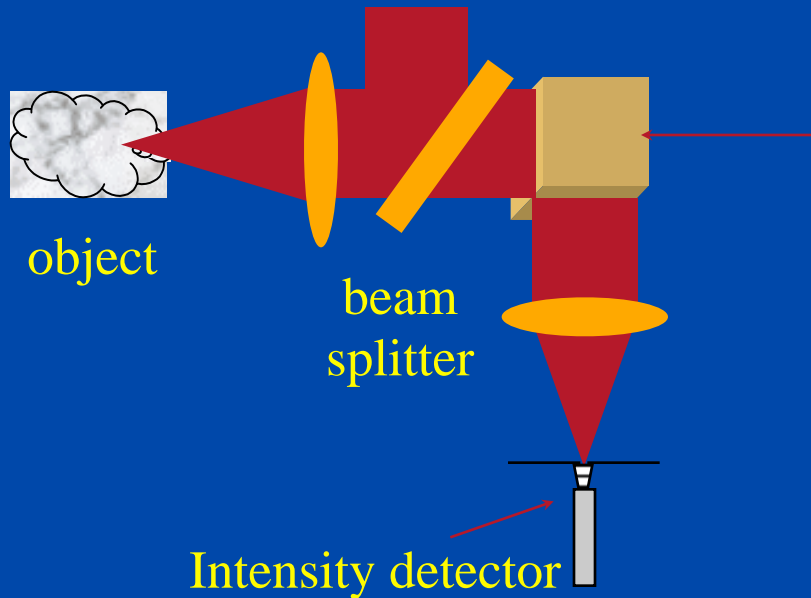


(3+)D Optical Engineering



in Dave Brady's group, circa ~98-99... co-conspirators: Bob Plemmons, Sudhakar Prasad, the late Dennis Healy ...

Confocal microscope with volume holographic filter



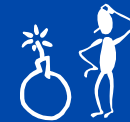
The volume hologram acts as a depth-selective filter through the *Bragg pinhole* effect.

Barbastathis, Balberg, and Brady *Opt. Lett.* **24** (12) 811-813, 1999.

Matched filtering is better suited to propagation properties of light



3D scanning is still required to acquire the entire object



Hologram does not diffract 100% of the light \Rightarrow potential photon collection deficiency

in Dave Brady's group, circa ~98-99... co-consipators: Bob Plemmons, Sudhakar Prasad, the late Dennis Healy ...



Compressive phase retrieval

George Barbastathis,^{1,2,3}

Yi Liu,² Wensheng Chen,³ Lei Tian,⁵ Jon Petrucci,⁶

Zhengyun Zhang,³ Shakil Rehman,³ Chen Zhi,³ Justin W. Lee,⁴ Adam Pan⁴

¹University of Michigan - Shanghai Jiao Tong University Joint Institute

中国；上海市；闵行区；上海交通大学密西根大学学院

Massachusetts Institute of Technology

²Department of Mechanical Engineering

³Singapore-MIT Alliance for Research and Technology (SMART) Centre

⁴Health Sciences and Technology Program

⁵University of California, Berkeley

⁶State University of New York, Albany

Today's talk is about

- Compressive measurements (sparsity priors)
- Coherent light
 - Digital holography and particle localization
- Partially coherent light
 - Phase space and mutual intensity retrieval

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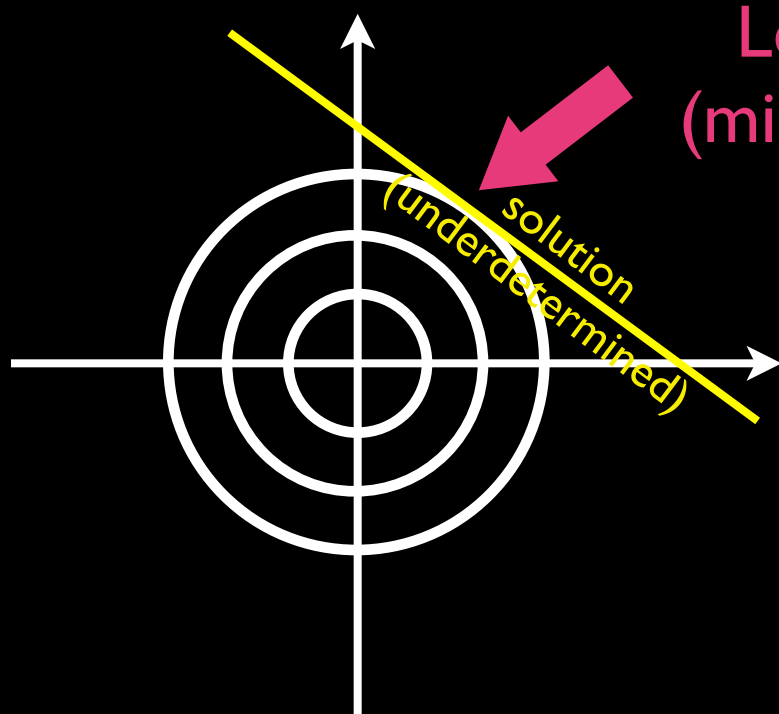


Compressive sensing: a simple-minded example



- You drive by a farm with chicken and sheep and count a total of 8 legs:
- how many chicken and sheep are there?
- **underdetermined** - *need another equation like “the total number of heads I count is...”*
- alternatively we can use a “sparsity prior” in the total # of types of animals.
- *either chicken or sheep*

L2



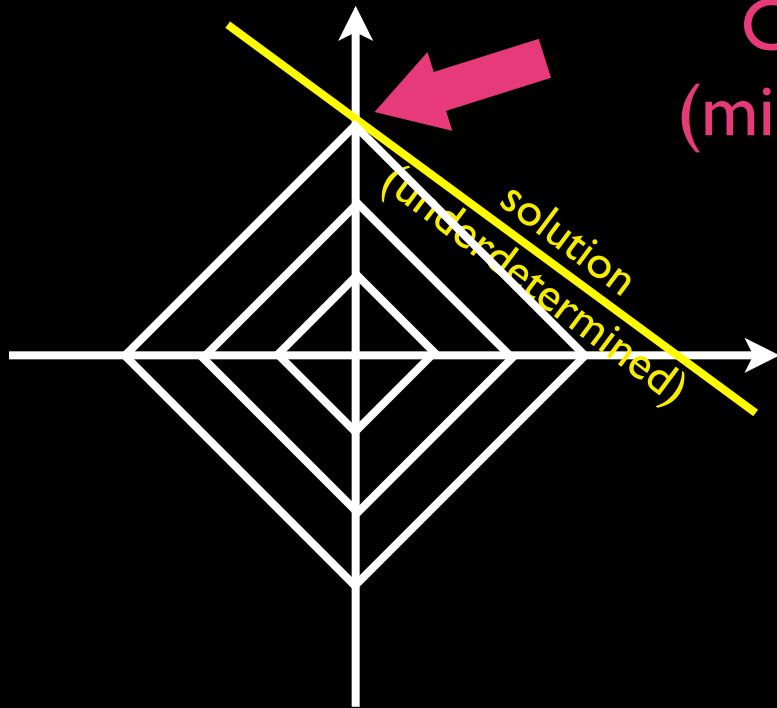
Least squares solution
(minimizes L2 on the line)

$$\begin{pmatrix} 2 & 4 \end{pmatrix} \begin{pmatrix} c \\ s \end{pmatrix} = \begin{pmatrix} 8 \end{pmatrix}$$

$$\text{s.t. } c^2 + s^2 = \min$$

NOT Sparse

LI



Compressive solution
(minimizes LI on the line)

$$\begin{pmatrix} 2 & 4 \end{pmatrix} \begin{pmatrix} c \\ s \end{pmatrix} = \begin{pmatrix} 8 \end{pmatrix}$$

$$\text{s.t. } |c| + |s| = \min$$

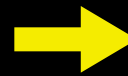
Sparse

Generally, of the form $(0, \dots, 0, \xi, 0, \dots, 0)$

Sparse spiky signals

$$f(t) = \sum_{\tau \in T} f(\tau) \delta(t - \tau)$$

$$\hat{f}(\omega) = \sum_{x=0}^{N-1} f(x) e^{i2\pi\omega x/N}$$

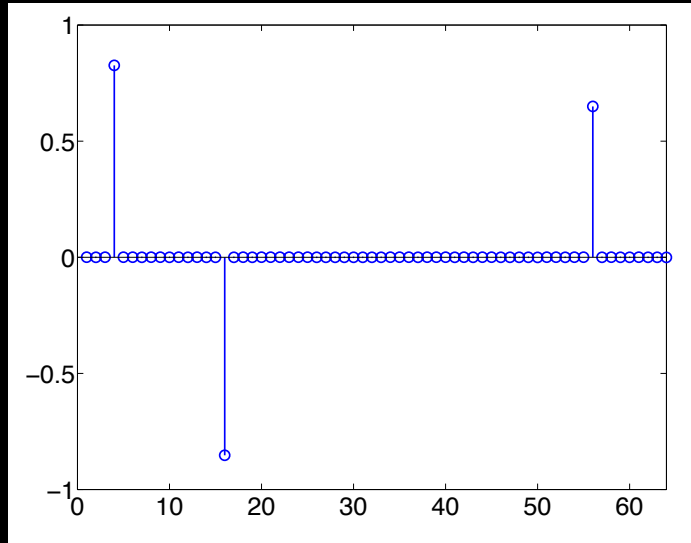


$$\hat{f}(\omega) = \sum_{\tau \in T} f(\tau) e^{-i2\pi\omega\tau}$$

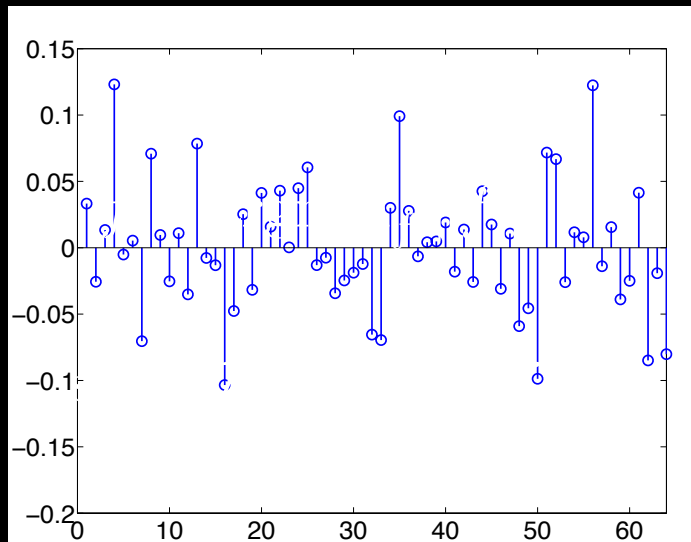
- ➔ Native space: spiky signal \Rightarrow Nyquist sampling necessary
- ➔ Fourier space: smooth signal (superposition of a few sinusoids only) \Rightarrow fewer than Nyquist samples perhaps suffice
- ➔ To make up for missing samples: L1 minimization

Compressive sensing example

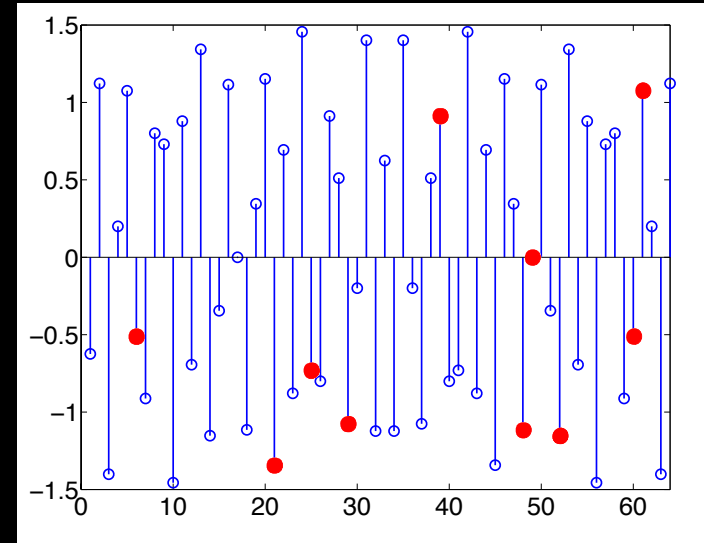
Original signal with 3 spikes (total length=64)



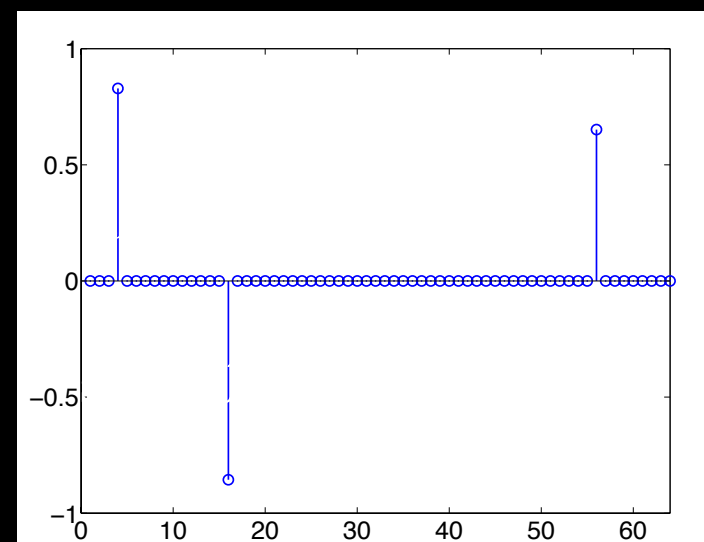
Conventional (L2) reconstruction



DFT measurements (# of samples=12)



Compressive (L1) reconstruction



Reconstruction success is subject to sparsity

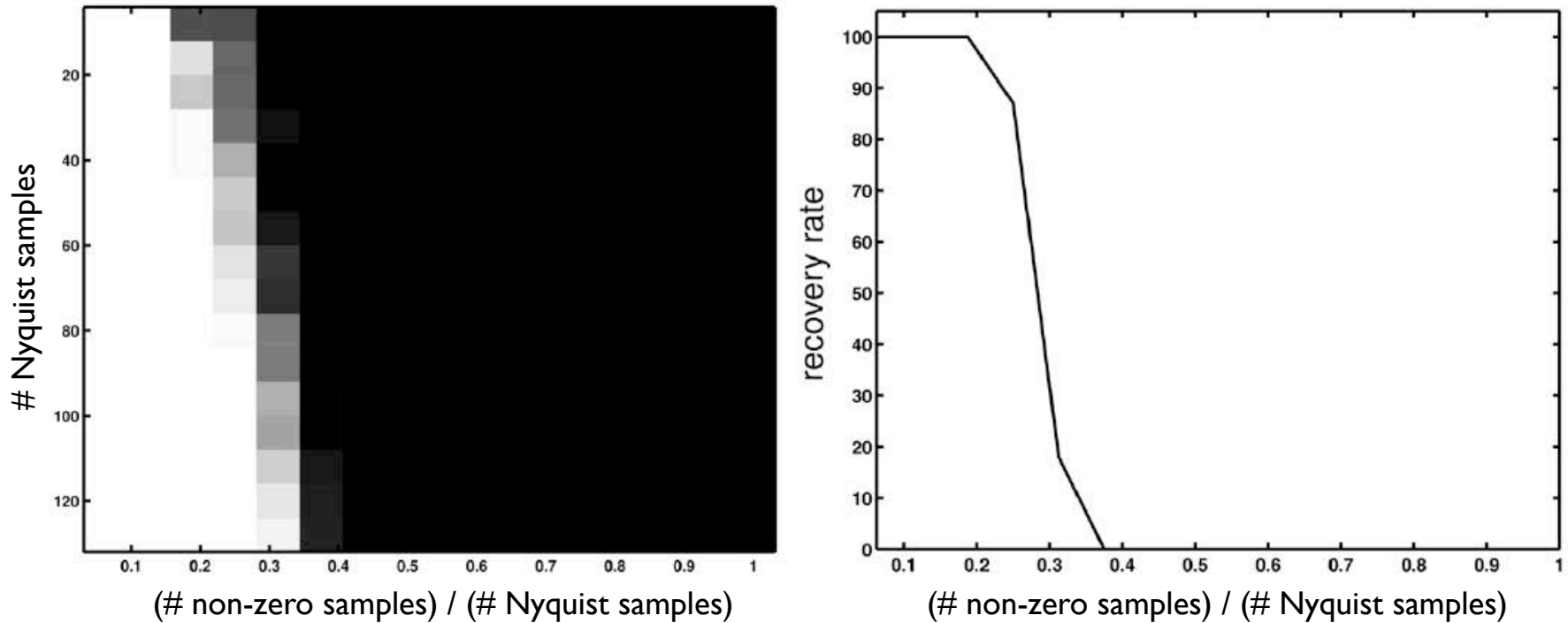


Fig. 2. Recovery experiment for $N = 512$. (a) The image intensity represents the percentage of the time solving (P_1) recovered the signal f exactly as a function of $|\Omega|$ (vertical axis) and $|T|/|\Omega|$ (horizontal axis); in white regions, the signal is recovered approximately 100% of the time, in black regions, the signal is never recovered. For each $|T|, |\Omega|$ pair, 100 experiments were run. (b) Cross section of the image in (a) at $|\Omega| = 64$. We can see that we have perfect recovery with very high probability for $|T| \leq 16$.

The premise of compressive sensing

- Nyquist criterion is too restrictive because it takes no priors into account
- Most signals are **sparse** if expressed in an appropriate basis, e.g.
 - sparse in time - “spiky”
 - sparse in frequency - “beaty”
- Far **fewer** samples than Nyquist may suffice to completely reconstruct, provided
 - the appropriate basis has been selected
 - sufficient **signal mixing** by the measurement operator
 - “measurement must be **incoherent**”
- Sparsity can then be enforced as a prior (regularizer) by \mathcal{L}_1 norm minimization

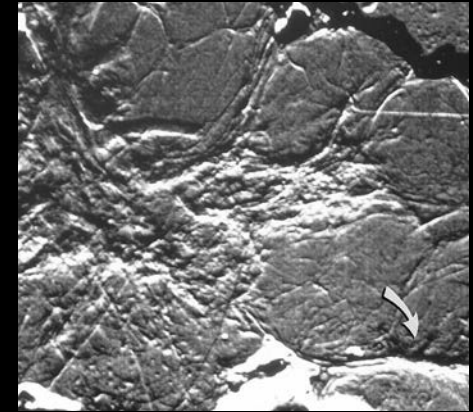
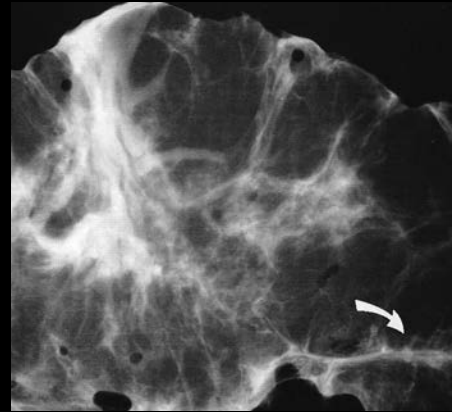
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The significance of phase

Visible

X-ray



intensity image

phase-contrast image

attenuation image

phase-contrast image

(F. Zernike, Science 121, 1955)

(human breast cancer specimen)

(E. D. Pisano et al., Radiology 214, 2000)

$$\phi(x_o) = k \int_{\Gamma} n(\mathbf{r}) dl \Rightarrow \rho \propto \frac{n^2 - 1}{n^2 + 2}$$

Refractive index

Density

$$\Rightarrow \rho \propto \begin{matrix} \text{temperature} \\ \text{pressure} \\ \text{humidity} \end{matrix}$$

Phase is irrelevant!

Optical Path Length (OPL)

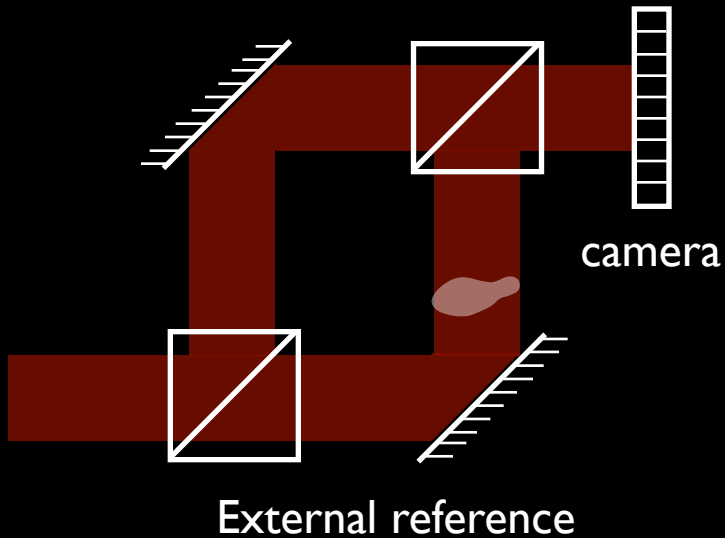
is relevant

(and works with partially coherent light)

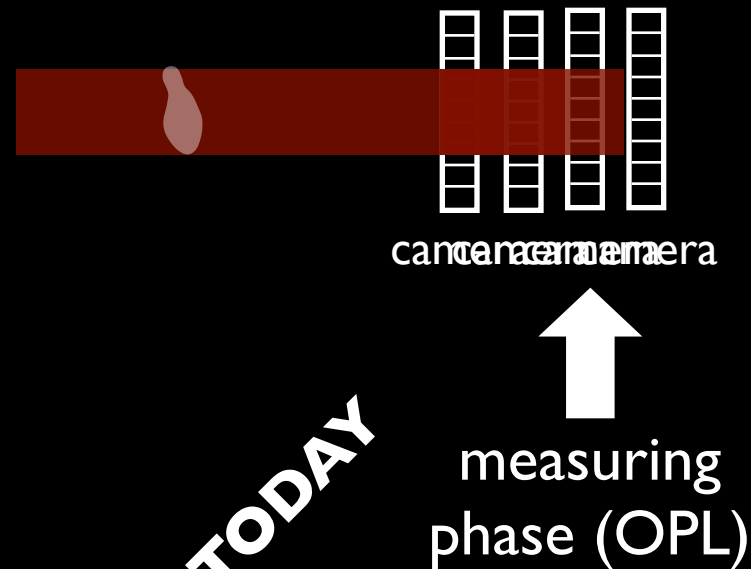
J. C. Petruccelli, et al, *Opt. Express*
21:14430, 2013

Phase Imaging

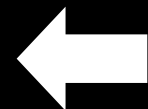
- Interferometric



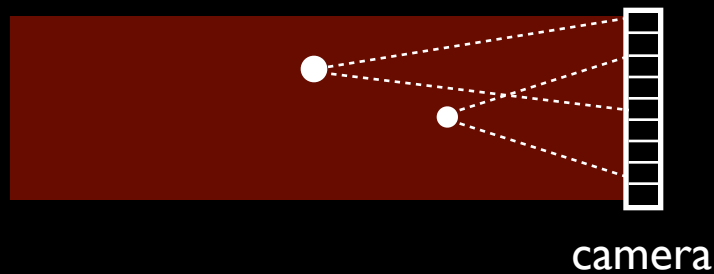
- Axial stack



TODAY

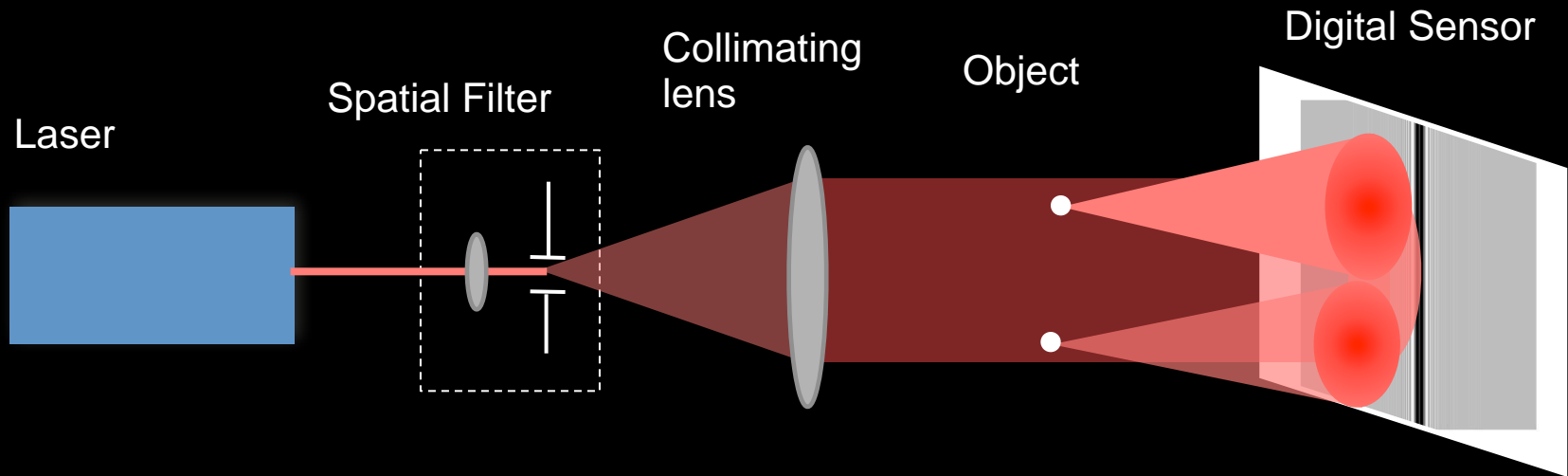


object/particle
localization



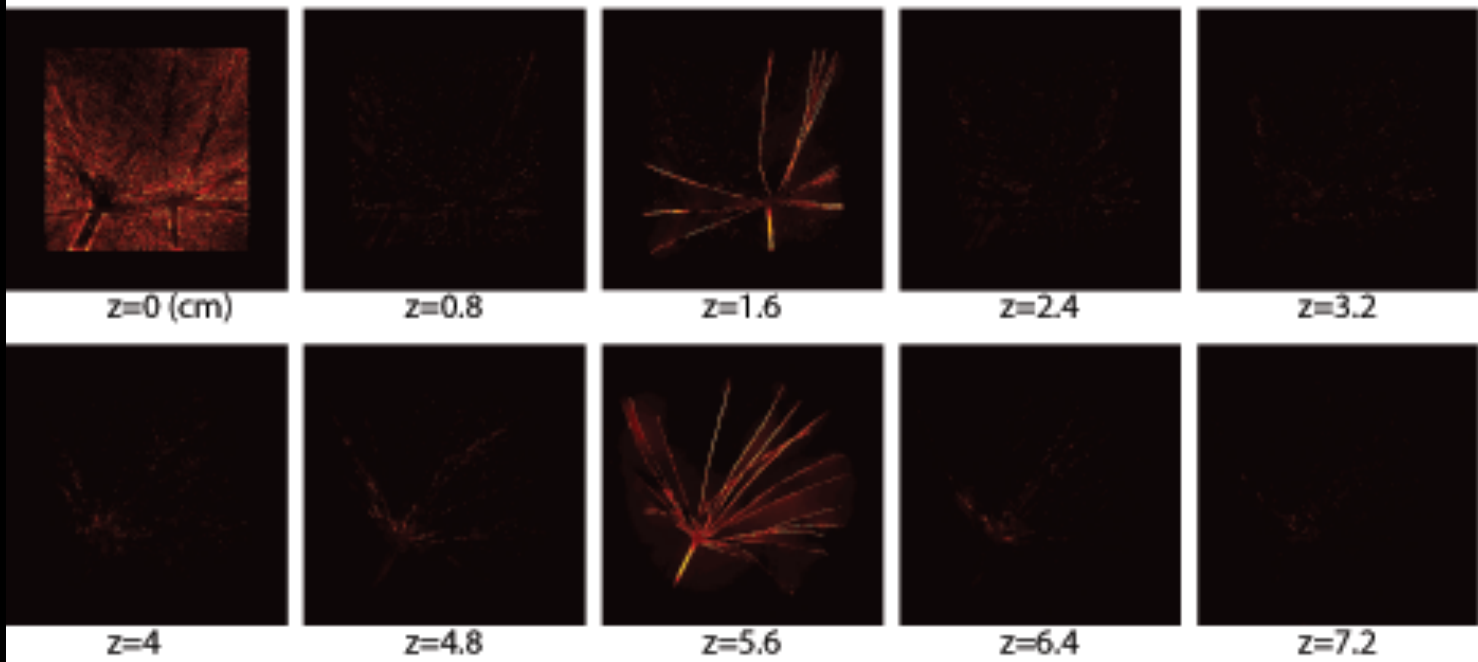
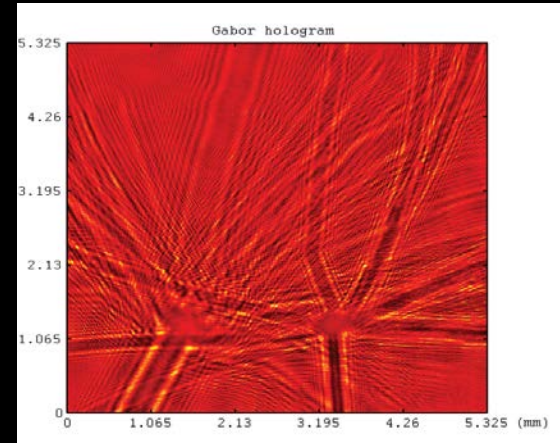
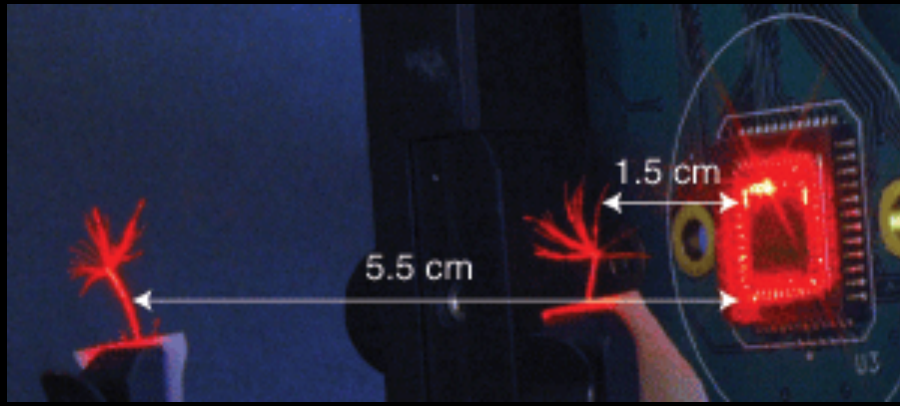
Self-referenced (in-line digital holography)

Digital Holography: Measurement is “incoherent” !



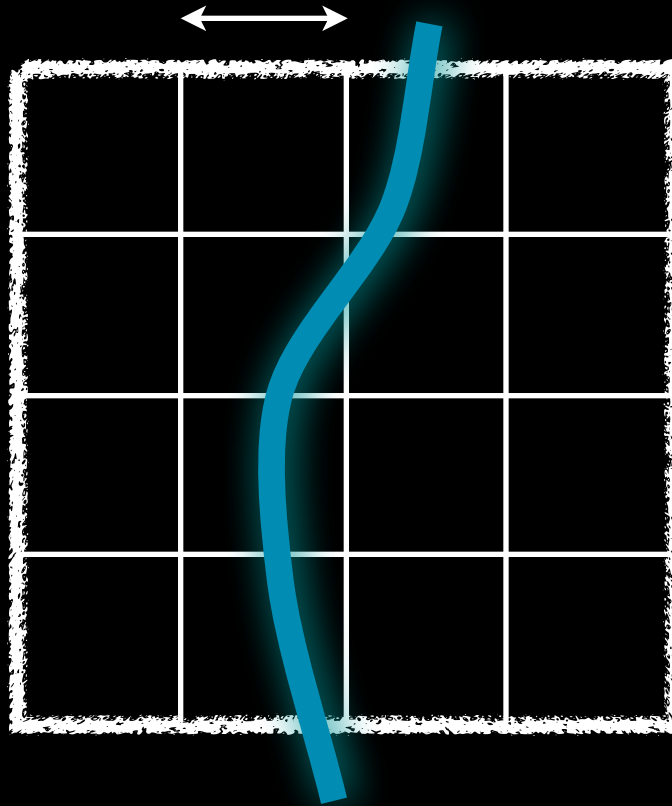
- ➔ According to Statistical Optics, a digital hologram is formed by interfering spatially and temporally *coherent* beams (e.g. originating from a HeNe laser)
- ➔ According to Compressive Sensing theory, this measurement is “*incoherent*” because light scattered from the object spreads out over several pixels

Compressive Holography



Compressive localization

Desirable accuracy: < 1 pixel

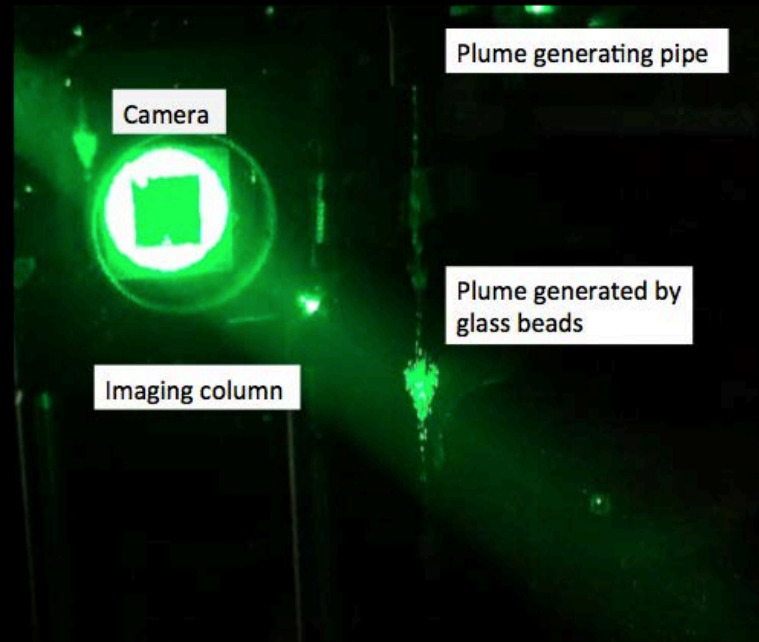


Prior: sparsity of object(s) within the field of view

Localization examples

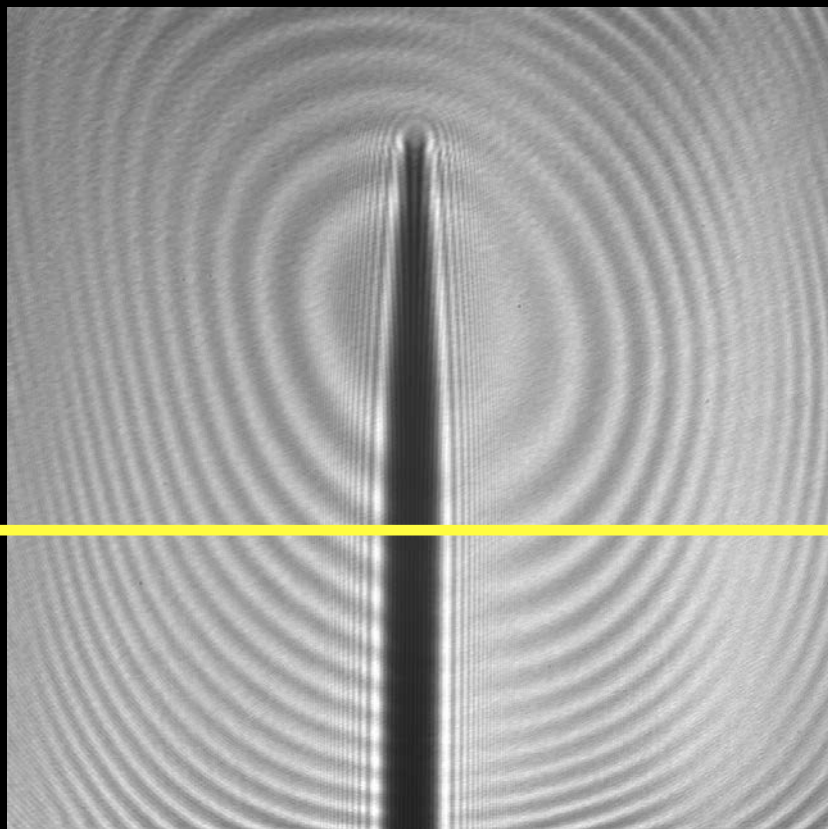


Quantitative measurement
of seal whisker motion

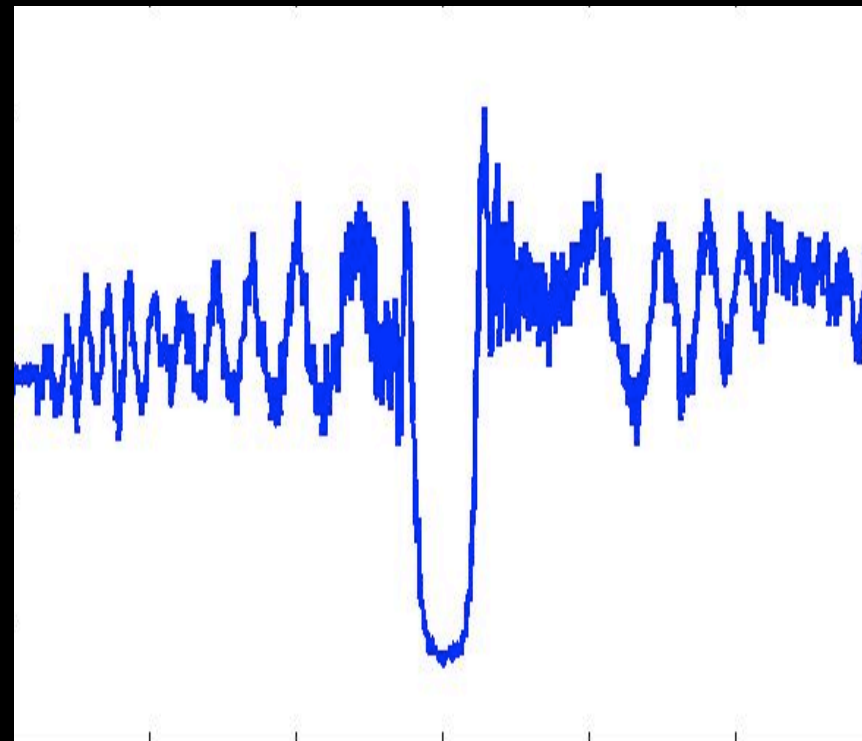


Quantitative analysis
of bubbles and plumes
(multi-phase flows)

Emulating a 1D whisker: pin object



Digital hologram across 1 row



Algorithm diagram

Object

Edge extraction

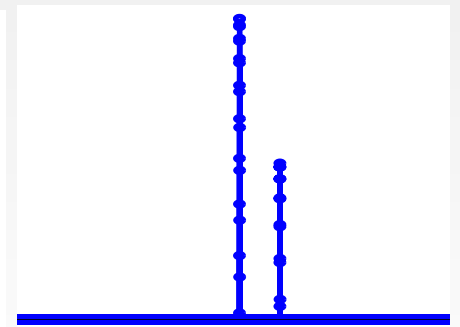
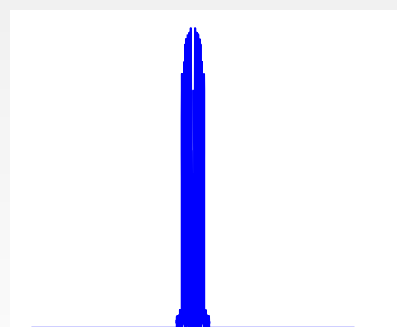
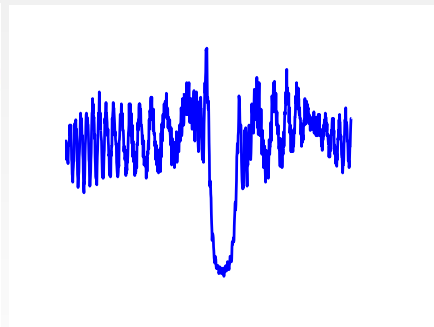
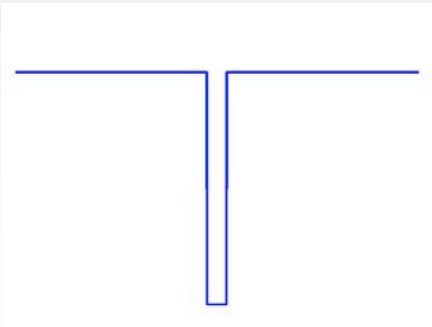
Free-space
Propagation

Compressive
reconstruction
(TwIST)

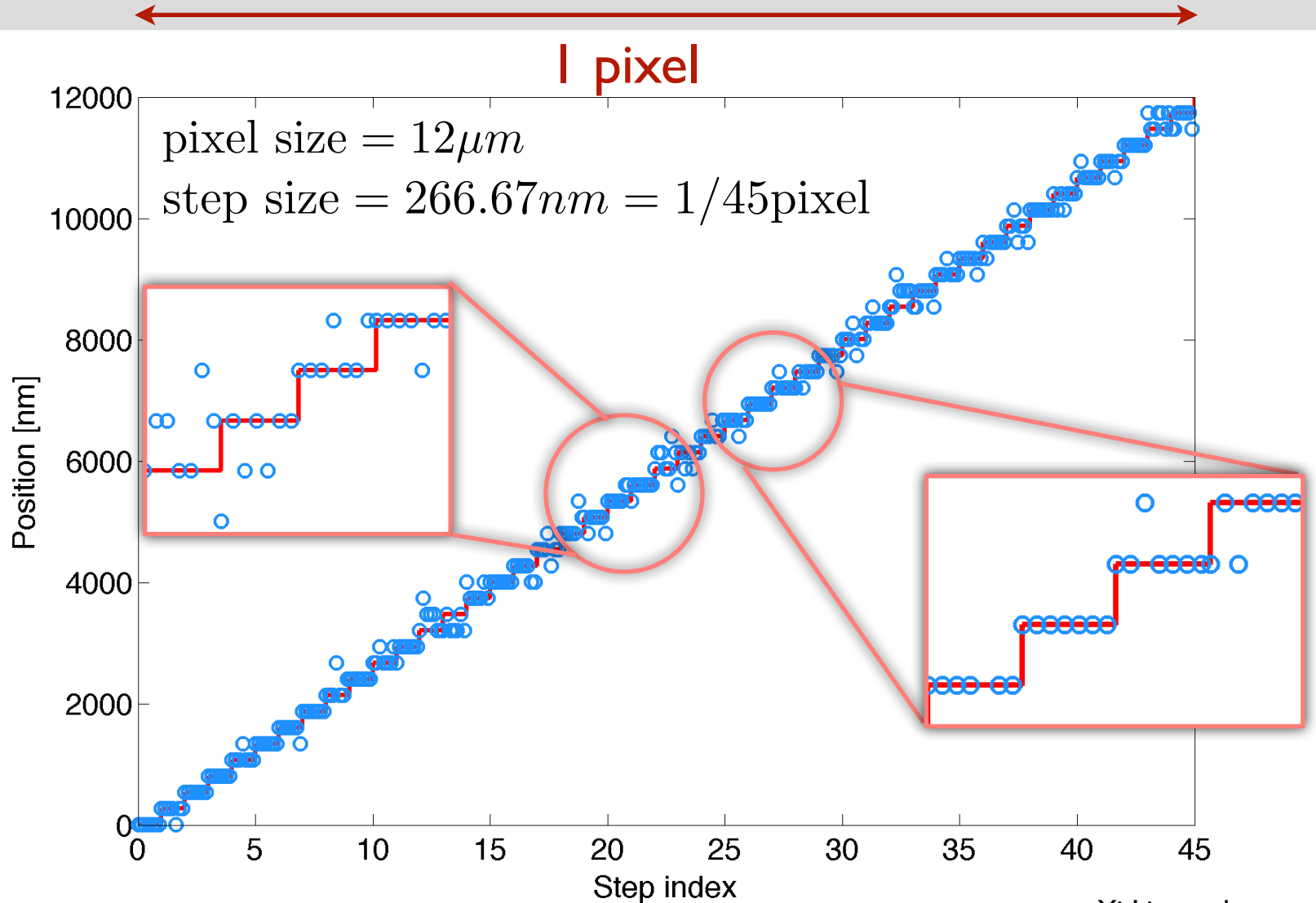
hologram

$F(I') = iu \cdot F(I)$
zero-padding *interpolation*

Interpolated
edge's spectrum



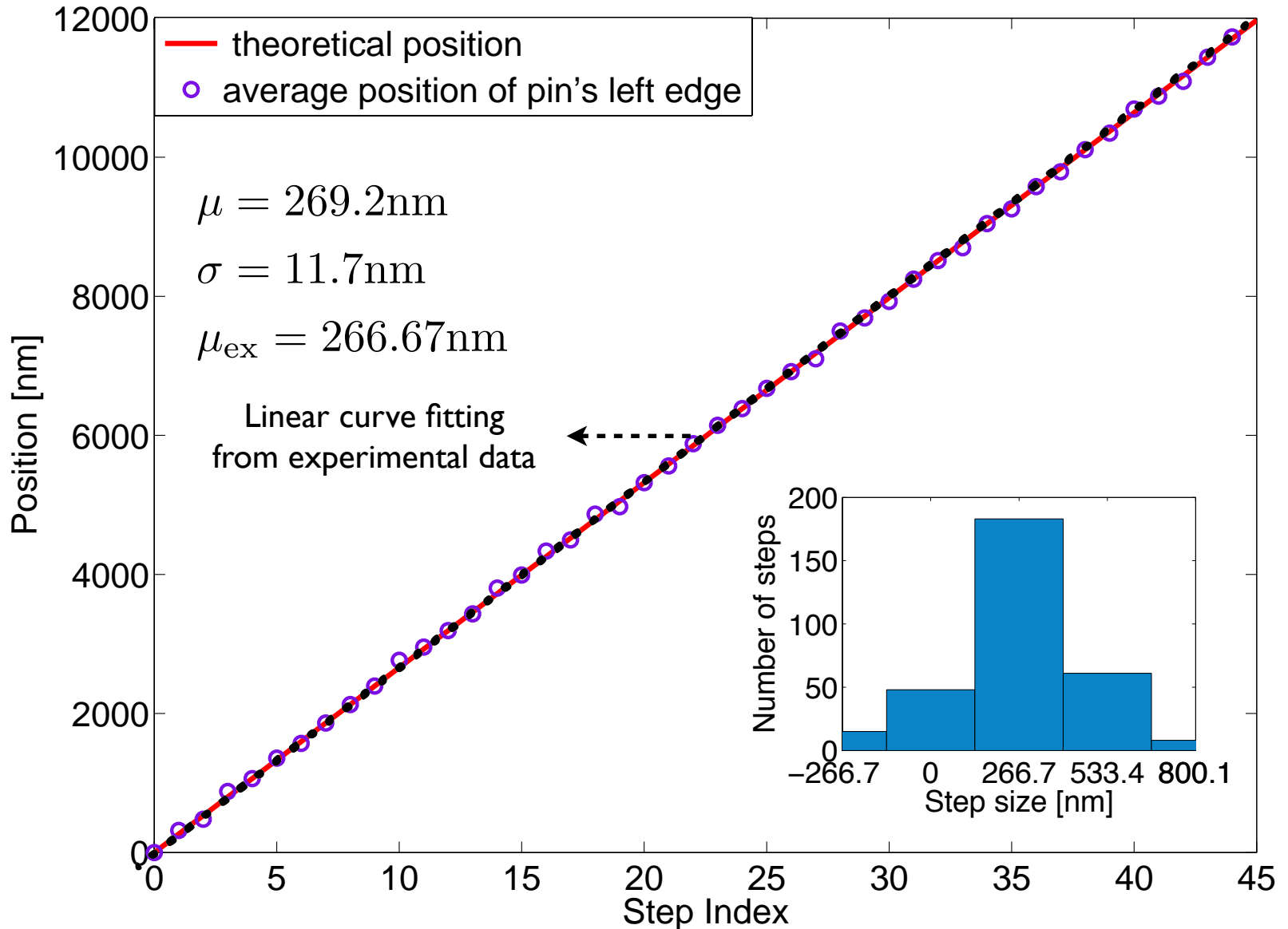
Experimental result



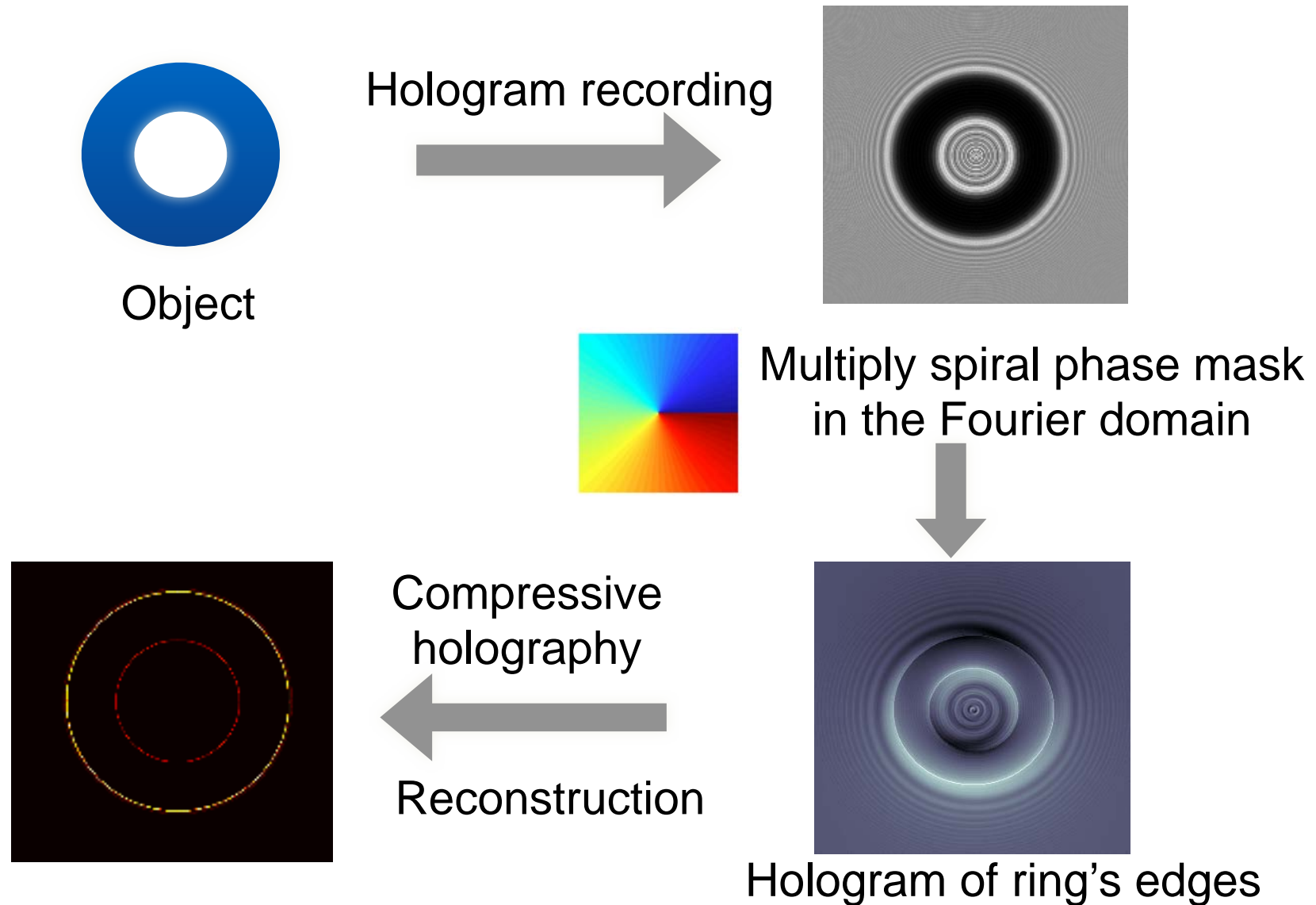
○ : position of the left point of one row on the pin at each step
Each step, tracing 7 rows.

Yi Liu et al, to appear in
Optics Letters issue
August 15, 2012

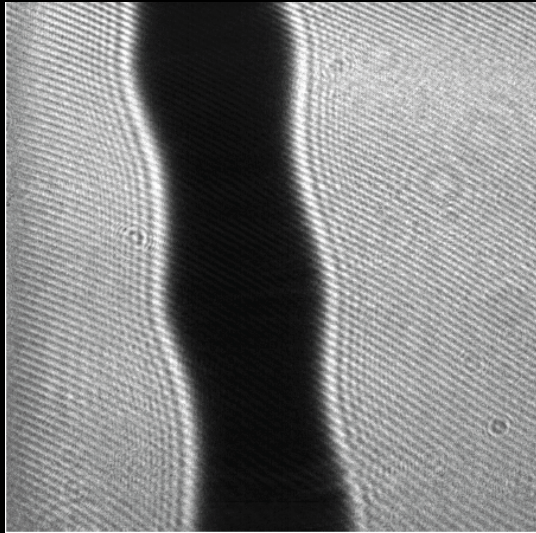
Experimental result



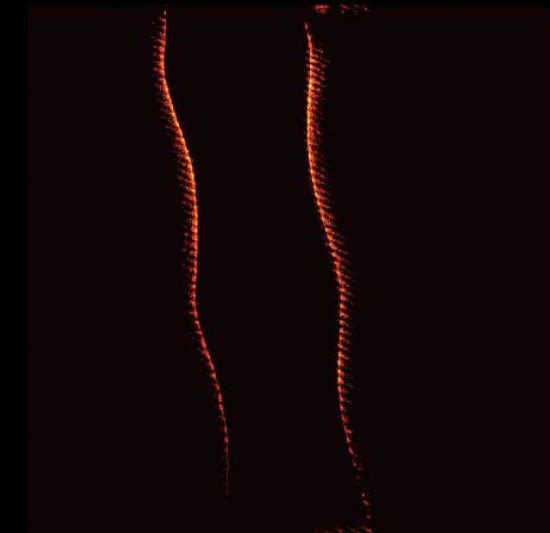
2D object localization



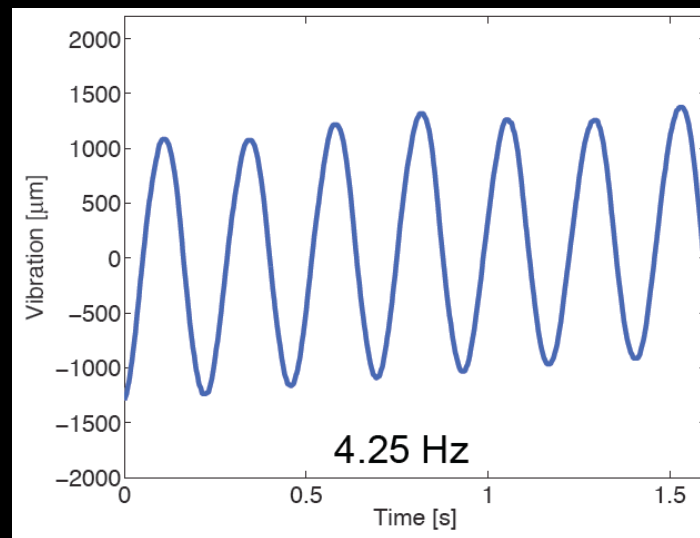
Whisker vibration experiments



Whisker hologram



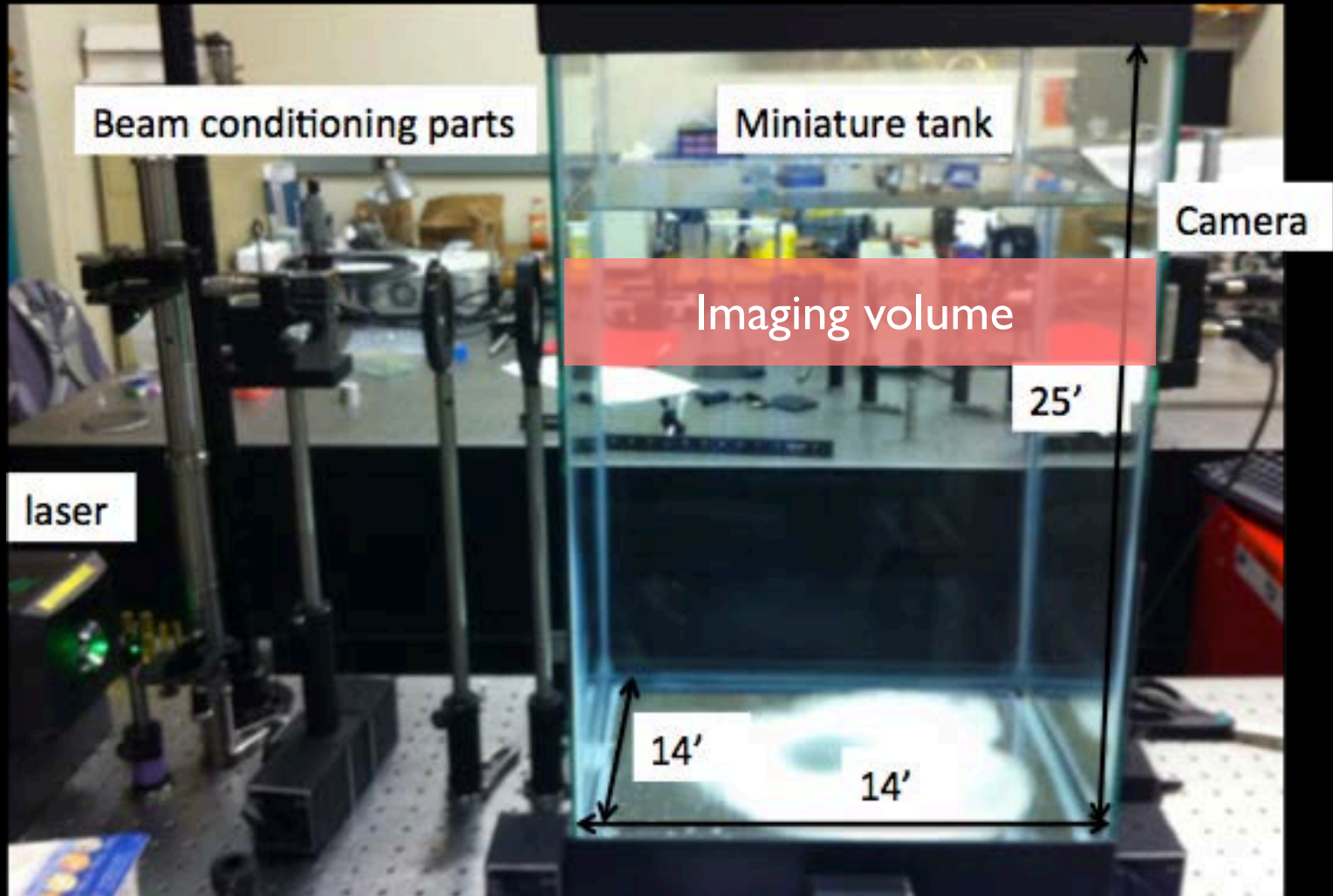
Whisker's edges extracted



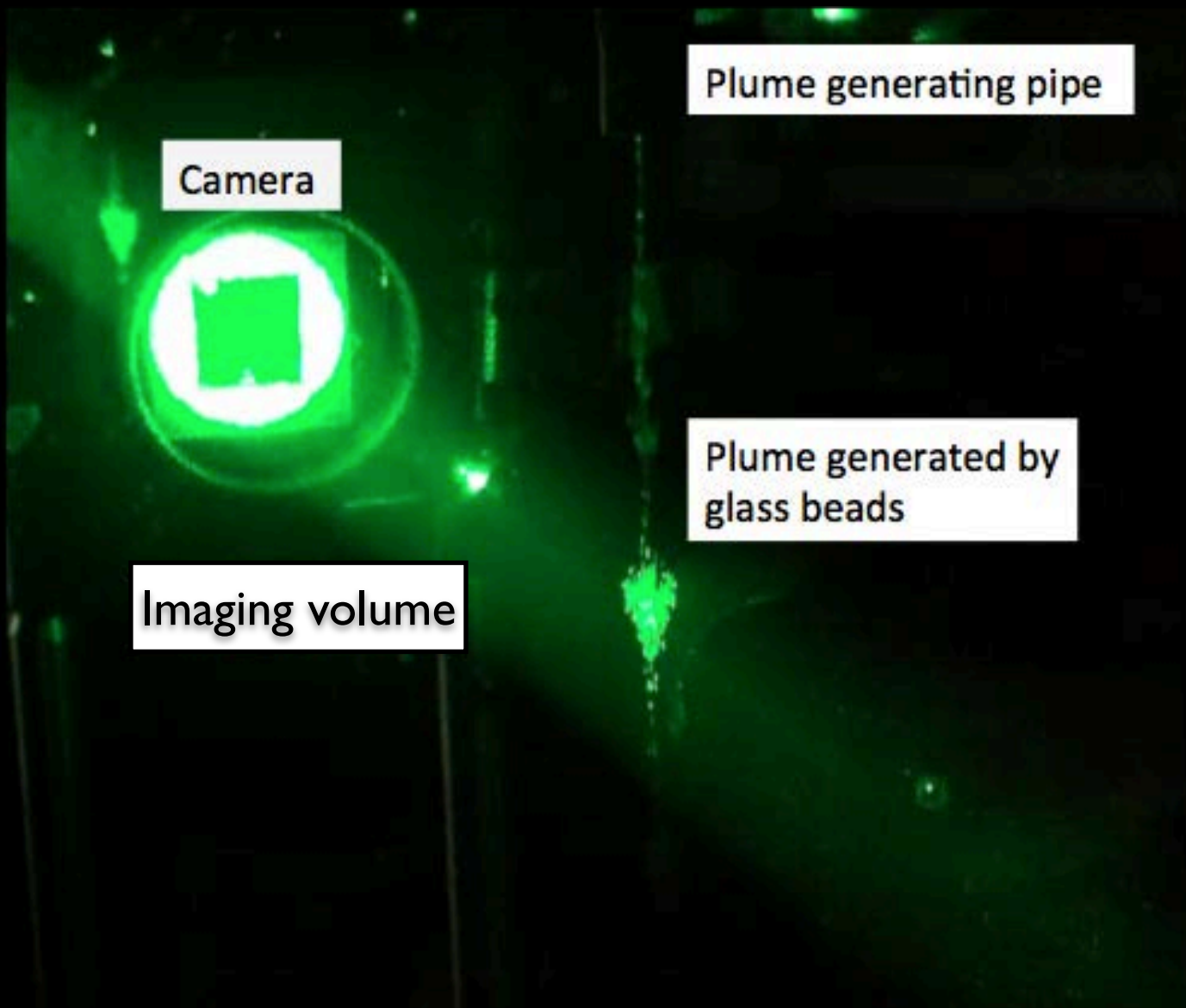
Whisker's motion reconstructed
(pixel size = $10\mu\text{m}$)

Acknowledgment
Heather Beem,
Michael Triantafyllou
MIT

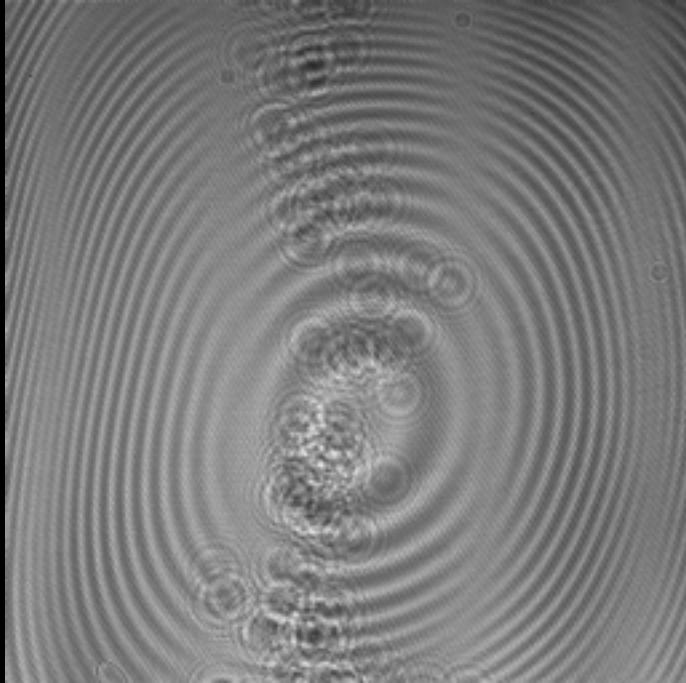
DH particle localization



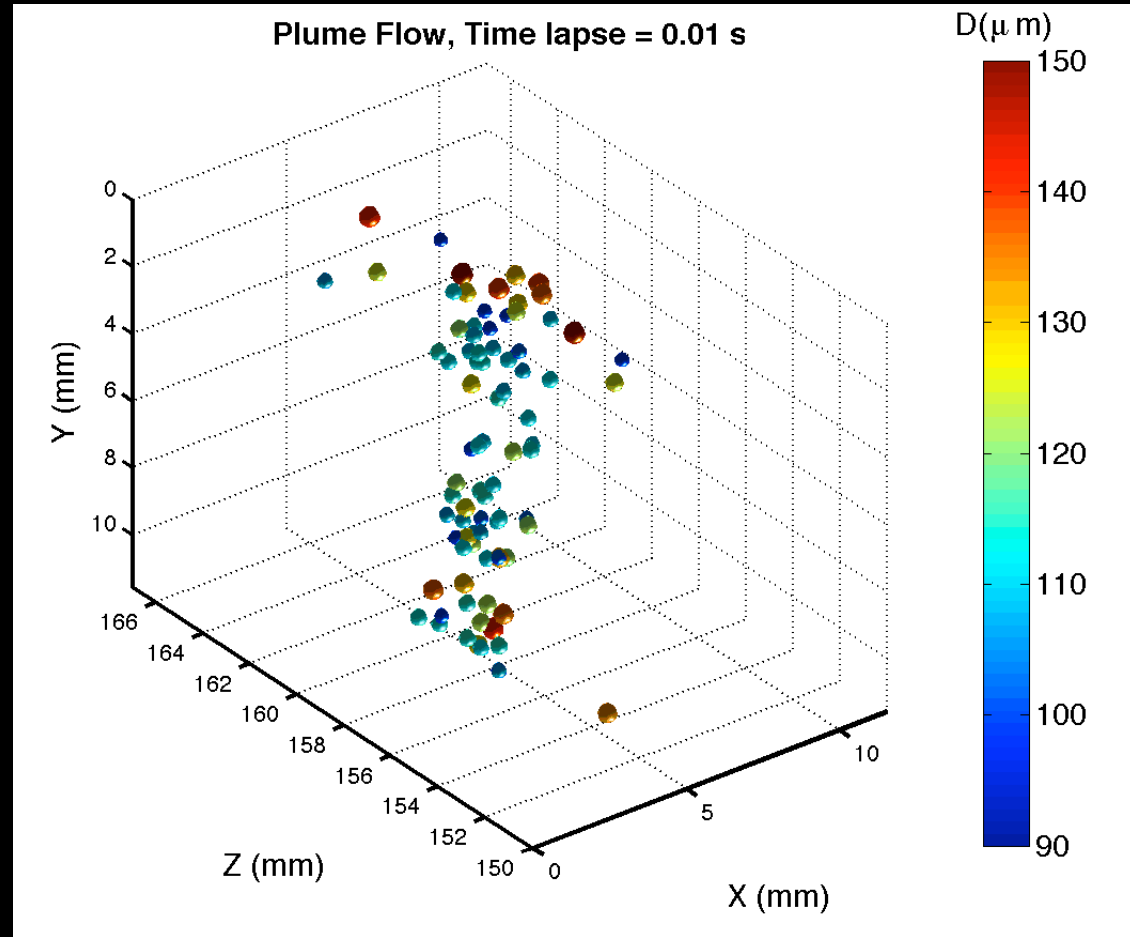
DH particle localization



3D reconstruction of a plume (standard back-propagation)



Original hologram

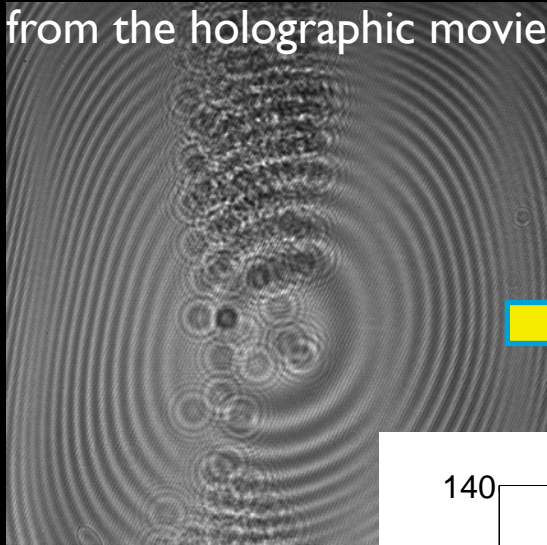


3D Reconstruction

Particle size not to scale

Size distribution analysis

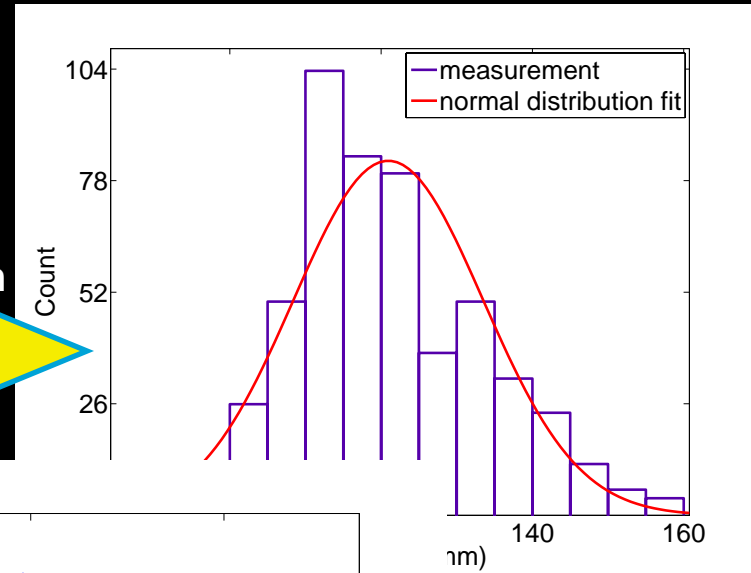
a single frame
from the holographic movie



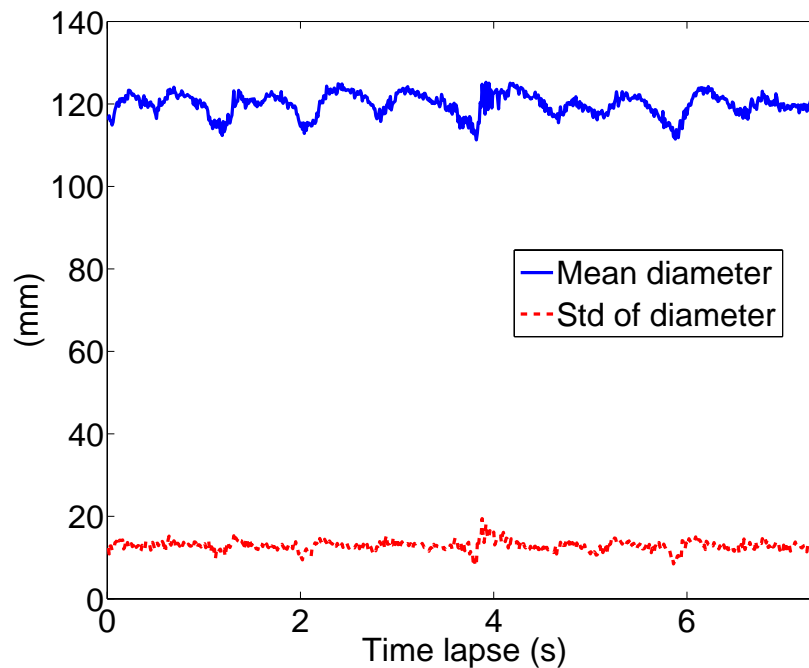
Reconstruction



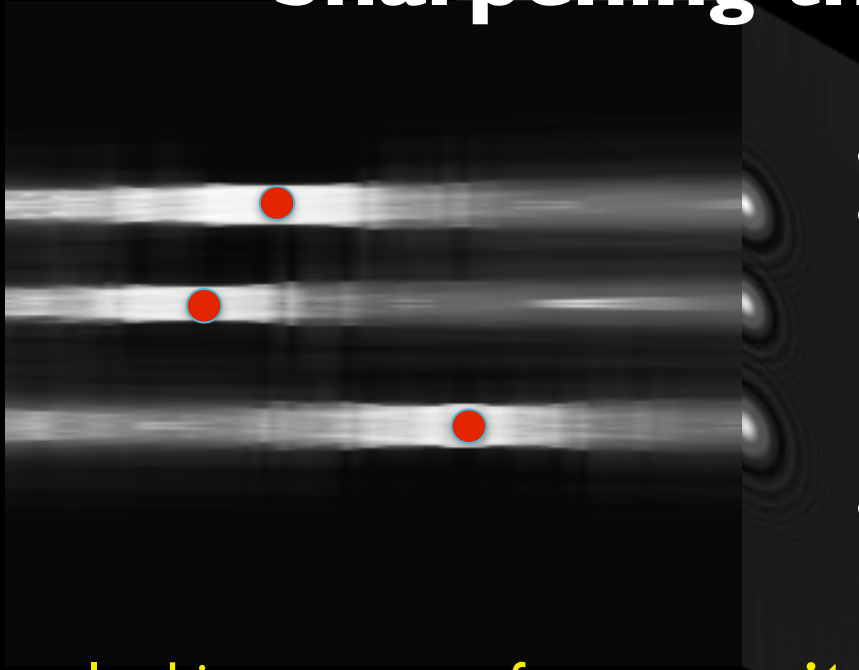
size distribution



repeat for
each frame



Sharpening the axial accuracy



In a given z plane:

- Sharp features mostly due to in-focus particles
- Smooth features due to:
 - Defocused particles
 - Twin image
 - Halo
 - Noise
- This essentially states a sparsity prior on the edge sharpness

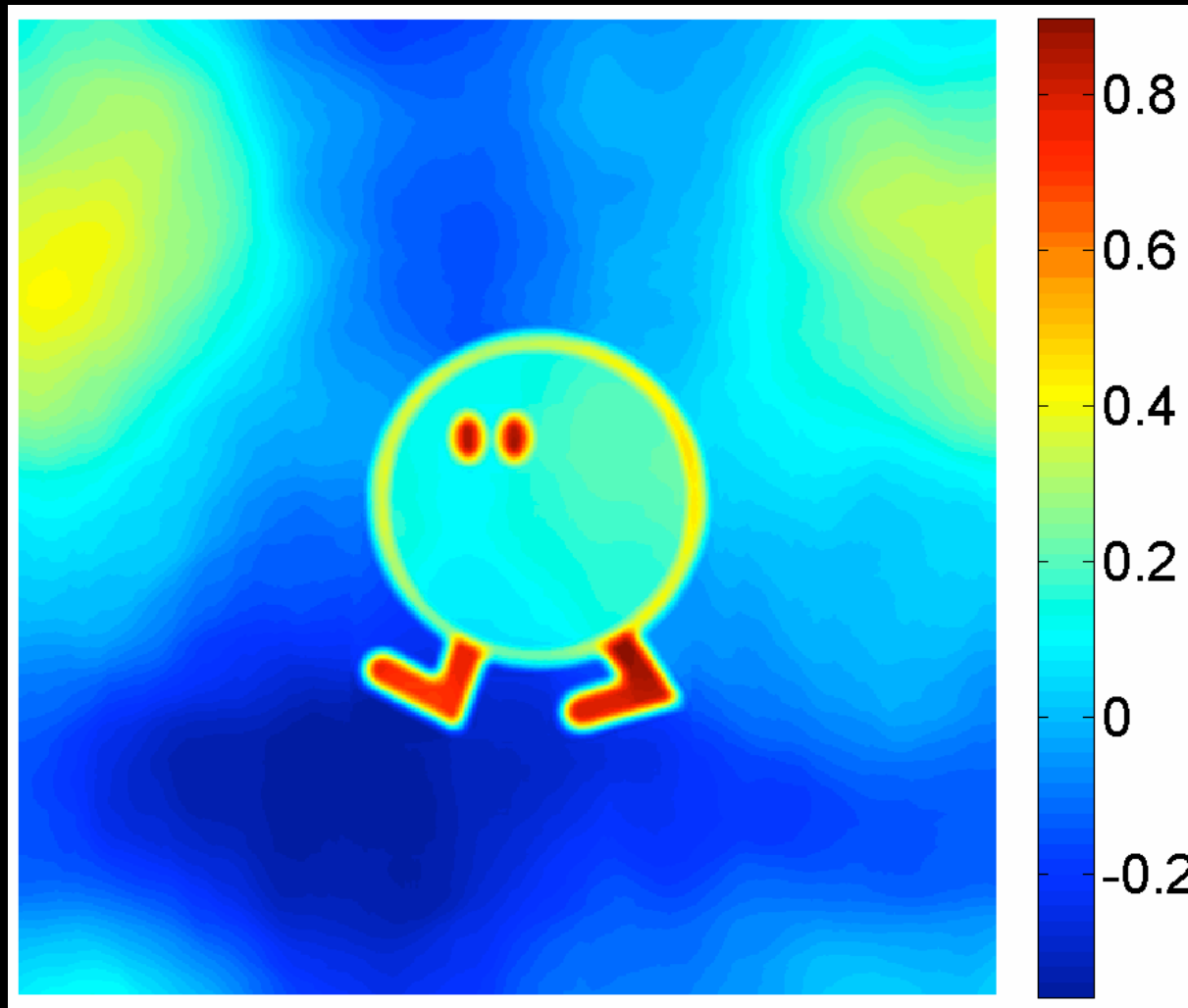
In this case we enforce **sparsity** by evolving the unknown radiance f to the steady state of a nonlinear diffusion equation

$$\frac{\partial f(\mathbf{x}, z; \tau)}{\partial \tau} = \alpha \nabla \cdot \left(F(|\nabla f|) \frac{\nabla f}{|\nabla f|} \right)$$

F : flux function

(notice $F=I \Rightarrow$ linear diffusion)

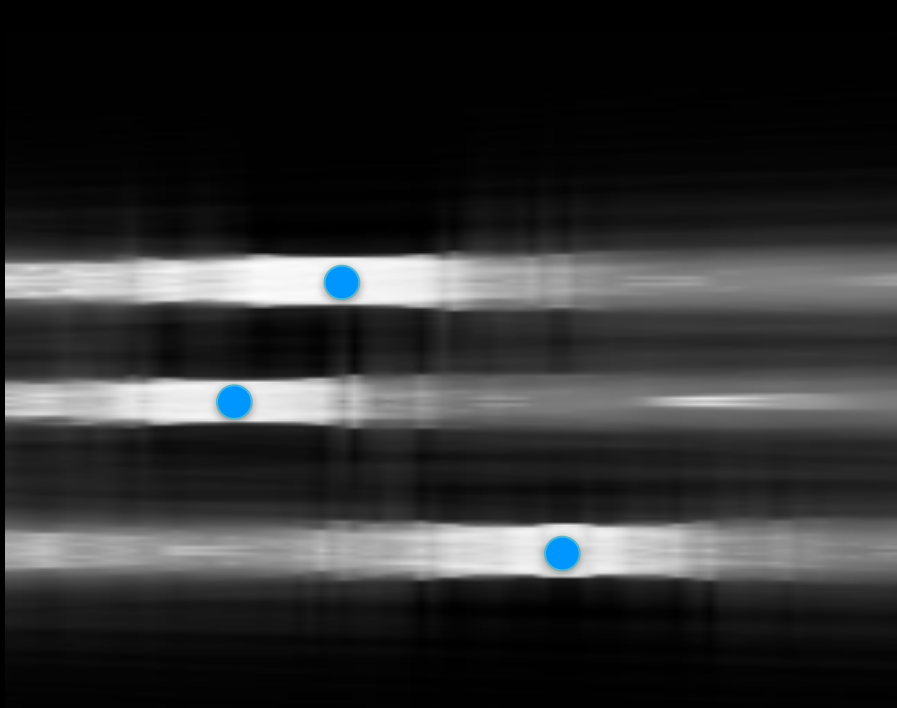
Nonlinear diffusion: animation



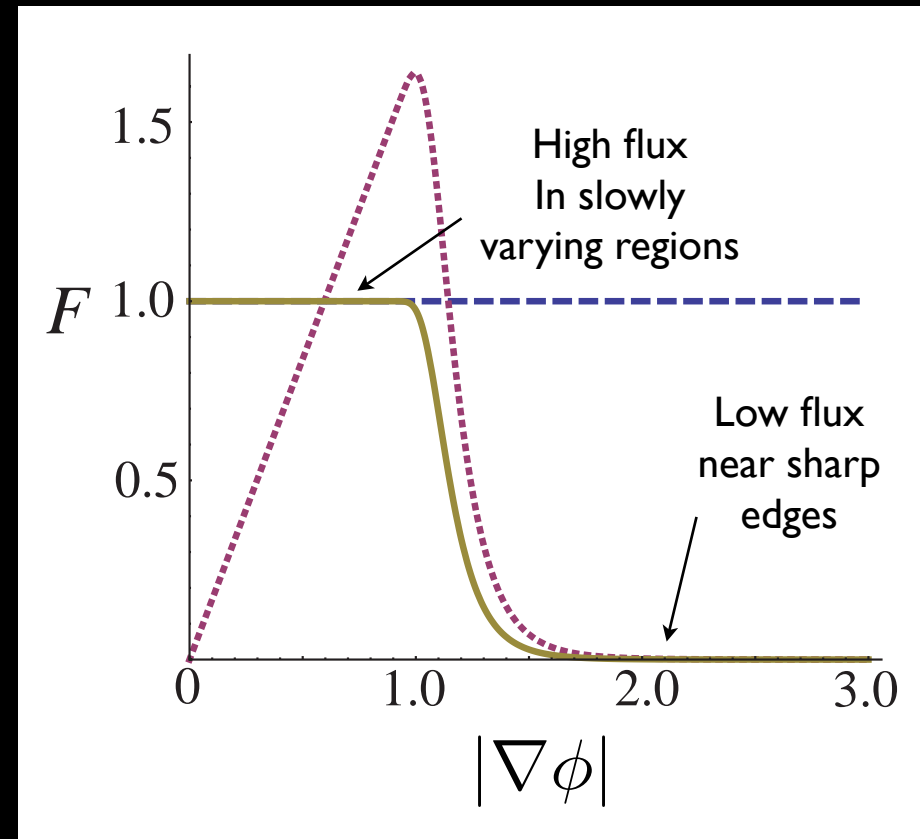
(movie shows output at every iteration)

Sharpening the axial accuracy by nonlinear diffusion

NLD in the *transverse* direction



Flux



Today's talk is about

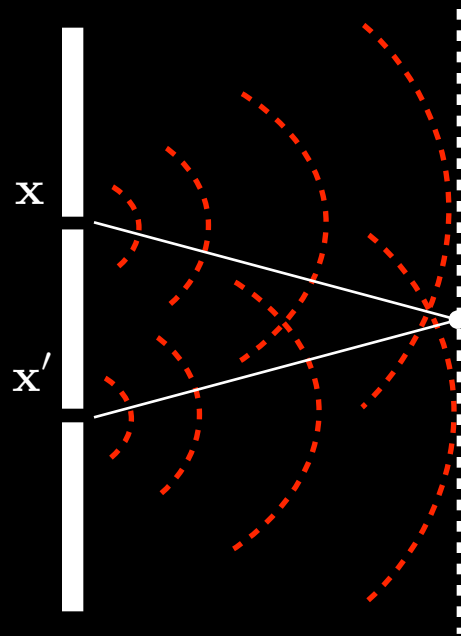
- Compressive measurements (sparsity priors)
- Coherent light
 - Digital holography and particle localization
- Partially coherent light
 - Phase space and mutual intensity retrieval

Partially coherent light

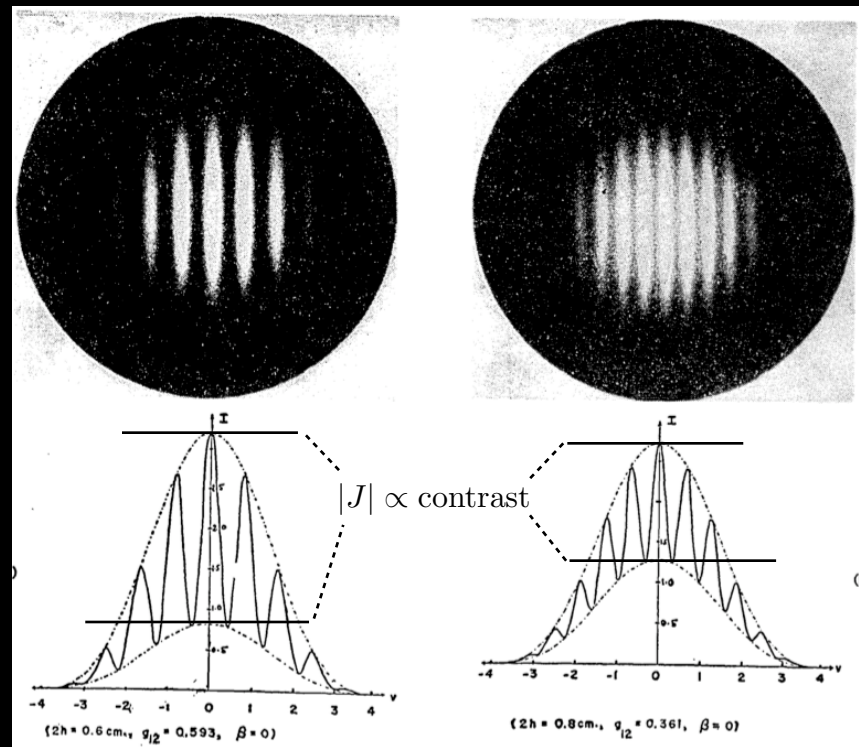
Random field $U(\mathbf{x})$

Correlation function (mutual intensity)

$$J(\mathbf{x}, \mathbf{x}') \equiv \langle U(\mathbf{x})U^*(\mathbf{x}') \rangle$$



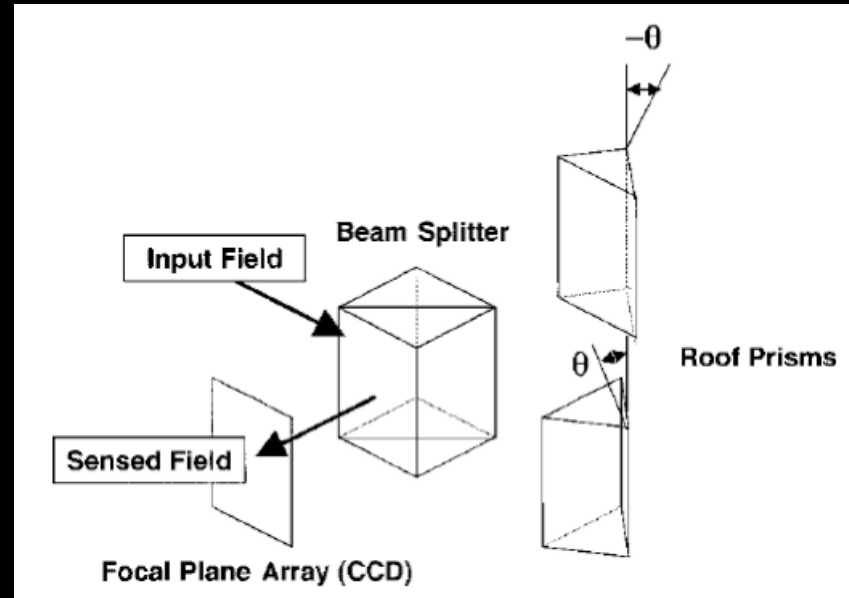
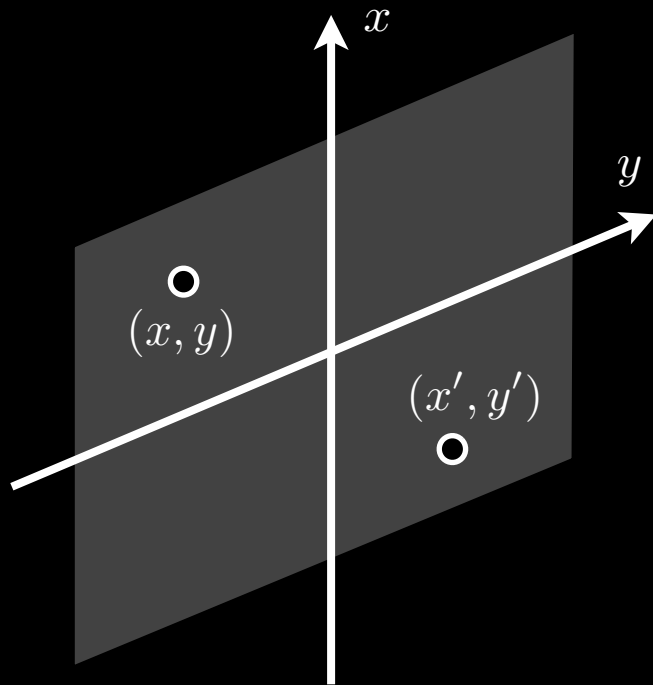
Young's two-slit experiment



B. J. Thompson and E. Wolf, *J. Opt. Soc. Am.*, 47:895, 1957.

The mutual intensity

$$J(\mathbf{x}, \mathbf{x}') \equiv \langle U(\mathbf{x})U^*(\mathbf{x}') \rangle$$



The mutual intensity

$$J(\mathbf{x}, \mathbf{x}') \equiv \langle U(\mathbf{x})U^*(\mathbf{x}') \rangle$$

- completely characterizes the (quasi-monochromatic) partially coherent field,
- in particular, the Optical Path Length (OPL);
 - J. C. Petruccelli, L. Tian, and G. Barbastathis, *Opt. Express* 21:14430, 2013
- is analogous to the density matrix in quantum mechanics;
- is semi-positive definite (eigenvalues ≥ 0);
- can be decomposed into **coherent modes**

$$J(\mathbf{x}, \mathbf{x}') = \sum_j c_j \phi_j(\mathbf{x}) \phi_j^*(\mathbf{x}')$$

- often, a case can be made that only few $c_j \neq 0$

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The Phase Space

- Wigner distribution function

$$W(x, u) = \int \psi \left(x + \frac{x'}{2} \right) \psi^* \left(x - \frac{x'}{2} \right) \exp(-i2\pi ux') dx'$$

$\langle \cdot \rangle$ \langle \dots \rangle

$$W(x, u) = \int J \left(x + \frac{x'}{2}, x - \frac{x'}{2} \right) \exp(-i2\pi ux') dx'$$

- Ambiguity function

$$A(u', x') = \int J \left(x + \frac{x'}{2}, x - \frac{x'}{2} \right) \exp(-i2\pi u'x) dx$$



\mathcal{F}
 $x \leftrightarrow u'$
 $u \leftrightarrow x'$

- By the way, $W(x, u)$ is real.

Phase space (Wigner space)

Temporal frequency

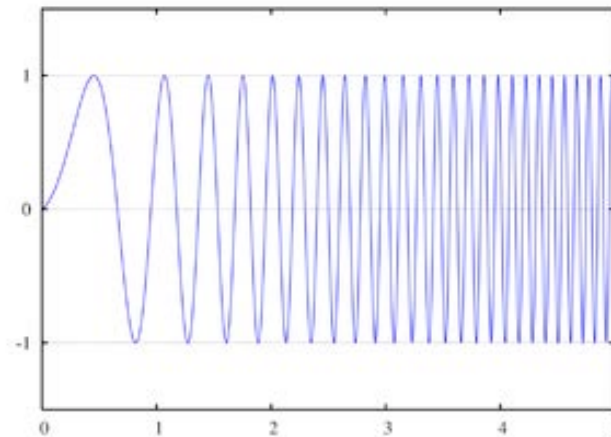


- Instantaneous frequency
- Local spatial frequency

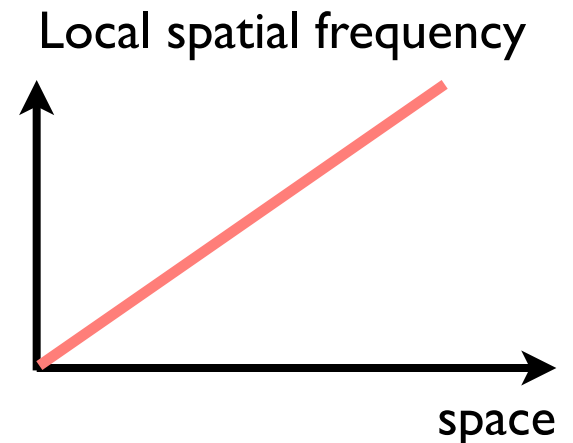
Spherical wave



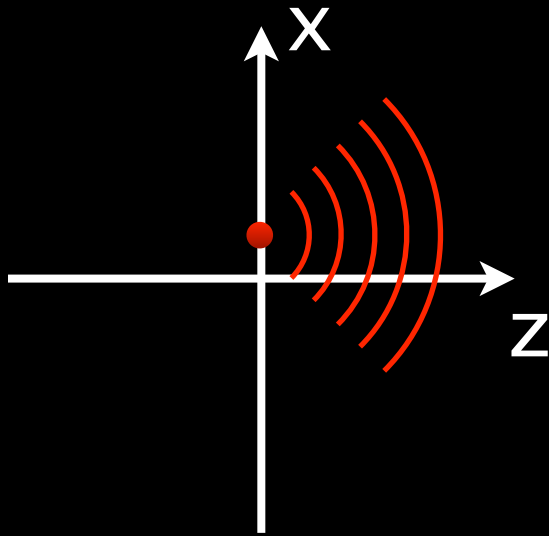
Chirp function



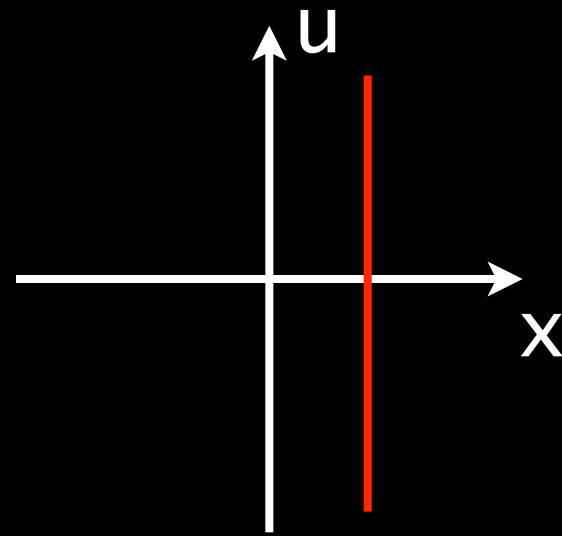
Phase space picture



point source

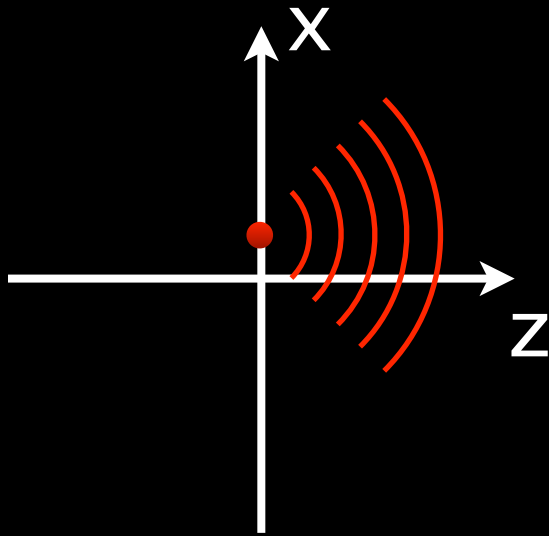


x-z space

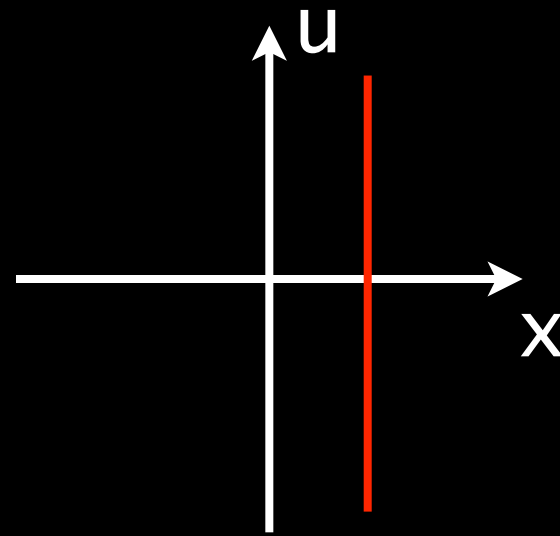


Wigner space
(x-u space)

spherical wave



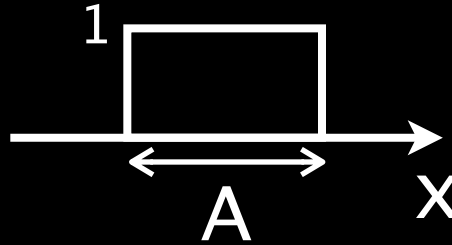
x-z space



Wigner space
(x-u space)

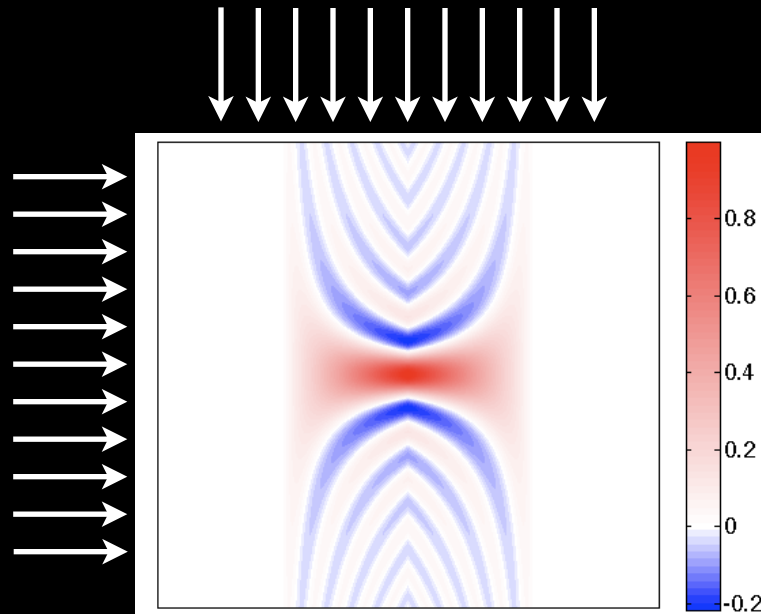
WDF *shears/rotates* upon propagation

boxcar (“rect”) function



integrate WDF along frequency axis

integrate WDF along space axis

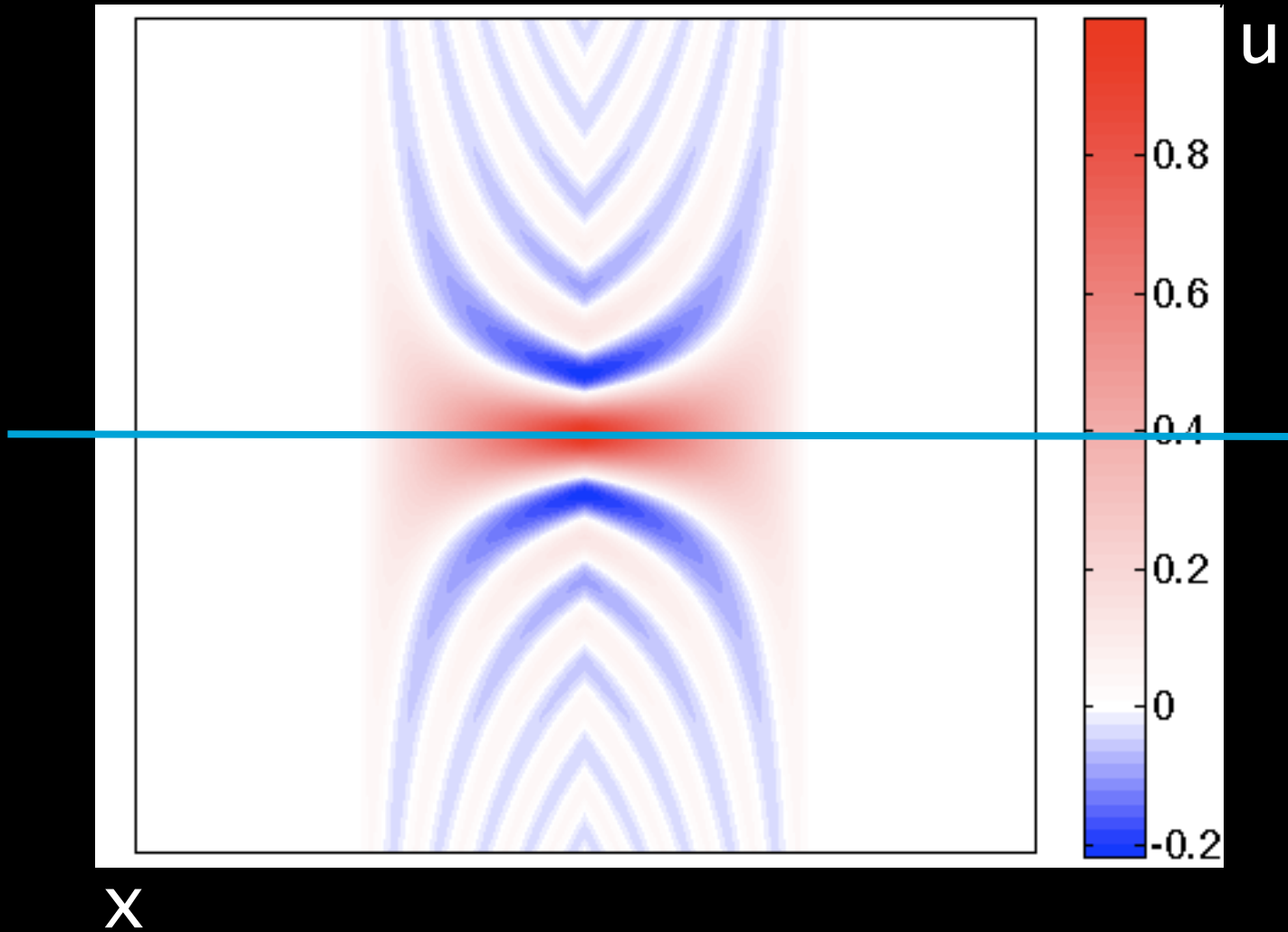


FT of original function



original function

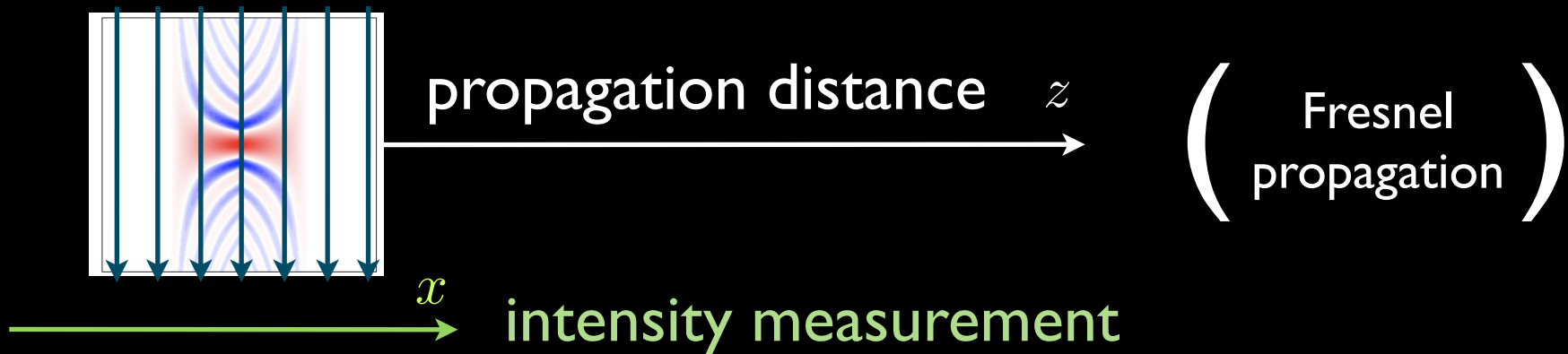
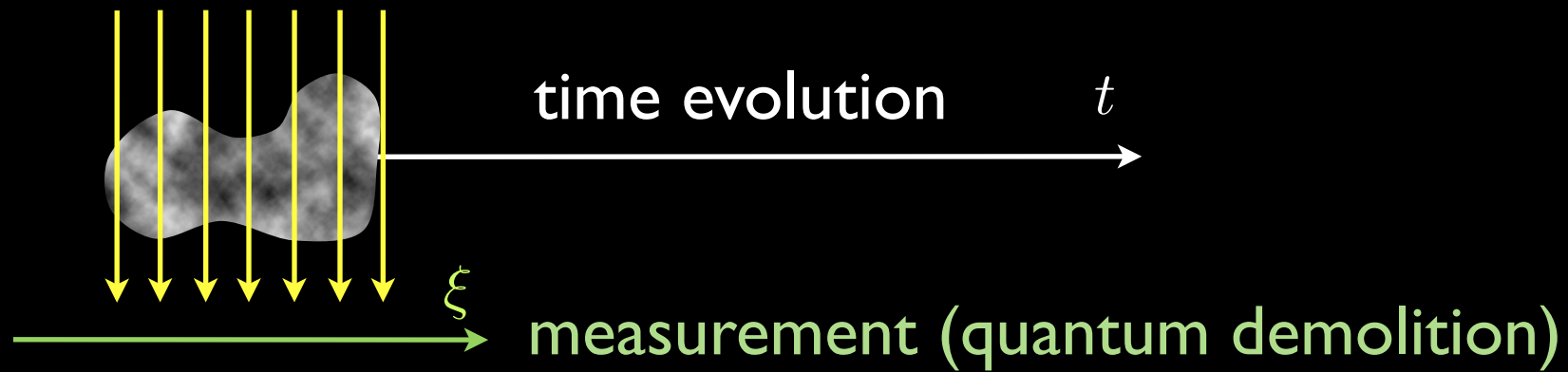
diffraction from a rectangular aperture



Wavefunction evolution and the WDF



Tomographic measurement from evolution/propagation



Phase-space tomography



z_0

z_1

z_2

z_3

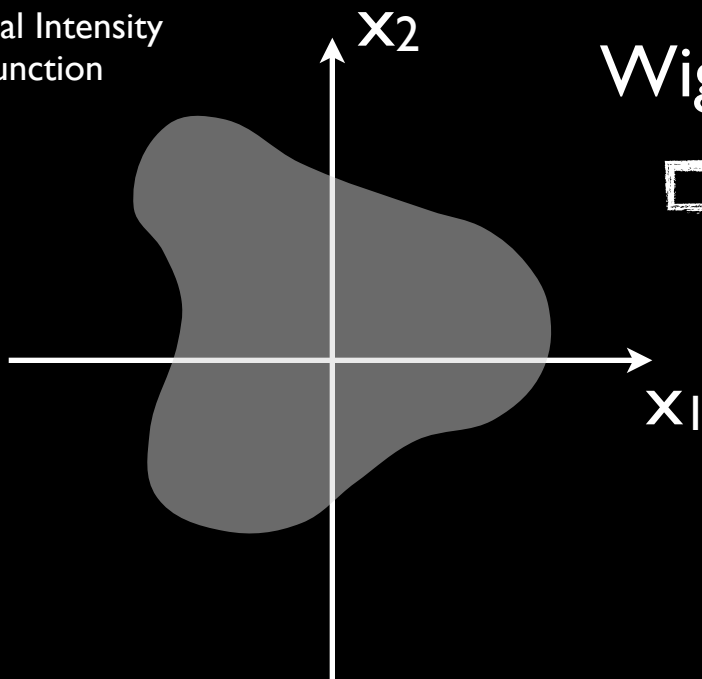
z_4

partially
coherent
field
(unknown)

camera
(intensity
measurement)

Phase-space tomography

Mutual Intensity
function

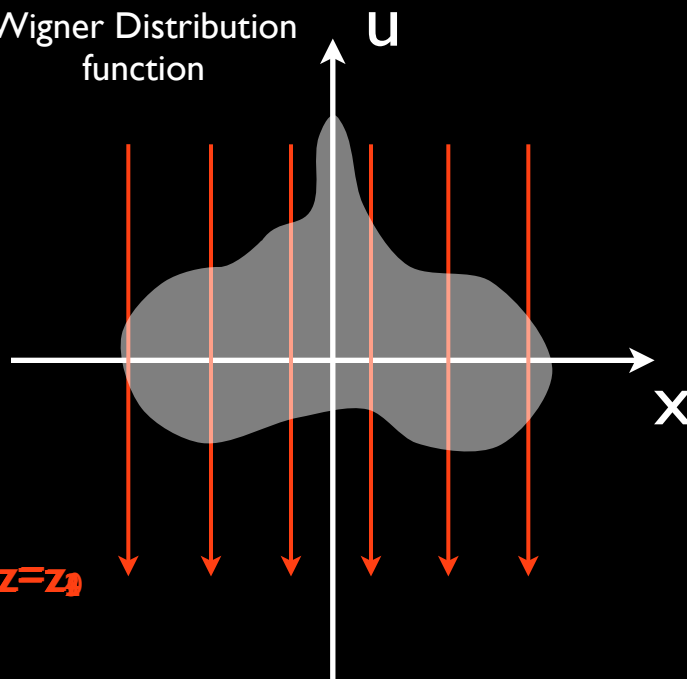


$$J(x_1, x_2)$$

Wigner

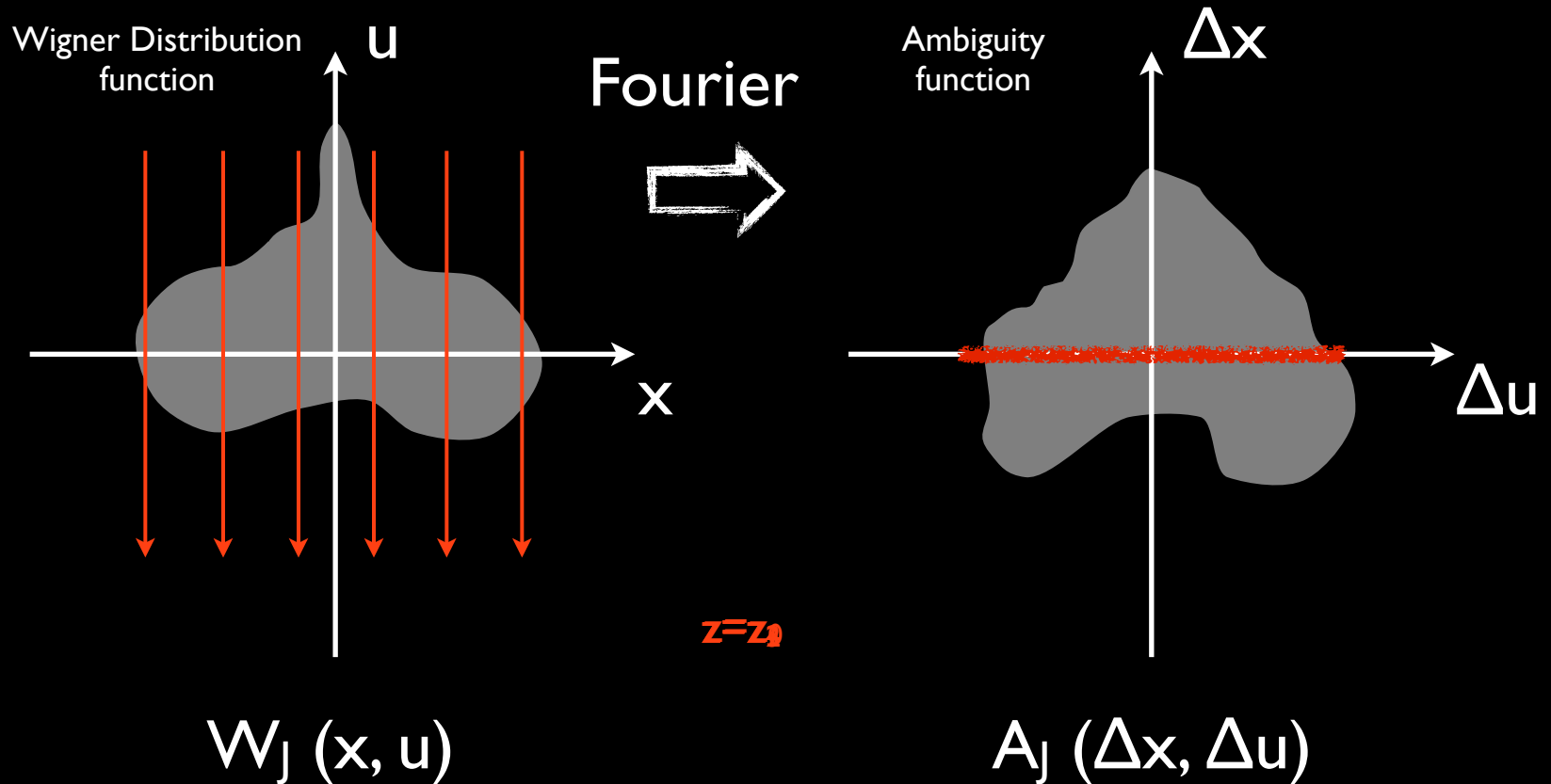


Wigner Distribution
function



$$W_J(x, u)$$

Phase-space tomography

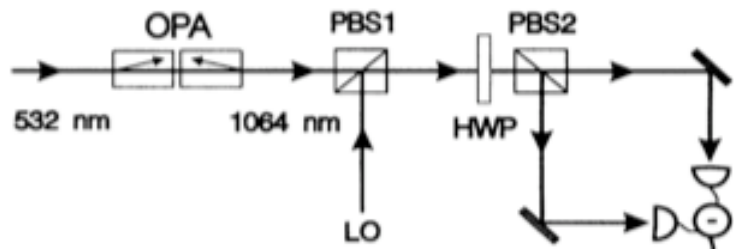


Quantum phase space tomography

Squeezed state recovery

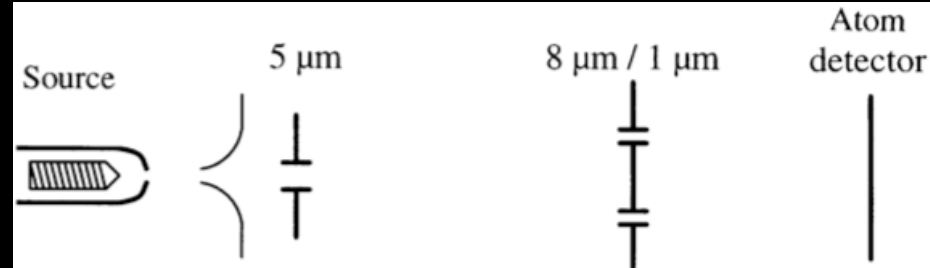
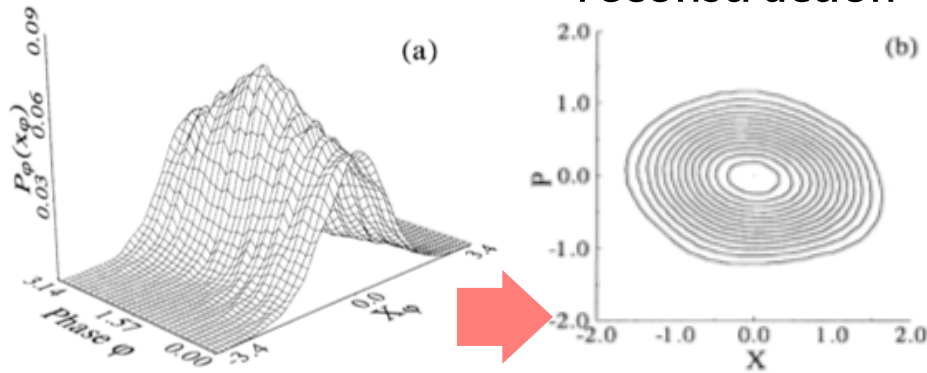
Matter wave interference

Optical Homodyne Tomography



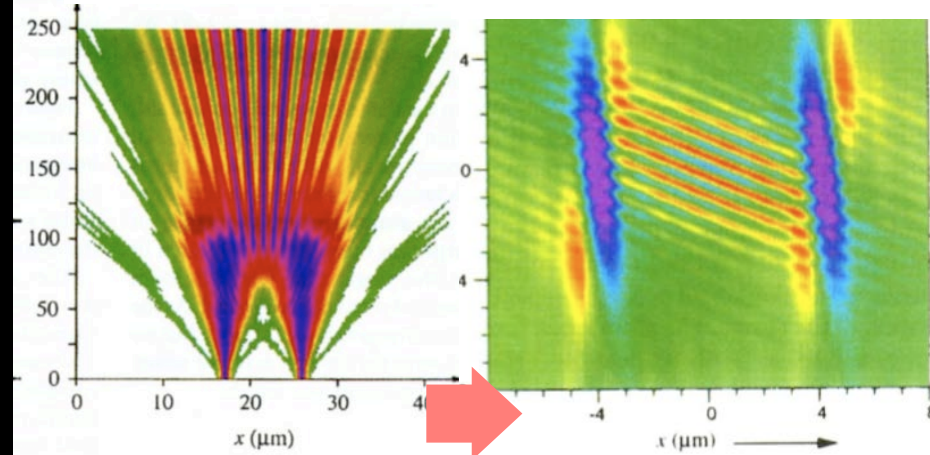
measurement

Tomographic reconstruction



measurement

Tomographic reconstruction



D. Smithey, and et al, *Phys. Rev. Lett.* 1993

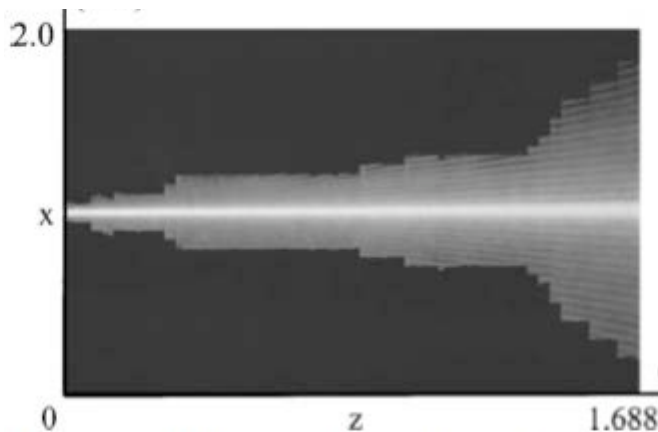
C. Kurtsiefer, and et al, *Nature*, 1997
 J. Itatani, and et al, *Nature*, 2000

Optical phase space tomography

- Non-interferometric technique

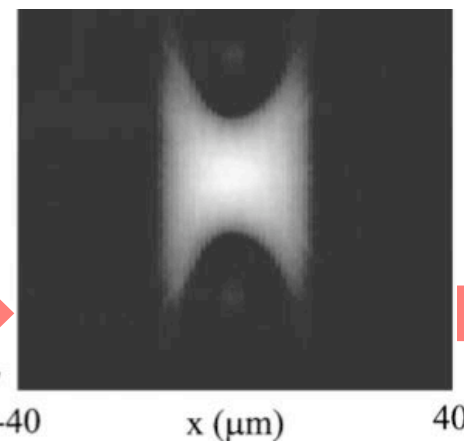
Spatial coherence measurements of a 1D soft x-ray beam

Axial intensity measurement

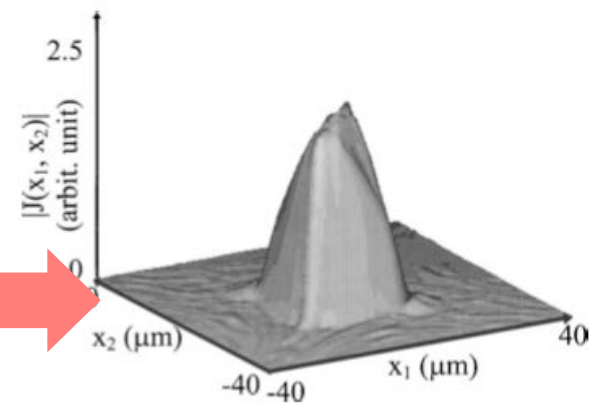


u (μrad)

Reconstructed WDF



Reconstructed MI



The problem of limited data



- Assume intensity symmetric about $z=0$

Compressive reconstruction of phase-space data

- Sparsity claim: number of *coherent modes*
- Low-Rank Matrix Recovery
 - E. J. Candés and B. Recht, *Found. Comp. Math.* 9:717, 2009
 - L. Tian, J. Lee, S. B. Oh, and G. Barbastathis, *Opt. Express* 20:8296, 2012
- Factored Form Descent
 - Z. Zhang, S. Rehman, C. Zhi, and G. Barbastathis, *Opt. Express* 21:5759, 2013
- Sparse Kalman filtering
 - J. Zhong, L. Tian, R. A. Claus, J. Dauwels, and L. Waller, *FiO 2013* paper FW6A.9 (post-deadline, today)

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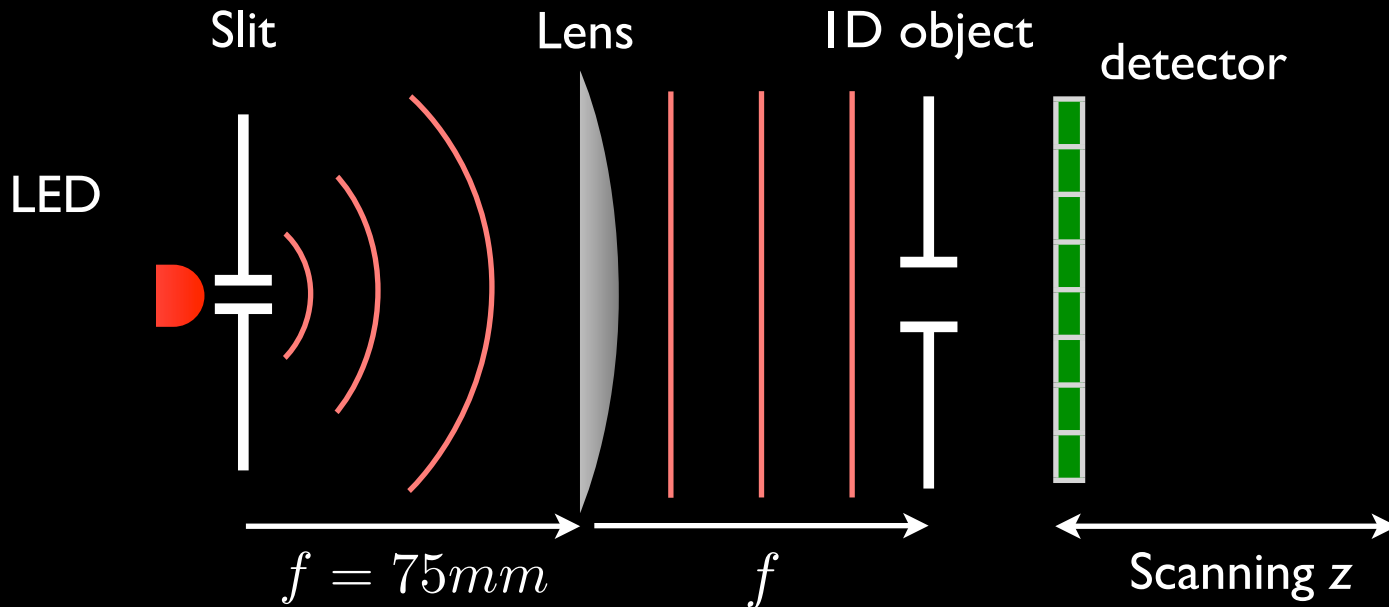
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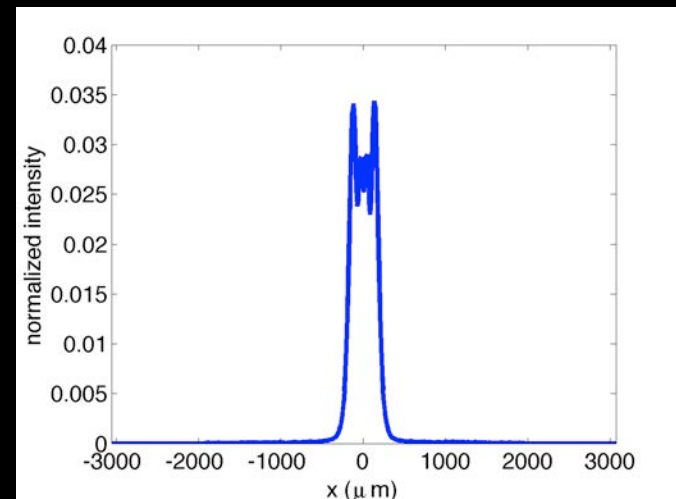
Low Rank Matrix Recovery (LRMR) of phase-space data

- Solution expressed as *few* coherent modes
- L0 minimization
 - sadly, intractable
 - however [Candes] the problem can be mapped onto an equivalent L1 minimization
- Semi-positive definiteness \Leftrightarrow mode coefficients ≥ 0
additionally enforced as constraint

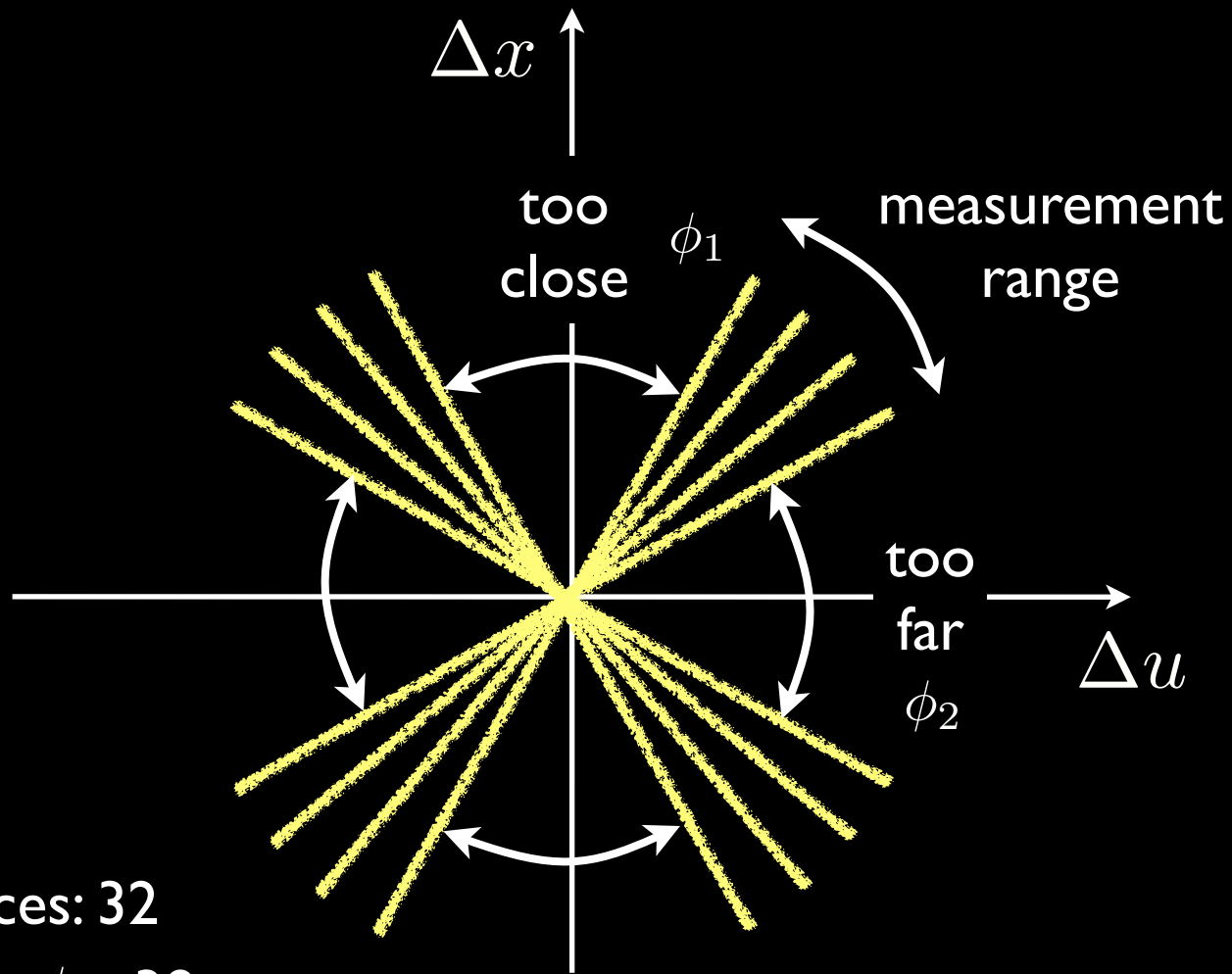
Experimental compressive phase-space tomography



- Illumination central wavelength: 620nm; bandwidth: 20nm
- Width of illumination slit: 300 μm
- Coherence length: 93 μm
- Width of object slit: 400 μm
- 32 measurements (axial positions)

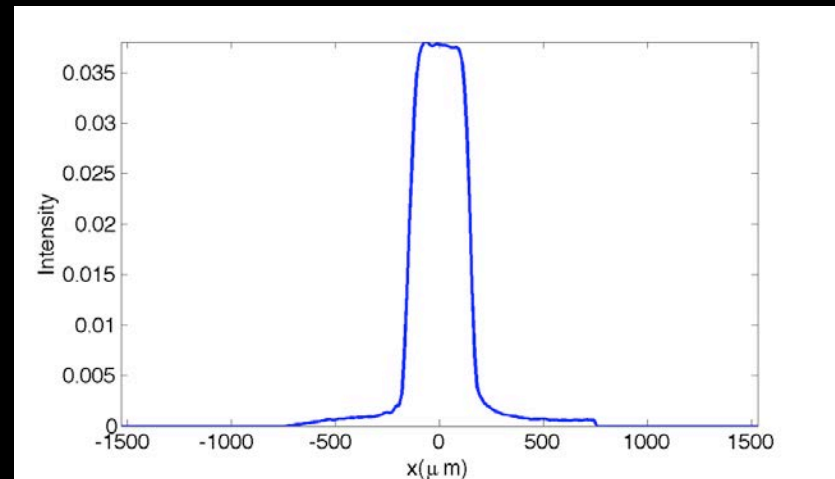
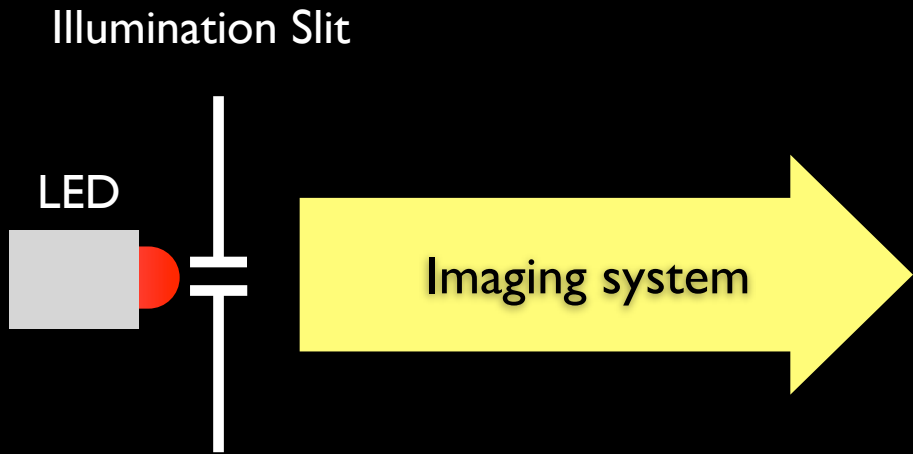


Limited data in our experiment



- Total # of slices: 32
- Missing angle ϕ_1 : 38°
- Missing angle ϕ_2 : 22°

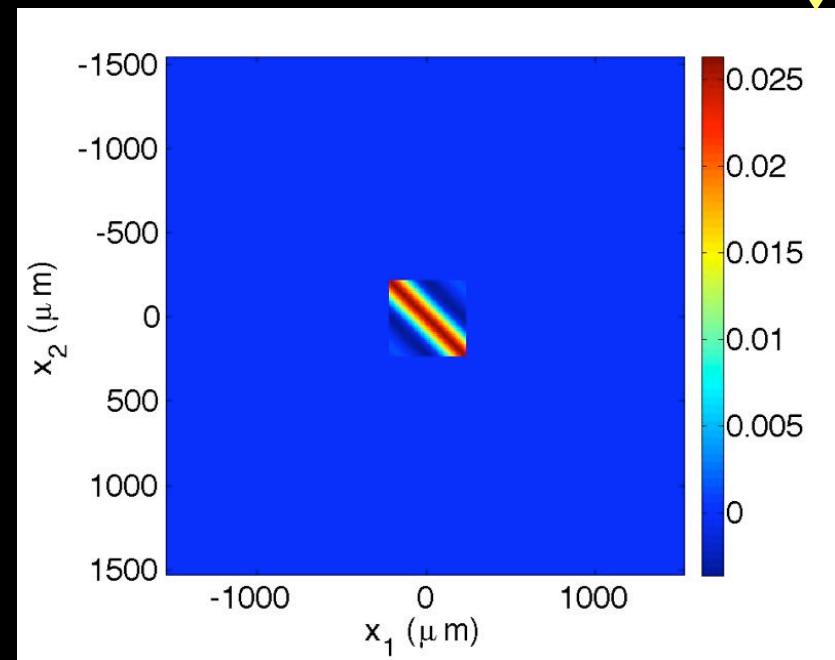
Ground Truth



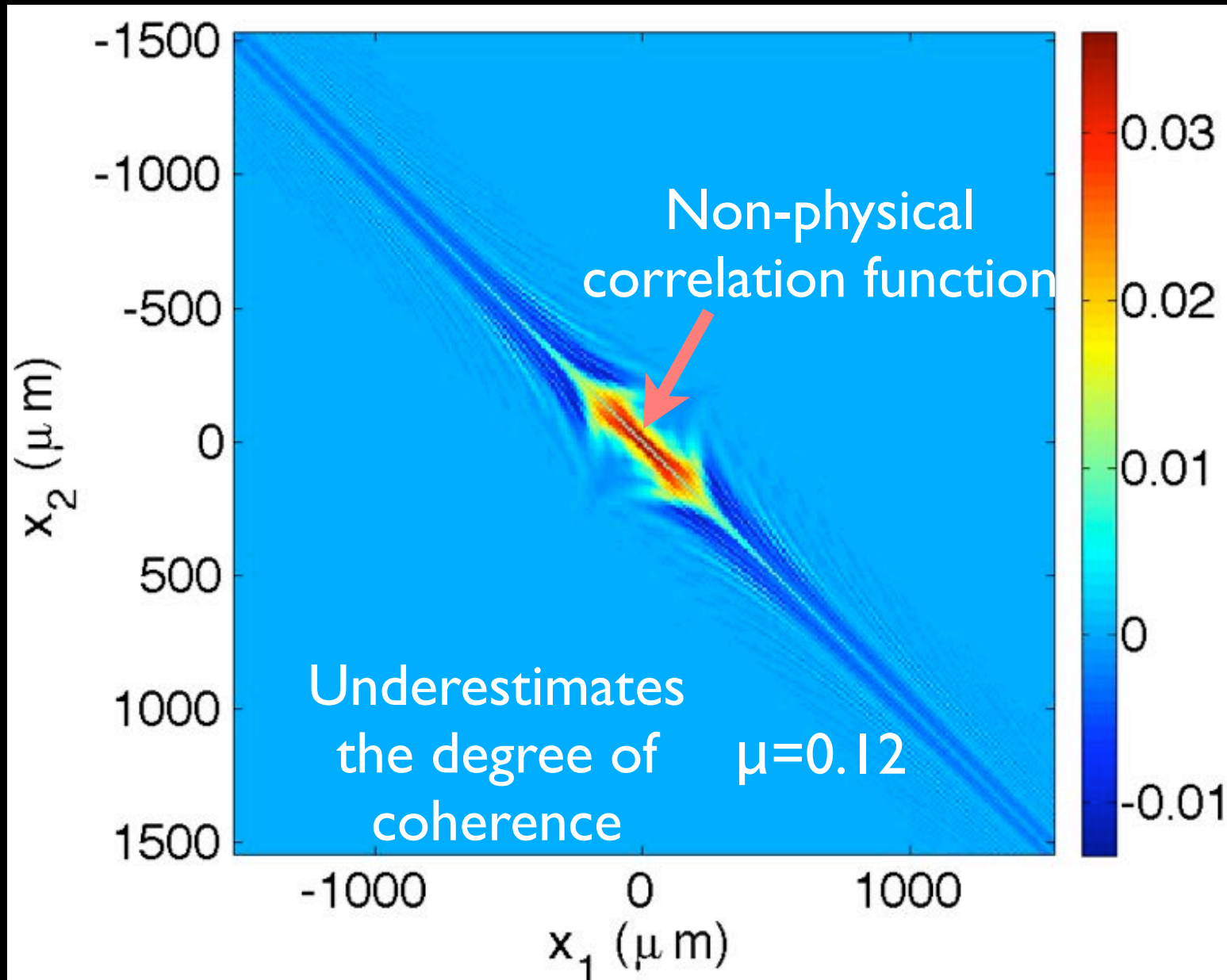
van Cittert-Zernike theorem

Global Degree of Coherence

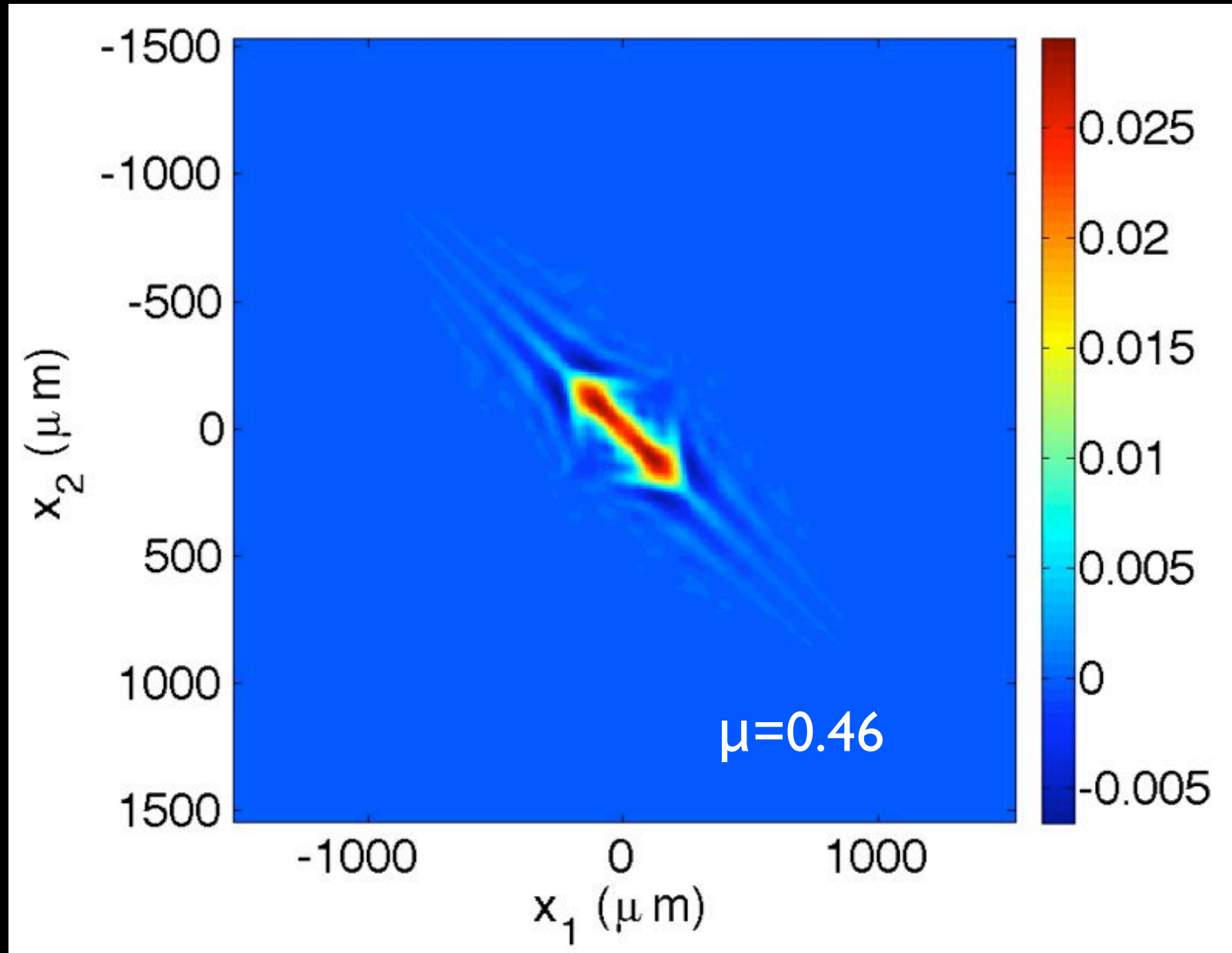
$$\mu=0.49$$



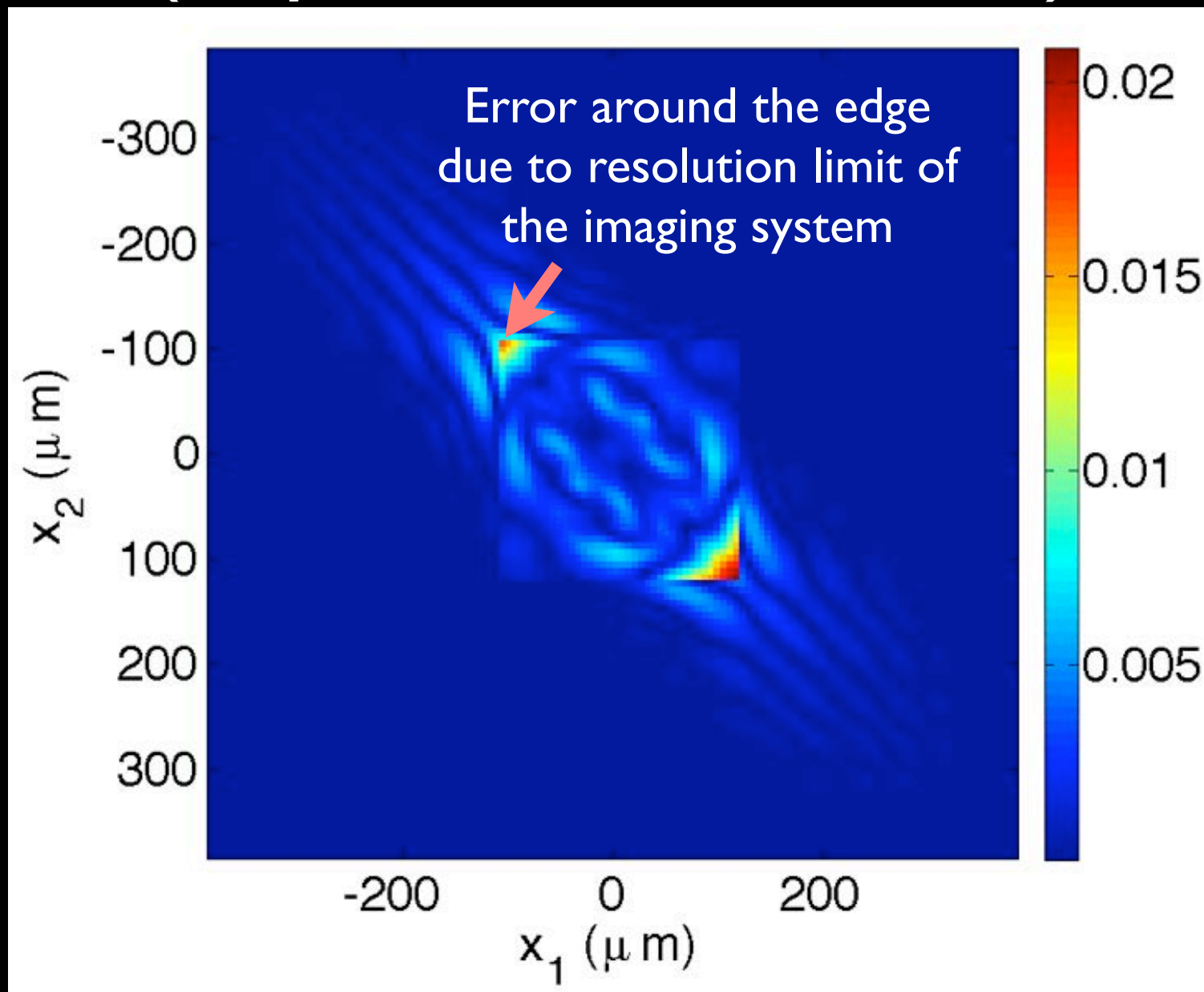
Filtered back-projection fails



Compressive reconstruction



Error in compressive reconstruction (compared to Van Cittert-Zernike)



Compressive reconstruction of phase-space data

- Sparsity claim: number of *coherent modes*

- 
- Low-Rank Matrix Recovery

- E. J. Candés and B. Recht, *Found. Comp. Math.* 9:717, 2009

- L. Tian, J. Lee, S. B. Oh, and G. Barbastathis, *Opt. Express* 20:8296, 2012

- Factored Form Descent

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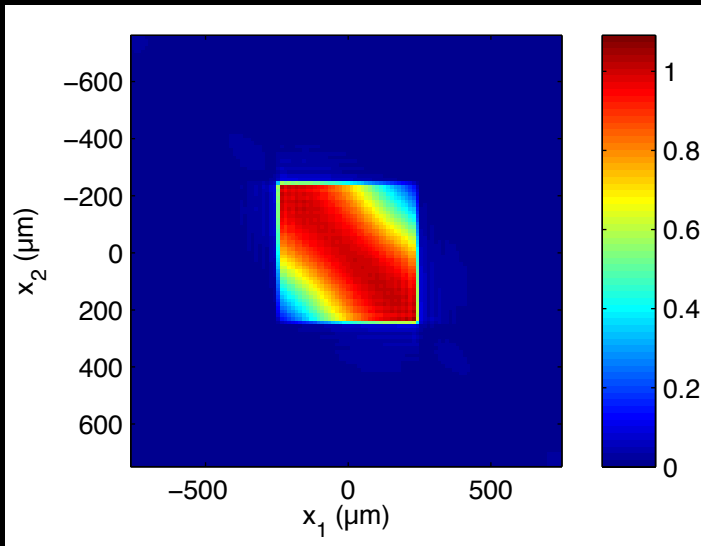
Factored Form Descent

- coherence mode decomposition: $J = UU^H$
- solve for U instead of J means semi-positive definiteness is automatically ensured
- quartic, need iterated descent algorithm
 - initialize with fully incoherent guess
 - find search direction, e.g. $\Delta U = ADA^H U$ (steepest descent, NL conjugate gradient)
 - perform (global) line search (solve for roots of cubic polynomial)
 - mutual intensity from singular value decomposition

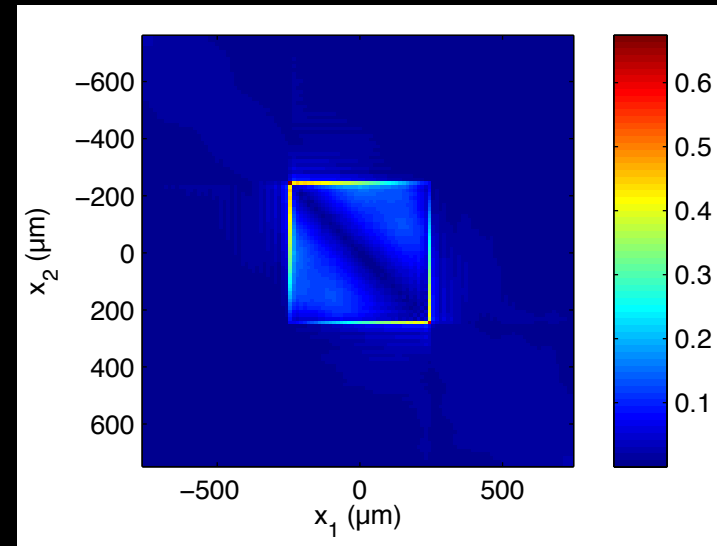
Wolf, JOSA 1982
Ozaktas et al.,
JOSAA 2002

Experimental Test

51×401 intensity measurements
to compute 200×200 mutual intensity matrix



experimental



error compared to theory
(theory assumes perfect lenses, paraxial
propagation, uniform LED, infinitesimal pixel size)

Phase from partially coherent fields

- Compressive reconstruction of the mutual intensity [Tian *Opt. Exp.* 20:8296]
- Factored Form Descent [Zhang *Opt. Exp.* 21:5756]
- Wigner distribution function recovery from **lenslet arrays** [Tian *Opt. Exp.* 21:10511]
- Optical Path Length (OPL) recovery with partially coherent illumination [Petruccelli *Opt. Exp.* 21:14430]

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 - Laura Waller, David Brady, Colin Sheppard
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