

# Phase Retrieval: From Computational Imaging to Machine Learning

Jonathan Dong<sup>1</sup>, Lorenzo Valzania<sup>2</sup>, Antoine Maillard<sup>3</sup>, Thanh-an Pham<sup>1</sup>, Sylvain Gigan<sup>2</sup>, Michael Unser<sup>1</sup>

<sup>1</sup>Biomedical Imaging Group, EPFL, Lausanne, Switzerland

<sup>2</sup>Laboratoire Kastler Brossel, ENS-PSL, Paris, France

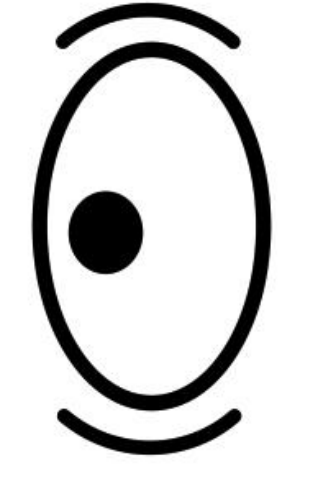
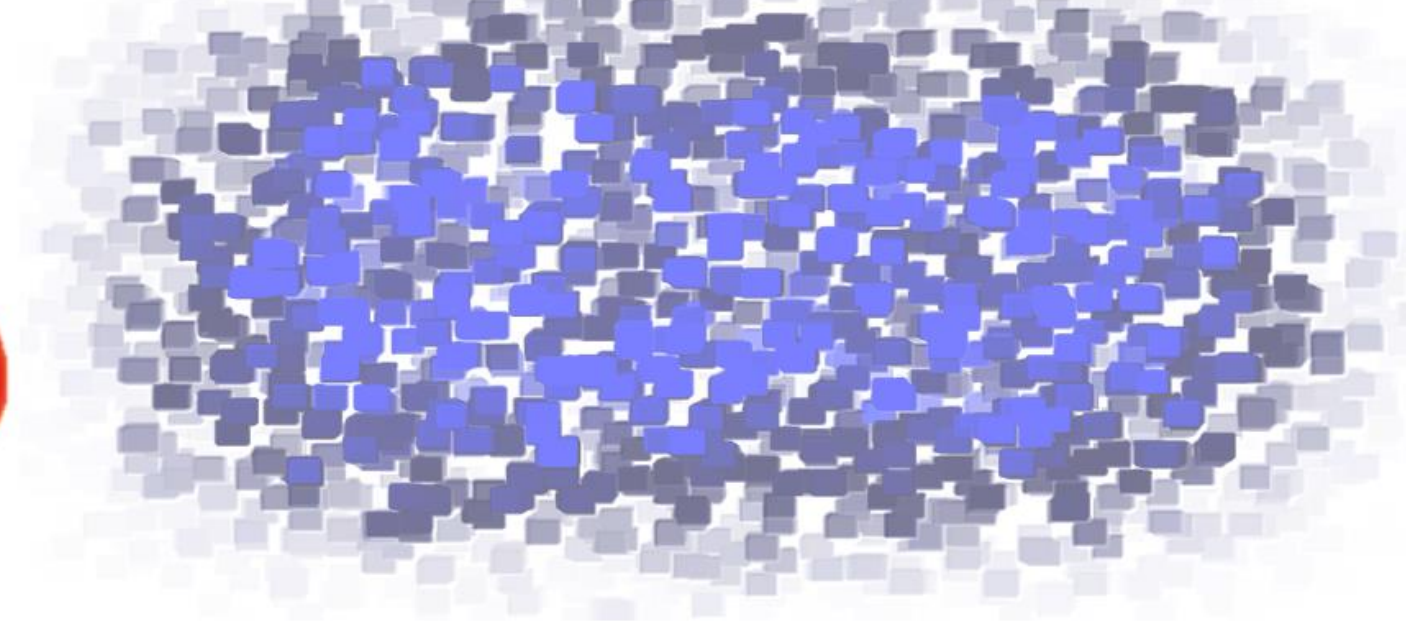
<sup>3</sup>Department of Mathematics, ETHZ, Zürich, Switzerland

## Phase Retrieval

Find  $x$  in  
 $y = |Ax|^2$

$$\begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = \begin{bmatrix} a_{11} & \dots & a_{1d} \\ \vdots & \ddots & \vdots \\ a_{n1} & \dots & a_{nd} \end{bmatrix} \begin{bmatrix} x_1^* \\ \vdots \\ x_d^* \end{bmatrix}$$

## 2 Computational imaging



unknown object

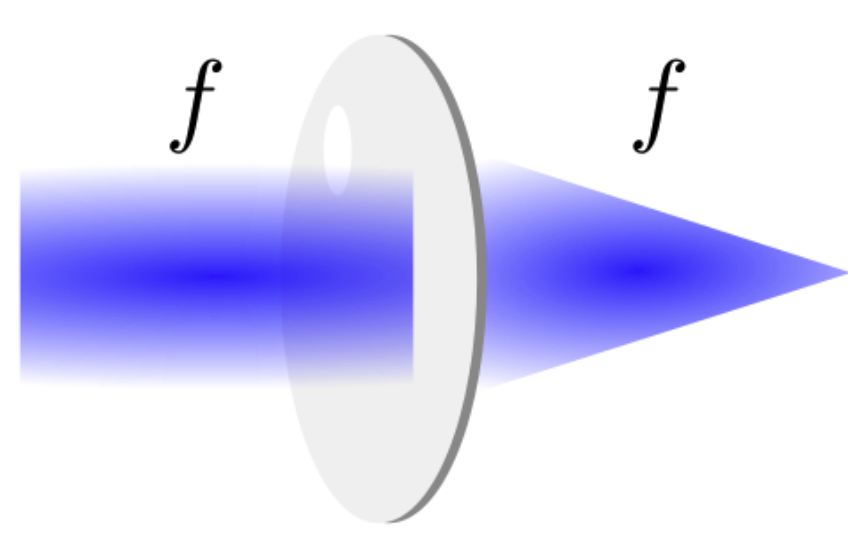
sensing matrix

measurements

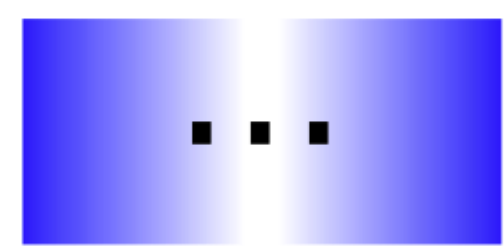
### Fourier phase retrieval

$$A = F$$

by a lens



by free-space propagation

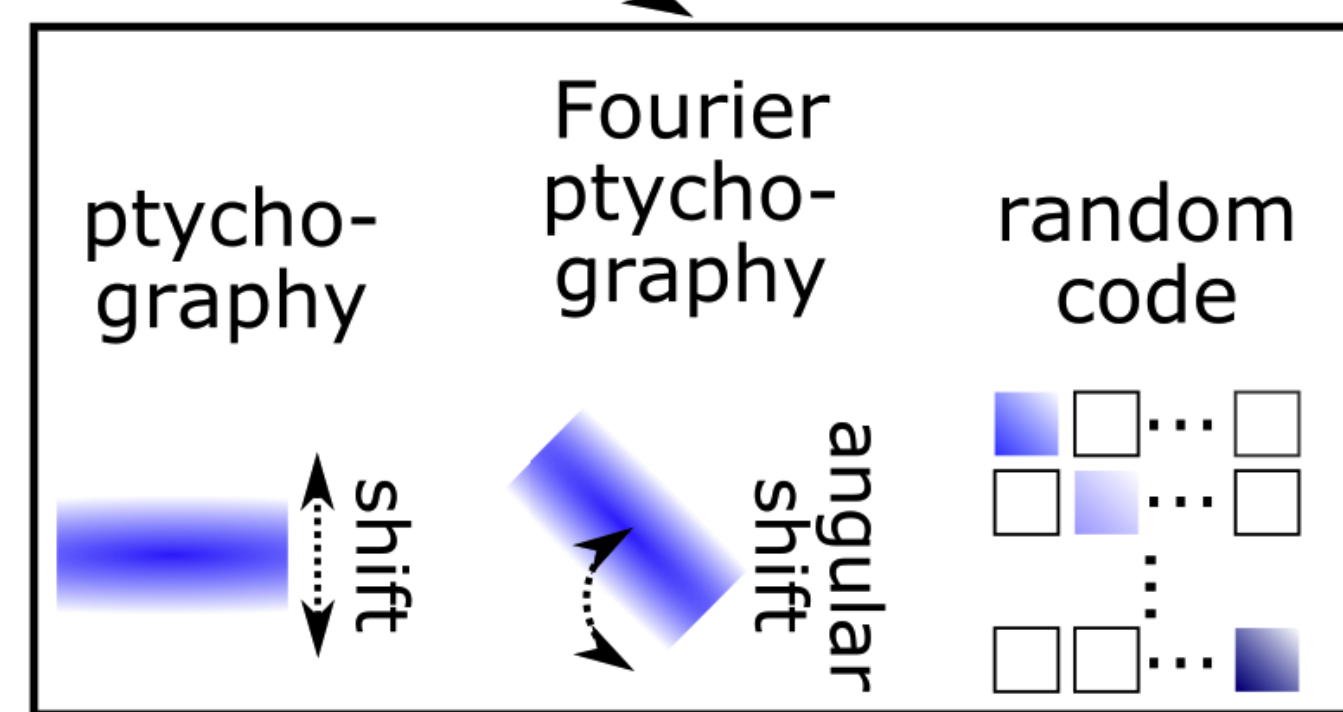


### Coded illumination

$$A_l = FD_l$$

imaging system

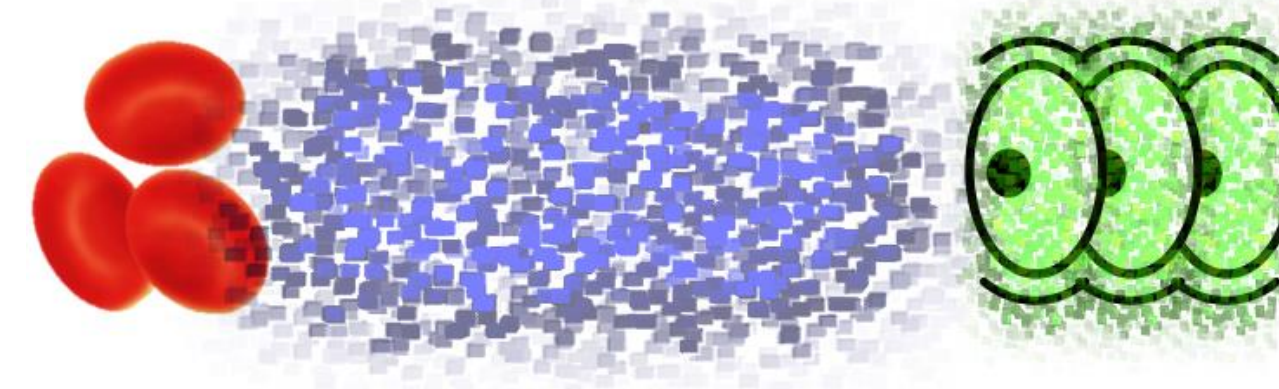
code



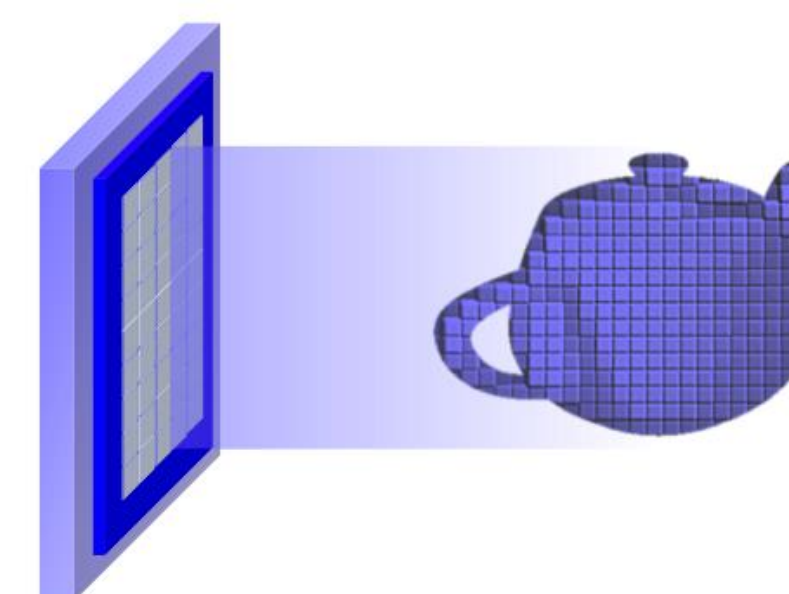
### Coded detection

$$A_l = FD_l F^H$$

phase diversity

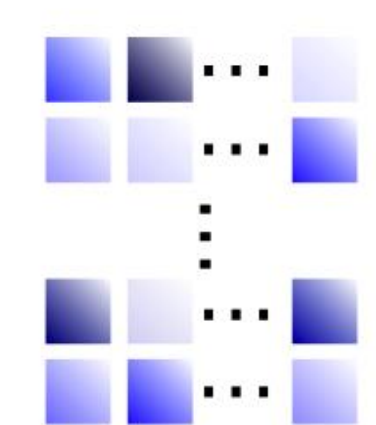


computer-generated holography

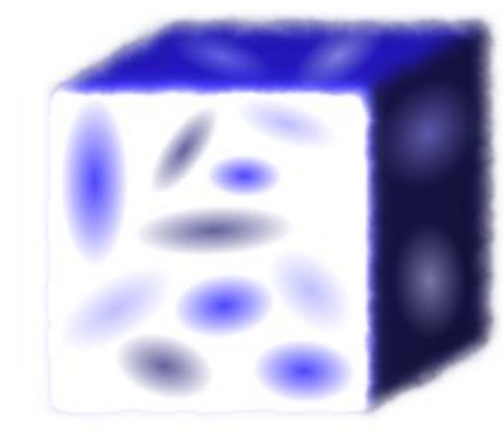


### Random

by random projections



by propagation through complex media



## Reconstruction algorithms

↔ Alternating projections

First algorithms  
Gerchberg-Saxton, Fienup, etc.  
Still used today

↘ Gradient-based optimization

Non-linear optimization  
Simple and flexible  
Few convergence guarantees

🛏 Convex relaxation

Rewrite the original equation  
Closer to convex optimization  
A few variants

🧠 Bayesian AMP

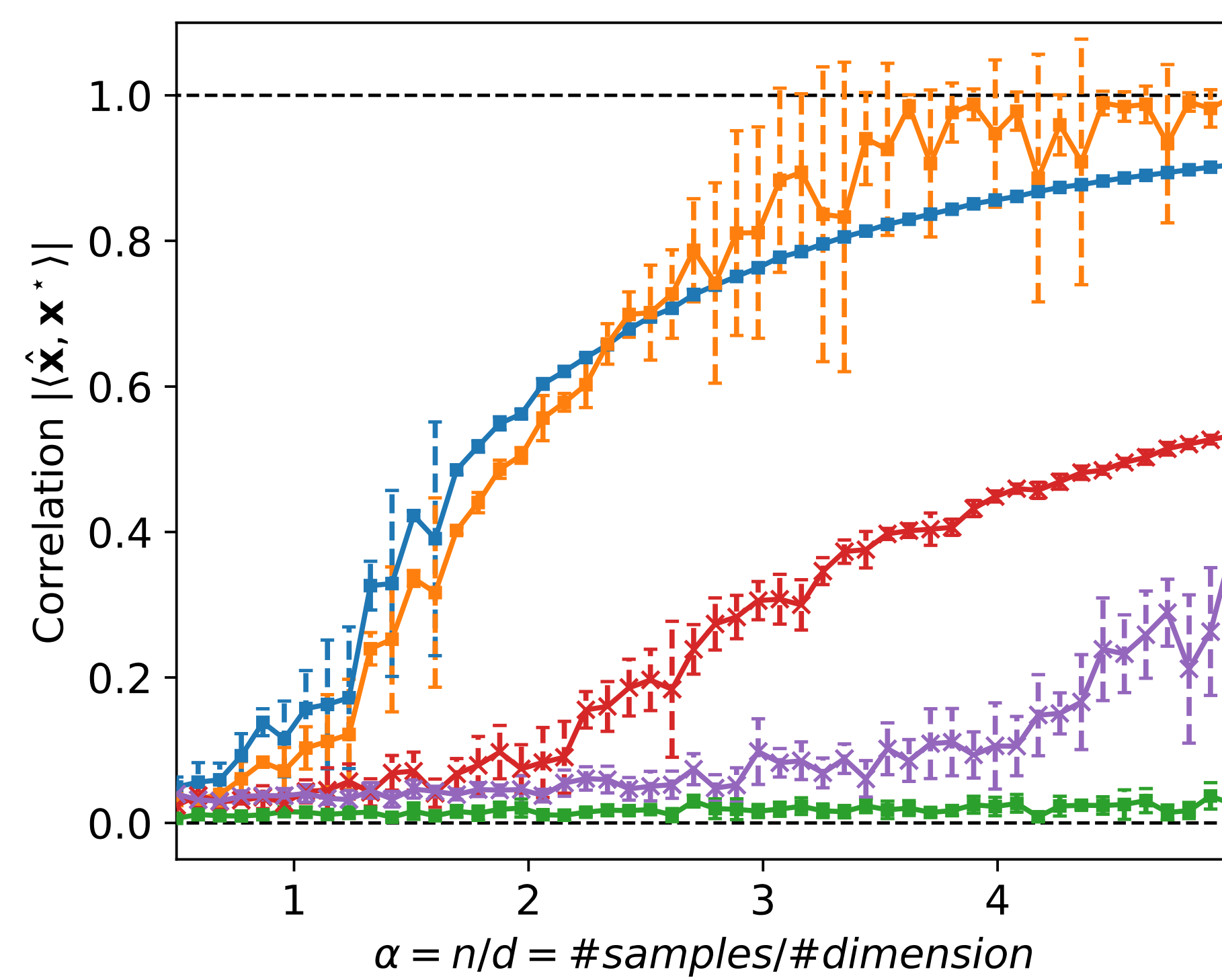
Based on Bayesian framework  
Approximate Message Passing  
Optimal among algorithms?

👻 Spectral methods

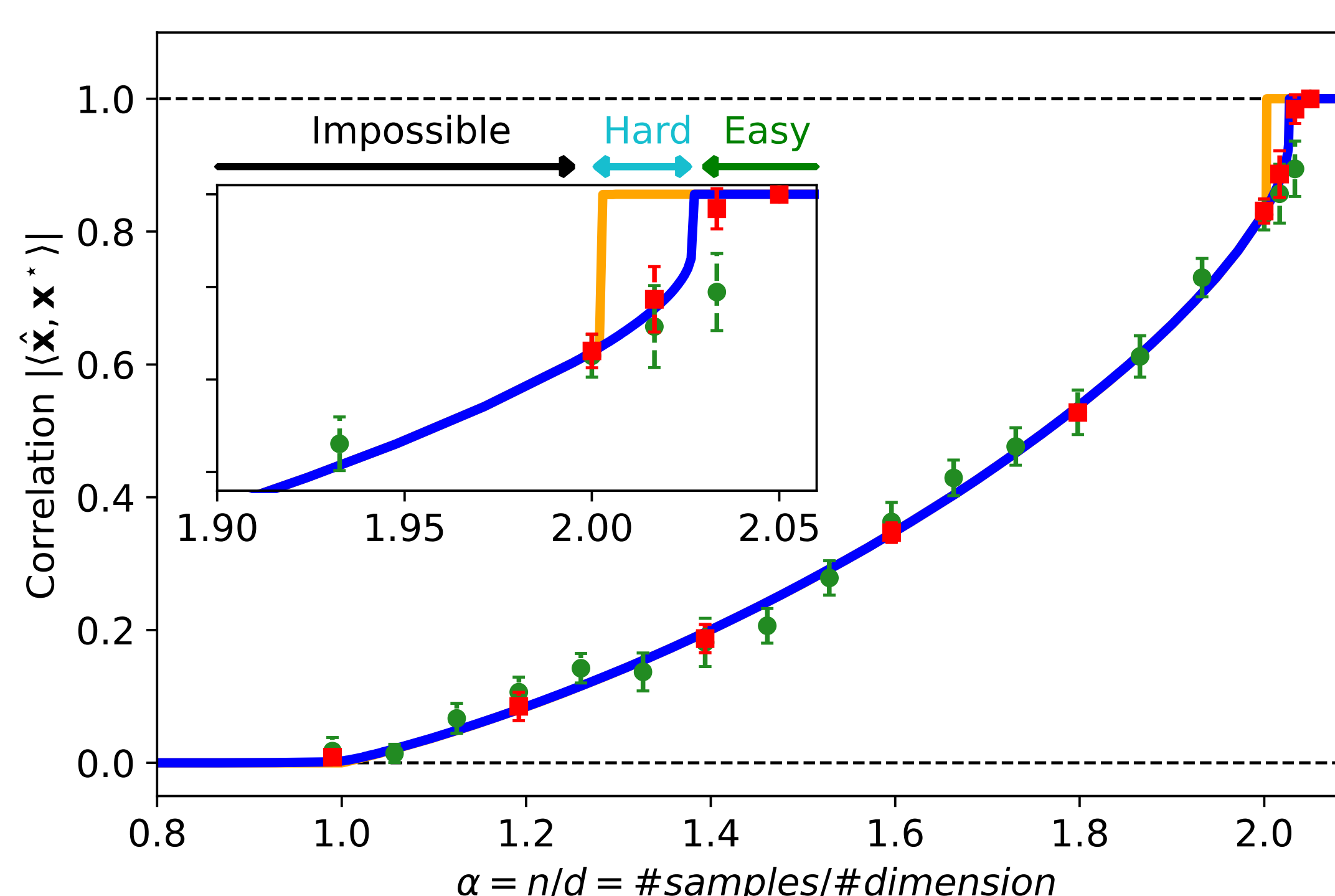
Based on power iterations  
Fast to compute  
Provide good initial guess

## Algorithm benchmark in the random setting

Correlation: 0 → random guess, 1 → perfect recovery



Legend:  
 - Gradient descent (GD) - Random initialization  
 - Trimming spectral method (t = 5.0)  
 - Trimming spectral method (t = 10.0)  
 - Optimal Spectral method (OS)  
 - OS + Gradient descent



Legend:  
 - IT  
 - AMP (asymptotic)  
 - AMP (synthetic, d = 5000)  
 - AMP (image)

## Machine Learning regularization

Direct inversion with NN

Train neural network to map measurements to reconstruction

Regularization using CNN denoisers

Denoising introduced at each iteration  
RED, PnP, etc.

Generative models as priors

Train a GAN to generate images  
Restrict search space to GAN outputs

Deep Image Prior

Use untrained CNN to generate estimate  
Optimize weights to fit measurements

Others?