

System of quadratic equations

- Quadratic measurements obtained with high-rank measurement matrices arise in applications such as unassigned distance geometry problem.
- Most prior works focus on rank-1 psd measurement matrices or real measurements.

Measurement Model:

$$y_i = \mathbf{x}^* \mathbf{A}_i \mathbf{x}, \quad i = 1, \dots, m. \quad (1)$$

- $\mathbf{x} \in \mathbb{C}^n$ is the complex signal.
- $y_i \in \mathbb{C}$ is the i -th complex quadratic measurement.
- $\mathbf{A}_i \in \mathbb{C}^{m \times n}$ is the i -th complex random Gaussian measurement matrix.

Problem formulation

We minimize the following objective function $f(\mathbf{z})$:

$$f(\mathbf{z}) = \frac{1}{m} \sum_{i=1}^m |\mathbf{z}^* \mathbf{A}_i \mathbf{z} - y_i|^2, \quad (2)$$

using gradient descent:

$$\mathbf{z}^{(t+1)} = \mathbf{z}^{(t)} - \eta \nabla f(\mathbf{z}), \quad (3)$$

where $\eta > 0$ is the step size.

- Nonconvex optimization problem.
- $\mathbf{x} e^{j\phi}$ is a global minimum solution for all $\phi \in [0, 2\pi)$.
- The distance between the recovered \mathbf{z} and a global minimum solution \mathbf{x} is

$$\text{dist}(\mathbf{z}, \mathbf{x}) = \min_{\phi \in [0, 2\pi)} \|\mathbf{z} - \mathbf{x} e^{j\phi}\|_2 \quad (4)$$

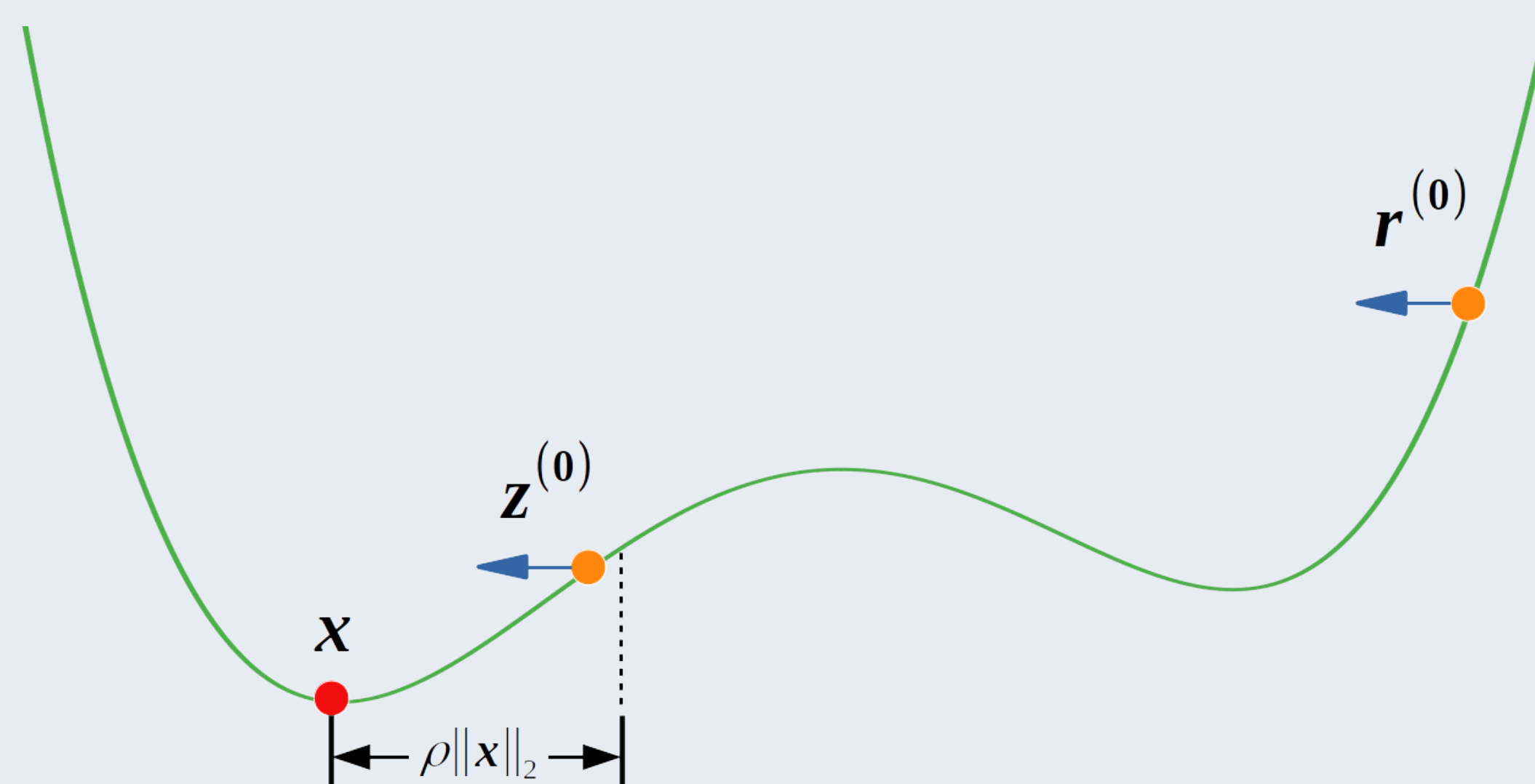


Figure 1: A good initialization is needed to solve a nonconvex optimization problem via gradient descent.

Spectral initialization

The left or right singular vector $\mathbf{z}^{(0)}$ of the following matrix:

$$\mathbf{S} = \frac{1}{m} \sum_{i=1}^m \bar{y}_i \cdot \mathbf{A}_i = \frac{1}{m} \sum_{i=1}^m \mathbf{x}^* \mathbf{A}_i^* \mathbf{x} \cdot \mathbf{A}_i. \quad (5)$$

- For sufficiently large m , with high probability \mathbf{S} concentrates around its expectation in terms of spectral norm.

$$\mathbb{E}[\mathbf{S}] = 2\mathbf{x}\mathbf{x}^* \quad (6)$$

- The spectral initializer $\mathbf{z}^{(0)}$ is close to a global optimum \mathbf{x} [1,2].

$$\text{dist}(\mathbf{z}^{(0)}, \mathbf{x}) \leq \delta \|\mathbf{x}\|_2 \quad (7)$$

holds with high probability.

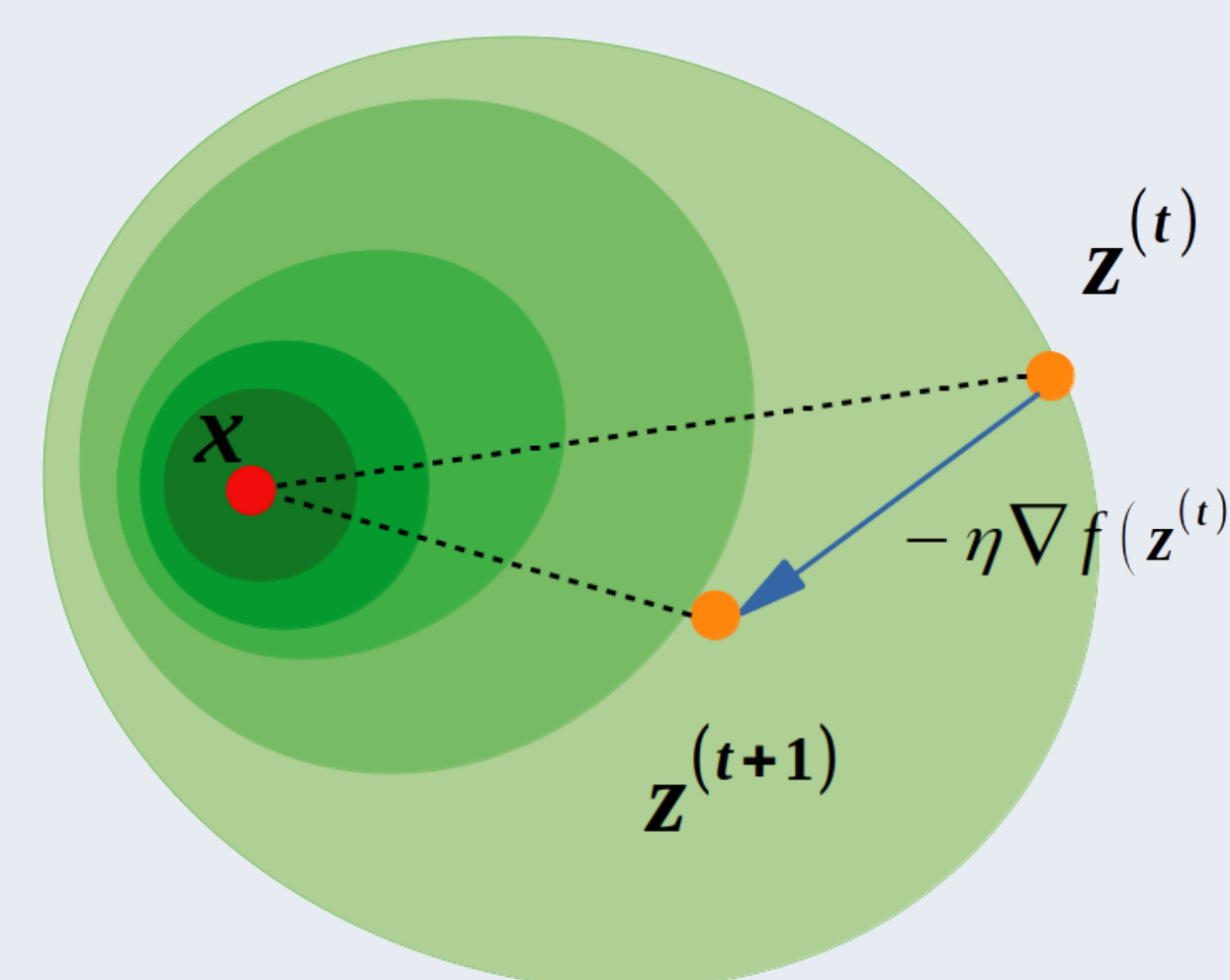
Lemma

When $m \geq Cn$ for some sufficiently large constant C , for all $\mathbf{p}, \mathbf{q} \in \mathbb{C}^n$ satisfying $\|\mathbf{p}\|_2 = 1$, $\|\mathbf{q}\|_2 = 1$ and every $\nu > 0$, the following

$$\left\| \frac{1}{m} \sum_{i=1}^m \mathbf{p}^* \mathbf{A}_i^* \mathbf{q} \cdot \mathbf{A}_i - 2\mathbf{q}\mathbf{p}^* \right\| \leq \nu, \quad (8)$$

holds with high probability.

Linear convergence



$$\text{dist}(\mathbf{z}^{(t+1)}, \mathbf{x}) \leq \mu \cdot \text{dist}(\mathbf{z}^{(t)}, \mathbf{x}), \quad \mu \in (0, 1) \quad (9)$$

- Guaranteed if $f(\mathbf{x})$ satisfies the regularity condition $RC(\alpha, \beta, \rho)$ [2] for all $\mathbf{z} \in E(\rho) = \{\mathbf{z} \mid \text{dist}(\mathbf{z}, \mathbf{x}) \leq \rho \|\mathbf{x}\|_2\}$.

$$\text{Re}(\langle \nabla f(\mathbf{z}), \mathbf{z} - \mathbf{x} e^{j\phi_{\min}} \rangle) \geq \frac{1}{\alpha} \text{dist}^2(\mathbf{z}, \mathbf{x}) + \frac{1}{\beta} \|\nabla f(\mathbf{z})\|_2^2, \quad (10)$$

where $\alpha > 0$, $\beta > 0$, $1 > \rho > 0$ are some constants.

Main theorem

When $m \geq Cn$ for some sufficiently large constant C ,

- There exists a choice of $1 > \nu > 0$, $1 > \rho > 0$, $\alpha > 0$, $\beta > 0$ such that $RC(\alpha, \beta, \rho)$ holds on $E(\rho)$ with high probability.
- Furthermore, if the step size $0 < \eta \leq \frac{2}{\beta}$, the gradient descent with the spectral initializer $\mathbf{z}^{(0)}$ converges linearly to \mathbf{x}

$$\text{dist}(\mathbf{z}^{(t)}, \mathbf{x}) \leq \left(\sqrt{1 - \frac{2\eta}{\alpha}} \right)^t \rho \cdot \|\mathbf{x}\|_2, \quad (11)$$

with high probability.

Experimental results

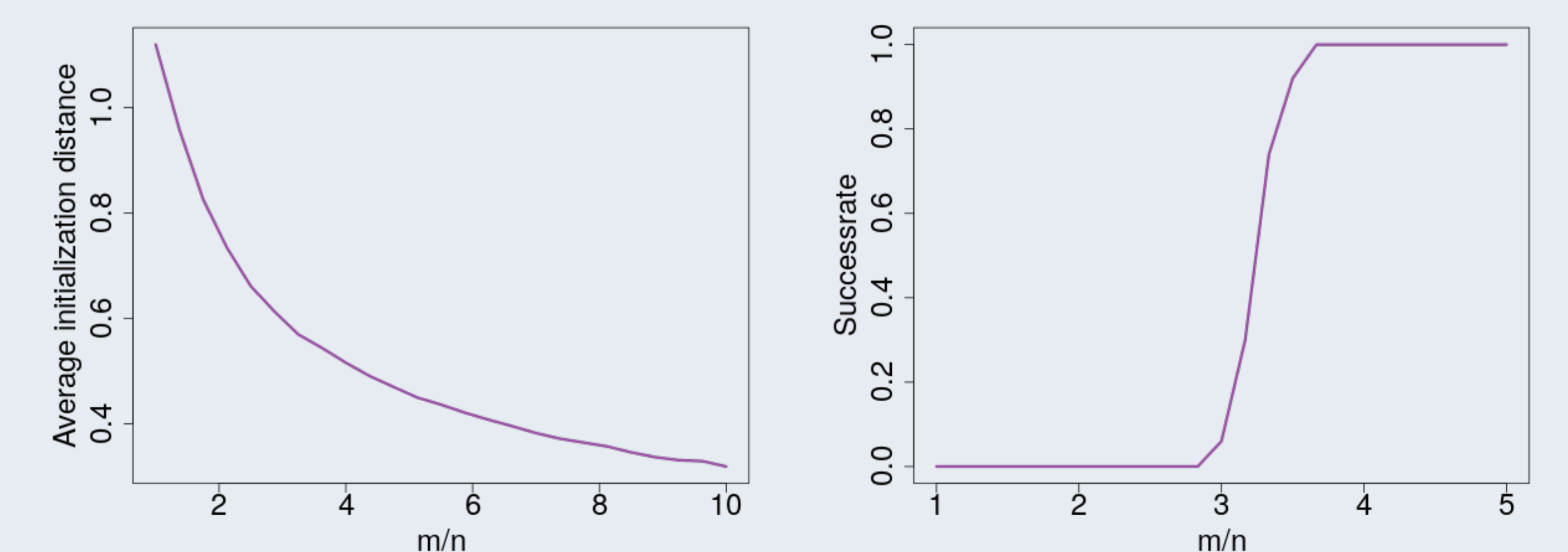


Figure 2: Left: $\text{dist}(\mathbf{z}^{(0)}, \mathbf{x})$; Right: Phase transition of the success rate.

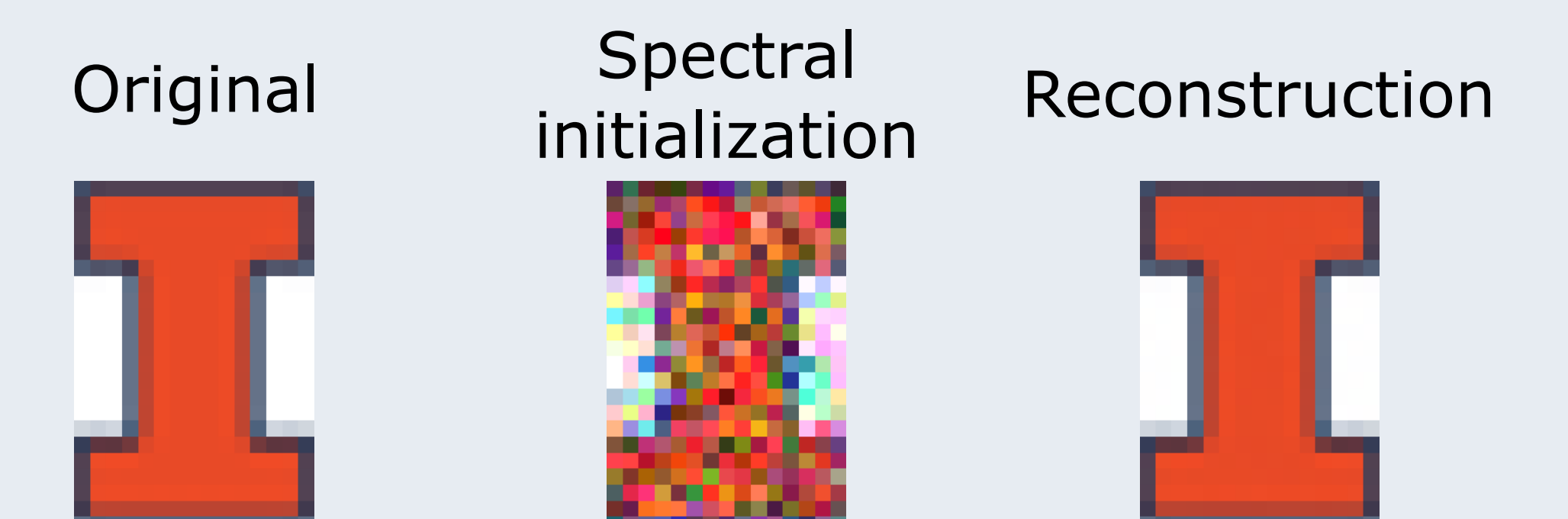


Figure 3: Recovery of the UIUC logo.

References

- Netrapalli et al., Phase retrieval using alternating minimization, IEEE Trans. Signal Process., 2015
- Candès et al., Phase retrieval via Wirtinger flow: Theory and algorithms, IEEE Trans. Inf. Theory, 2015
- S. Huang et al., Solving Complex Quadratic Systems with Full-Rank Random Matrices, arXiv, vol.abs/1902.05612, 2019