



arXiv:2404.05616

Quantum tomography of structured light patterns from simple intensity measurements



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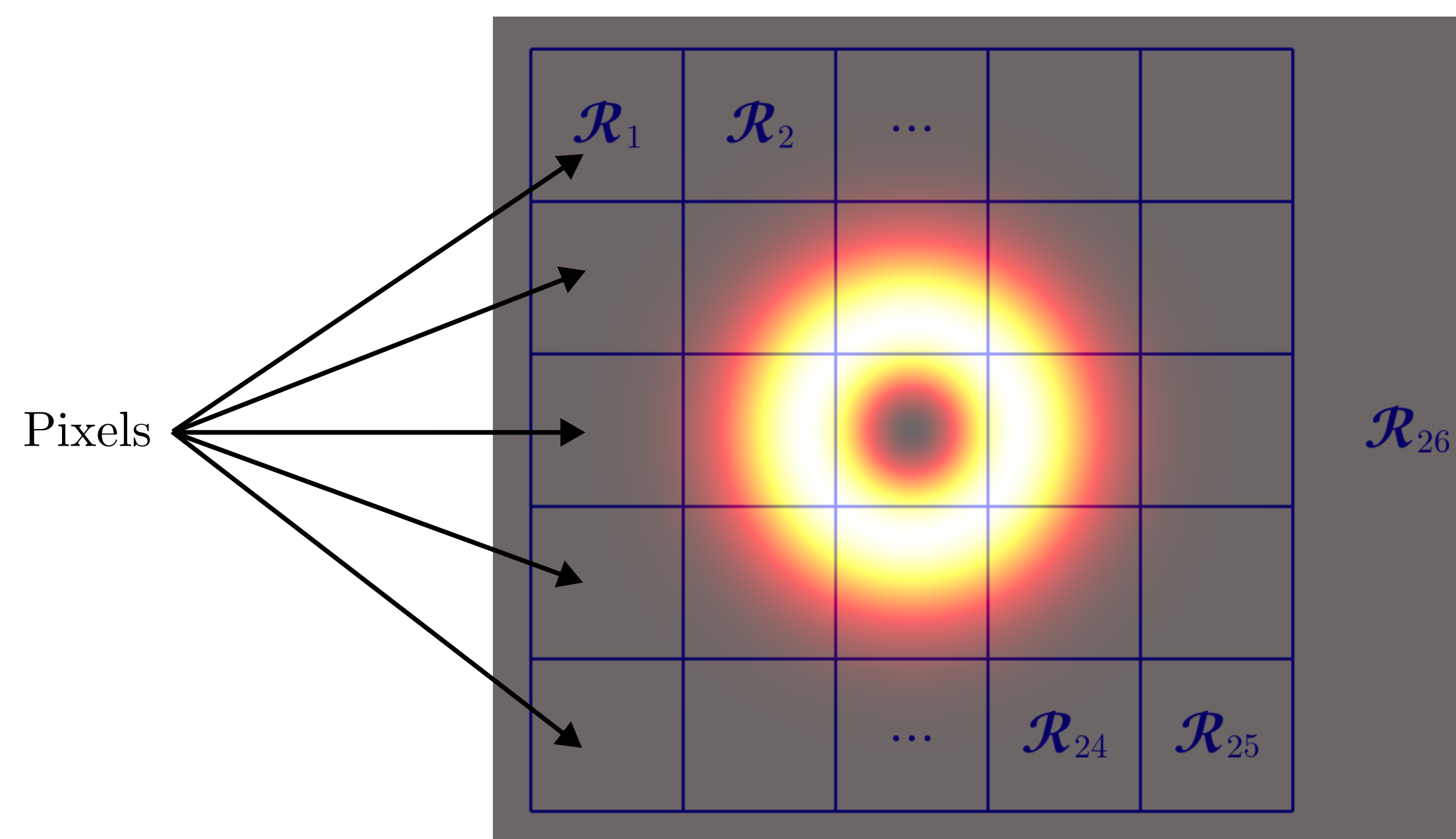
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Introduction

- The transverse structure of light has many important applications in areas such as quantum cryptography and optical tweezing, so its characterization is extremely important.
- We want to produce a linear combination of modes and recover the coefficients that specify it.
- **We strive for simplicity:** only use the camera as a measuring device.
- We identify the problem as a standard Quantum State Tomography, which gives us access to the tools of this well-developed field.
- We experimentally show the power of our method in both the intense and photocount regimes, including even mixed states.

What does a camera measure?



A photocount will happen at a pixel with probability

$$P(\mathcal{R}) = \int_{\mathcal{R}} \Gamma(\mathbf{r}, \mathbf{r}) d\mathbf{r}; \quad \text{Hermite-Gaussian modes: } u_j = HG_{j,N-j}$$

$$\Gamma(\mathbf{r}, \mathbf{r}') = \frac{\langle E^-(\mathbf{r}) E^+(\mathbf{r}') \rangle}{\int_{\mathbb{R}^2} \langle E^-(\mathbf{r}) E^+(\mathbf{r}) \rangle d\mathbf{r}} = \sum_{j,k=0}^N \rho_{jk} u_j^*(\mathbf{r}') u_k(\mathbf{r})$$

It is as if we have a quantum particle with position density matrix Γ and we measured

$$P(\mathcal{R}) = \langle F(\mathcal{R}) \rangle; \quad F(\mathcal{R}) = \int_{\mathcal{R}} |\mathbf{r}\rangle \langle \mathbf{r}| d\mathbf{r}$$

It's is just quantum state tomography!

A Positive Operator Valued Measure (POVM) is a set of positive operators $\{F_m\}$ that satisfy $\sum_m F_m = I$. They specify the probability of outcomes in an experiment: $p(m) = \text{Tr } \rho F_m$. We can see that our set $\{F_m = F(\mathcal{R}_m)\}$ is a POVM! Consider the linear operator

$$T : \text{Her}(\mathcal{H}) \rightarrow \mathbb{R}^M; \quad T(\Omega) = (\text{Tr } \Omega F_1, \dots, \text{Tr } \Omega F_M)$$

Space of modes of fixed order

Hermitian operator

POVM elements

Then, quantum state tomography reduces to the solution of the linear system

$$T\rho = \mathbf{p} \longrightarrow \begin{array}{l} \text{Detection probability at each pixel} \\ \text{Experimentally determined in the intense regime} \end{array}$$

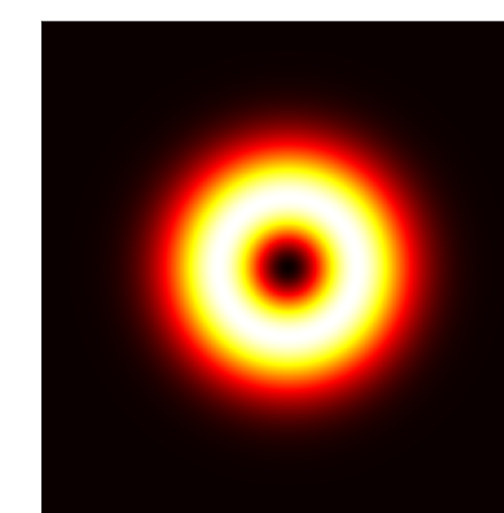
One needs a second image

We want that two different states produce two different experimental outcomes, so we may be able to distinguish them. In other words, we want the transformation T to be injective. We know that it isn't: **Laguerre Gaussian modes with opposite l have the same intensity.**

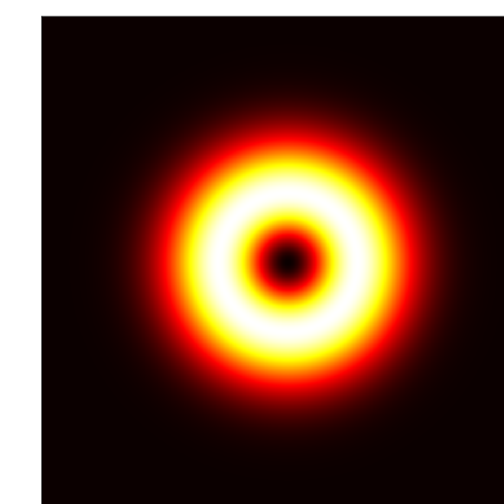
$$Y_{jk} = \frac{i}{\sqrt{2}} (|u_j\rangle \langle u_k| - |u_k\rangle \langle u_j|) \quad \begin{array}{l} \text{Hermite-Gaussian} \\ \text{modes are real-valued} \end{array} \quad \begin{array}{l} T \text{ is not} \\ \text{injective} \end{array}$$

$$\text{Tr } F(\mathcal{R}) Y_{jk} = \sqrt{2} \Im \int_{\mathcal{R}} u_j^*(\mathbf{r}) u_k(\mathbf{r}) d\mathbf{r} \stackrel{\downarrow}{=} 0 \Rightarrow T(Y_{jk}) \stackrel{\downarrow}{=} \mathbf{0}$$

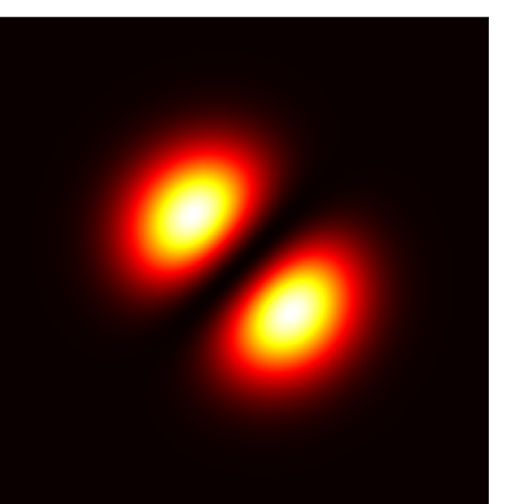
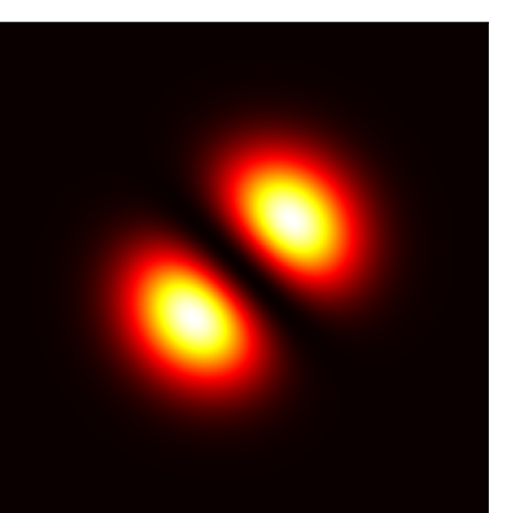
Solution: a mode converter implemented by a tilted cylindrical lens.



$$U_\theta = \sum_{m,n=0}^{\infty} e^{i(m-n)\theta} |HG_{mn}\rangle \langle HG_{mn}|$$



The angle can be adjusted experimentally. It is usually taken as $\pi/2$ but we will need other values.



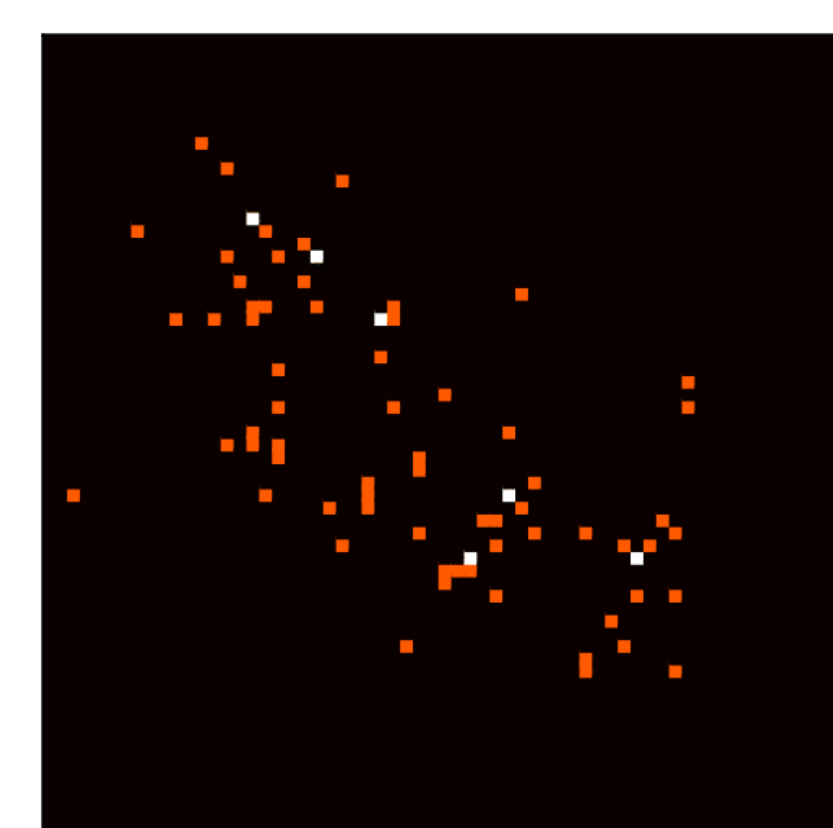
This generates a larger POVM $\{F_m/2, U_\theta^\dagger F_m U_\theta/2\}$ for which the transformation is shown to be injective.

Experimental results

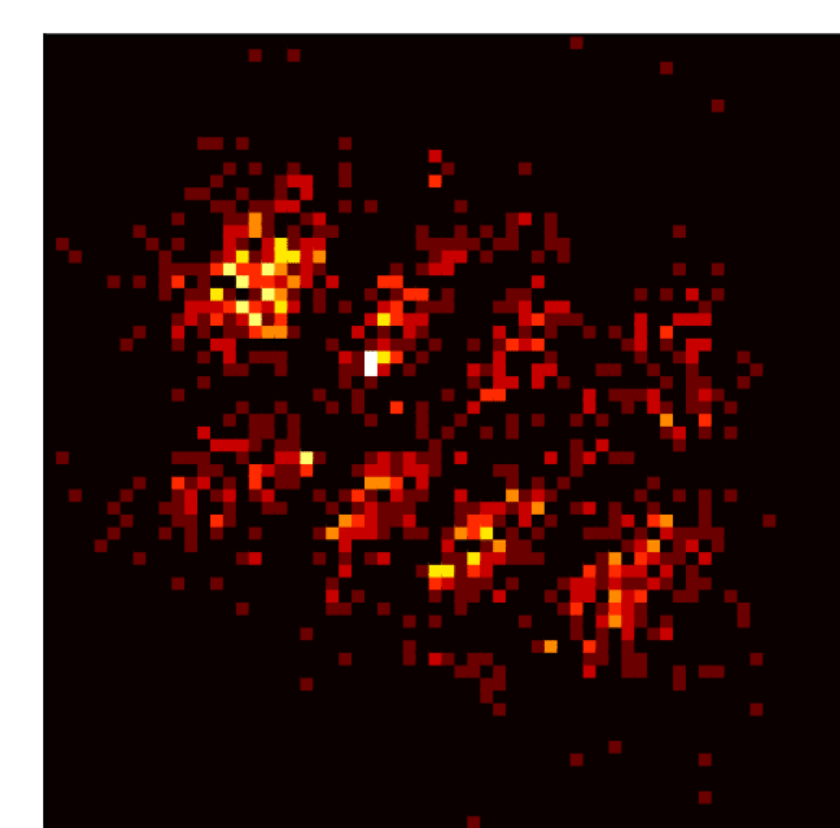
- Mixed states; intense regime; $\pi/6$ mode converter:

Method \ Order	1	2	3	4	5
Linear Inversion	0.998	0.995	0.992	0.989	0.983
Machine Learning	0.995	0.988	0.985	0.979	0.968

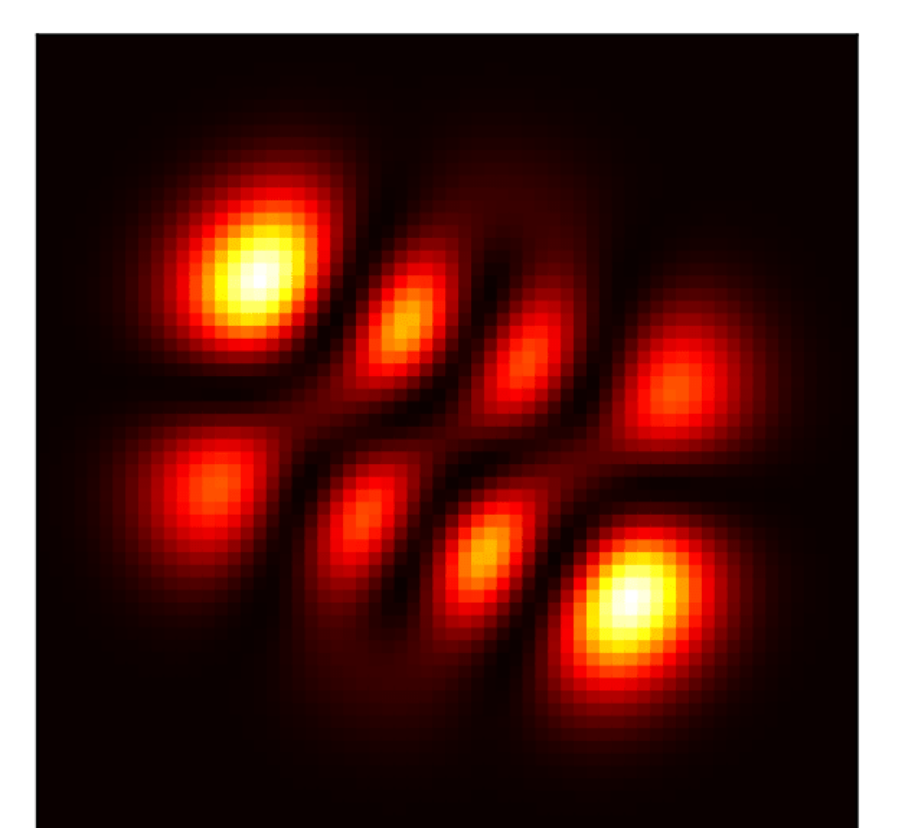
- Pure states; photocount regime: $\pi/2$ mode converter



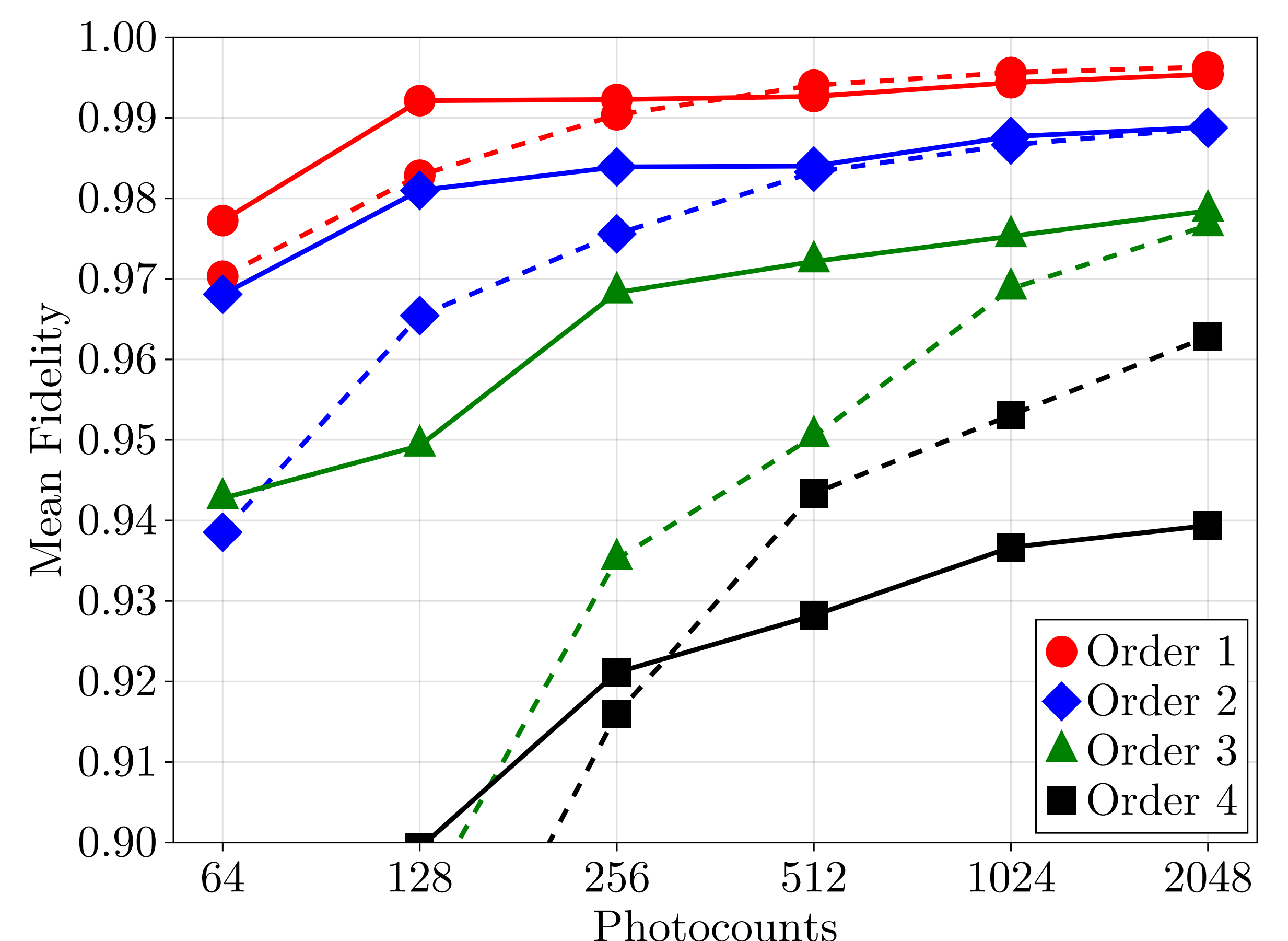
128 counts



2048 counts



Intense simulation



Full lines: machine learning; dashed lines: Bayesian tomography

Funding:

