

Information-Driven Design of Imaging Systems

Henry Pinkard Leyla Kabuli Eric Markley Tiffany Chien Jiantao Jiao Laura Waller

UC Berkeley Department of Electrical Engineering and Computer Sciences

Why?

Modern imaging systems computationally process raw measurements
So capturing lots of information is more important than capturing a pretty, human-interpretable picture

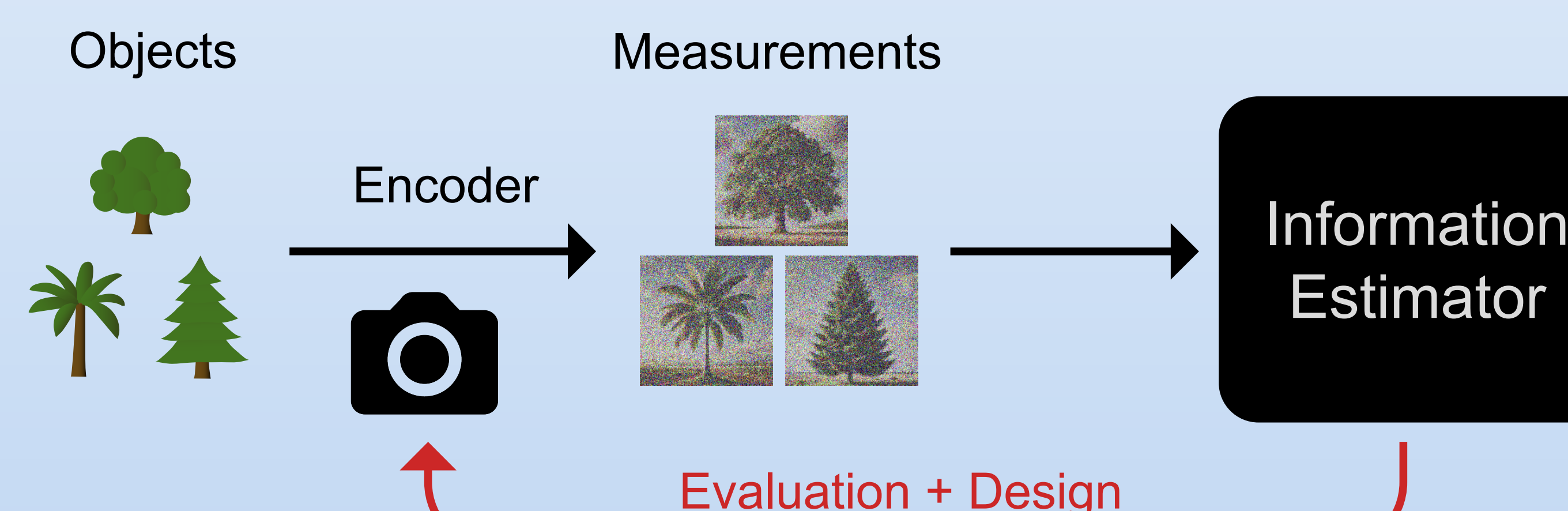
Traditional: human interpretation

Modern: Computational processing



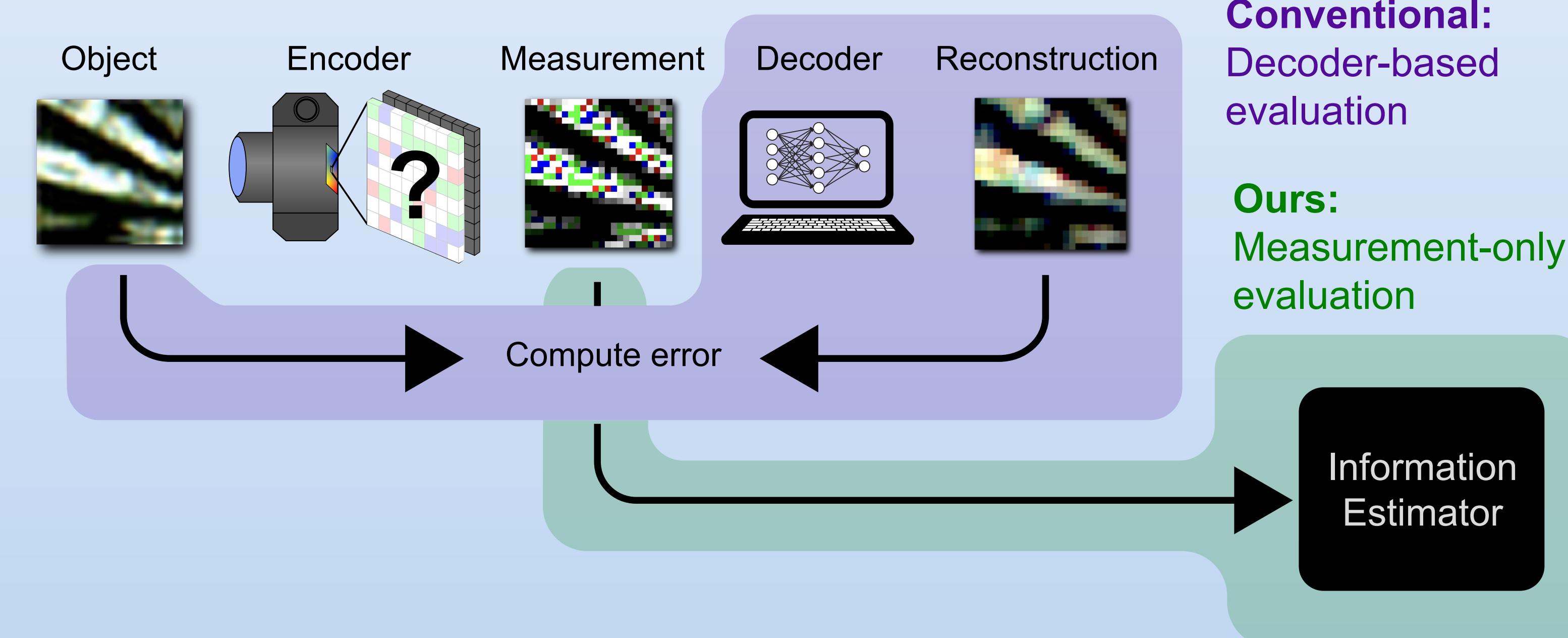
What?

We developed an **Information estimator**, which enables the evaluation and design of imaging based on information capture, not visual appearance



For example?

What is the best color filter mask for photography?



How?

Decompose estimation into tractable subproblems

$$I(\mathbf{X}; \mathbf{Y}) = H(\mathbf{Y}) - H(\mathbf{Y} | \mathbf{X})$$

Information in measurements = Diversity of measurements - Diversity of noise alone

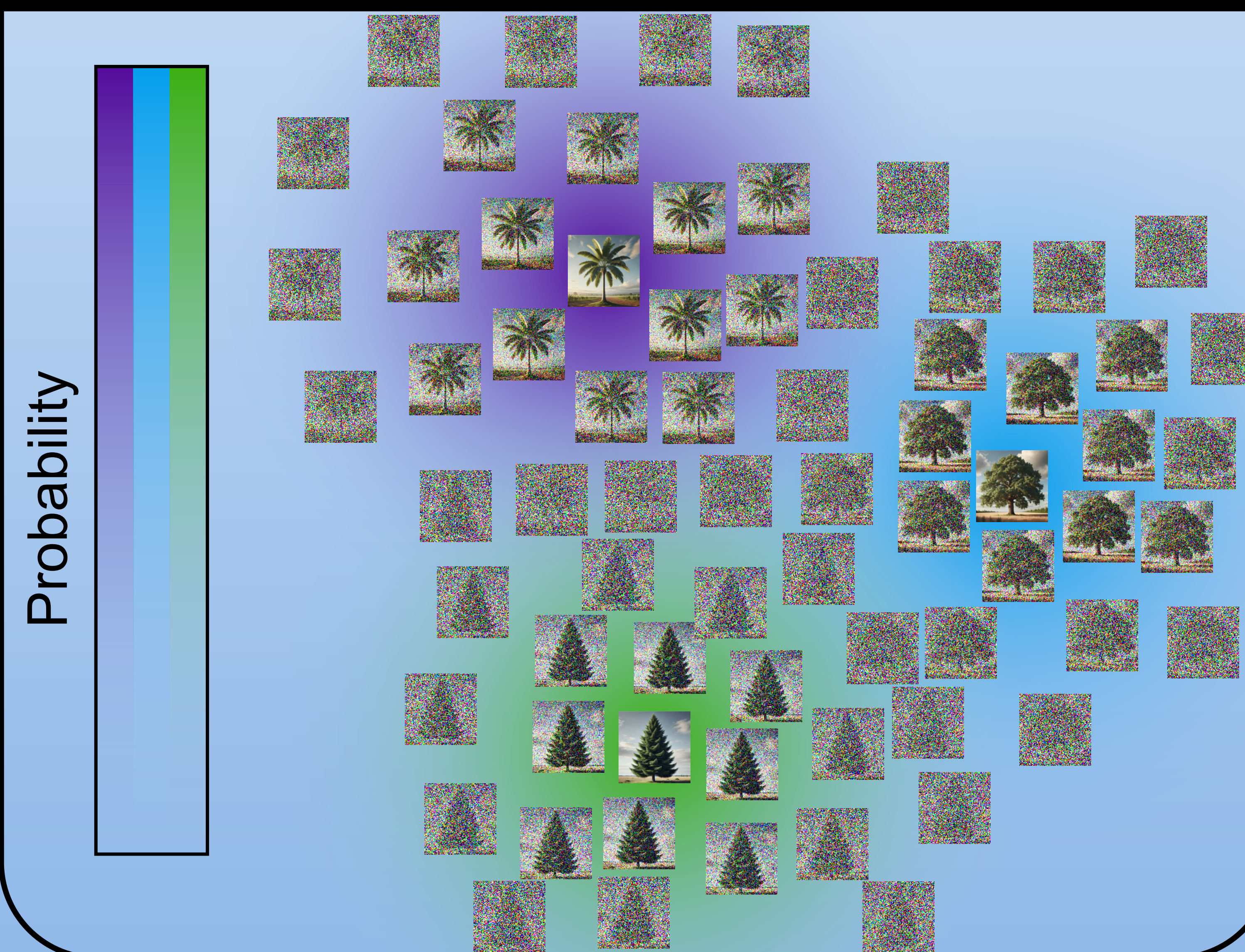
Upper bound by fitting probabilistic model

$$H(\mathbf{Y}) = \mathbb{E}[-\log p_{\theta}(\mathbf{Y})] \geq H(\mathbf{Y}) = \mathbb{E}[-\log p(\mathbf{Y})]$$

Analytically calculate using physics-based noise model (e.g. Poisson noise)

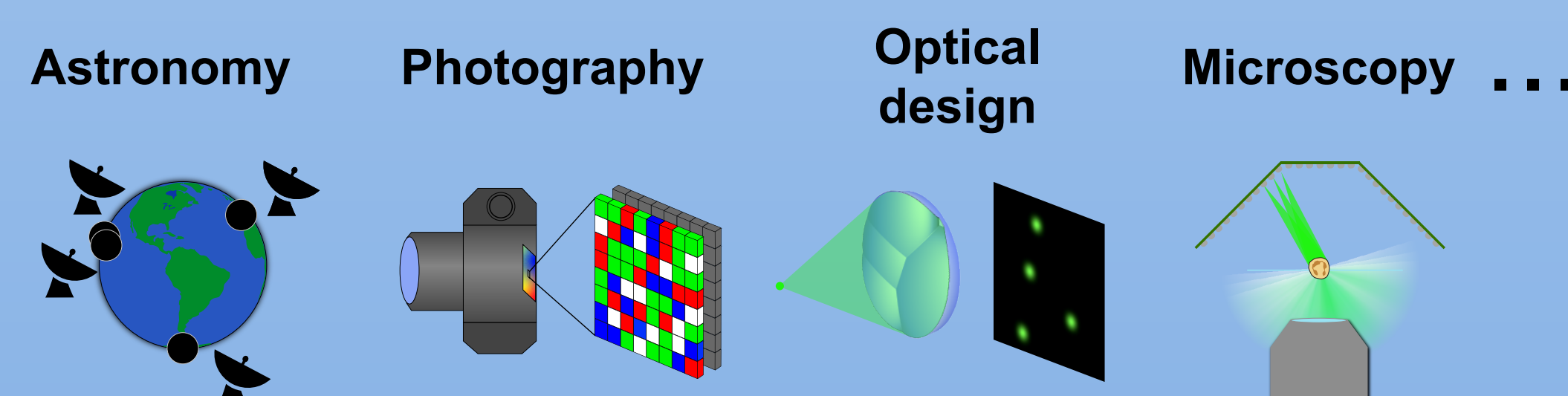
$$H(\mathbf{Y} | \mathbf{X}) \approx \frac{1}{N} \sum_{i=1}^N \sum_{k=1}^D \frac{1}{2} \log_2 (2\pi e x_k^{(i)})$$

Information quantifies discernibility

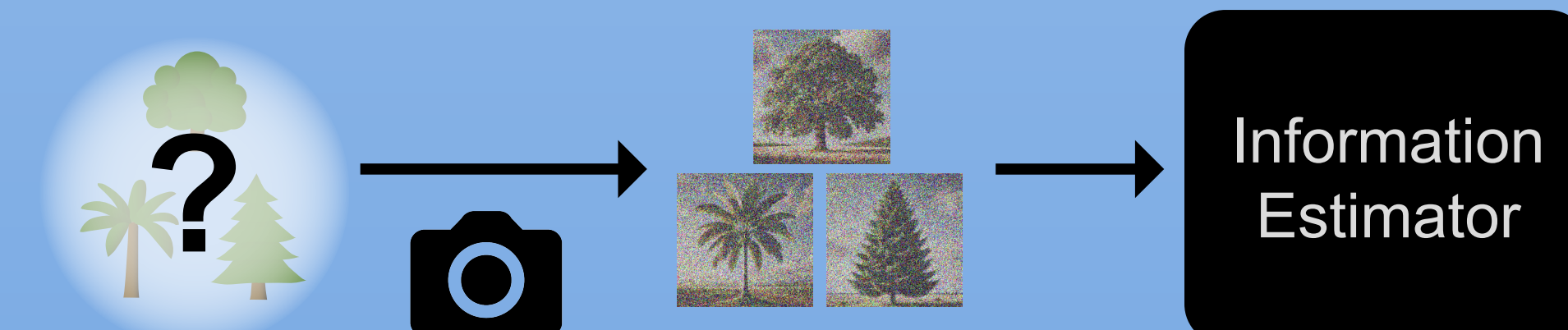


Advantages?

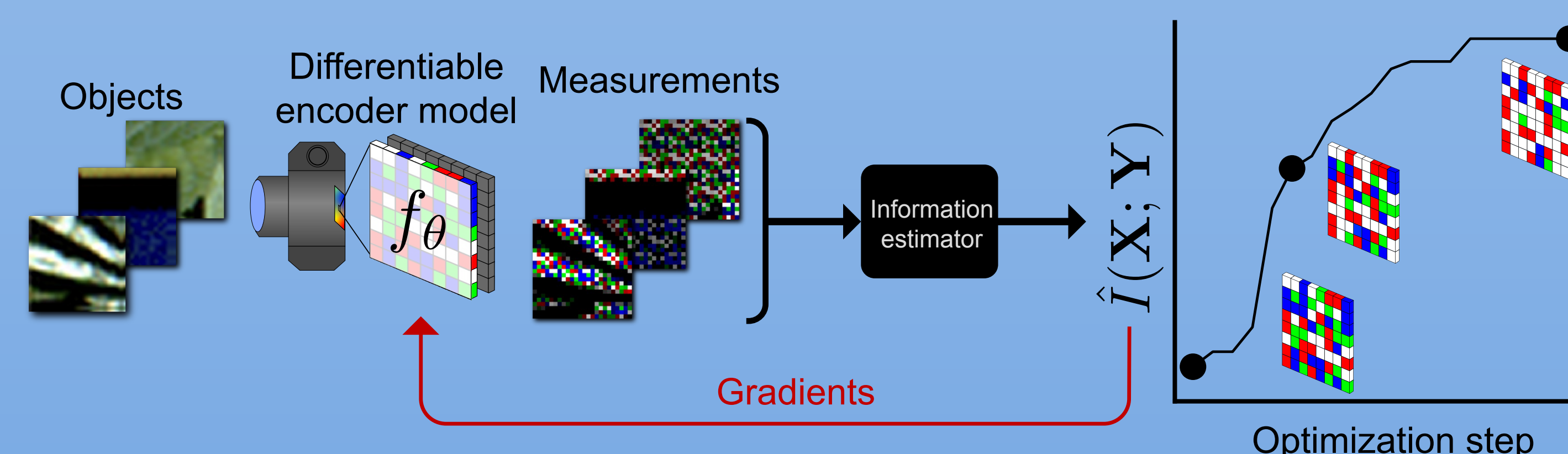
1) Broadly applicable



2) Ground truth-free → Field-deployable



3) Optimizable + computationally efficient



Does it work?

Yes! Information estimates consistently predicted task performance across applications

