

Information-Driven Design of Imaging Systems

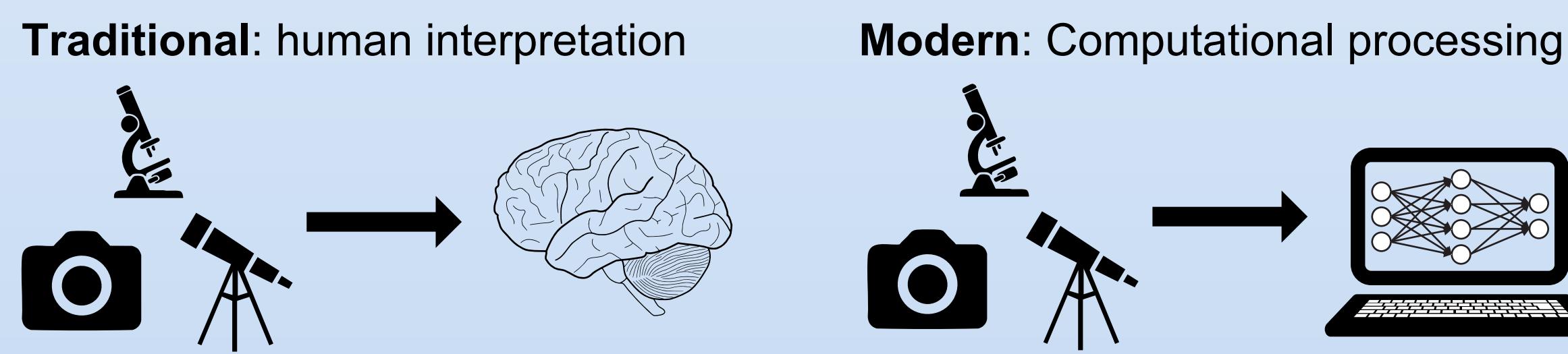
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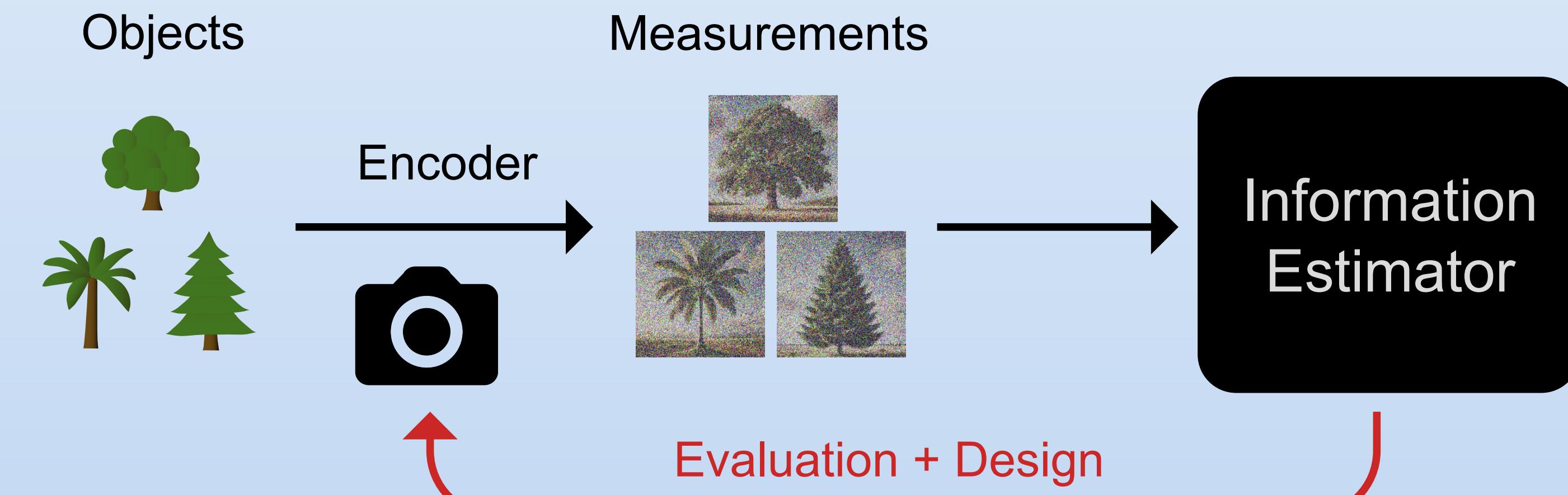
Why?

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



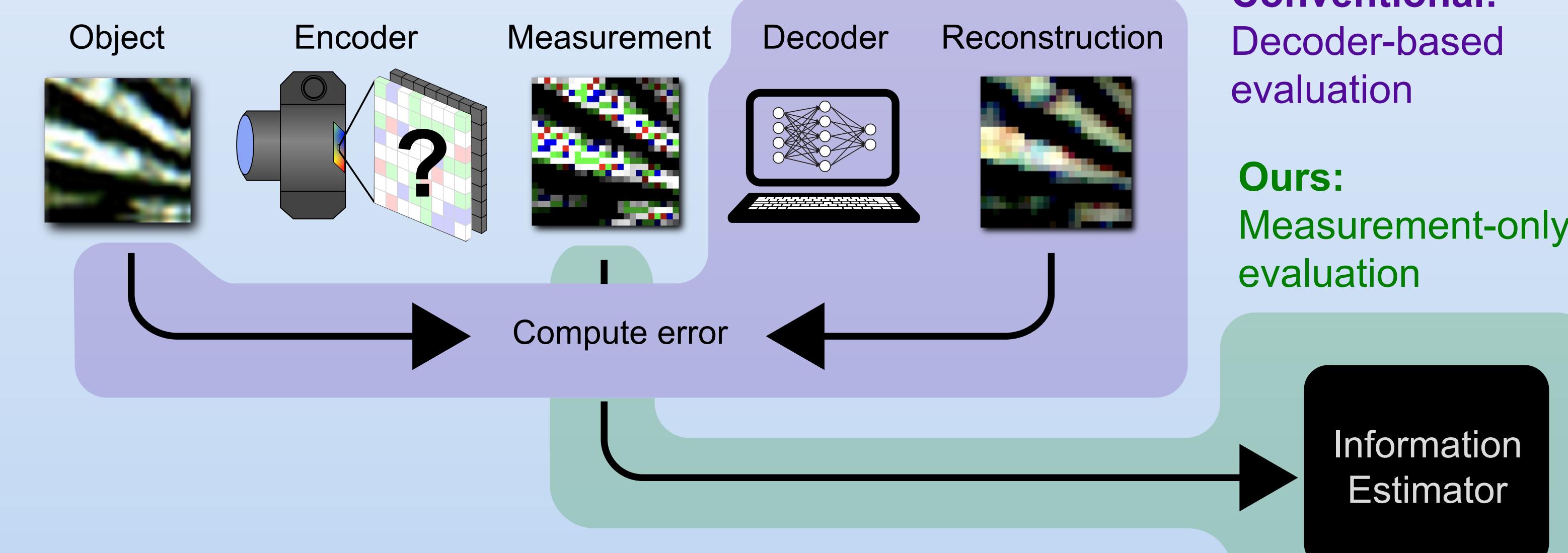
What?

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

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

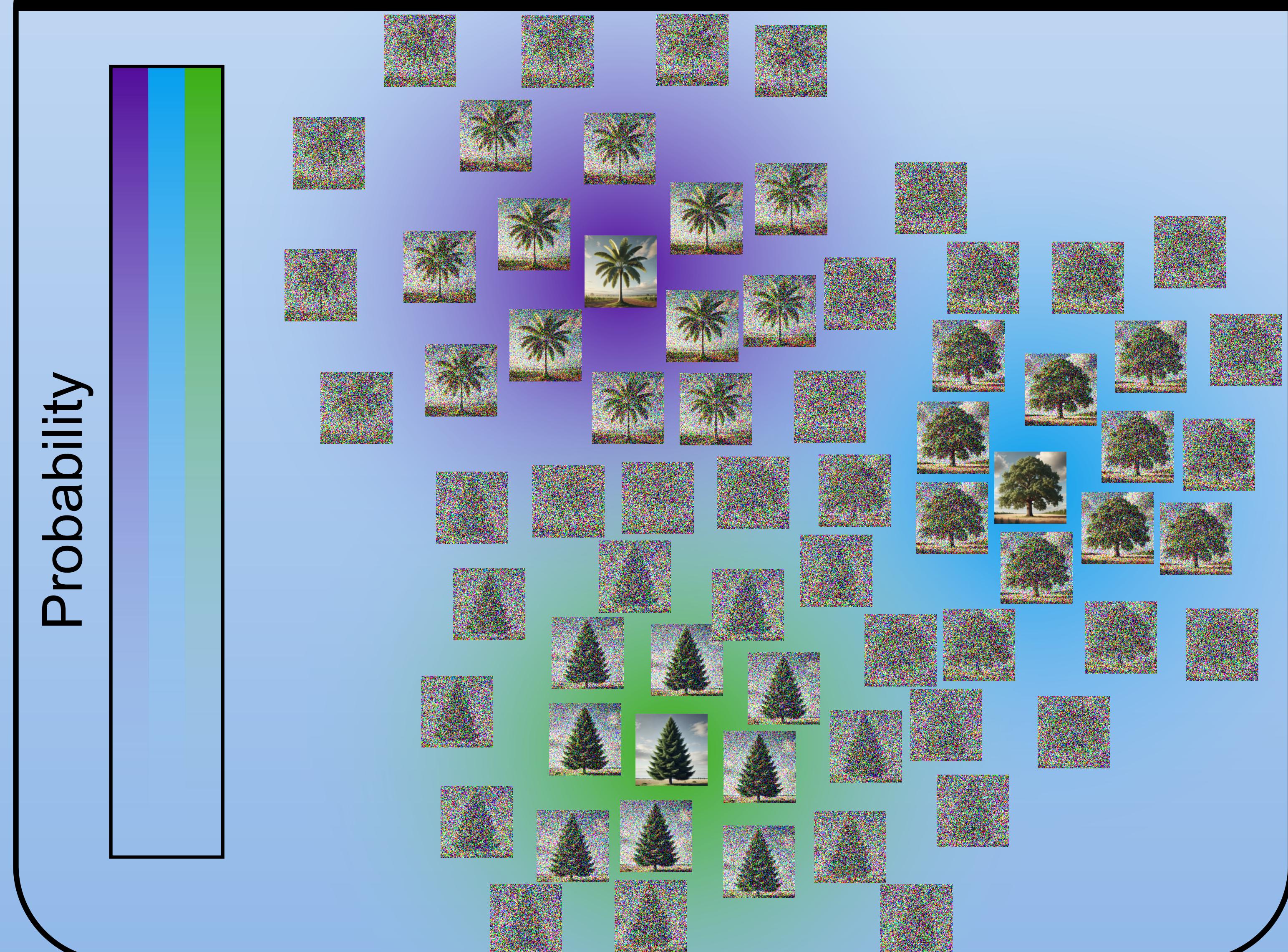
Information in measurements Diversity of measurements Diversity of noise alone

Upper bound by fitting probabilistic model Analytically calculate using physics-based noise model (e.g. Poisson noise)

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

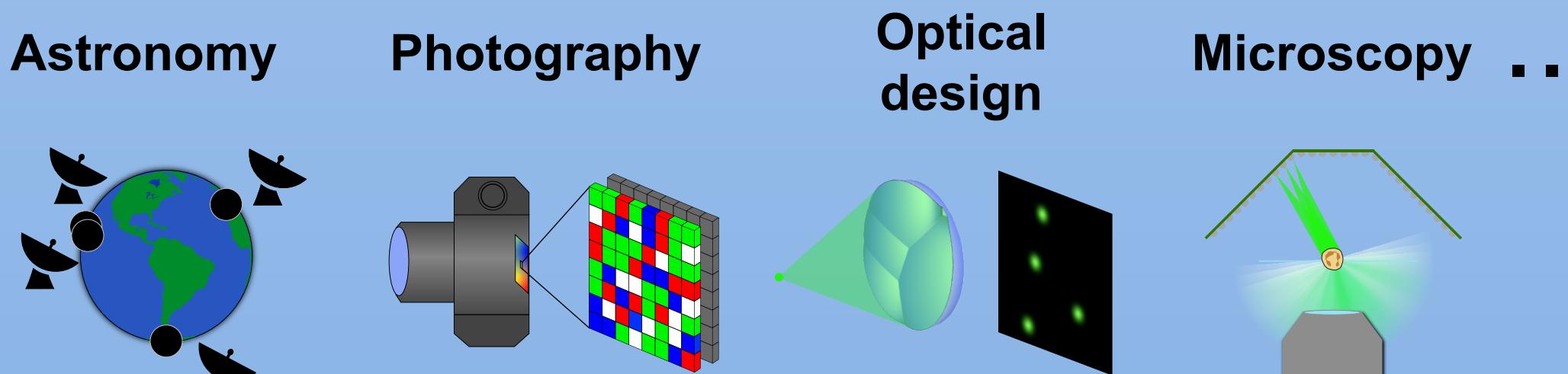
$E[-\log p_\theta(\mathbf{Y})] \geq H(\mathbf{Y}) = E[-\log p(\mathbf{Y})]$

Information quantifies discernibility



Advantages?

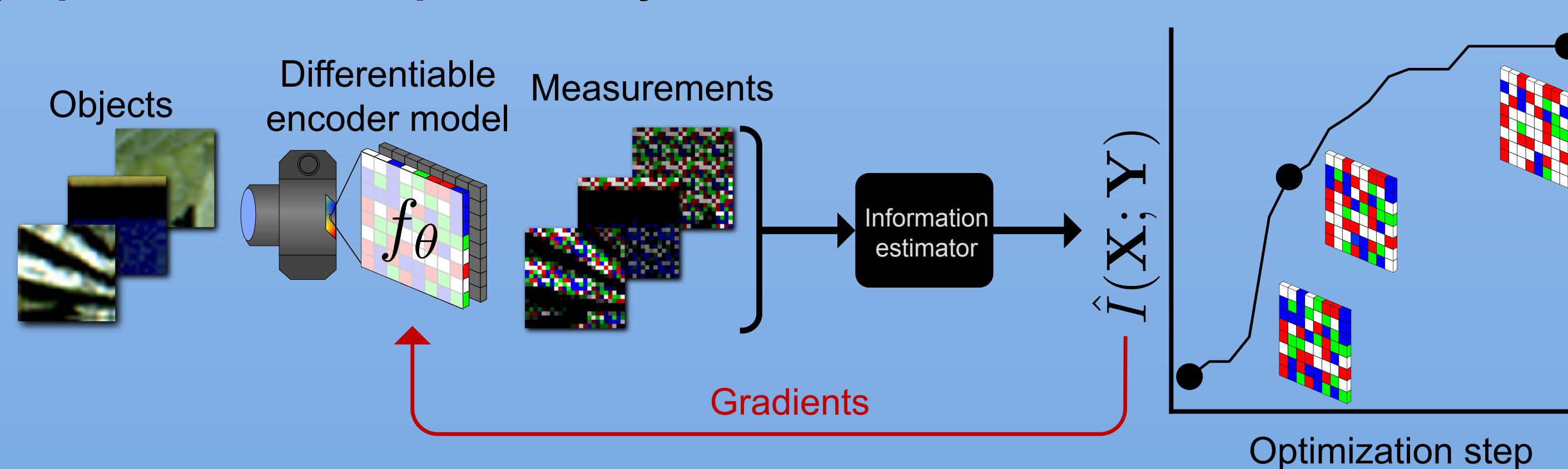
1) Broadly applicable



2) Ground truth-free → Field-deployable



3) Optimizable + computationally efficient



Does it work?

Yes! Information estimates consistently predicted task performance across applications

