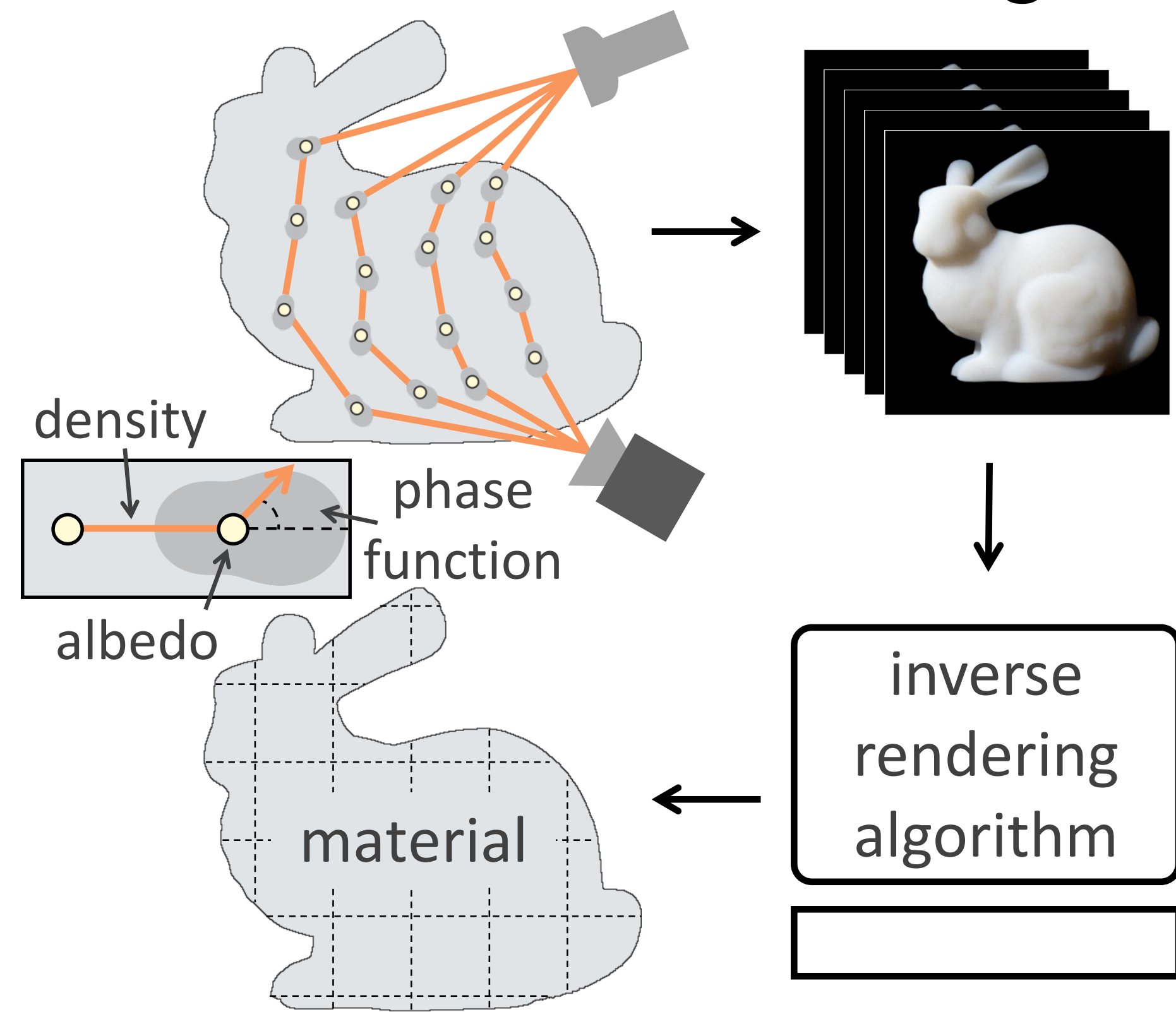


An Evaluation of Computational Imaging Techniques for Heterogeneous Inverse Scattering

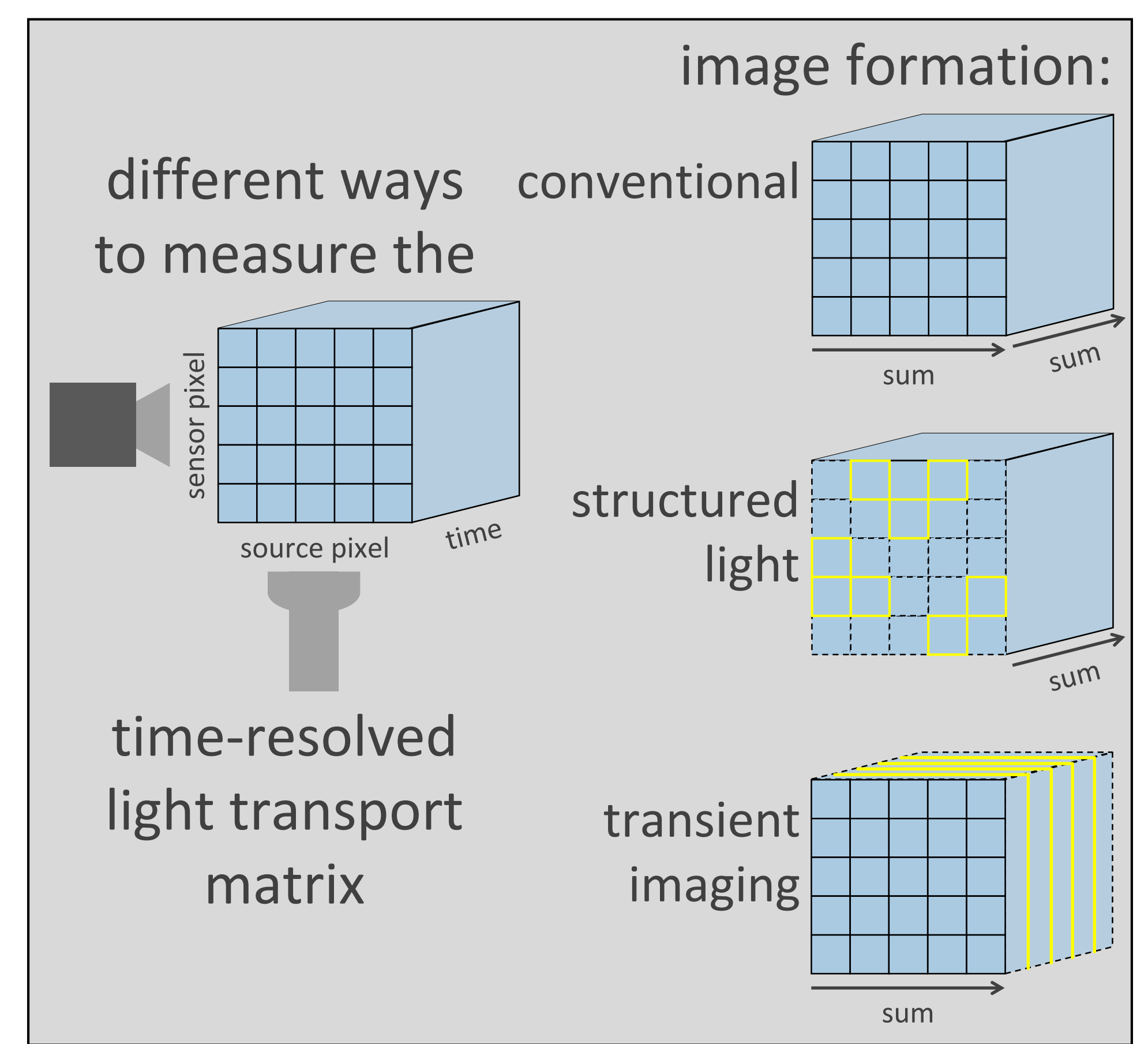
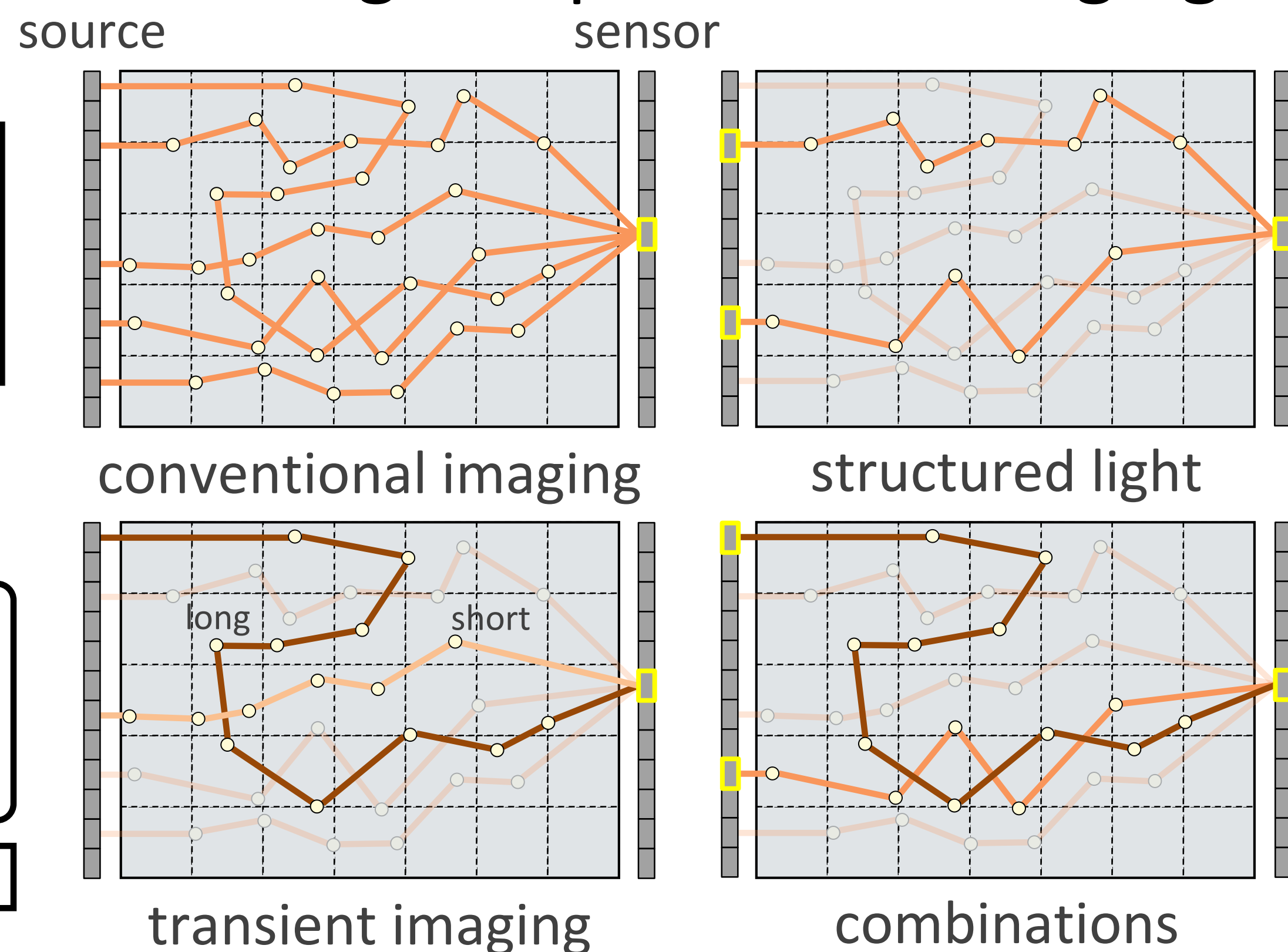


Ioannis Gkioulekas (Harvard), Anat Levin (Technion, Weizmann Institute), Todd Zickler (Harvard)

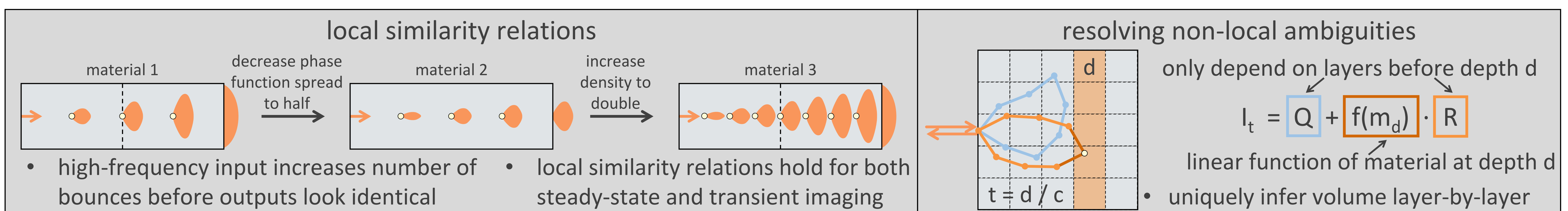
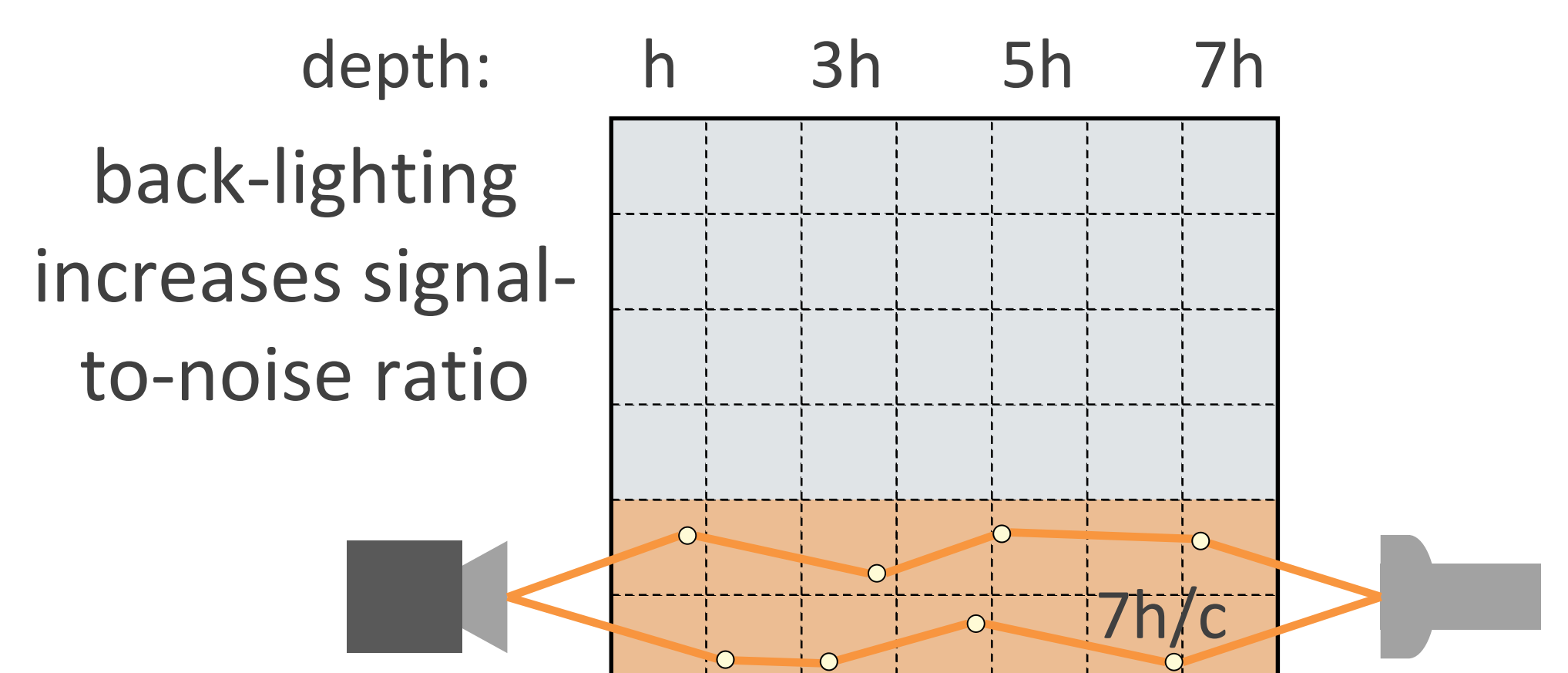
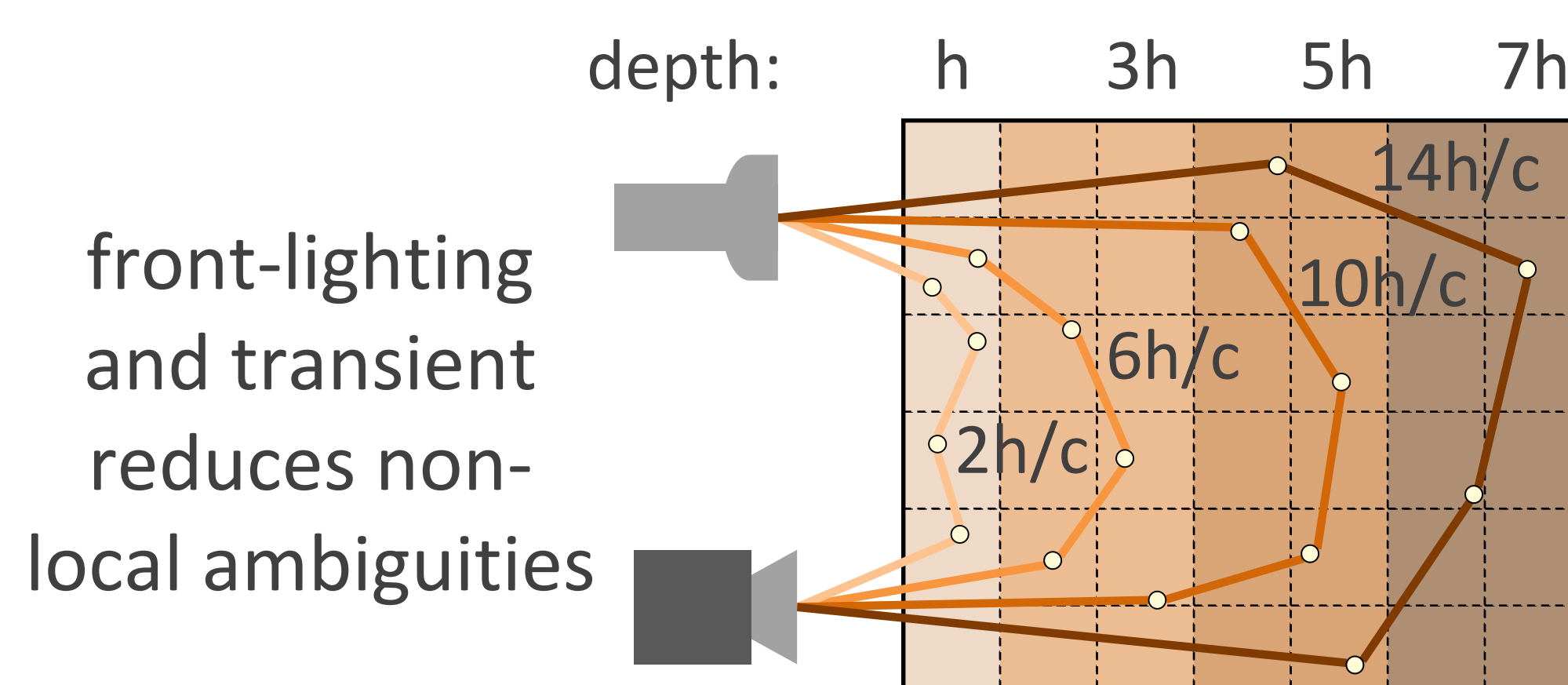
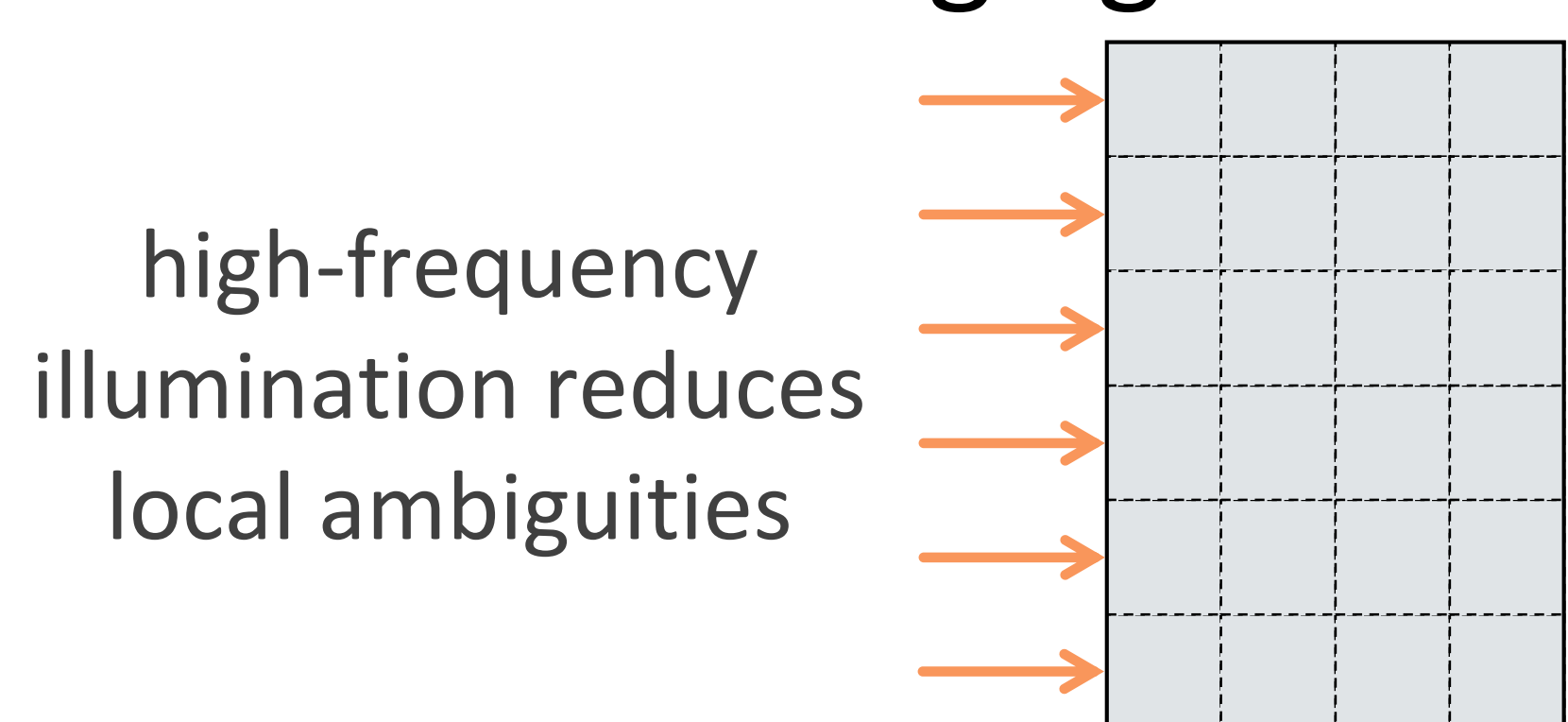
We do inverse scattering...



...using computational imaging



What kind of imaging to use?



Appearance matching optimization

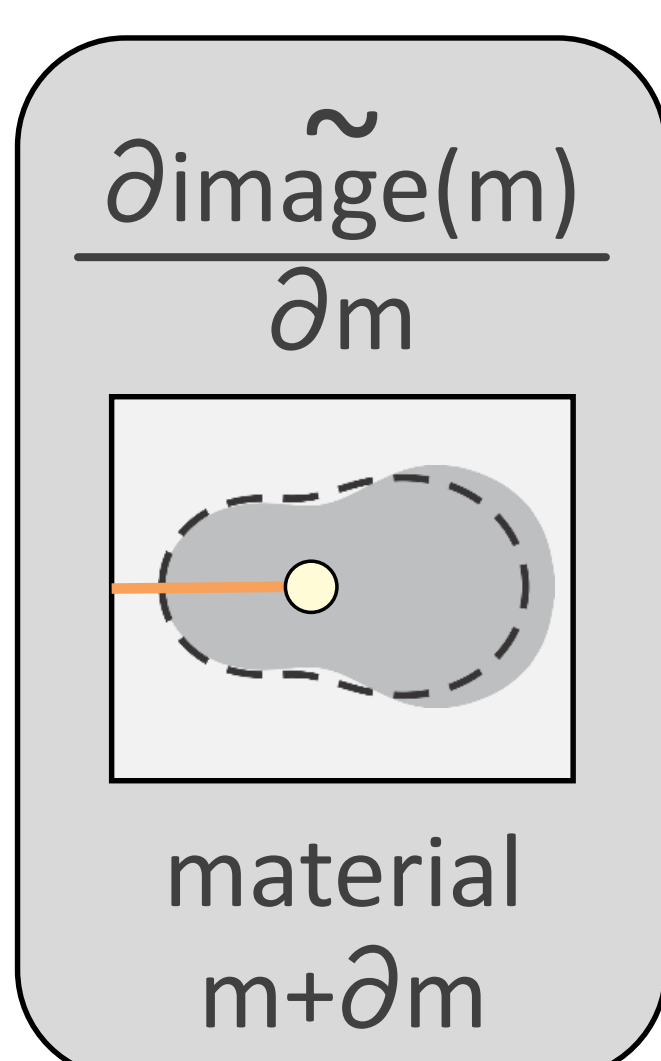
$$\min_m \| \text{image}(m) - \text{target} \|^2$$

challenging optimization problem:

- high-dimensional (10^5 unknowns)
- very nonlinear ("image" function)

stochastic gradient descent

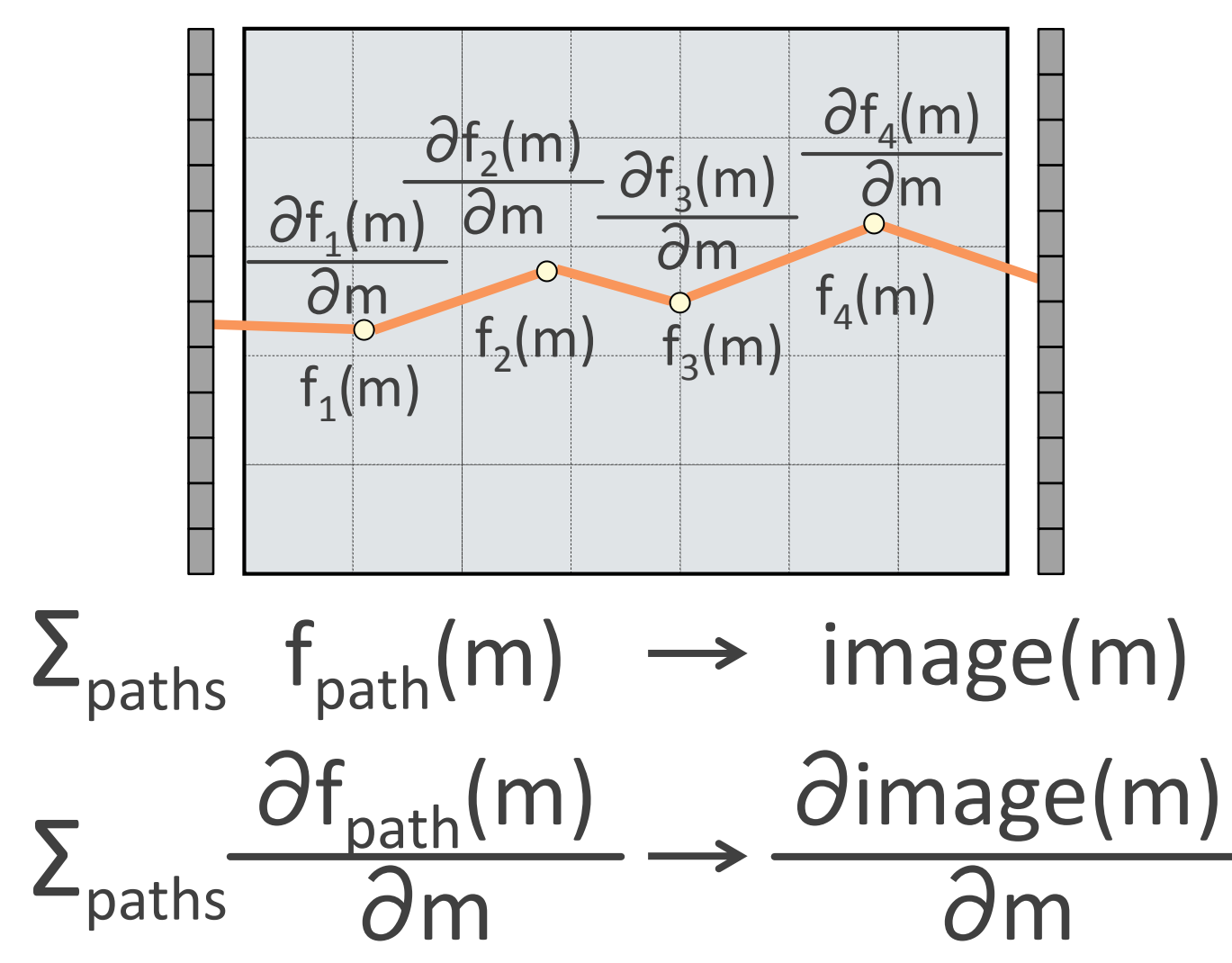
while (not converged)
update m using $\frac{\partial \text{loss}(m)}{\partial m}$



Monte Carlo gradient rendering

algorithm sketch

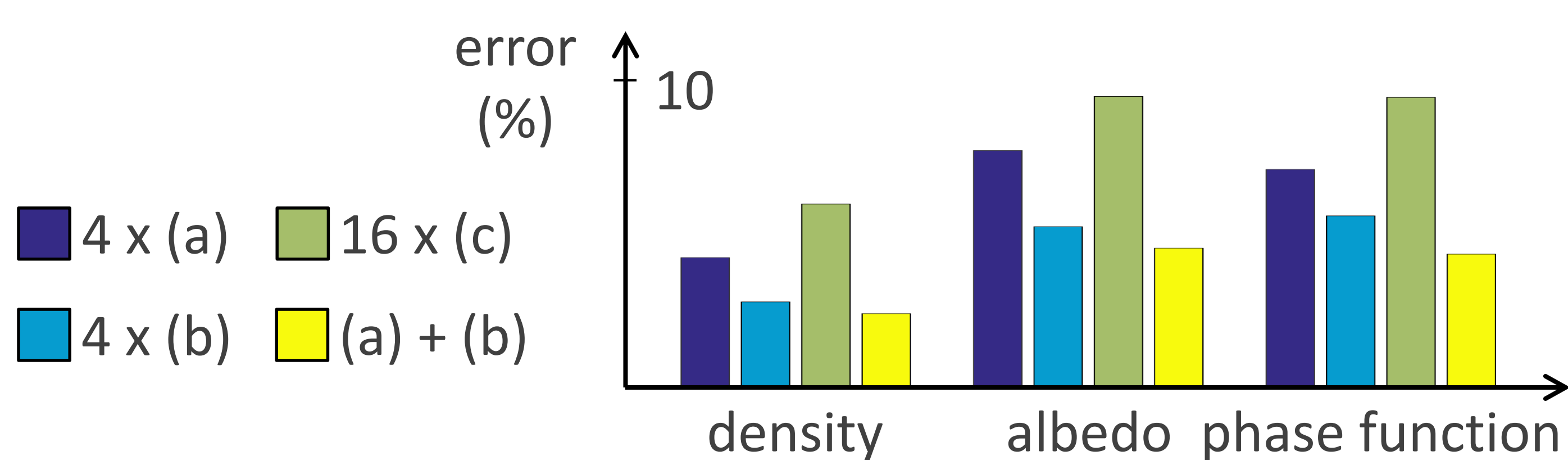
- while (not converged)
 - randomly sample a photon path
 - compute path-segment terms
 - radiance throughput when rendering images
 - throughput gradient when rendering gradients
 - aggregate terms into whole path contribution
 - update image estimate



computation-noise trade-off:
 10^6 paths 3 hours 10^2 paths 10 secs

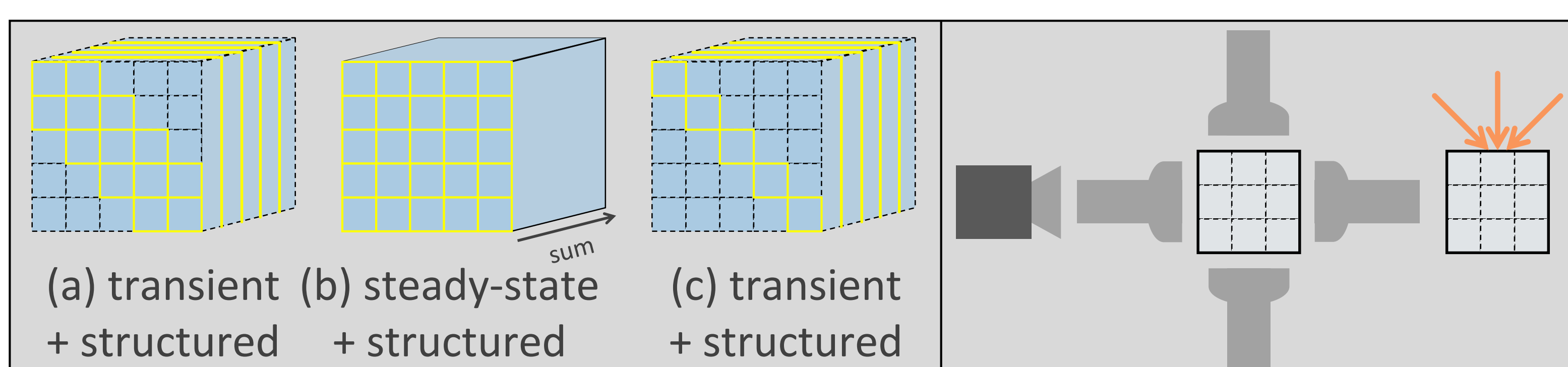
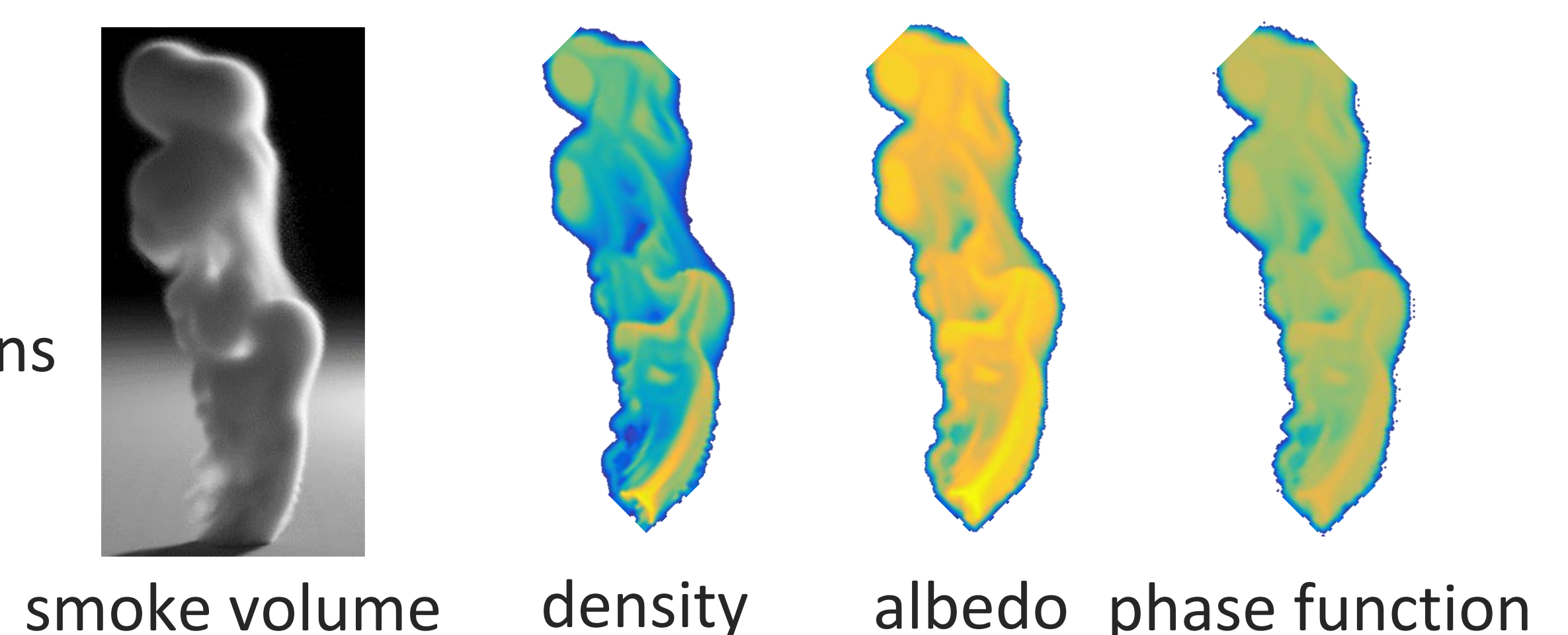
- physically accurate (accounts for all scattering)
- generally applicable (all above imaging types)
- scalable (highly parallelizable)

Empirical evaluation of imaging types



Large-scale inverse scattering simulation

- 10^4 cores
- 10^6 images
- 10^5 unknowns



Code online soon!

try with your own imaging configurations



http://vision.seas.harvard.edu/inverse_transient