

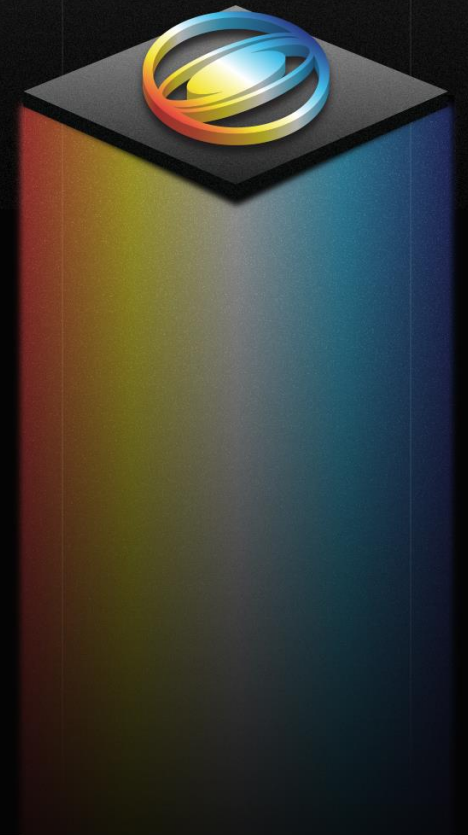


SIGGRAPH 2024
DENVER+ 28 JUL — 1 AUG

THE PREMIER CONFERENCE
& EXHIBITION ON
COMPUTER GRAPHICS &
INTERACTIVE TECHNIQUES

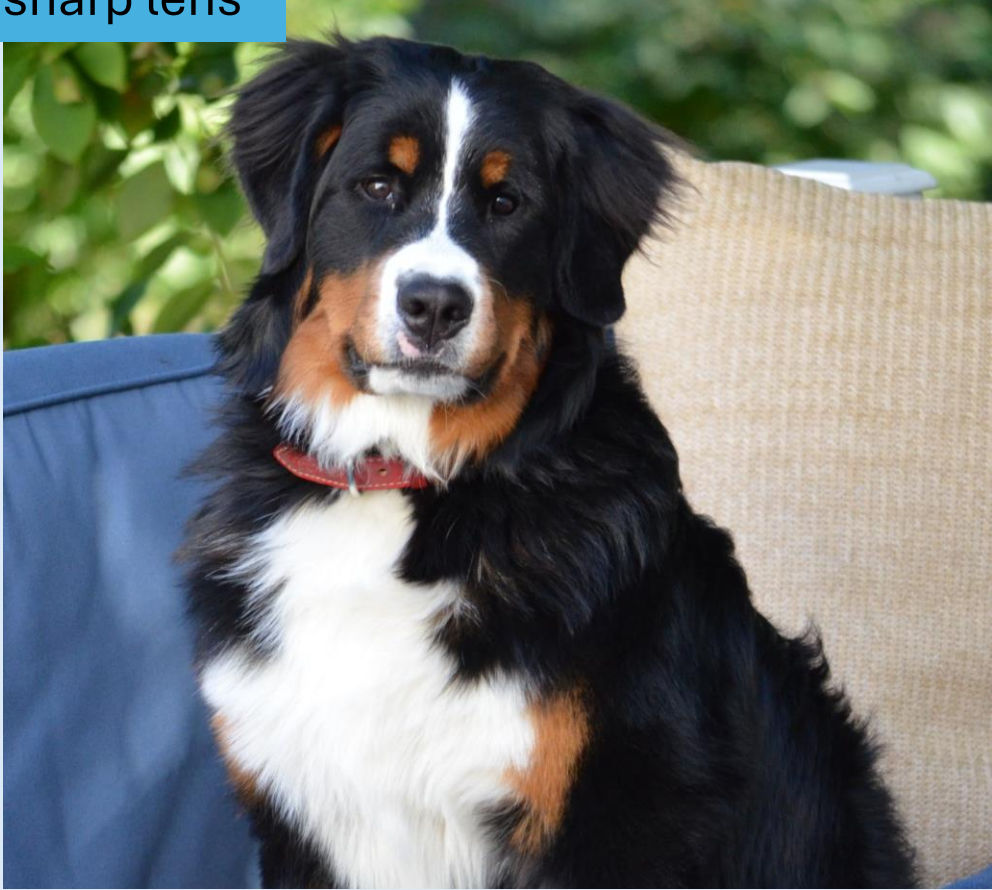
APERTURE-AWARE LENS DESIGN

ARJUN TEH
IOANNIS GKIOULEKAS
MATTHEW O'TOOLE

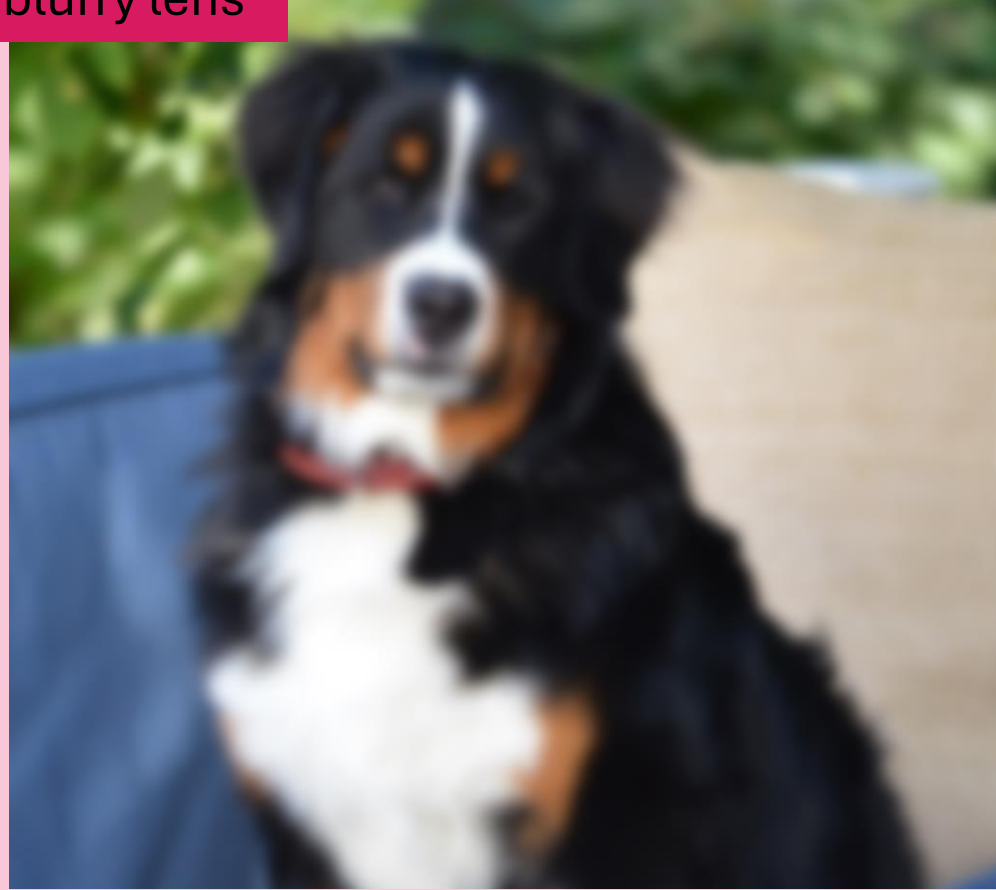


What makes a good lens?

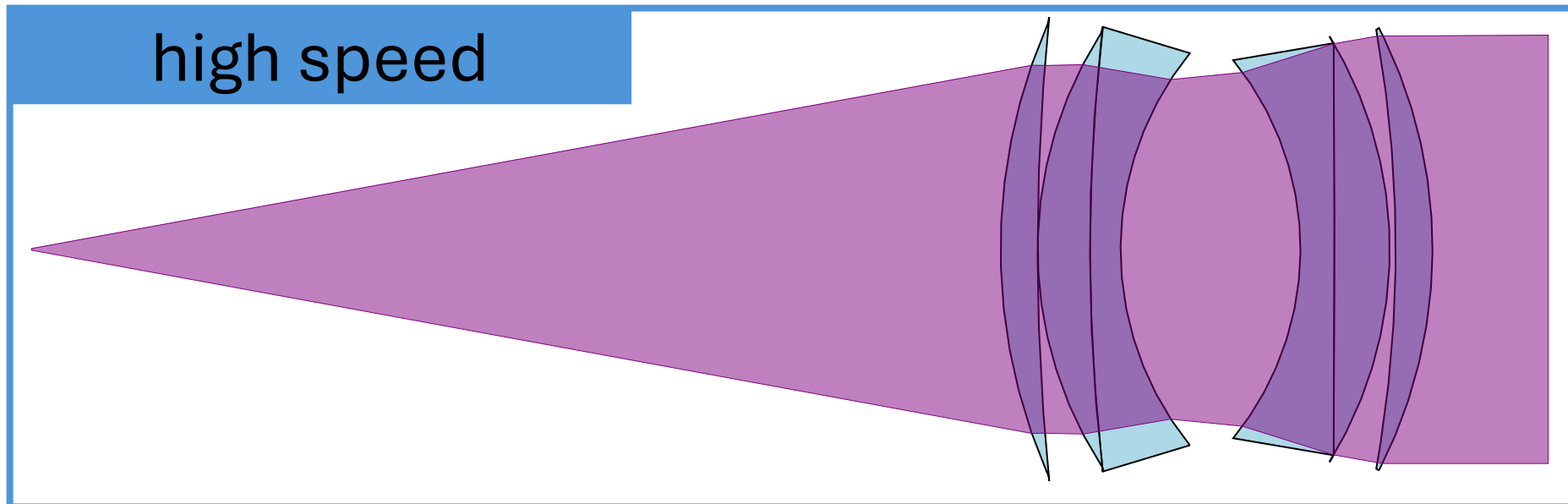
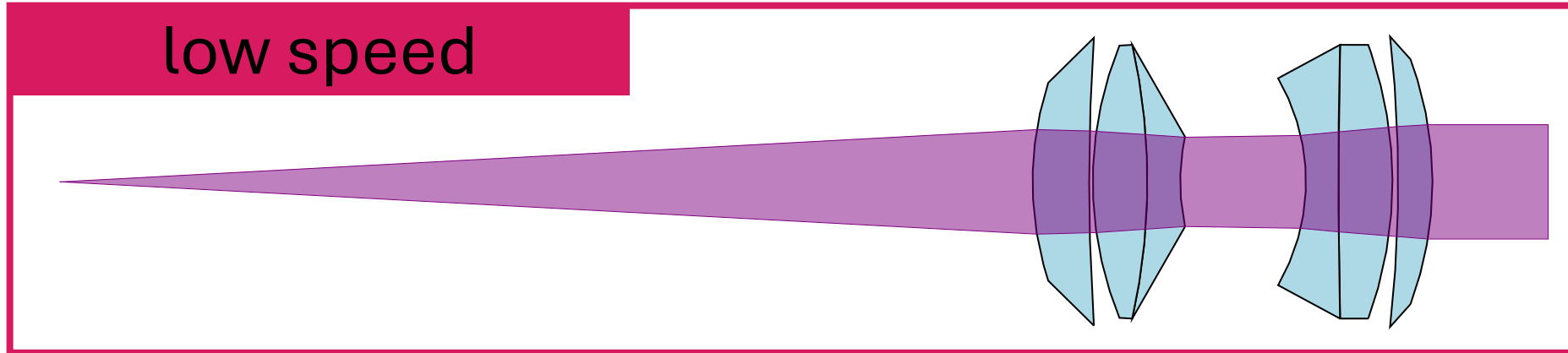
sharp lens



blurry lens



Lens speed is also important



Lens speed is also important

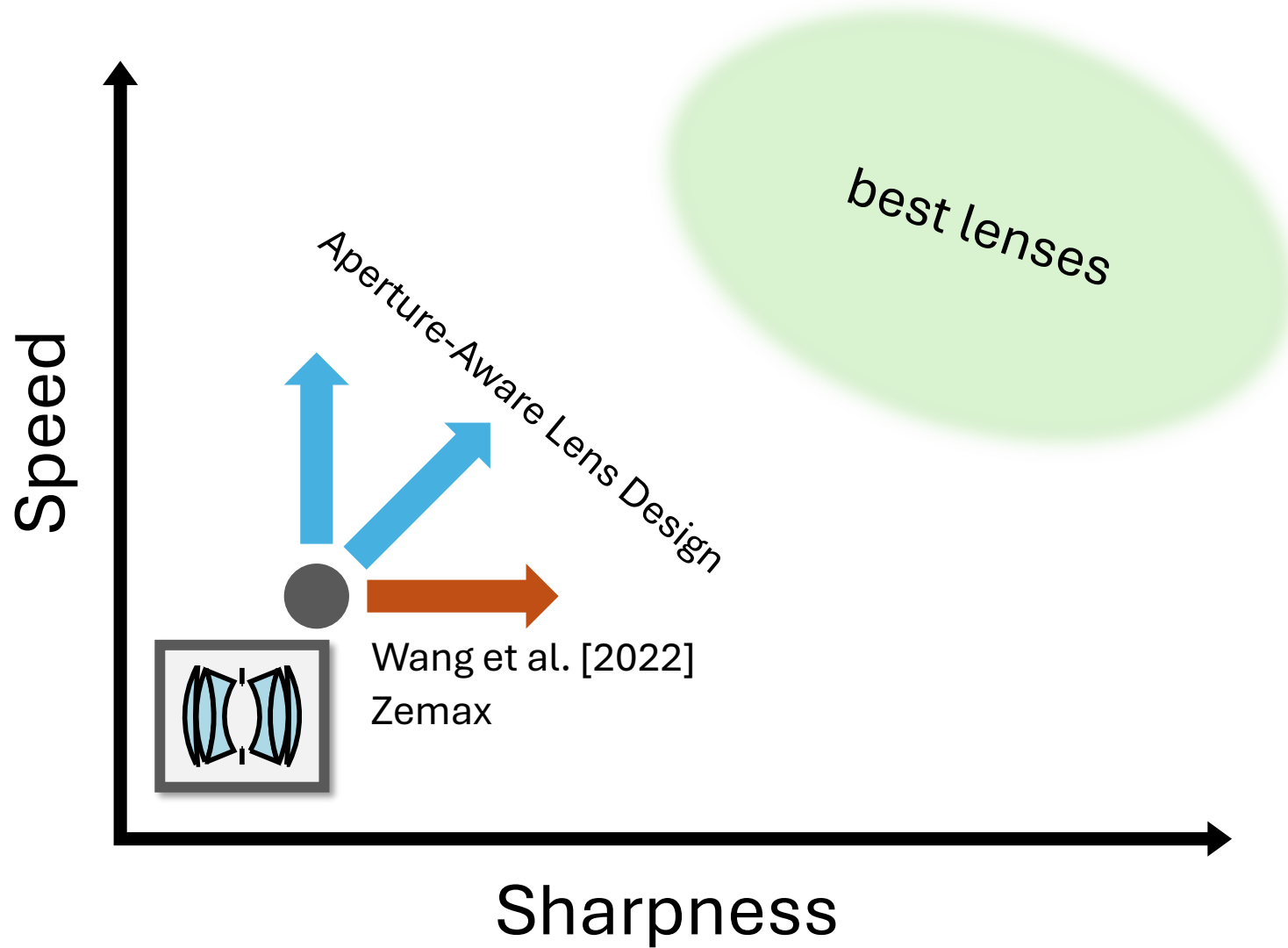


Motion blur

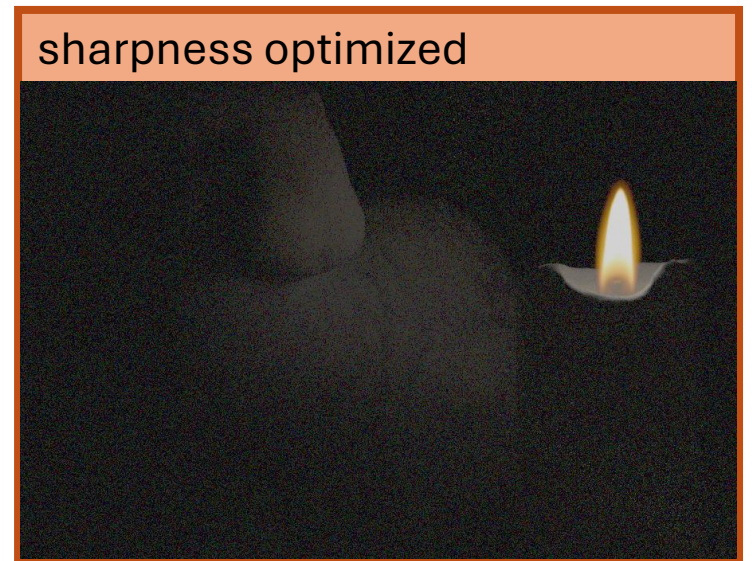
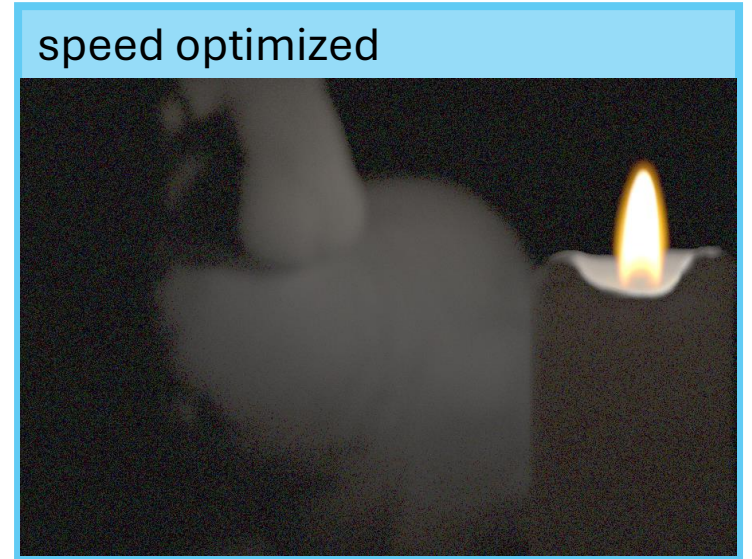
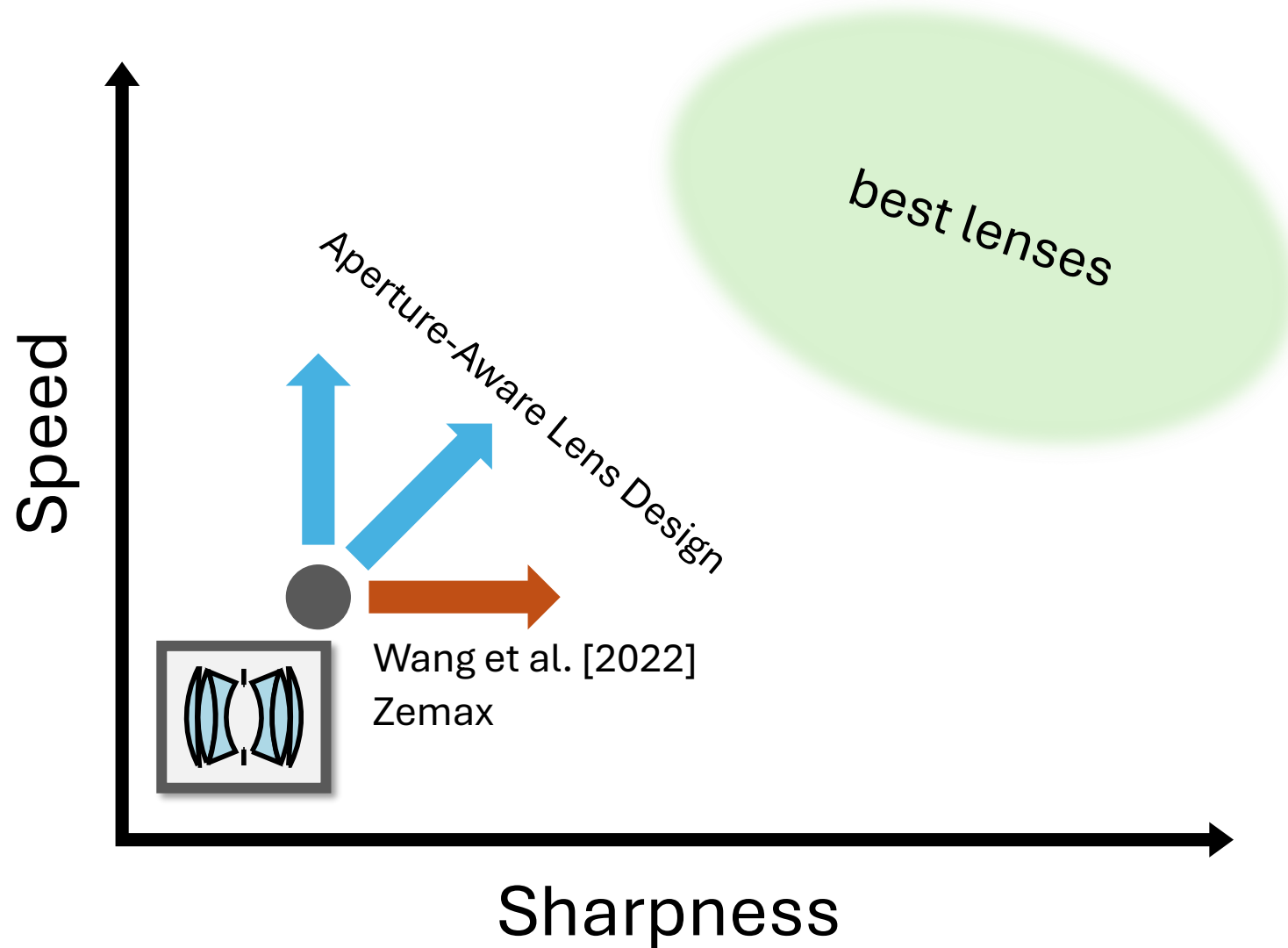


Low light scenes

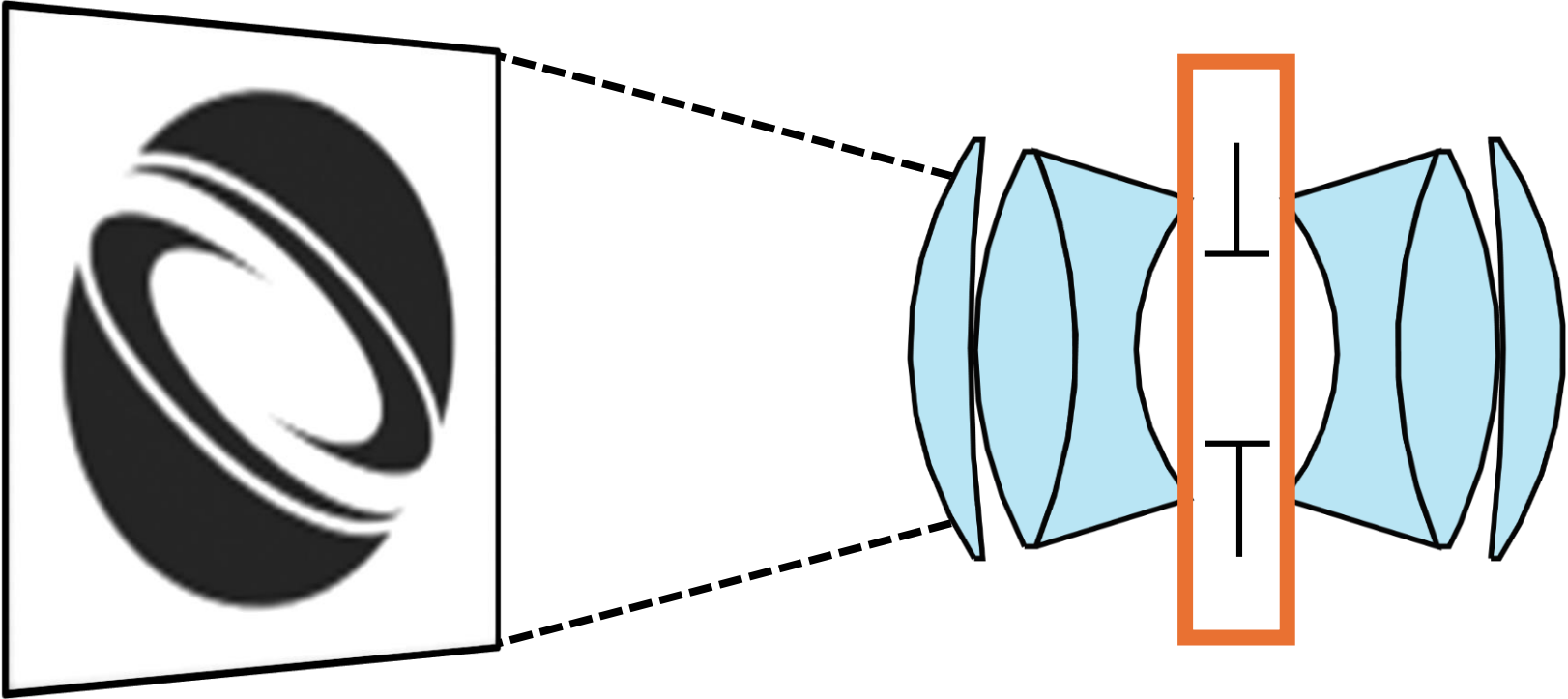
Sharpness vs speed



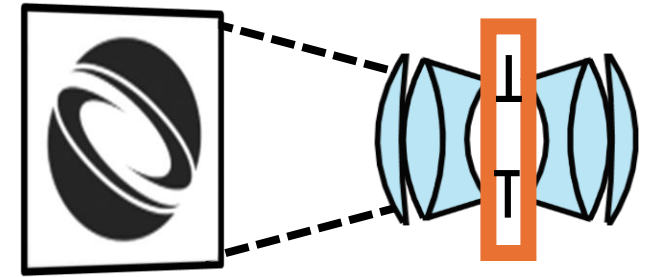
Sharpness vs speed



Autodiff can't capture lens speed



Autodiff can't capture lens speed



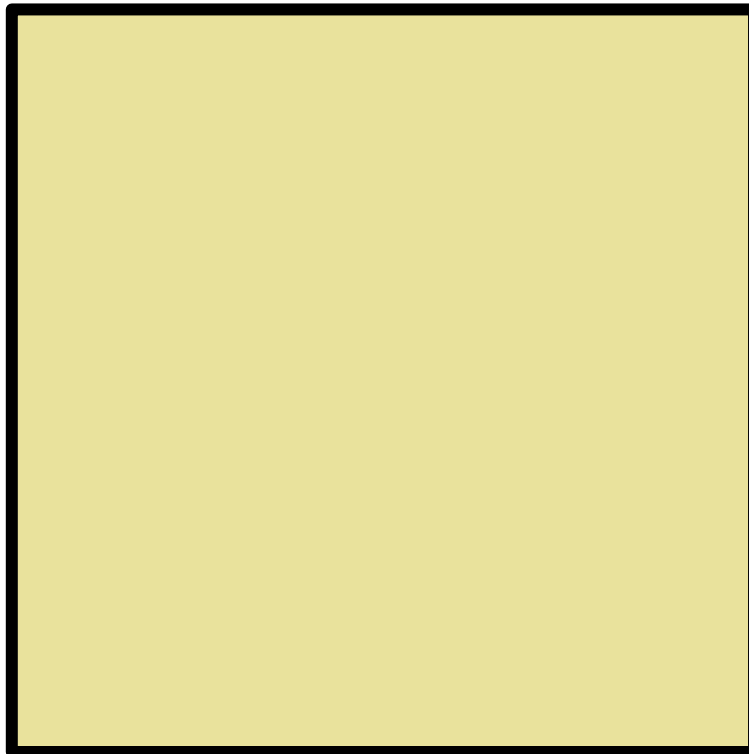
+

0

-



Finite differencing

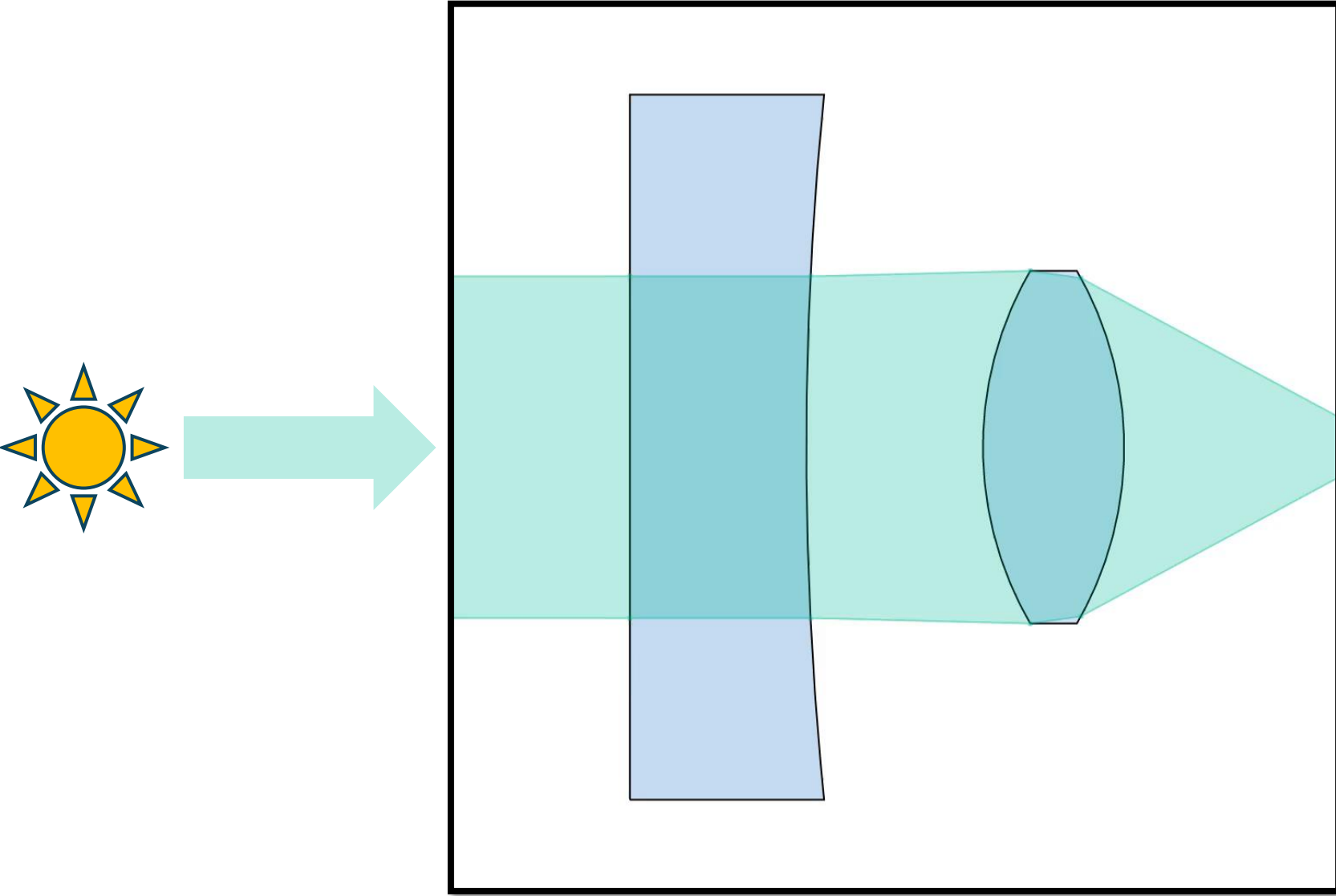


Automatic
differentiation

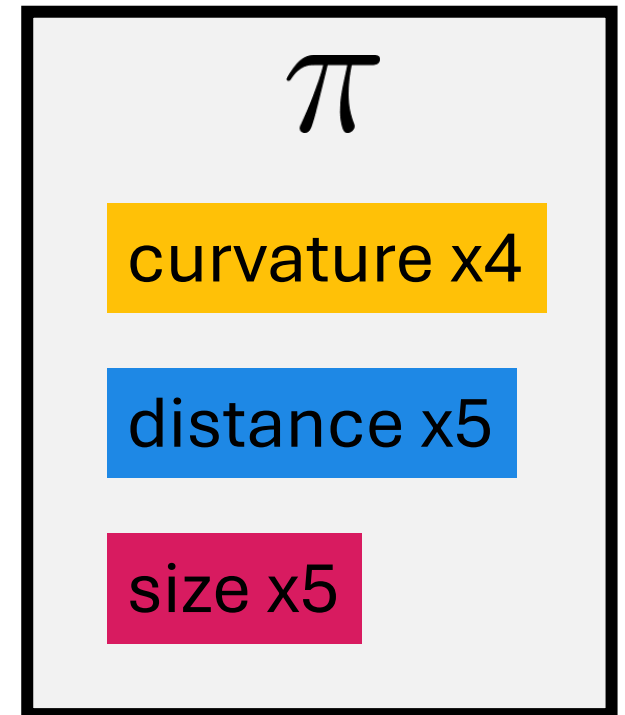
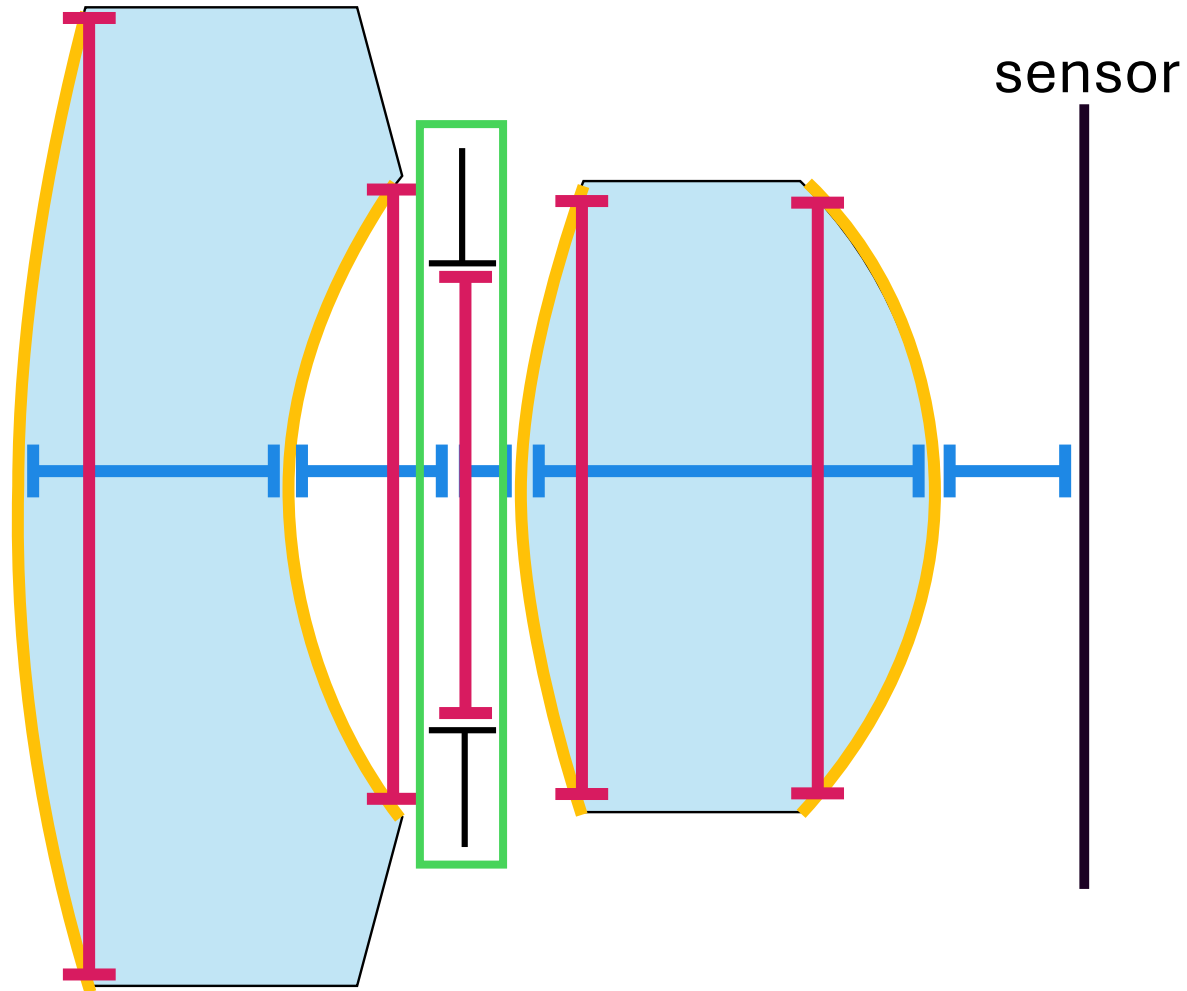


Our method

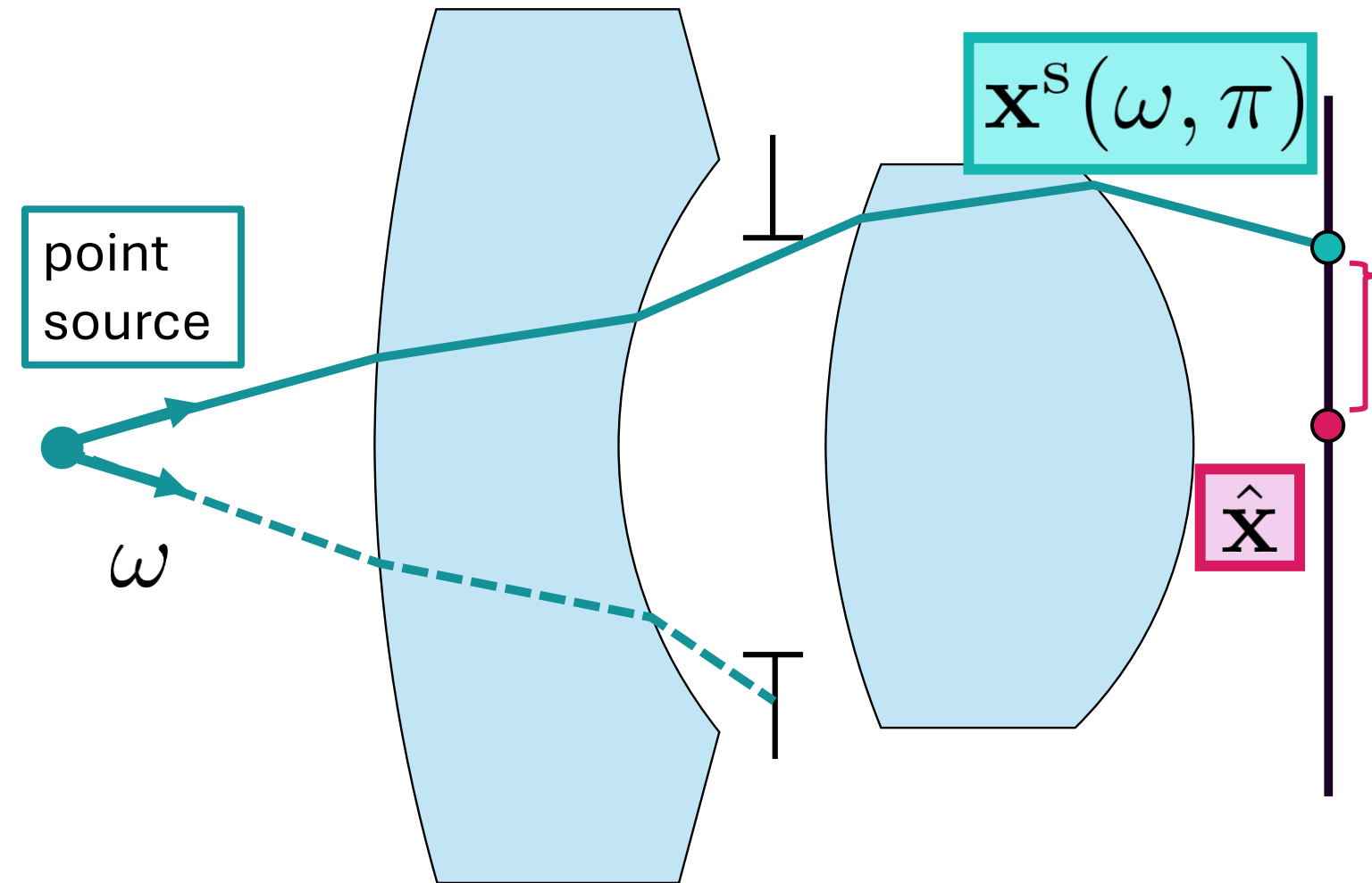
Shape affects speed as well



Representing a lens



Evaluating a lens



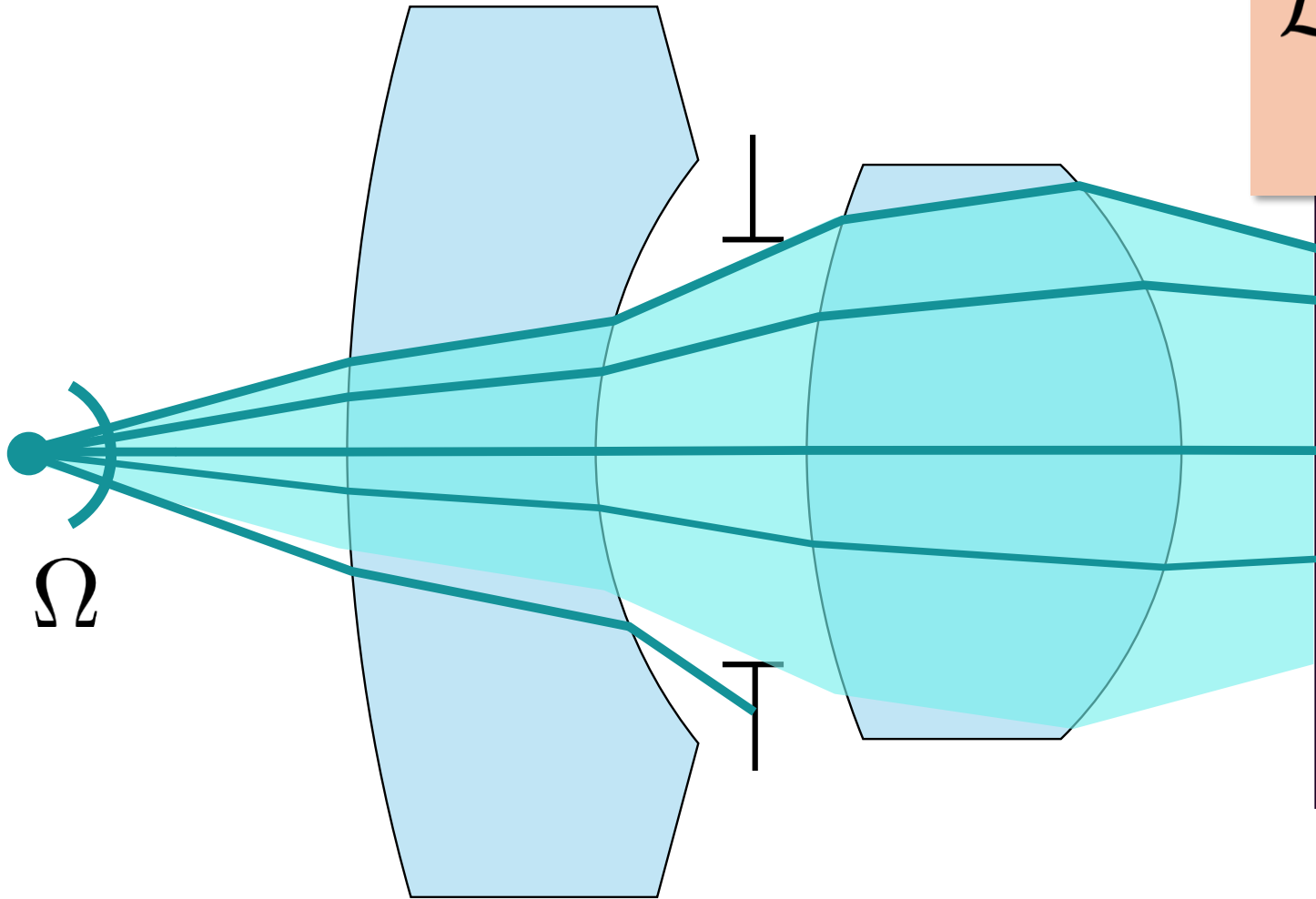
sharpness loss

$$f_{\text{sharpness}} = \|\mathbf{x}^s - \hat{\mathbf{x}}\|^2$$

speed loss

$$f_{\text{speed}} = 1$$

Evaluating a lens



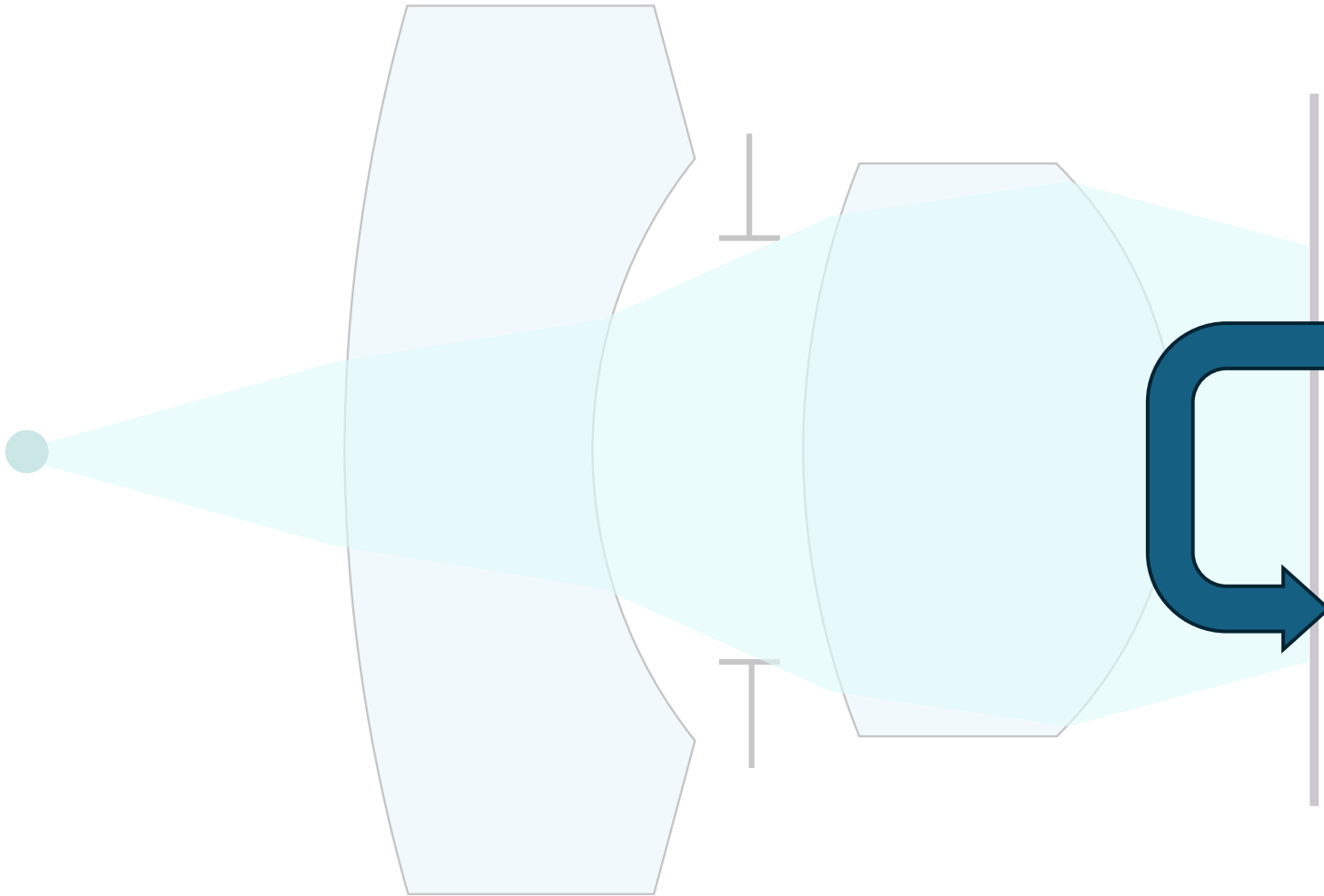
$$\mathcal{L} = \int_{\Omega(\pi)} f(\mathbf{x}^s(\omega, \pi)) d\omega$$

\approx

Monte Carlo estimator

$$\frac{1}{N} \sum_{i=0}^N \frac{f(\mathbf{x}^s(\omega_i))}{p(\omega_i)}$$

Optimizing a lens



$$\frac{d\mathcal{L}}{d\pi}$$

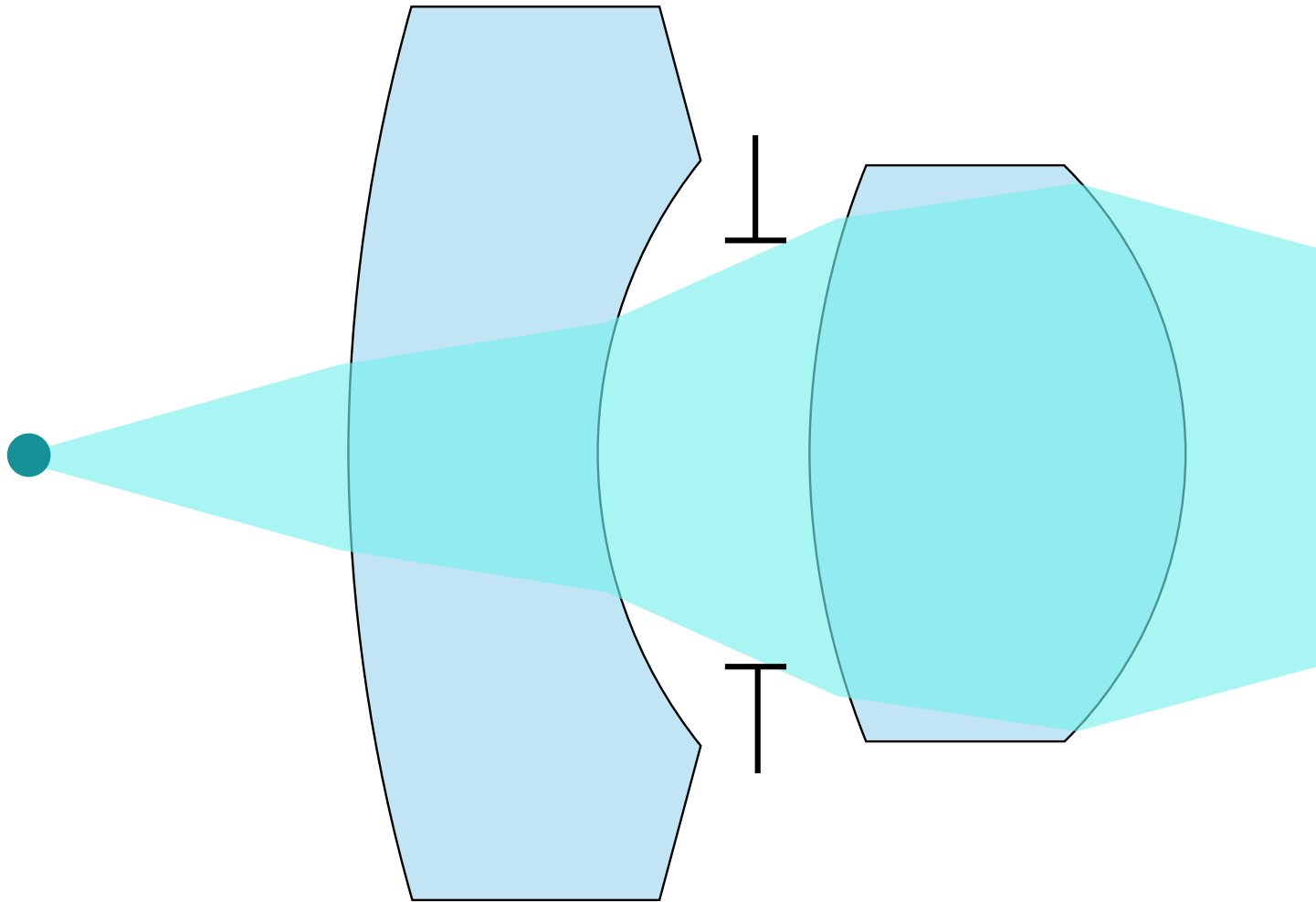
Apply gradient to estimator

$$\frac{d}{d\pi} \left(\frac{1}{N} \sum_{i=0}^N \frac{f(\mathbf{x}^s(\omega_i))}{p(\omega_i)} \right)$$

Monte Carlo gradient estimator (biased)

$$\frac{1}{N} \sum_{i=0}^N \frac{1}{p(\omega_i)} \frac{df}{d\mathbf{x}} \frac{d\mathbf{x}^s}{d\pi}$$

autodiff issue: aperture



Monte Carlo gradient estimator (biased)

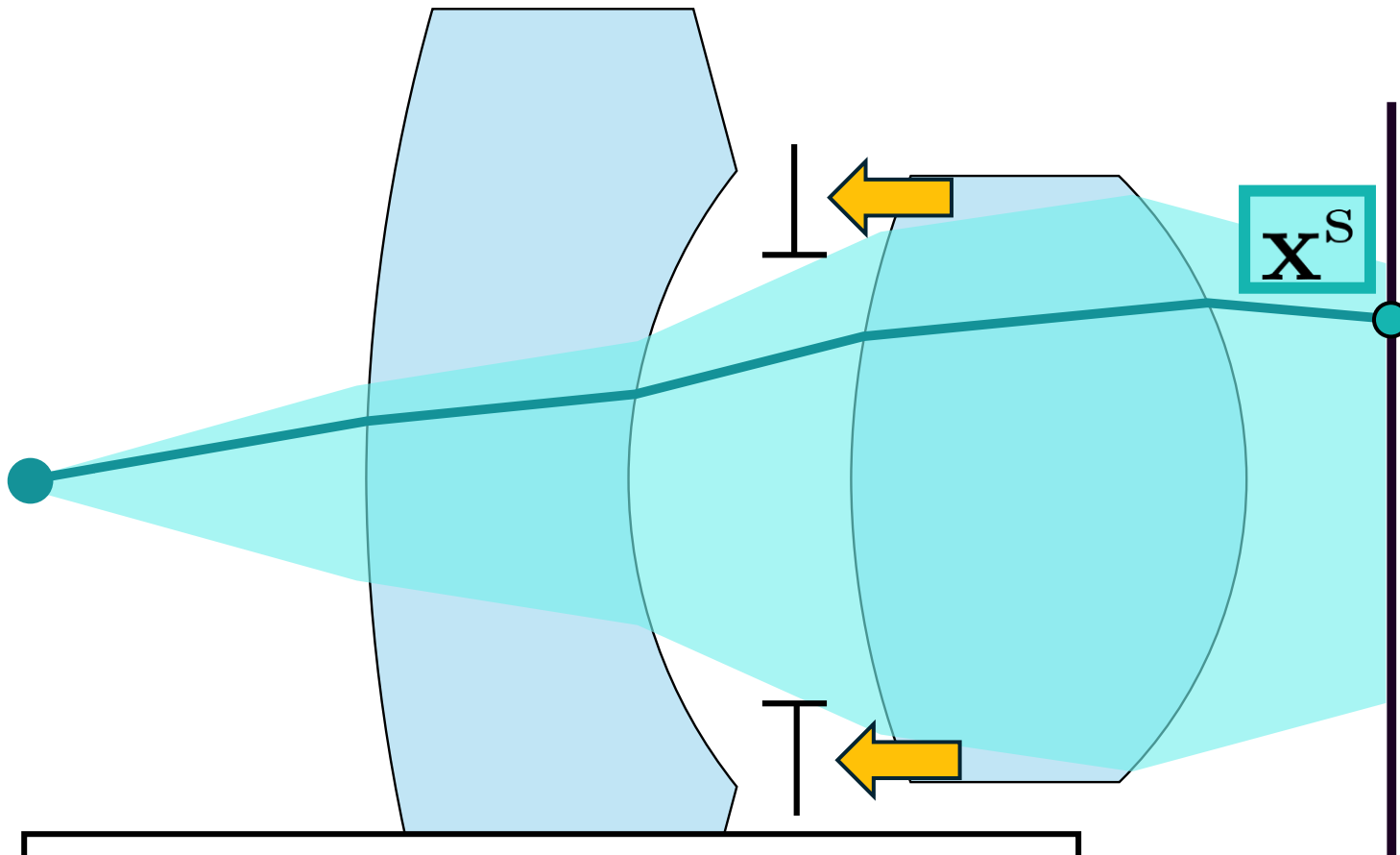
$$\frac{1}{N} \sum_{i=0}^N \frac{1}{p(\omega_i)} \frac{df}{d\mathbf{x}} \frac{d\mathbf{x}^s}{d\pi}$$

$$f_{\text{speed}} = 1$$



$$\frac{df}{d\mathbf{x}} = 0$$

autodiff issue: aperture



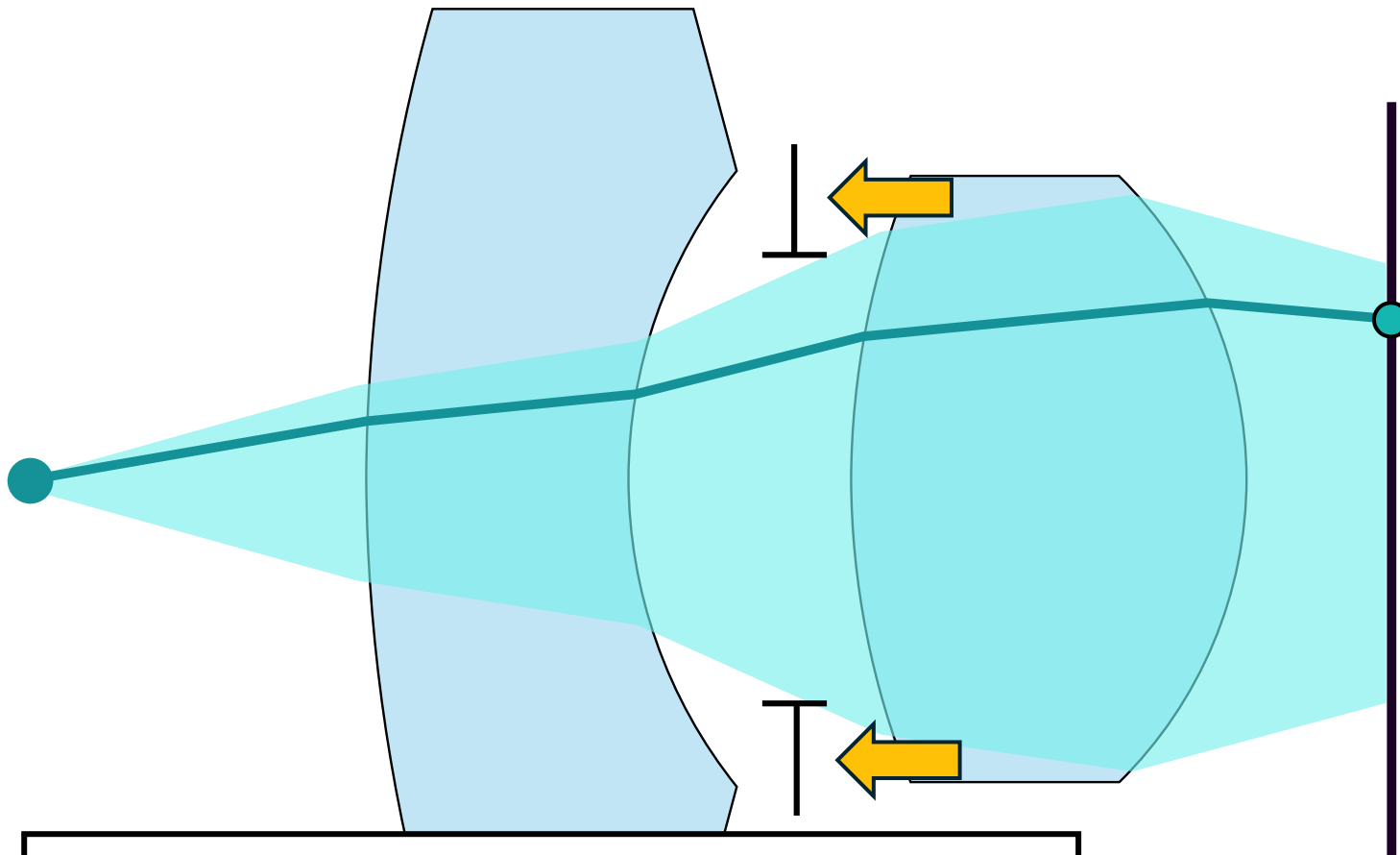
π : radius of aperture stop

Monte Carlo gradient estimator (biased)

$$\frac{1}{N} \sum_{i=0}^N \frac{1}{p(\omega_i)} \frac{df}{d\mathbf{x}} \frac{d\mathbf{x}^s}{d\pi}$$

$$\frac{d\mathbf{x}^s}{d\pi} = 0$$

autodiff issue: aperture



π : radius of aperture stop

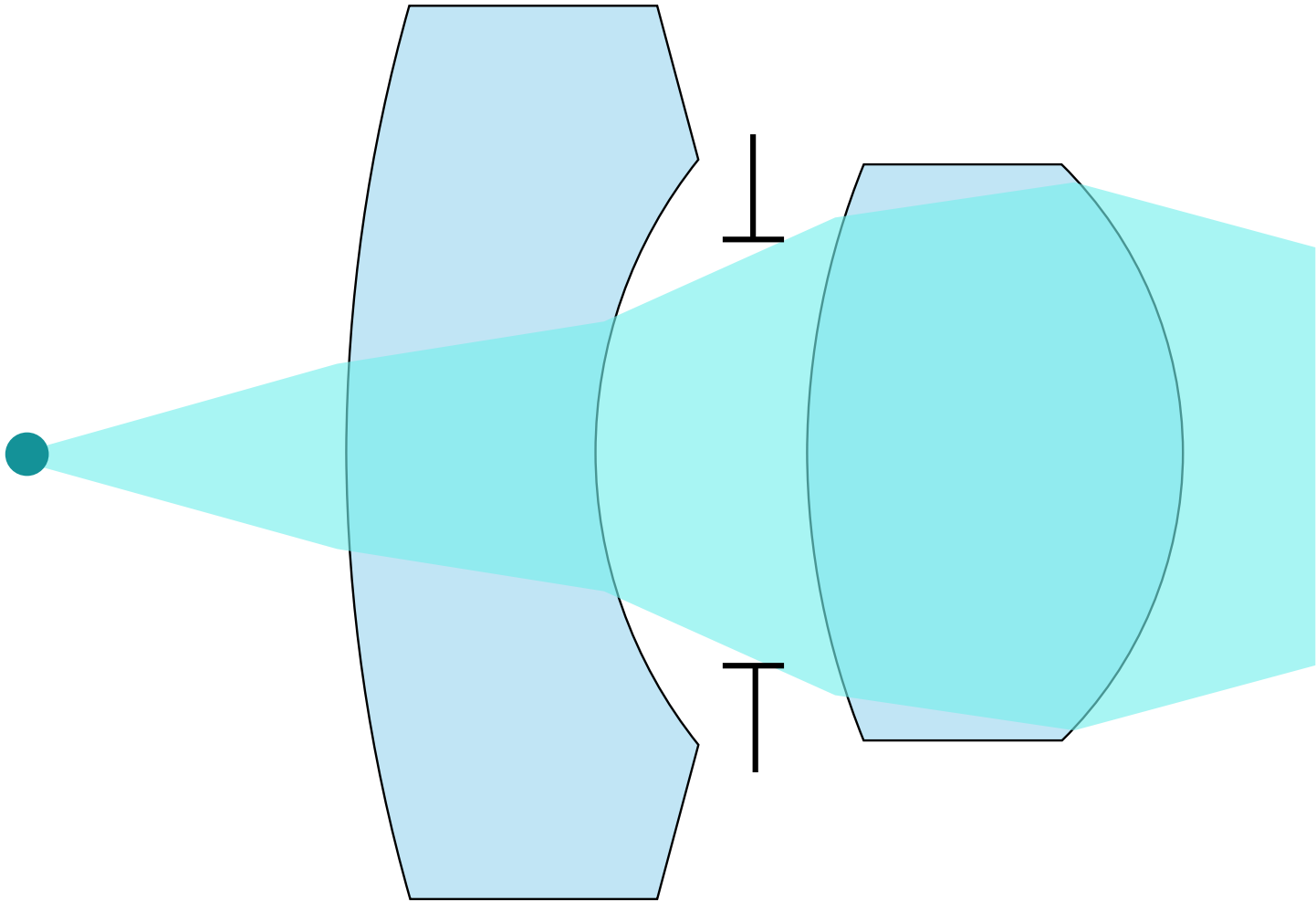
Monte Carlo gradient estimator (biased)

$$\frac{1}{N} \sum_{i=0}^N \frac{1}{p(\omega_i)} \frac{df}{d\mathbf{x}} \frac{d\mathbf{x}^s}{d\pi}$$

$$\frac{df_{\text{speed}}}{d\pi} = 0$$

$$\frac{df_{\text{sharpness}}}{d\pi} = 0$$

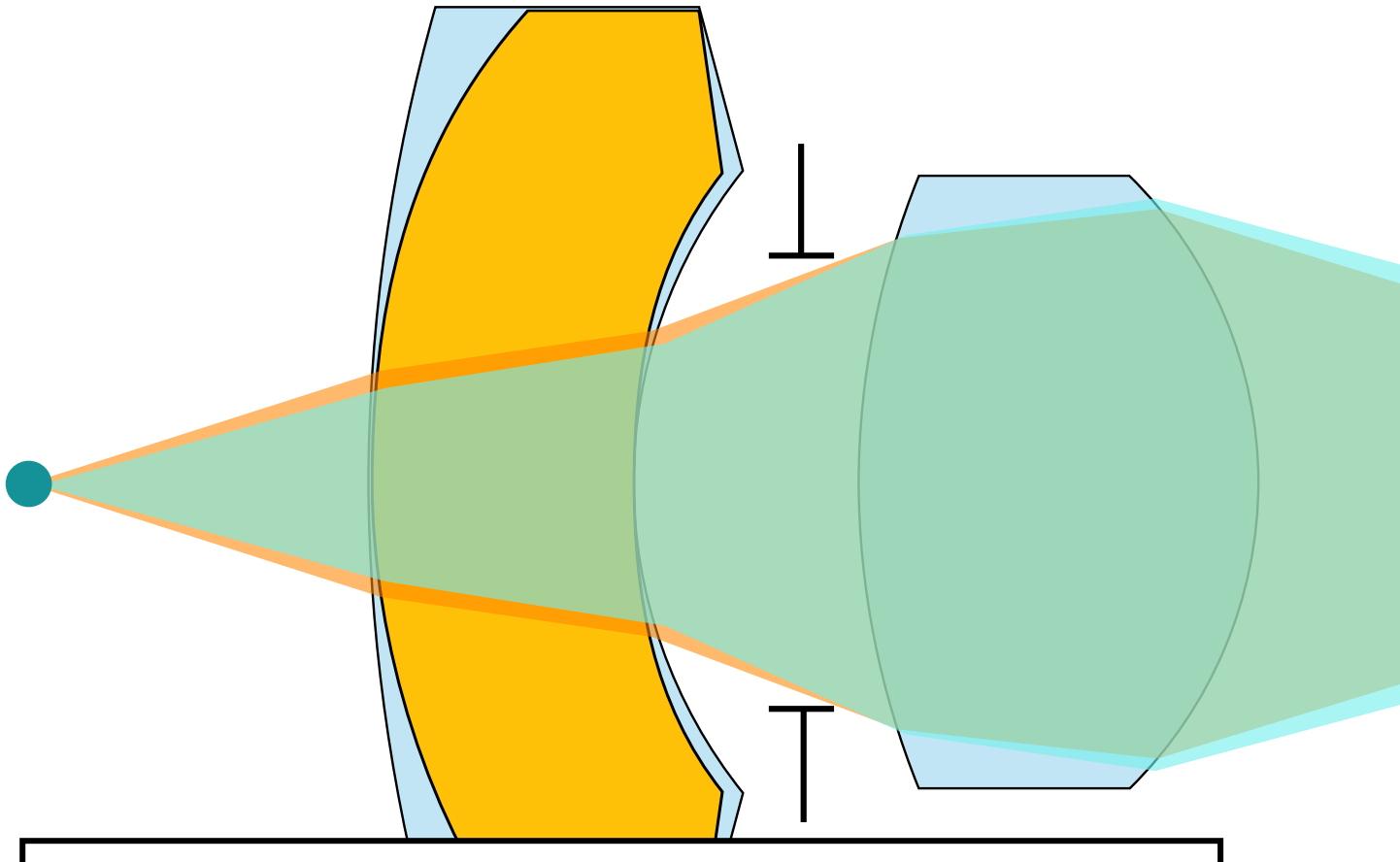
autodiff issue: curvature



Monte Carlo gradient estimator (biased)

$$\frac{1}{N} \sum_{i=0}^N \frac{1}{p(\omega_i)} \frac{df}{d\mathbf{x}} \frac{d\mathbf{x}^s}{d\pi}$$

autodiff issue: curvature



π : curvatures of first element

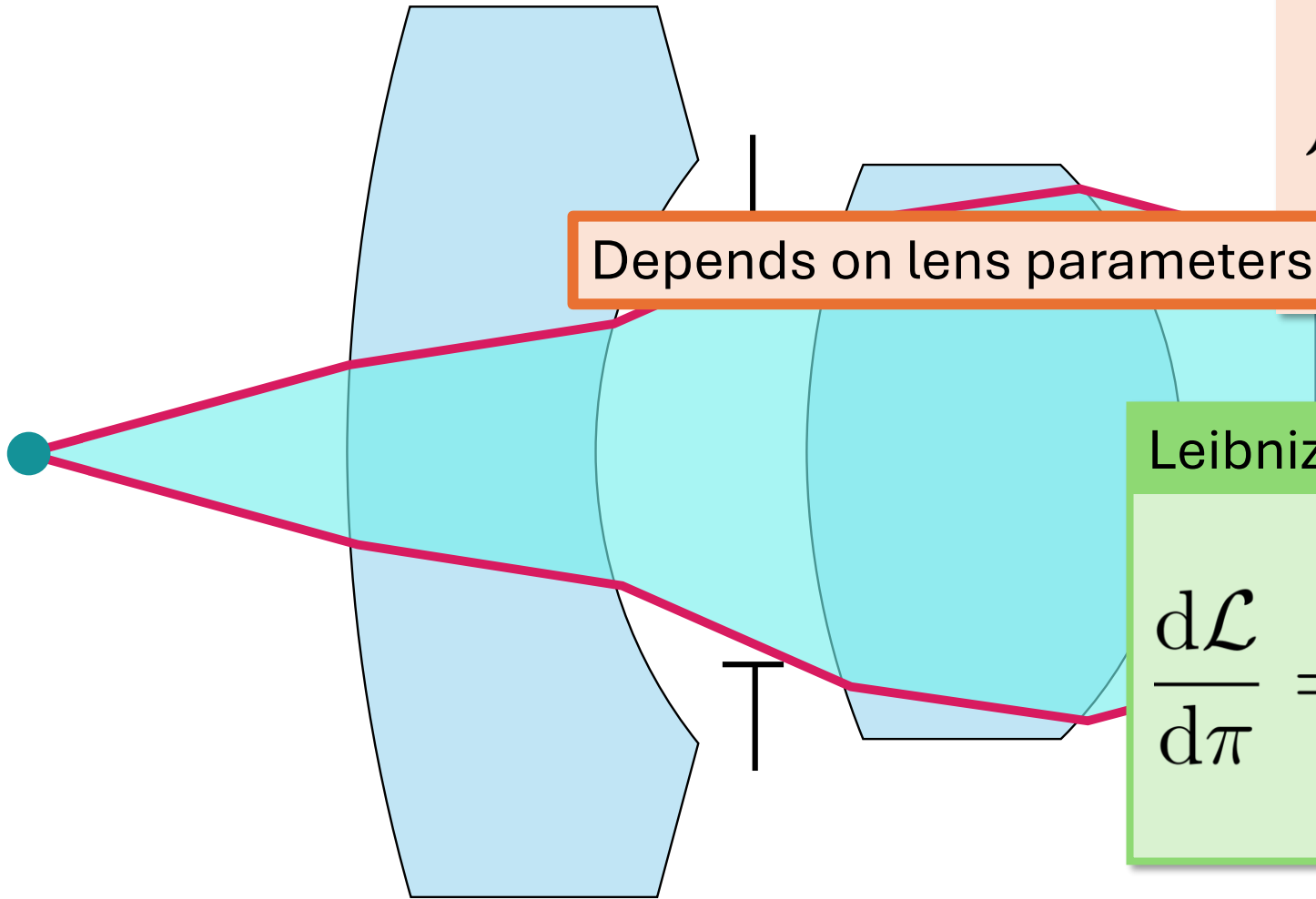
Monte Carlo gradient estimator (biased)

$$\frac{1}{N} \sum_{i=0}^N \frac{1}{p(\omega_i)} \frac{df}{d\mathbf{x}} \frac{d\mathbf{x}^s}{d\pi}$$

$$\frac{df_{\text{speed}}}{d\pi} = 0$$

$$\frac{df_{\text{sharpness}}}{d\pi} = \text{biased}$$

Key insight: variable domain of integration



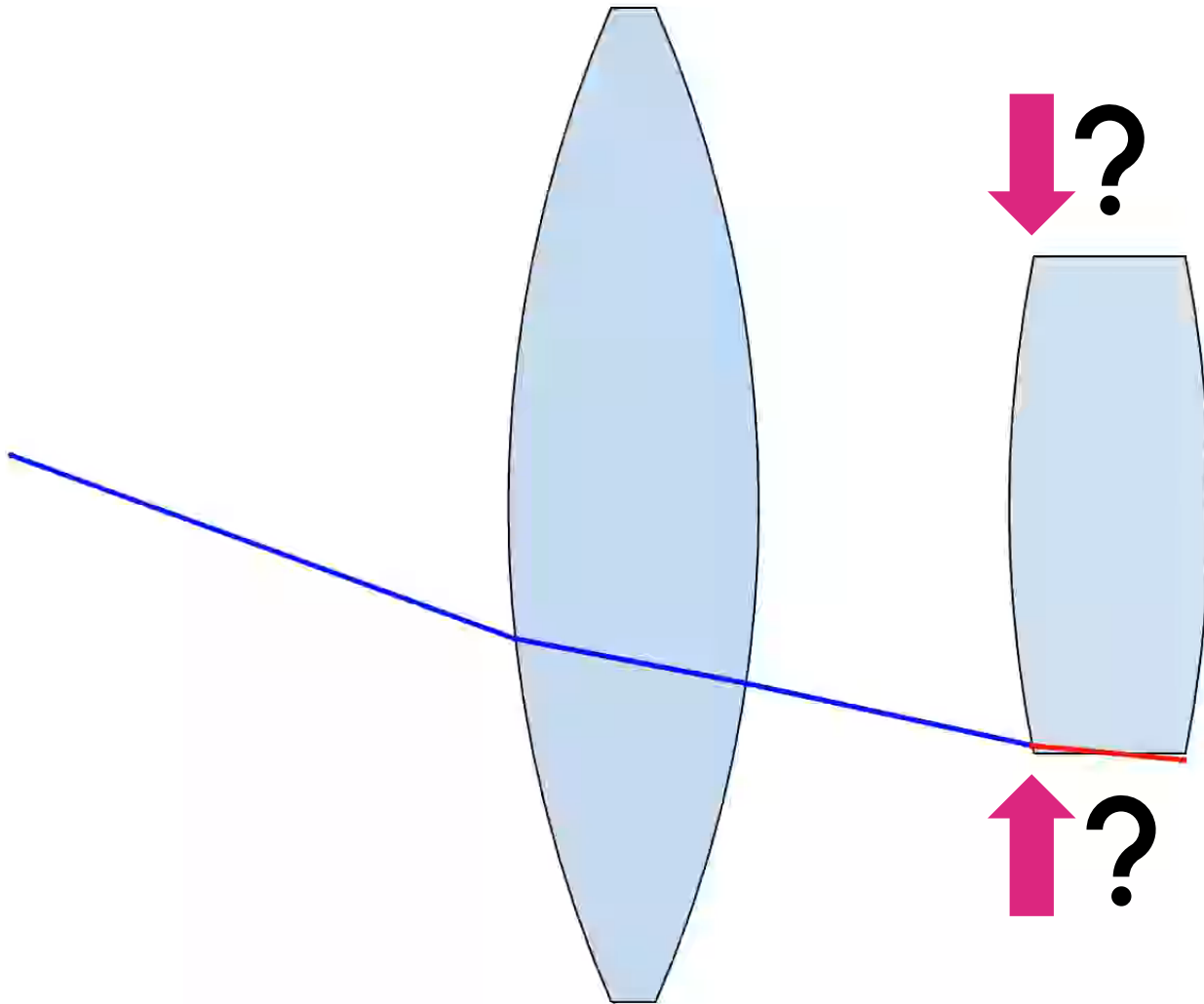
Depends on lens parameters!

$$\mathcal{L} = \int_{\Omega(\pi)} f(\mathbf{x}^s(\omega, \pi)) d\omega$$

Leibniz integral rule

$$\frac{d\mathcal{L}}{d\pi} = \int_{\Omega(\pi)} \frac{df}{d\pi} d\omega + \int_{\partial\Omega(\pi)} f \frac{dg}{d\pi} d\omega$$

Finding the boundary is hard



Leibniz integral rule

$$\frac{d\mathcal{L}}{d\pi} = \int_{\Omega(\pi)} \frac{df}{d\pi} d\omega + \int_{\partial\Omega(\pi)} f \frac{dg}{d\pi} d\omega$$

Use the reparameterization trick

$$\frac{d\mathcal{L}}{d\pi} = \int_{\Omega(\pi)} \frac{df}{d\pi} d\omega + \int_{\partial\Omega(\pi)} f \frac{dg}{d\pi} d\omega$$



Use the reparameterization trick

$$\frac{d\mathcal{L}}{d\pi} = \int_{\Omega(\pi)} \frac{df}{d\pi} d\omega + \int_{\partial\Omega(\pi)} f \frac{dg}{d\pi} d\omega$$



$$\frac{d\mathcal{L}}{d\pi} = \int_{\Omega(\pi)} \frac{df}{d\pi} d\omega + \int_{\Omega(\pi)} \nabla \cdot (f \mathcal{V}_g) d\omega$$



Unbiased Monte Carlo estimator

$$\frac{d\mathcal{L}}{d\pi} = \int_{\Omega(\pi)} \frac{df}{d\pi} d\omega + \int_{\Omega(\pi)} \nabla \cdot (f\mathcal{V}_g) d\omega$$

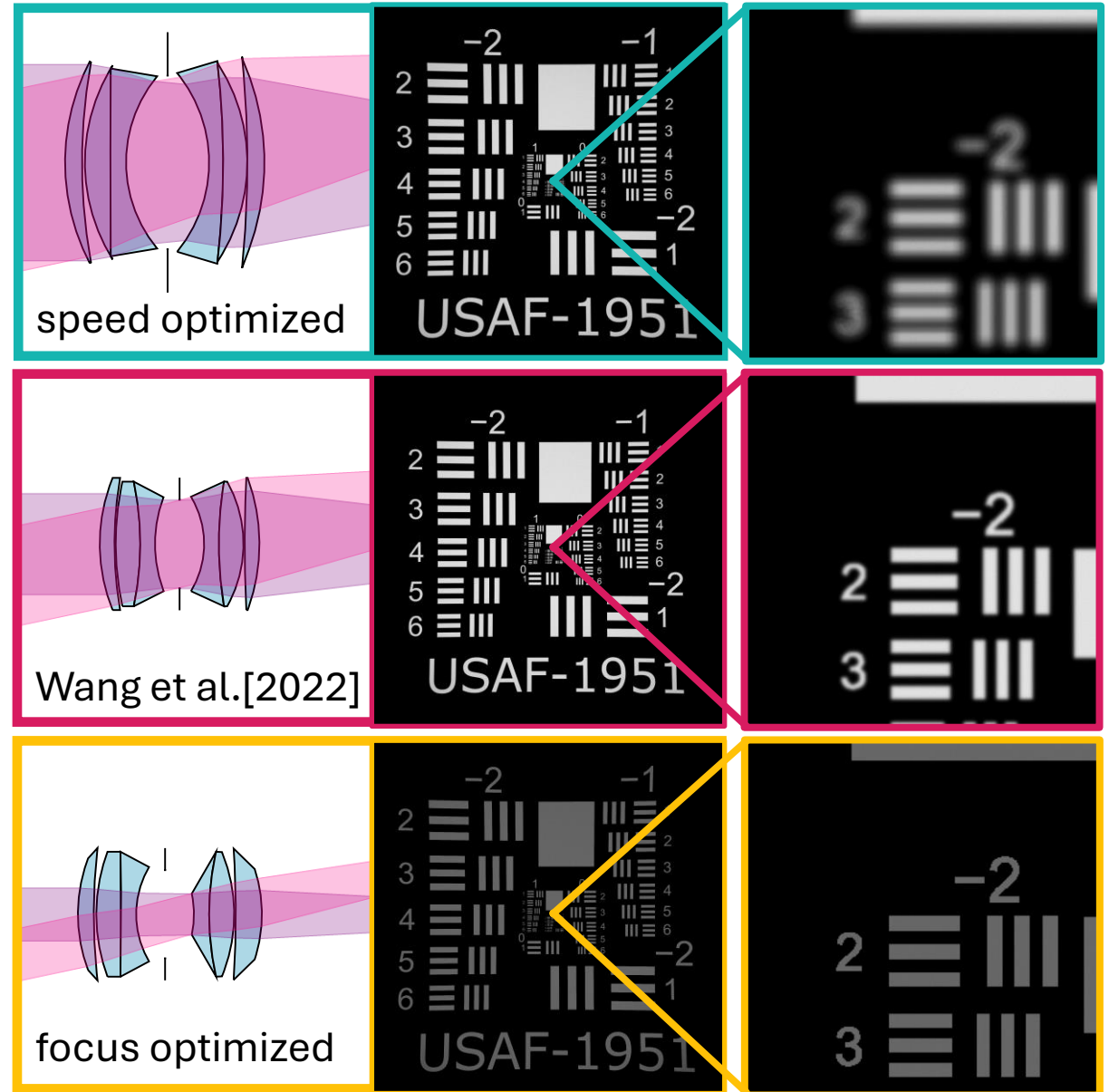
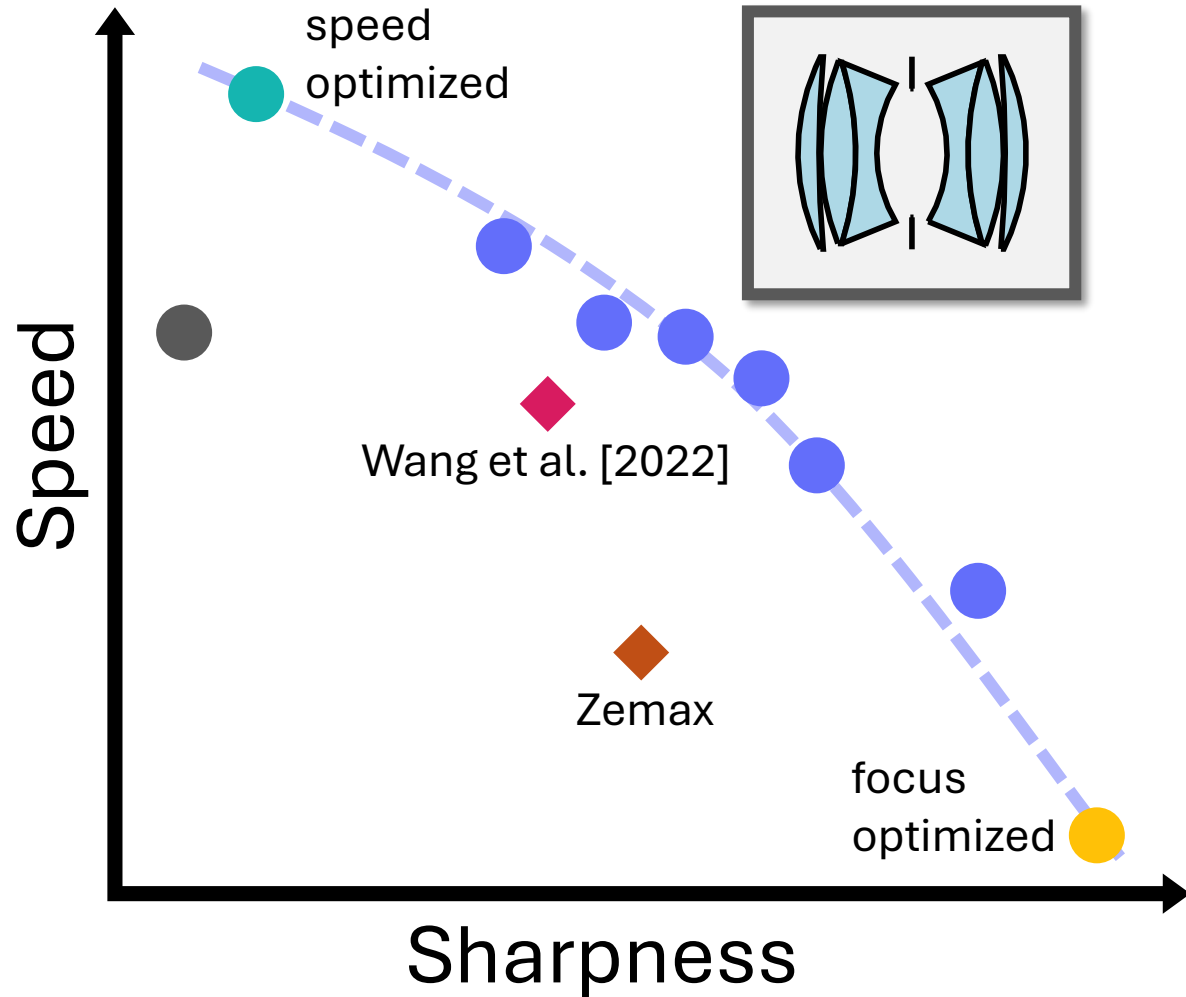
\approx

aperture-aware gradient estimator

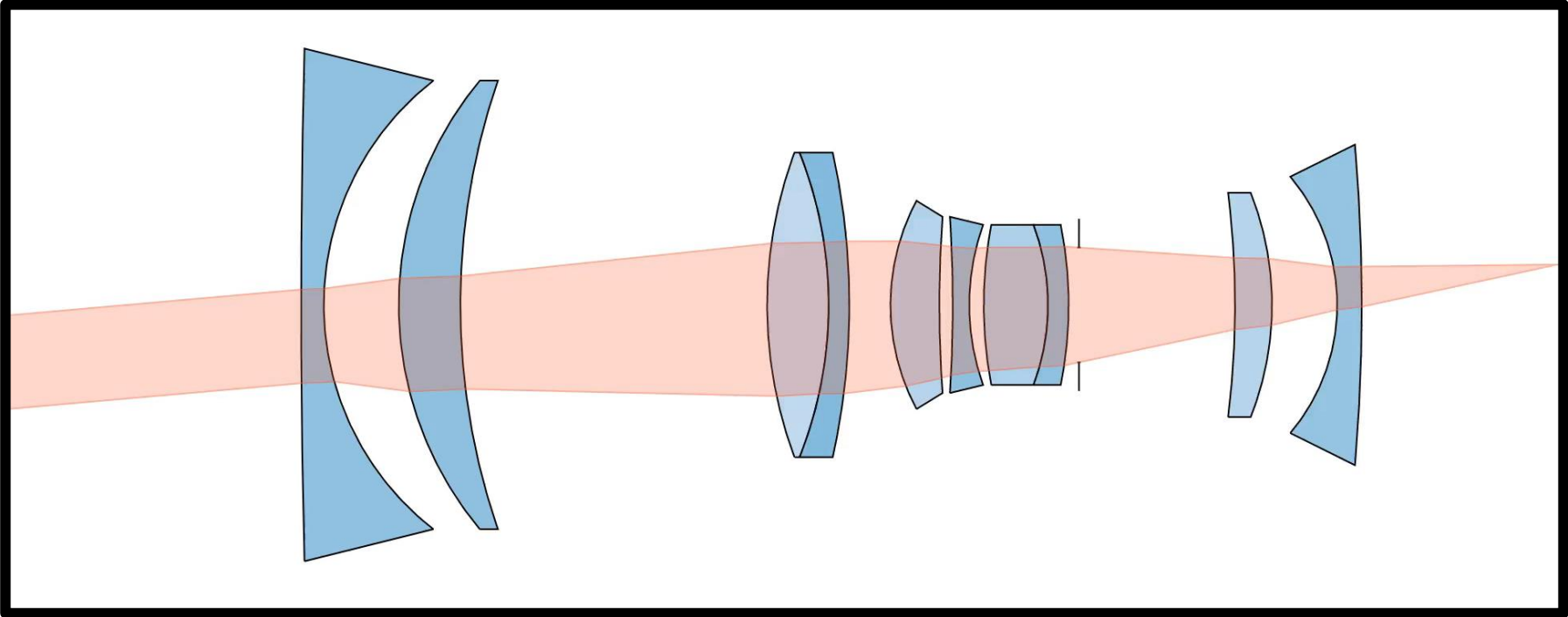
biased gradient estimator

$$\frac{1}{N} \sum_{i=0}^N \frac{1}{p(\omega_i)} \frac{df}{d\pi} + \nabla \cdot (f\mathcal{V}_g)$$

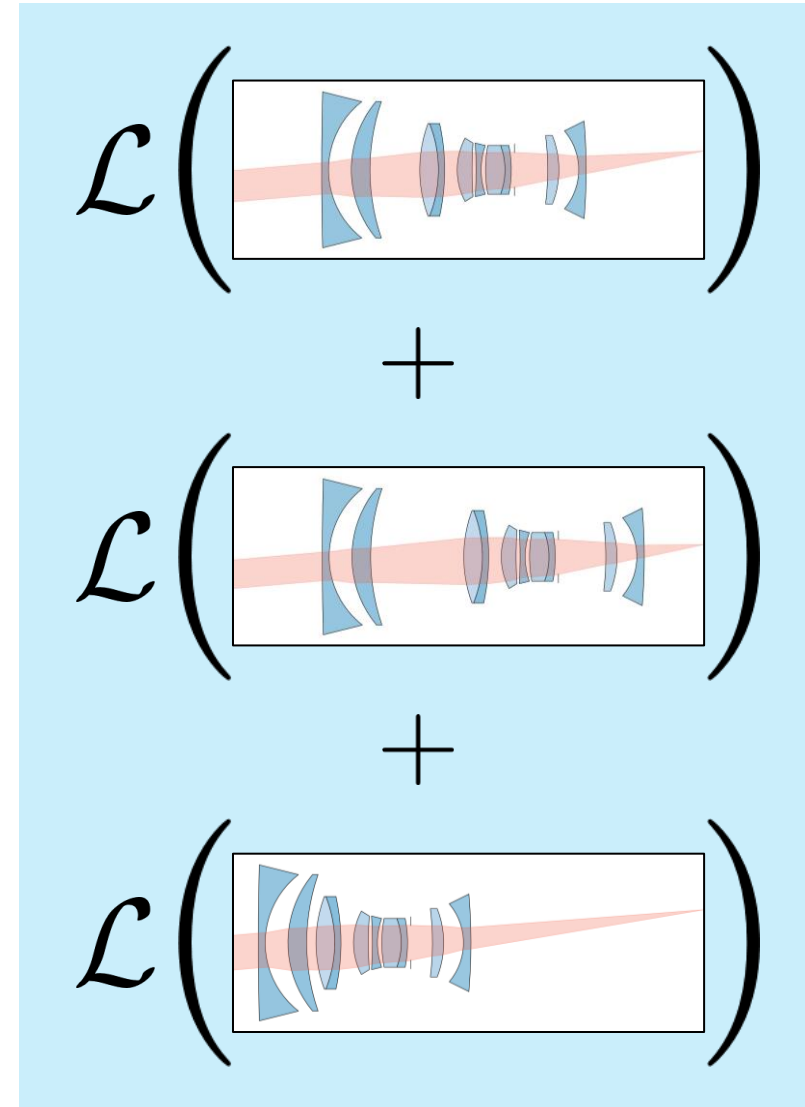
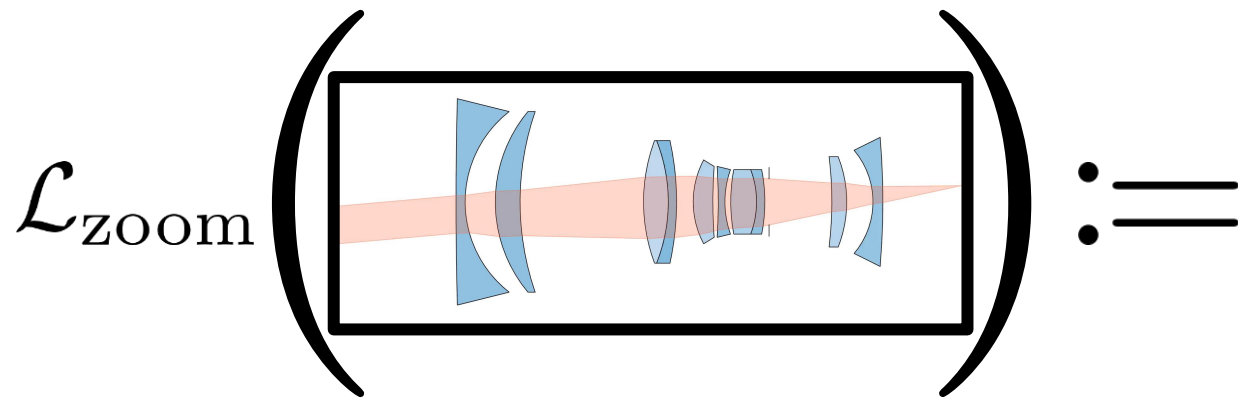
Results



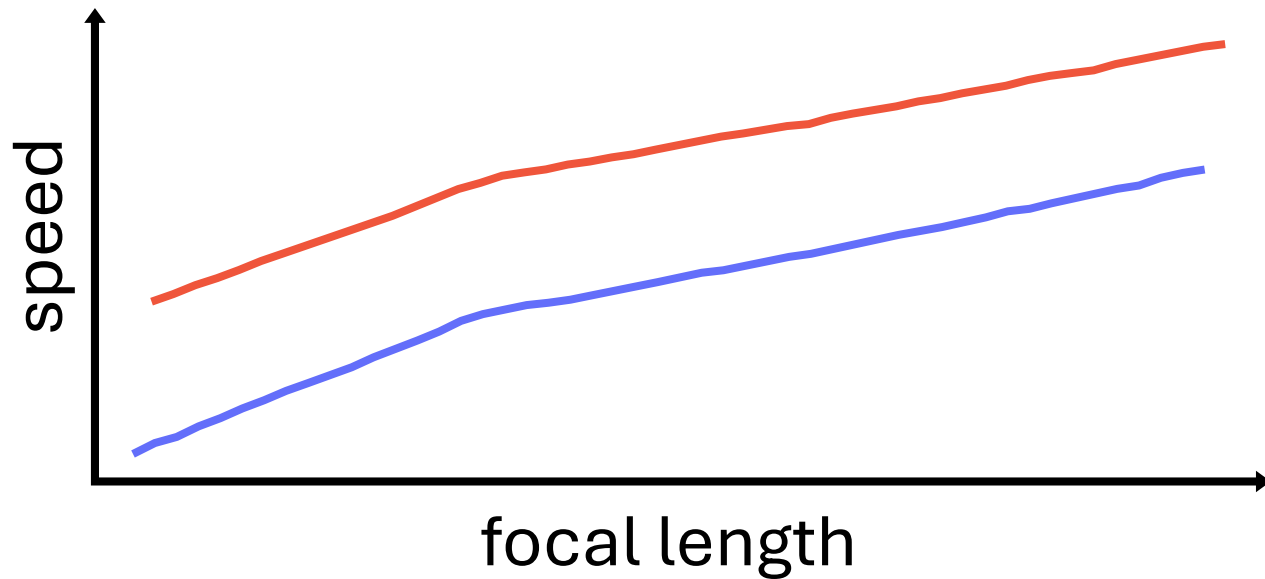
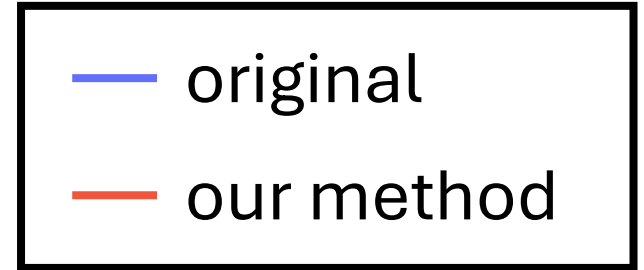
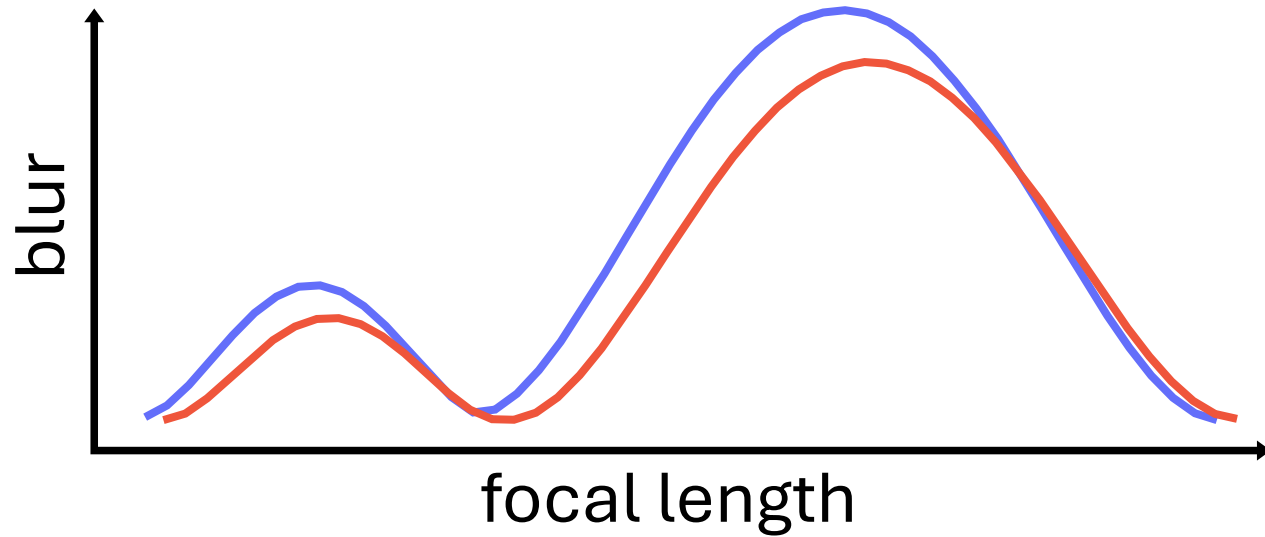
We can also optimize zoom lenses



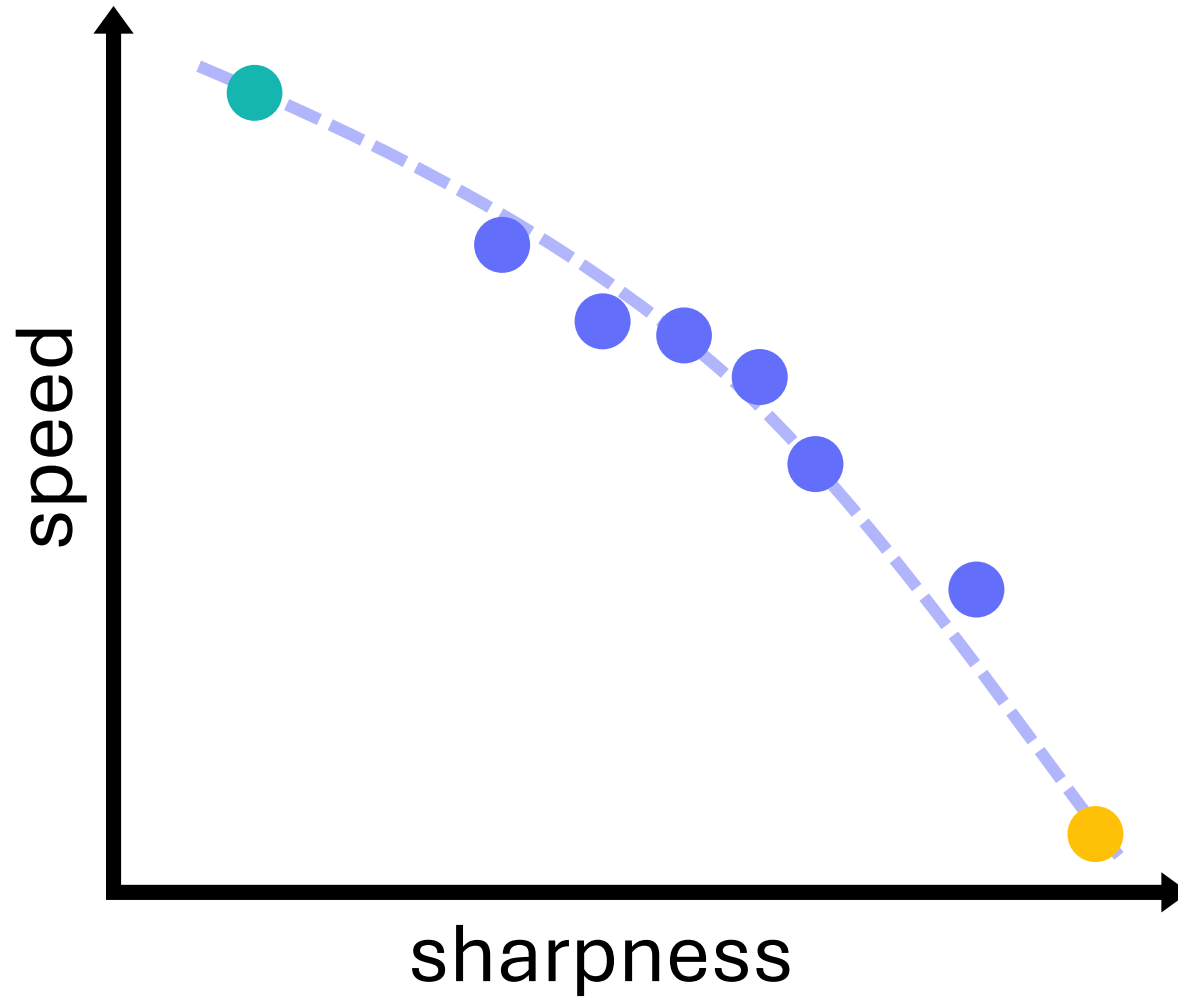
We can also optimize zoom lenses



We get better zoom lenses



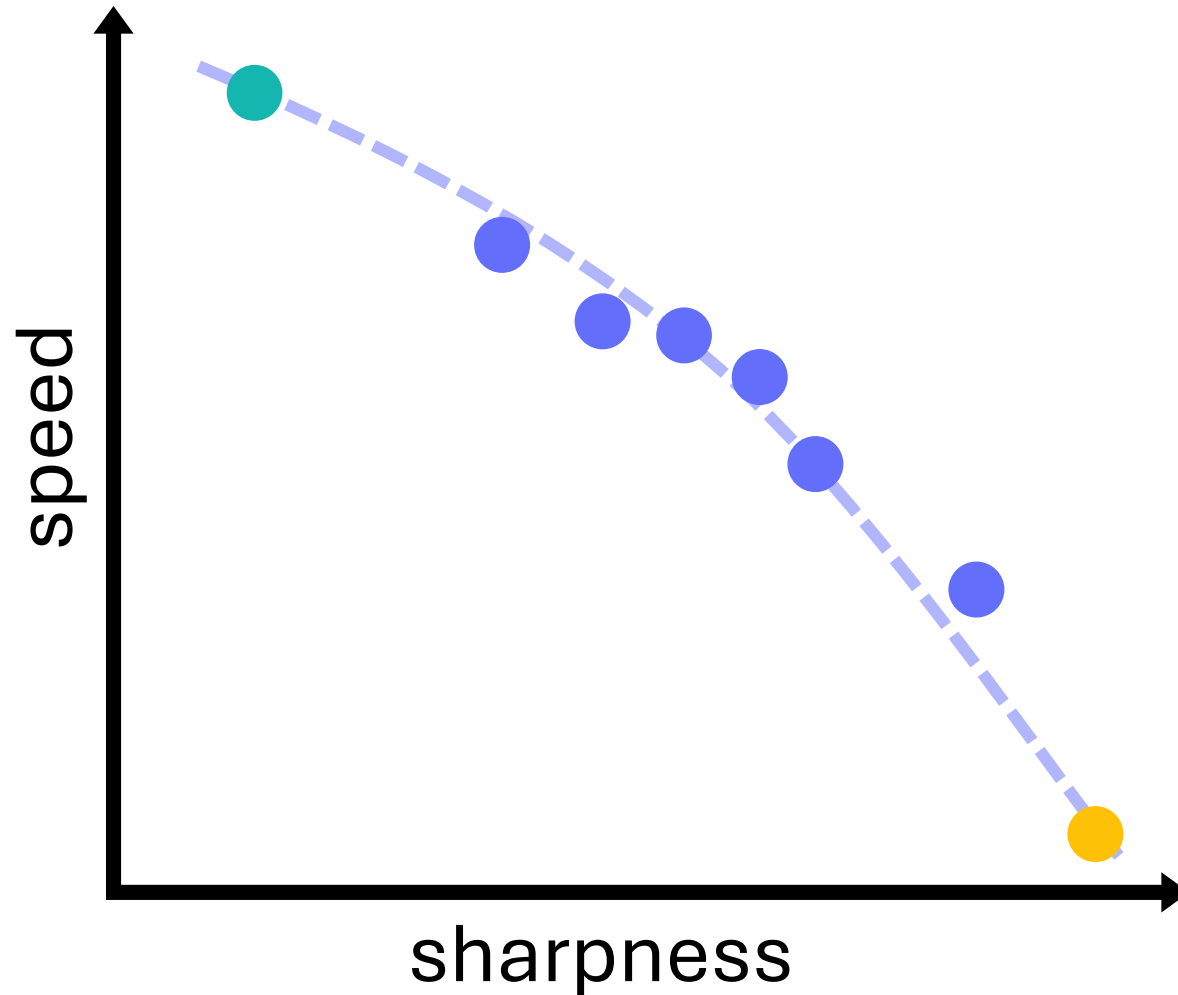
Summary



aperture-aware gradient estimator

$$\frac{1}{N} \sum_{i=0}^N \frac{1}{p(\omega_i)} \frac{df}{d\pi} + \nabla \cdot (f \mathcal{V}_g)$$

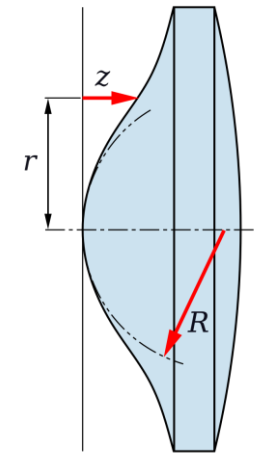
It's possible to optimize other types of lenses



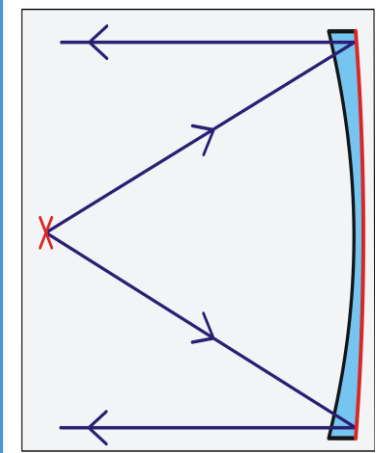
aperture-aware gradient estimator

$$\frac{1}{N} \sum_{i=0}^N \frac{1}{p(\omega_i)} \frac{df}{d\pi} + \nabla \cdot (f \mathcal{V}_g)$$

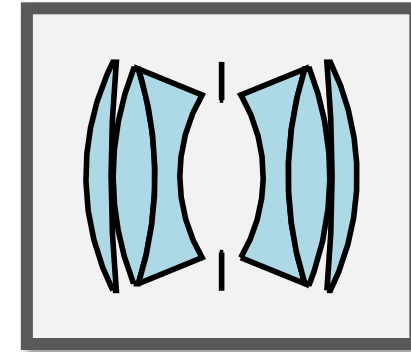
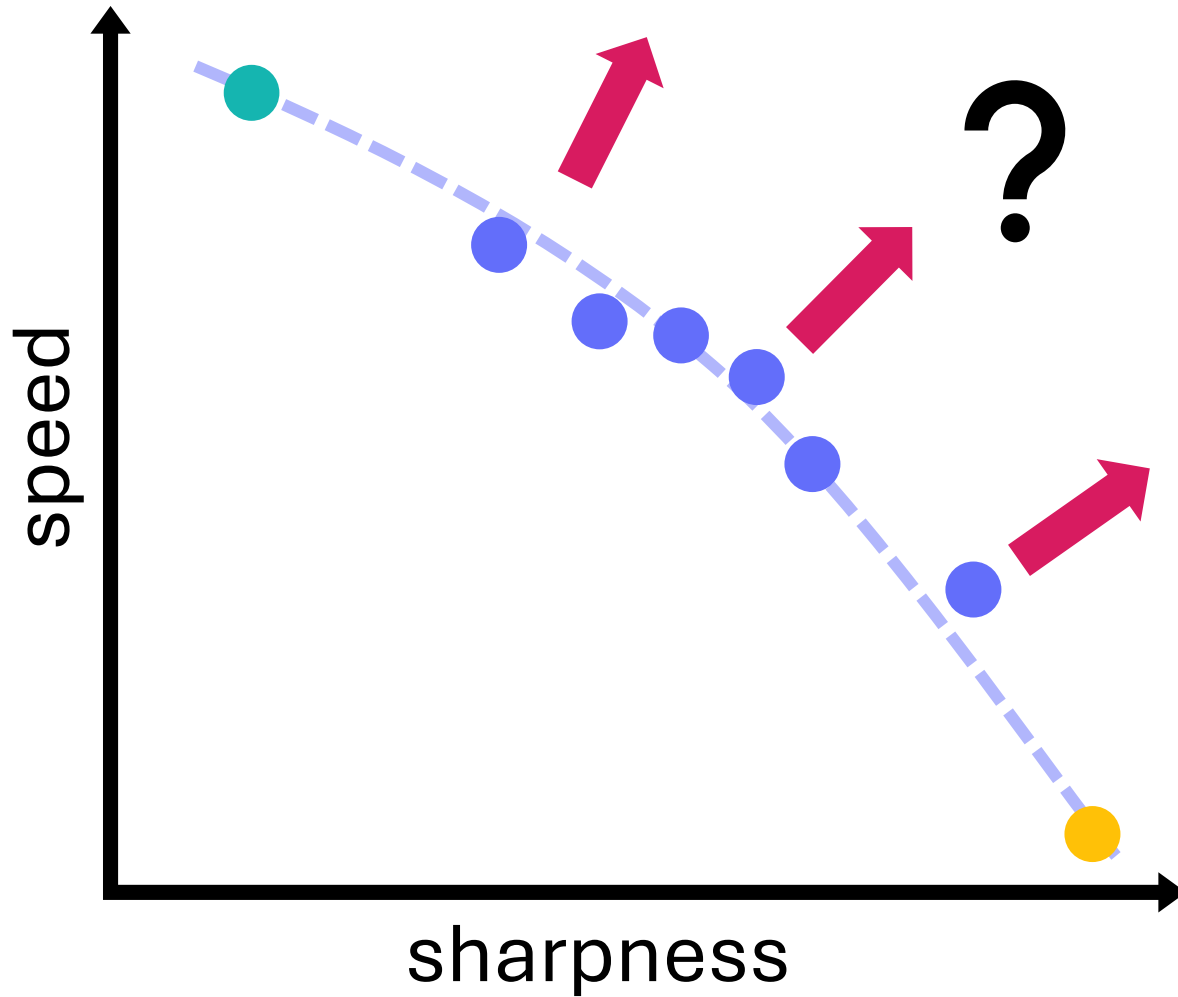
aspherics



mirrors

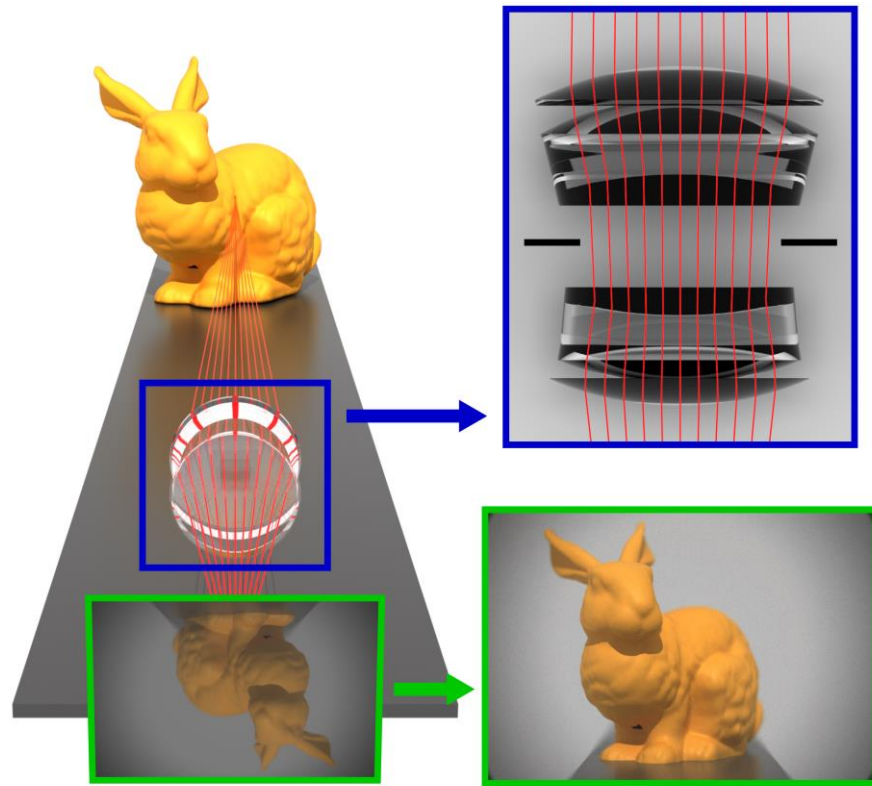


Gradient based methods are limited by the topology of the lens

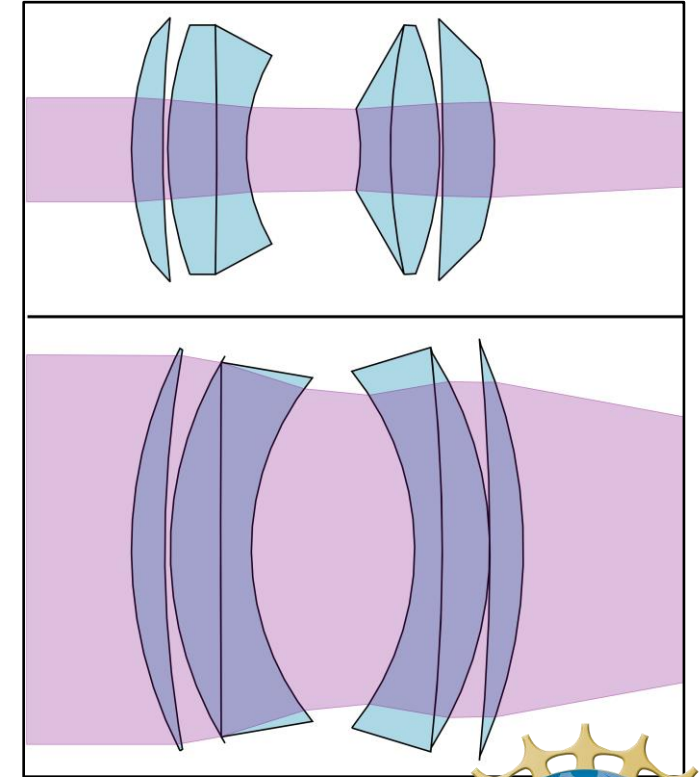


Aperture-Aware Lens Design

https://imaging.cs.cmu.edu/aperture_aware_lens_design/



code



ALFRED P. SLOAN
FOUNDATION

