

Vision Science Meets Computer Graphics:

Facts, Models, and Applications

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08/19/2021

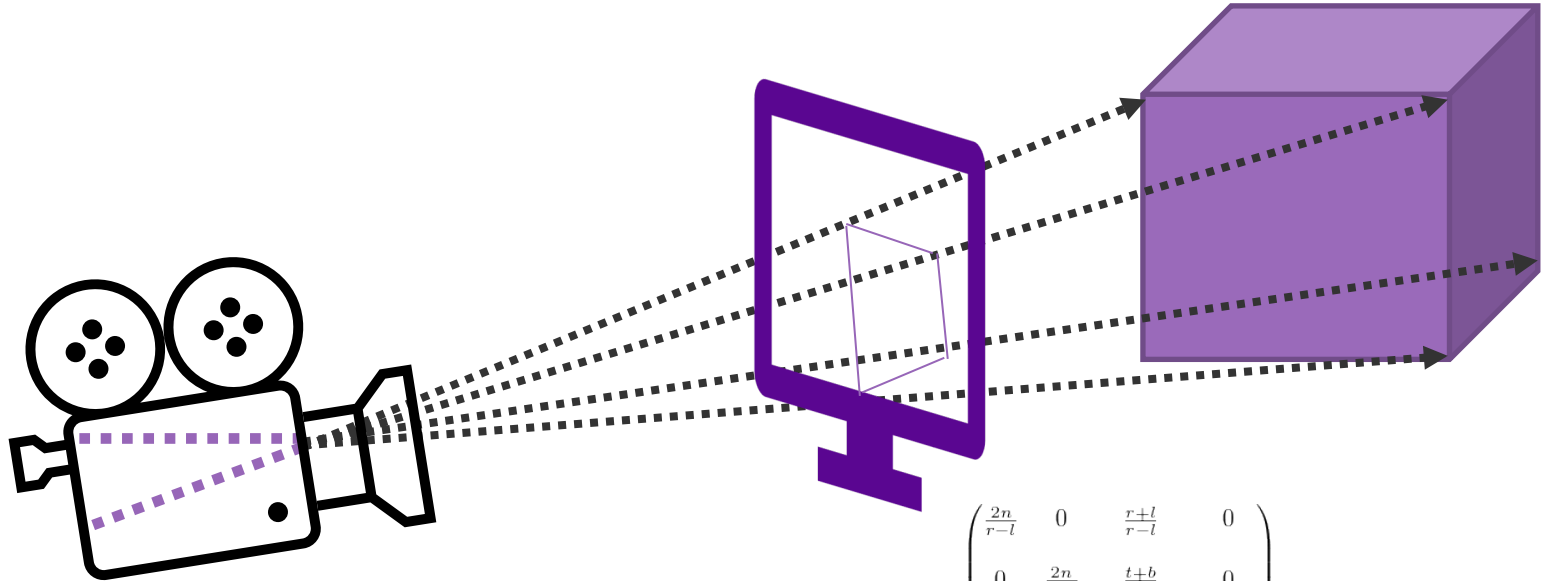


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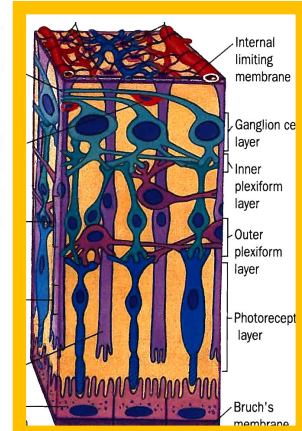
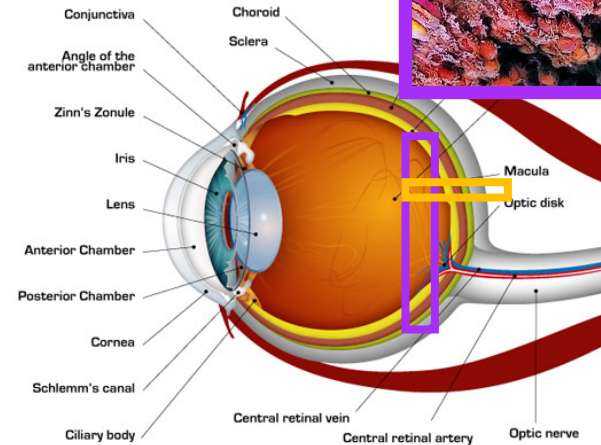
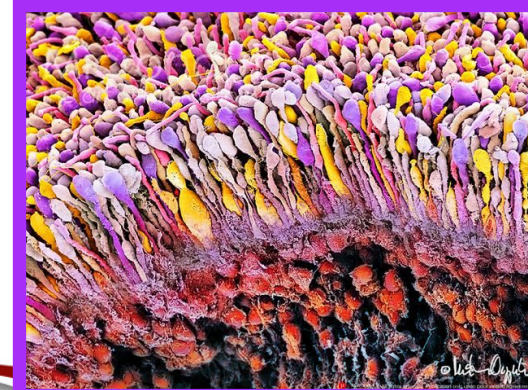
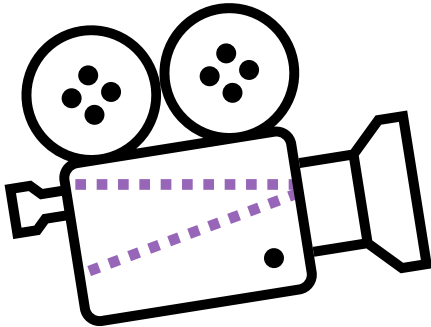
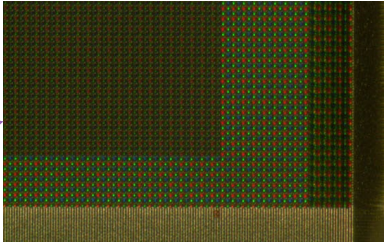
<https://www.immersivecomputinglab.org/>

The “Eye Model” in Graphics

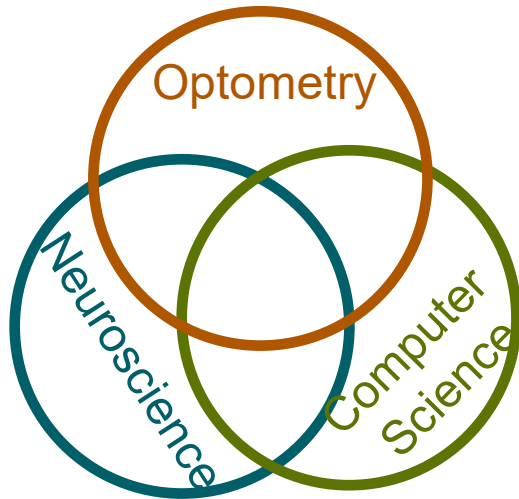


$$\begin{pmatrix} \frac{2n}{r-l} & 0 & \frac{r+l}{r-l} & 0 \\ 0 & \frac{2n}{t-b} & \frac{t+b}{t-b} & 0 \\ 0 & 0 & \frac{-(f+n)}{f-n} & \frac{-2fn}{f-n} \\ 0 & 0 & -1 & 0 \end{pmatrix}$$

The Biological Eye in Reality

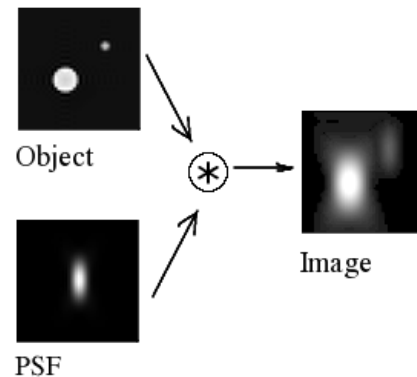
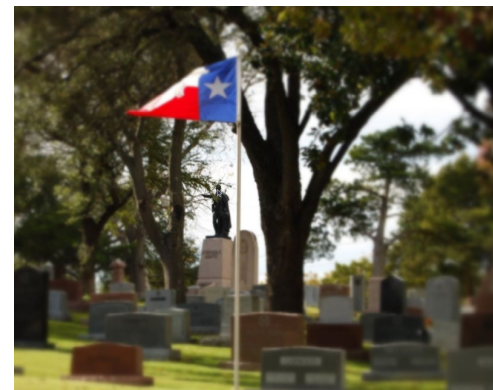
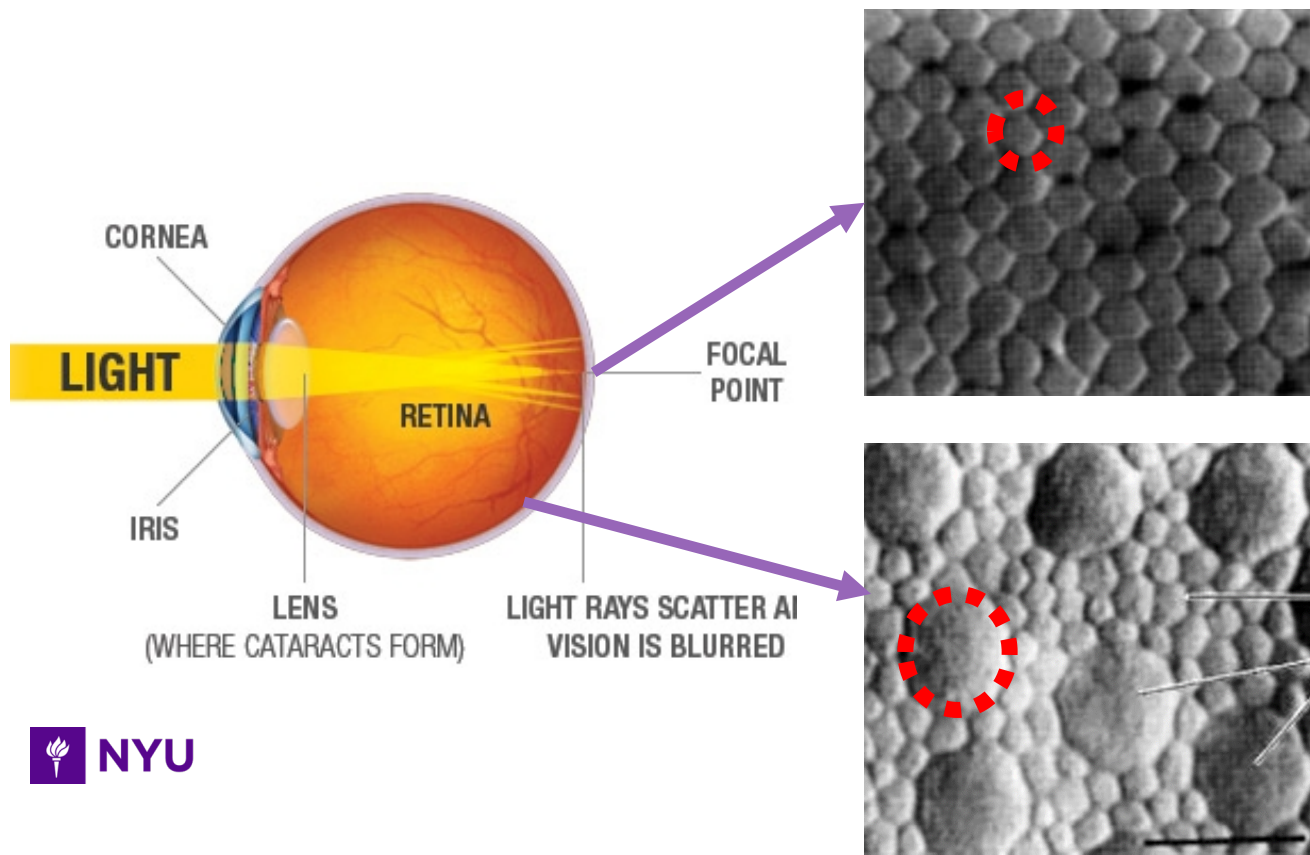


Vision Science



- Psychophysical studies on human visual system
- Understand visual perception mechanisms
- Derive computational models
- Advance human -inspired methods for computer vision, AI, etc.

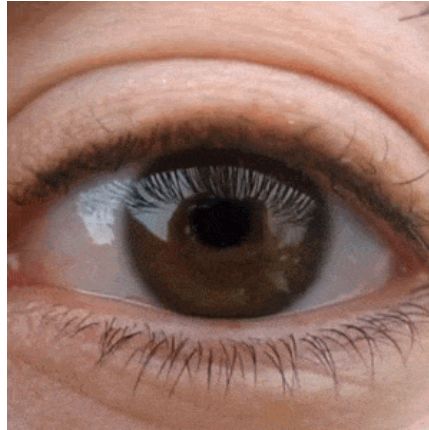
Optical Propagation



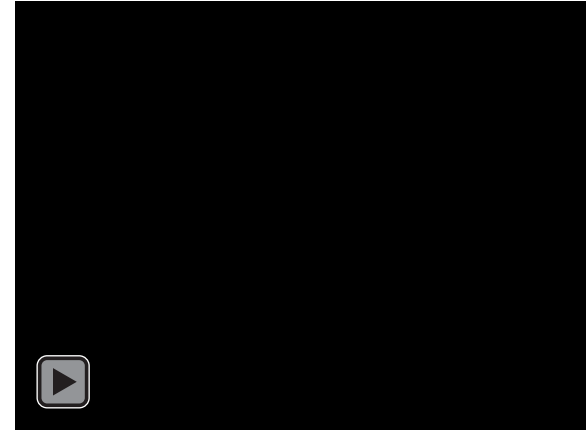
Dynamic Behaviors



vergence

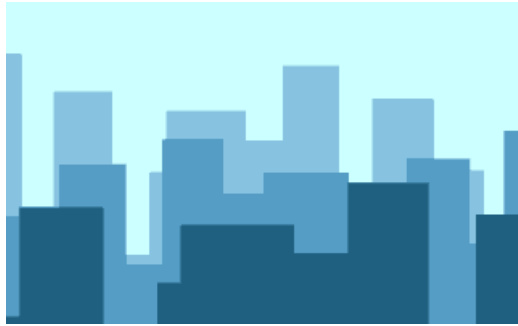


saccade



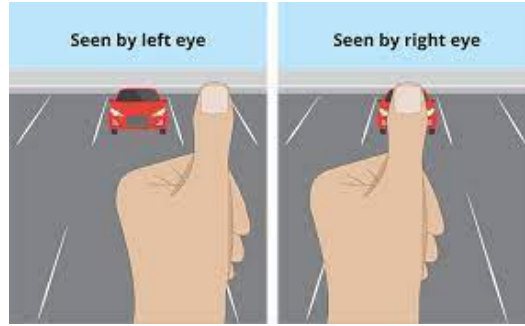
vestibulo-ocular reflex

Integration: Depth Perception



Motion

Motion Parallax



Binocular

left -right eye parallax



Monocular

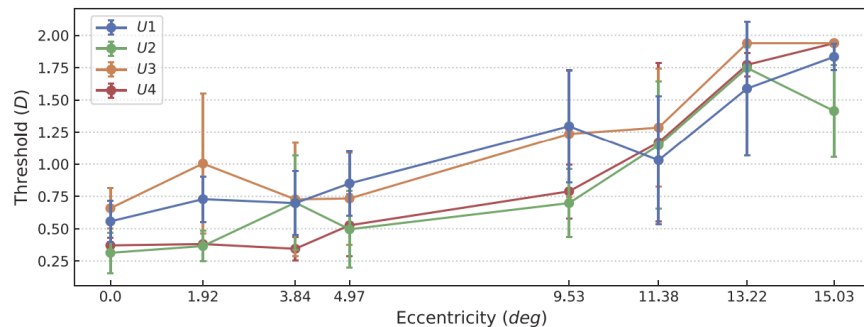
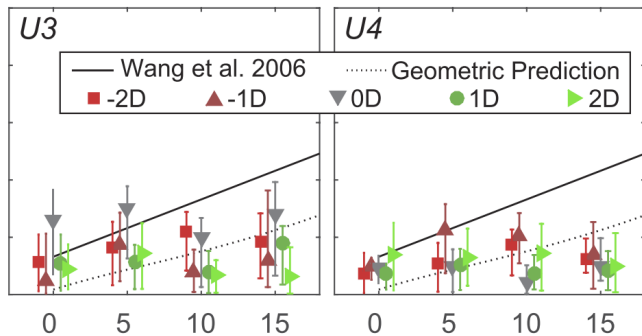
Accommodation



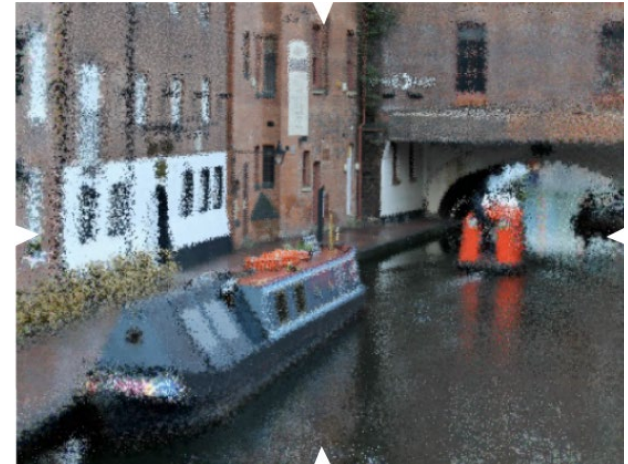
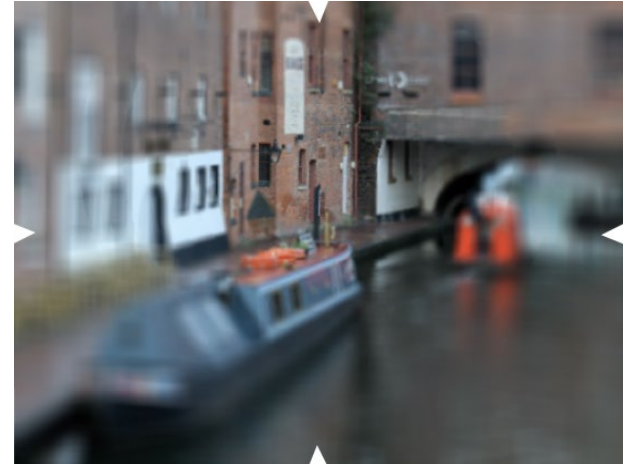
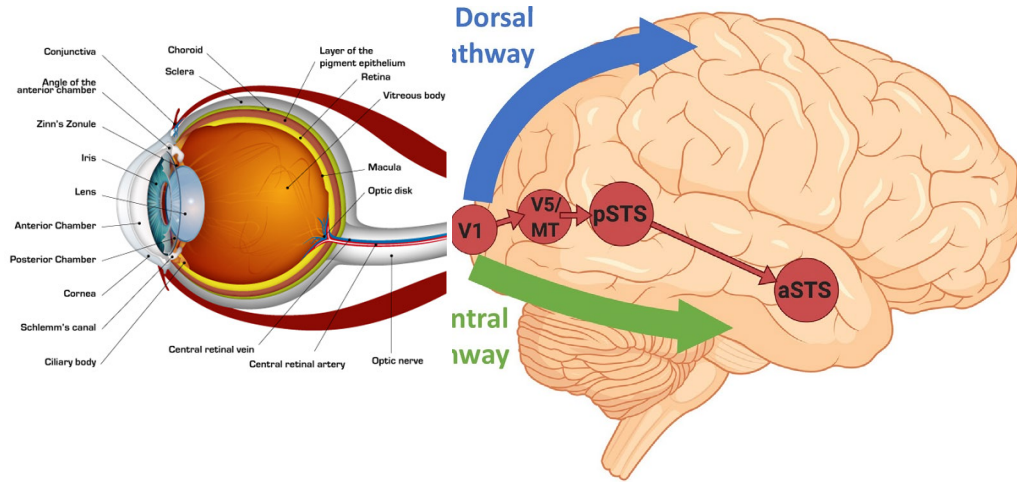




Accommodation – Visual or Muscular?



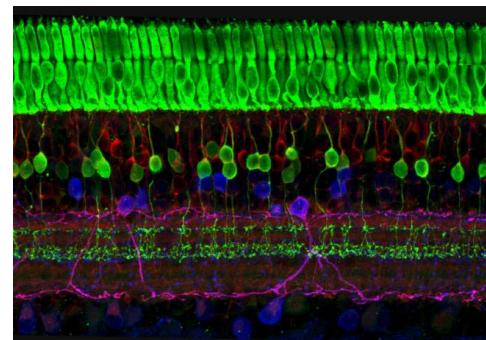
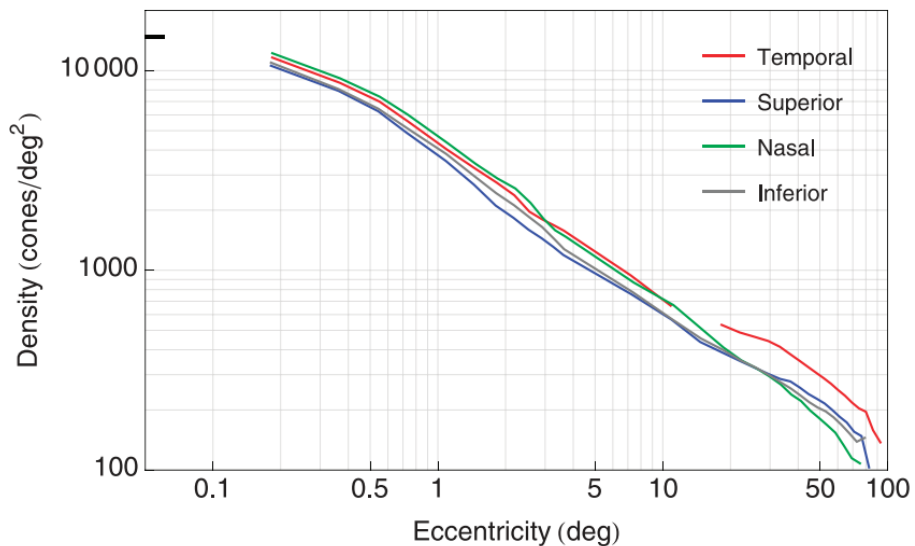
Visual Cortex



Computational Models

Adaptive Spatial Acuity

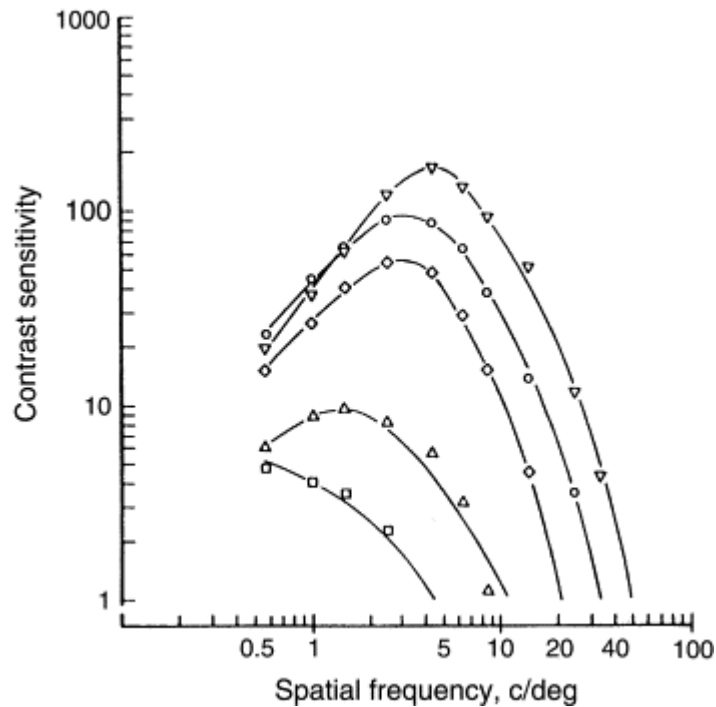
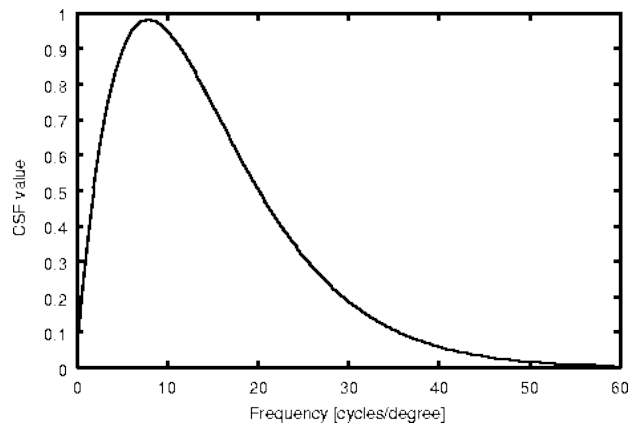
Anatomical Statistics



$$d_{gf}(r, k) = d_{gf}(0) \times \left[a_k \left(1 + \frac{r}{r_{2,k}} \right)^{-2} + (1 - a_k) \exp \left(-\frac{r}{r_{e,k}} \right) \right]$$

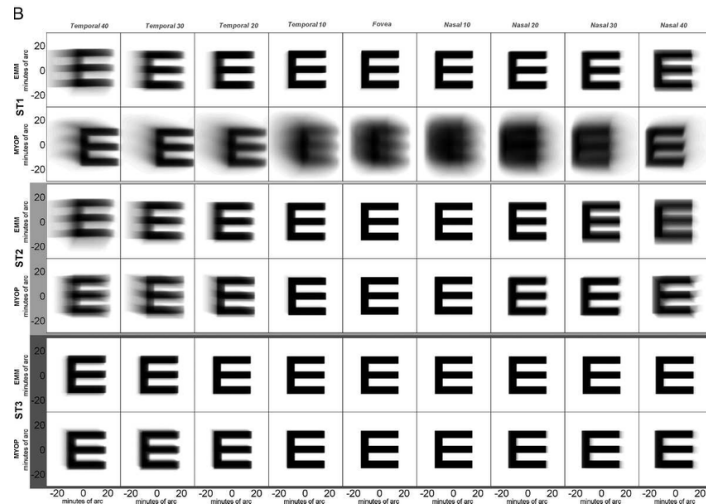
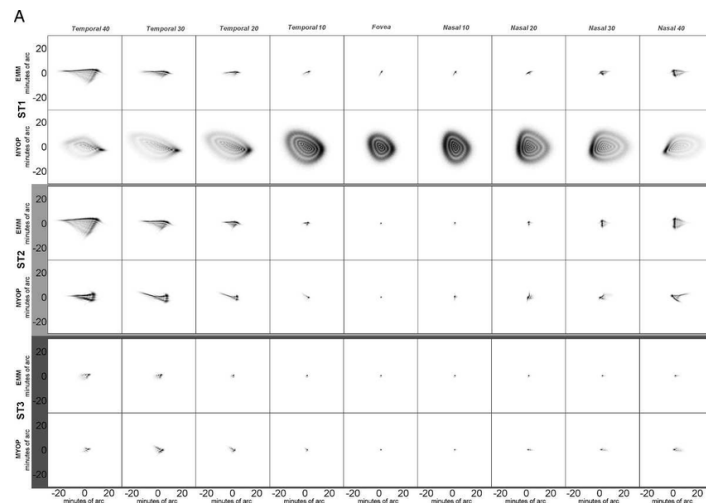
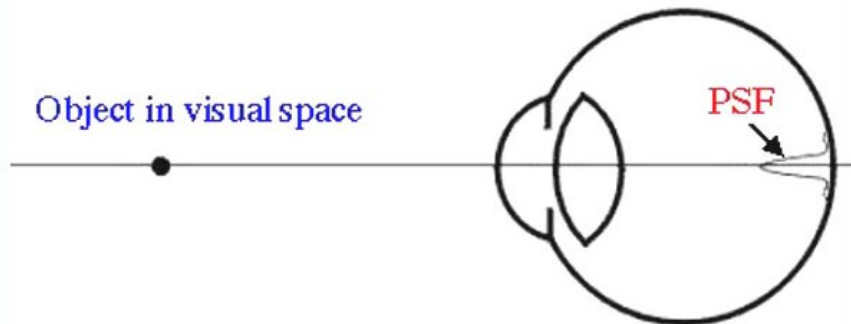
Adaptive Spatial Acuity

Contrast Sensitivity Function



Adaptive Spatial Acuity

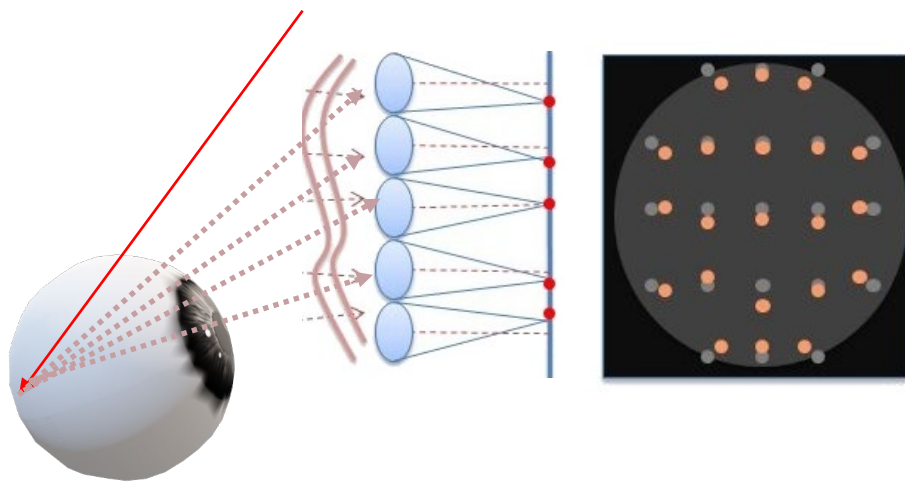
Point Spread Function



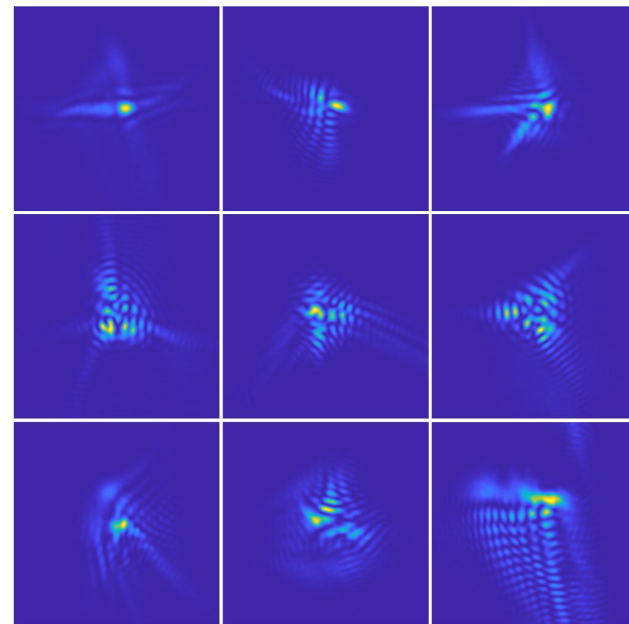


Adaptive Spatial Acuity

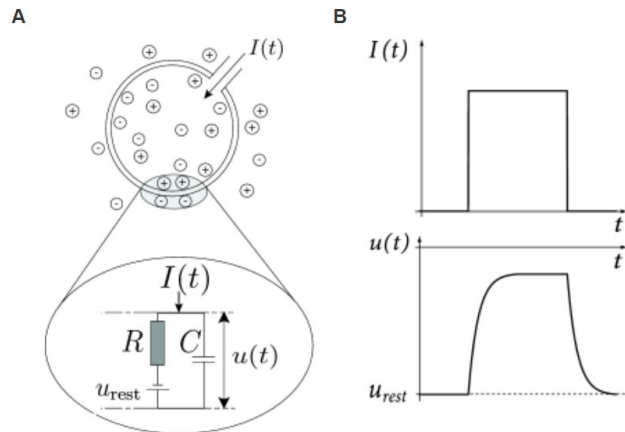
Point Spread Function



Shack-Hartmann wavefront sensor



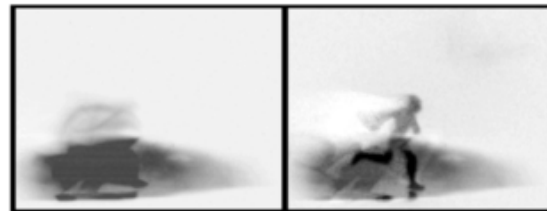
Integrate-and-Fire



$$\begin{aligned} I(t) &= I_R + I_C \\ &= \frac{u(t) - u_{\text{rest}}}{R} + C \frac{du}{dt} \\ \tau_m \frac{du}{dt} &= -[u(t) - u_{\text{rest}}] + R I(t) \end{aligned}$$



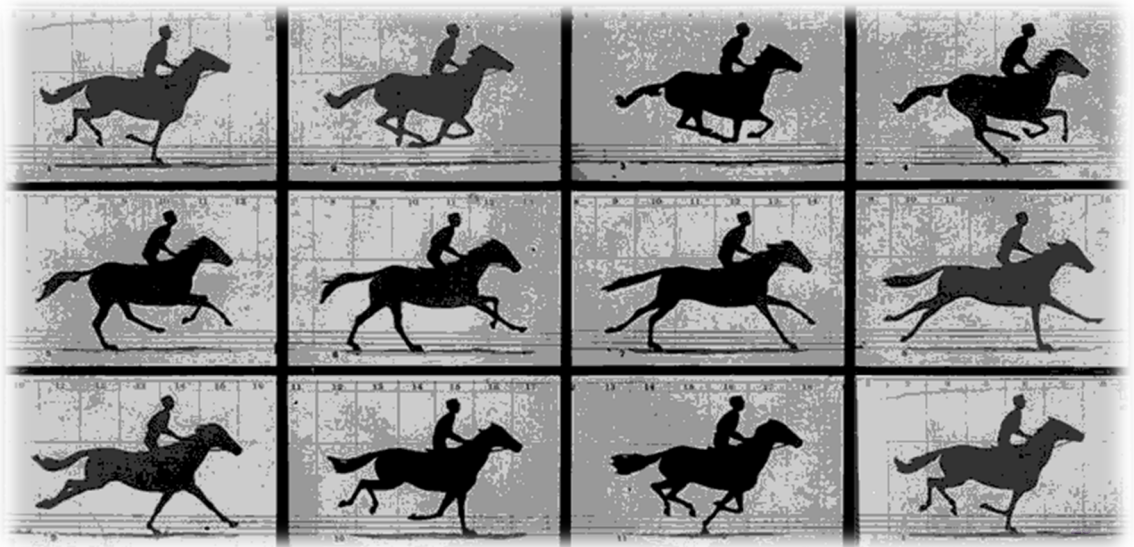
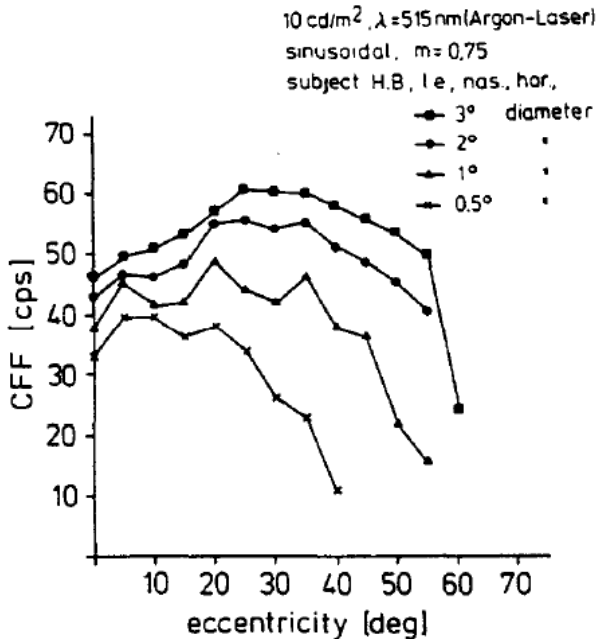
Conventional Camera Event Camera



Temporal Model: Our Eyes Sample @ 30FPS

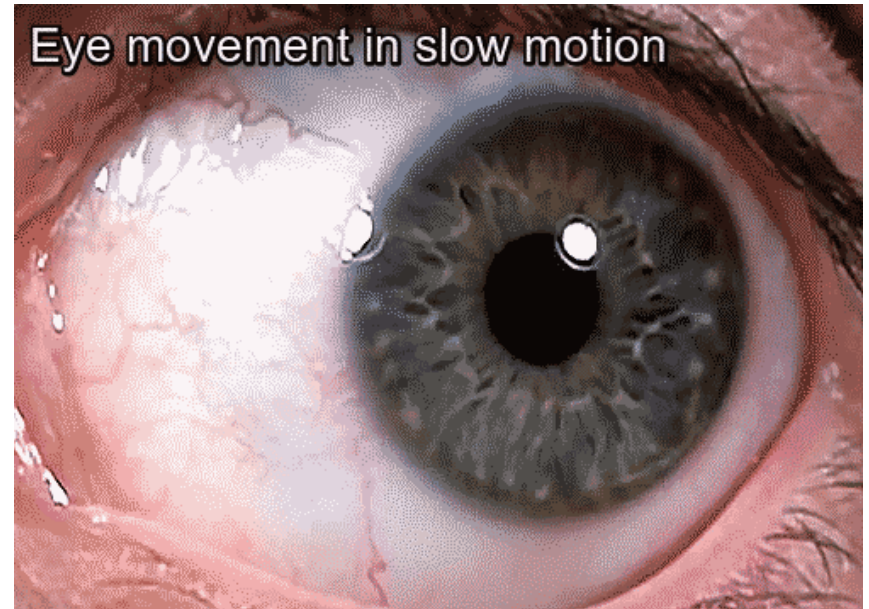
Persistence of vision/Flicker Fusion

$$FoV(\text{deg}) = n1 \log(l) + n2 \log(a) + n3 \log(r) + n4$$



Fun Biophysical Models

$$F(t) = At \exp\left[-\frac{t^\mu}{\tau^\mu}\right]$$

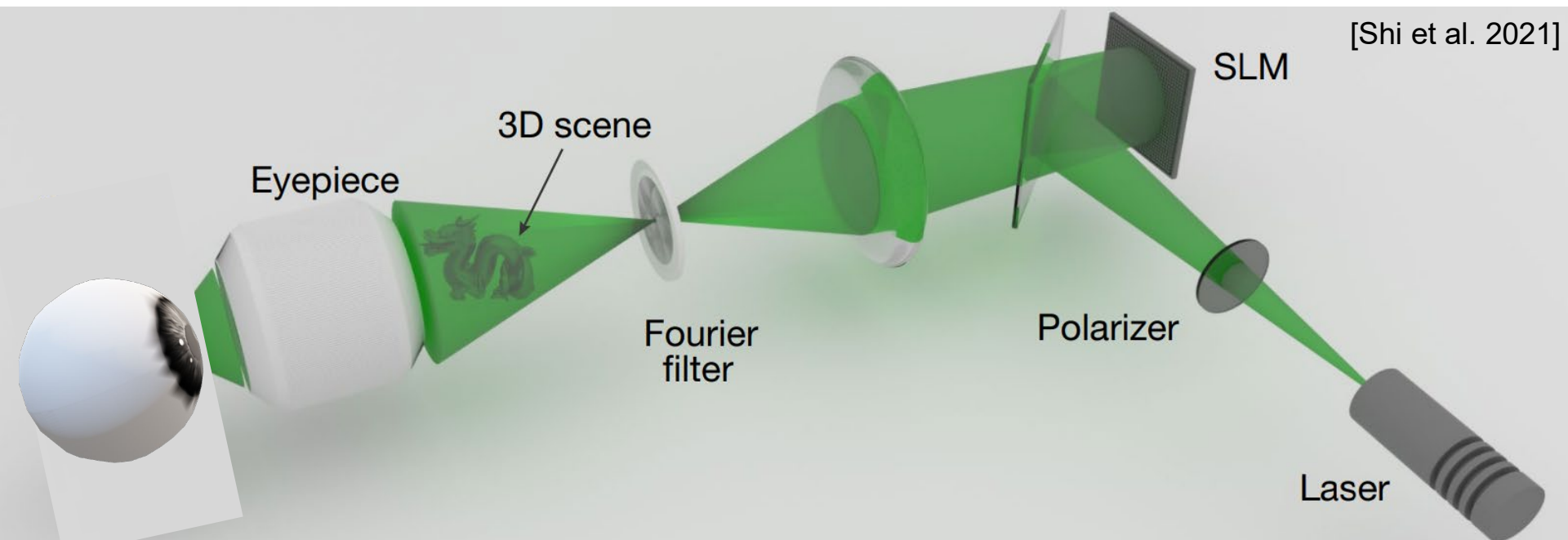


Example Applications

Computational Display and Rendering

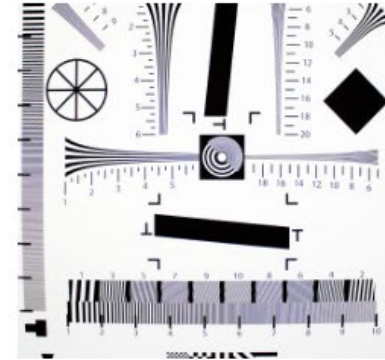
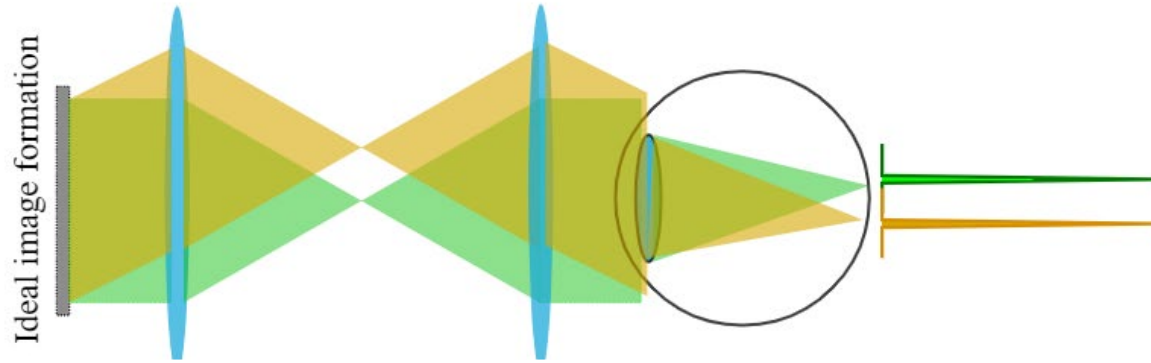


Application I – Holographic Display





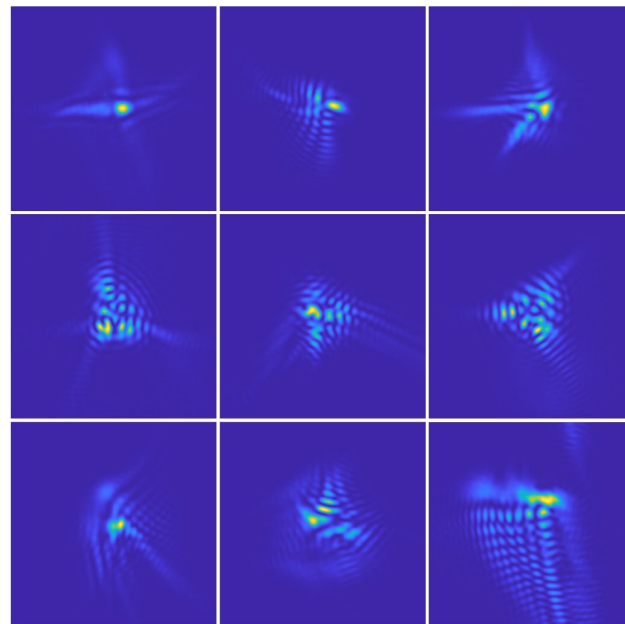
Applications – Holographic Display





Retina-in-the-Loop Holographic Display

$$G_{psf} \approx =$$



Gaze-Aware Holographic Optimization

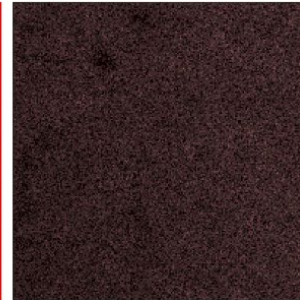
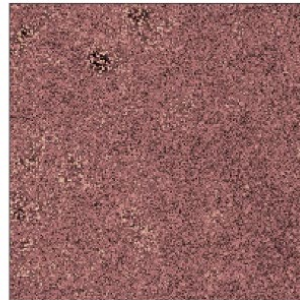
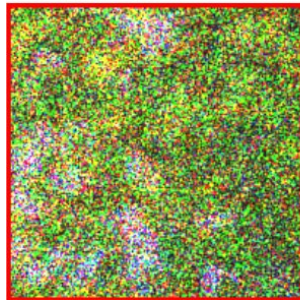
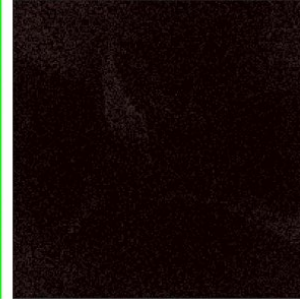
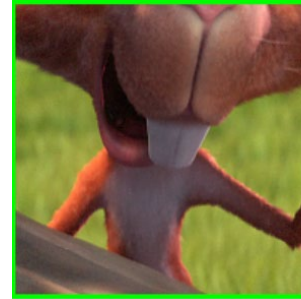
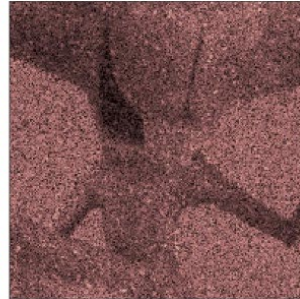
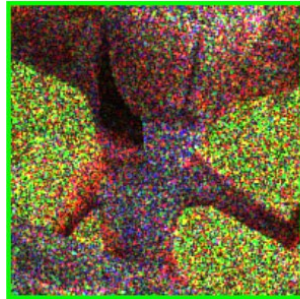
$$PSF(z) \triangleq z * G_{psf}$$

$$\mathcal{L}_{ours} = \sum_x \sum_y mask_{foveation}(x, y) ||PSF(z)_{x,y} - I_{x,y}||^2$$



Original

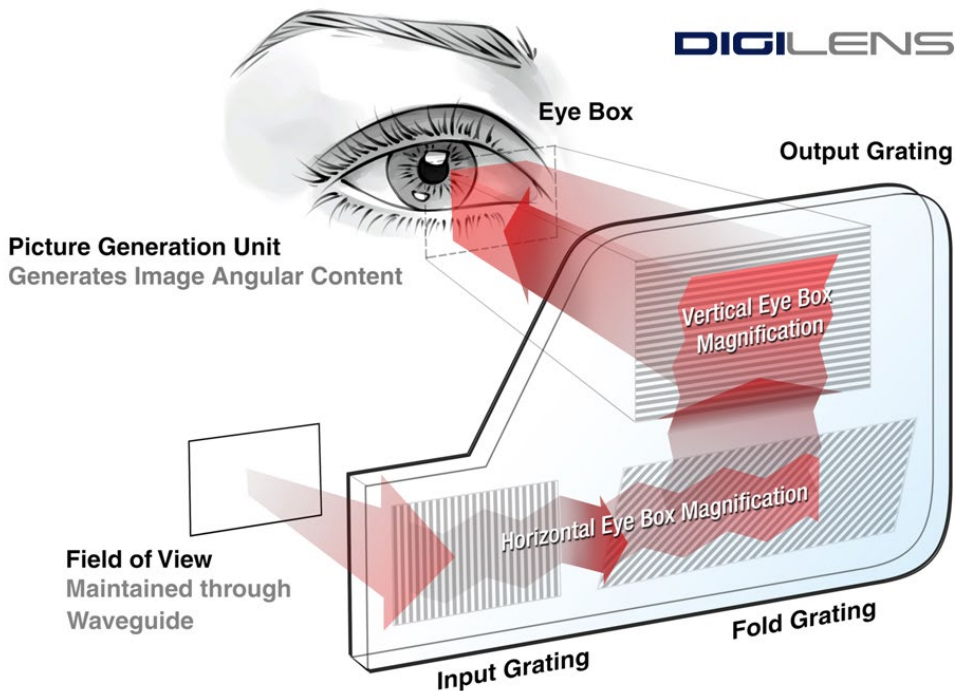
Optimized



Application II – Transparency Control



Why Transparency Matters?



Why Transparency Matters?

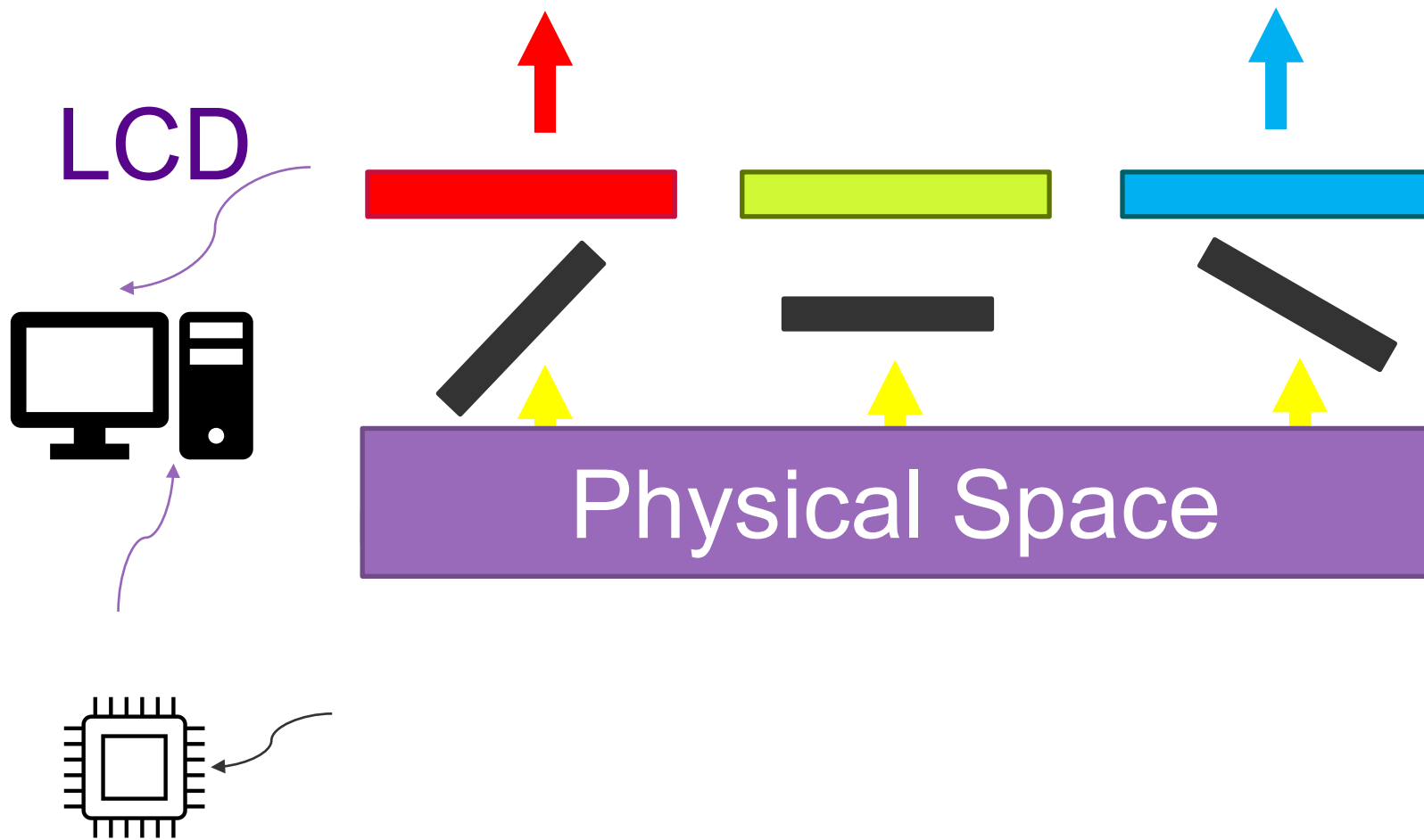


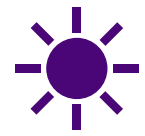
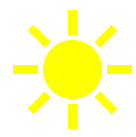
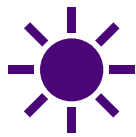
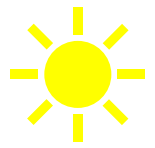
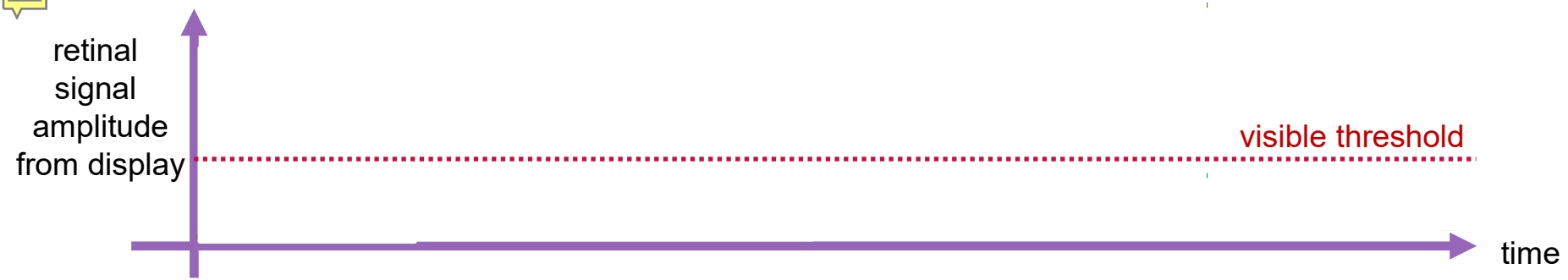
Why Transparency Matters?

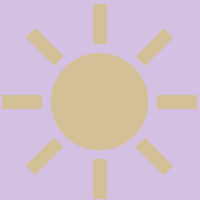


Why Transparency Matters?





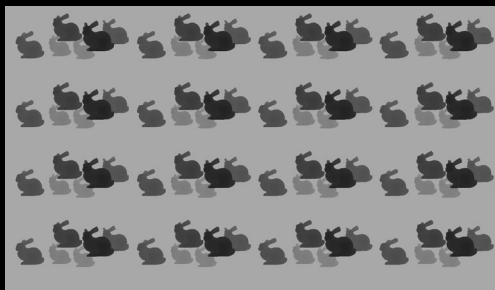




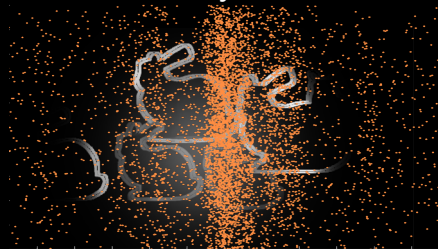


Application III – Optimizing Rendering

Low-Res 4D Rendering



Perceptual-Content Analysis



$$w_d^j(k) \approx \int_0^{B_s} \sum_{l=1}^2 -1^{l-1} a \omega_x^l \left(3a_3(a\omega_x^l(k - \hat{k}_1))^2 + 2a_2(a\omega_x^l(k - \hat{k}_1)) + a_1 \right) d\omega_x^l$$

$$\propto \int_0^{B_s} \omega_x^l \left(3a_3 a \omega_x^{l2} (2k - \hat{k}_1 - \hat{k}_2) + 2a_2 \omega_x^l \right) d\omega_x^l$$

$$\propto 9a_3 a (2k - \hat{k}_1 - \hat{k}_2) B_s + 8a_2$$

Using the estimation of w_d^j above, we obtain

$$W(k_1, k_2) \propto \iint \left(9a_3 a \left(2 \frac{\omega_x^i}{\omega_x^j} - \hat{k}_1 - \hat{k}_2 \right) B_s + 8a_2 \right) \left(t \left(\frac{\omega_x^i}{\omega_x^j}, \hat{k}_1, \omega_x^i \right) - t \left(\frac{\omega_x^i}{\omega_x^j}, \hat{k}_2, \omega_x^i \right) \right) d\omega_x^i d\omega_x^j$$

$$= \iint \frac{18a_3 a \omega_x^i}{\omega_x^j} B_s \left(\sum_{l=1}^2 -1^{l-1} \text{sinc}(a\omega_x^l - a\hat{k}_1 \omega_x^l) \right) d\omega_x^i d\omega_x^j$$

$$+ \iint \left(8a_2 - 9(\hat{k}_1 + \hat{k}_2) a_3 a B_s \right) \left(\sum_{l=1}^2 -1^{l-1} \text{sinc}(a\omega_x^l - a\hat{k}_1 \omega_x^l) \right) d\omega_x^i d\omega_x^j$$

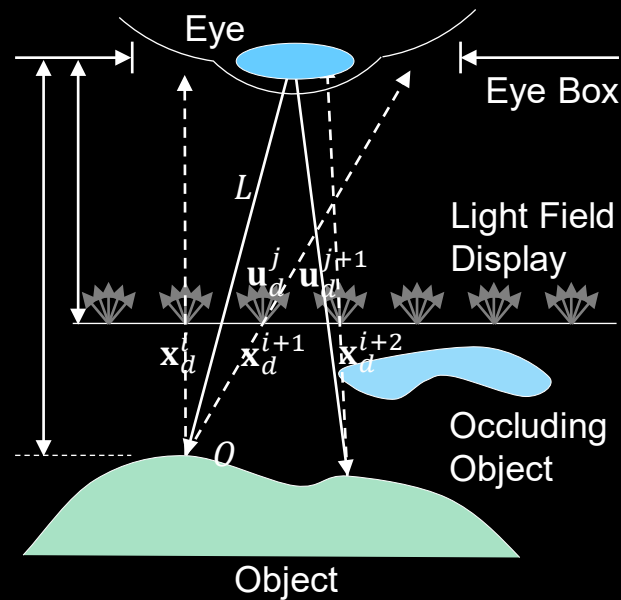
$$= 18a_3 B_s \int \left(\sum_{l=1}^2 -1^{l-1} \left(\text{sinc}(a\omega_x^l) (-\text{Ci}(a\hat{k}_1 \omega_x^l) + \text{Si}(a\omega_x^l - a\hat{k}_1 \omega_x^l) + \cos(a\omega_x^l) \text{Si}(a\hat{k}_1 \omega_x^l) \right) \right) d\omega_x^l$$

$$- \left(8a_2 - 9(\hat{k}_1 + \hat{k}_2) a_3 a B_s \right) \int \left(\sum_{l=1}^2 -1^{l-1} \frac{\text{Si}(a\omega_x^l - a\hat{k}_1 \omega_x^l)}{a\hat{k}_1} \right) d\omega_x^l$$

Sampling



Post-Filtering:
Separable 4D
RBF Occlusion-
aware
Reconstruction



Future Directions

Future

- Individualized model
- Predictive and temporal effects
- Visual metrics for realism
- Differentiable->optimize
- Multi-modality

NYU Immersive Computing Lab

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