

Reflectance fields and transport matrices

Computational Photography

the superposition principle



=



+



photo taken under two light sources =
sum of photos taken under each source individually

the superposition principle

Synthetic photo



Real photo



photo taken under two light sources =
sum of photos taken under each source individually

the superposition principle

Synthetic photo



Diff between synthetic and real photos

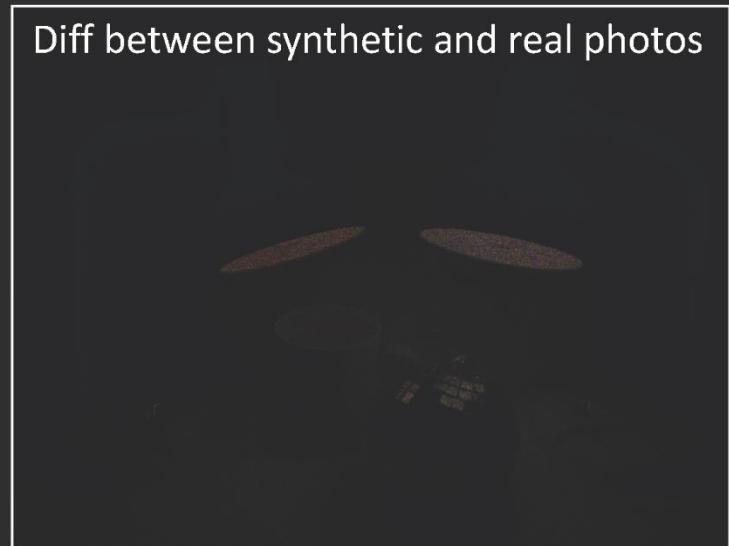


photo taken under two light sources =
sum of photos taken under each source individually

image-based relighting



=



image-based relighting



=



+



Weight 1

x 1

Weight 2

x 1

image-based relighting



=



+



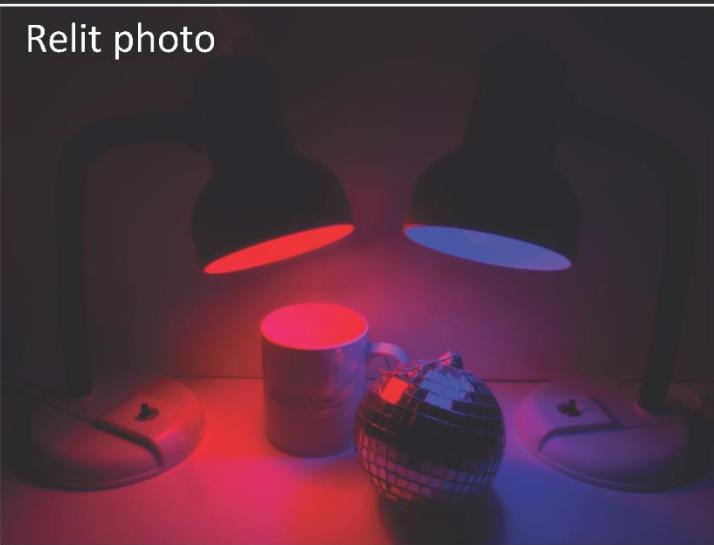
Weight 1

x 1

Weight 2

x 0

image-based relighting



=

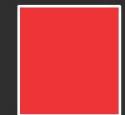


+



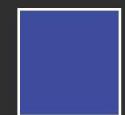
Weight 1

x



Weight 2

x



Relit photo **P**



=

photo with light 1 turned
on **T₁**



Weight 1
 $\times \boxed{\mathbf{l}_1} +$

photo with light 2 turned
on **T₂**



Weight 2
 $\times \boxed{\mathbf{l}_2}$

Relit photo **P**



=

photo with light 1 turned
on **T₁**



Weight 1
 $\times \boxed{\mathbf{l}_1}$ +

photo with light 2 turned
on **T₂**



Weight 2
 $\times \boxed{\mathbf{l}_2}$

$$\mathbf{p} = \sum_{i=1}^2 \mathbf{T}_i \times \boxed{\mathbf{l}_i}$$

Relit photo **P**



photo with light 1 turned
on **T₁**

Weight 1
 \times **l₁** +

photo with light 2 turned
on **T₂**

Weight 2
 \times **l₂**

$$\begin{matrix} \uparrow \\ n \\ \textbf{p} \\ \downarrow \\ n \text{ pixel values} \end{matrix} = \sum_{i=1}^2 \textbf{T}_i \times \textbf{l}_i$$

Relit photo \mathbf{p}

photo with light 1 turned
on \mathbf{T}_1

photo with light 2 turned
on \mathbf{T}_2

Weight 1

$$x \boxed{\mathbf{l}_1} +$$

Weight 2

$$x \boxed{\mathbf{l}_2}$$



$$\mathbf{p} = \sum_{i=1}^2 \mathbf{T}_i \times \boxed{\mathbf{l}_i}$$

n pixel values

Relit photo p

photo with light 1 turned
on T_1



Weight 1

$$\times \boxed{l_1}$$

photo with light 2 turned
on T_2



Weight 2

$$\times \boxed{l_2}$$

$$p = \sum_{i=1}^2 T_i \times \boxed{l_i}$$

n pixel values

Contribution of
the source



Relit photo p

photo with light 1 turned
on T_1



Weight 1
 $\times [l_1]$ +

photo with light 2 turned
on T_2



Weight 2
 $\times [l_2]$

Number of
controllable
sources



2



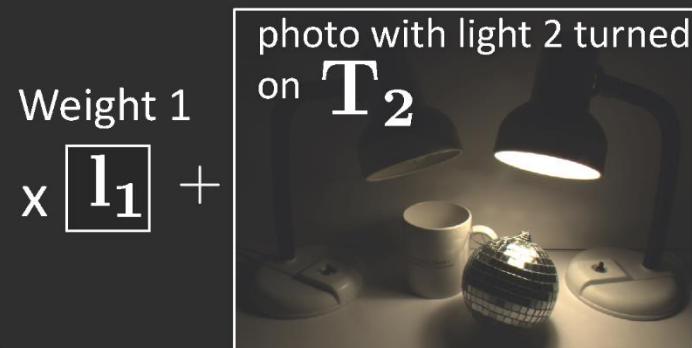
n pixel values

$$p = \sum_{i=1}^2 T_i \times [l_i]$$

Contribution of
each source



$[l_i]$



Weight 1
 $x \boxed{\mathbf{l}_1}$ +
Weight 2
 $x \boxed{\mathbf{l}_2}$

$$\mathbf{p} = \sum_{i=1}^n \mathbf{T}_i \times \boxed{\mathbf{l}_i}$$

Number of controllable sources \downarrow

Contribution of each source \downarrow

n pixel values



二



Weight 1
 x +



Weight 2
 x l_2

$$\mathbf{p} = \sum_{i=1}^m \mathbf{T}_i \mathbf{l}_i$$

Number of controllable sources m

Contribution of each source \mathbf{l}_i

n pixel values

Relit photo \mathbf{P}



=

photo with light 1 turned
on \mathbf{T}_1



Weight 1
 $\times \boxed{\mathbf{l}_1} +$

photo with light 2 turned
on \mathbf{T}_2



Weight 2
 $\times \boxed{\mathbf{l}_2}$

$$\begin{array}{c} \uparrow \\ \mathbf{p} = \\ \downarrow \\ n \text{ pixel values} \end{array}$$

$$\begin{array}{c} \uparrow \\ \mathbf{T} \\ \downarrow \\ n \times m \end{array}$$

$$\begin{array}{c} \uparrow \\ \mathbf{l} \\ \downarrow \\ m \end{array}$$

m independent
illumination
degrees of freedom

the light transport matrix

Sloan et al 02, Ng et al 03, Seitz et al 05, Sen et al 05, ...

transport matrix represents the set of photos under all possible (controllable) lighting conditions

$$\begin{array}{c} \uparrow \\ n \\ \downarrow \\ \mathbf{p} \\ \downarrow \\ n \text{ pixel values} \end{array}$$

$$= \begin{array}{c} \mathbf{T} \\ n \times m \end{array}$$

$$\begin{array}{c} \uparrow \\ m \\ \downarrow \\ \mathbf{l} \end{array}$$

m independent
illumination
degrees of freedom

Acquiring the Reflectance Field [Debevec et al. 2000]

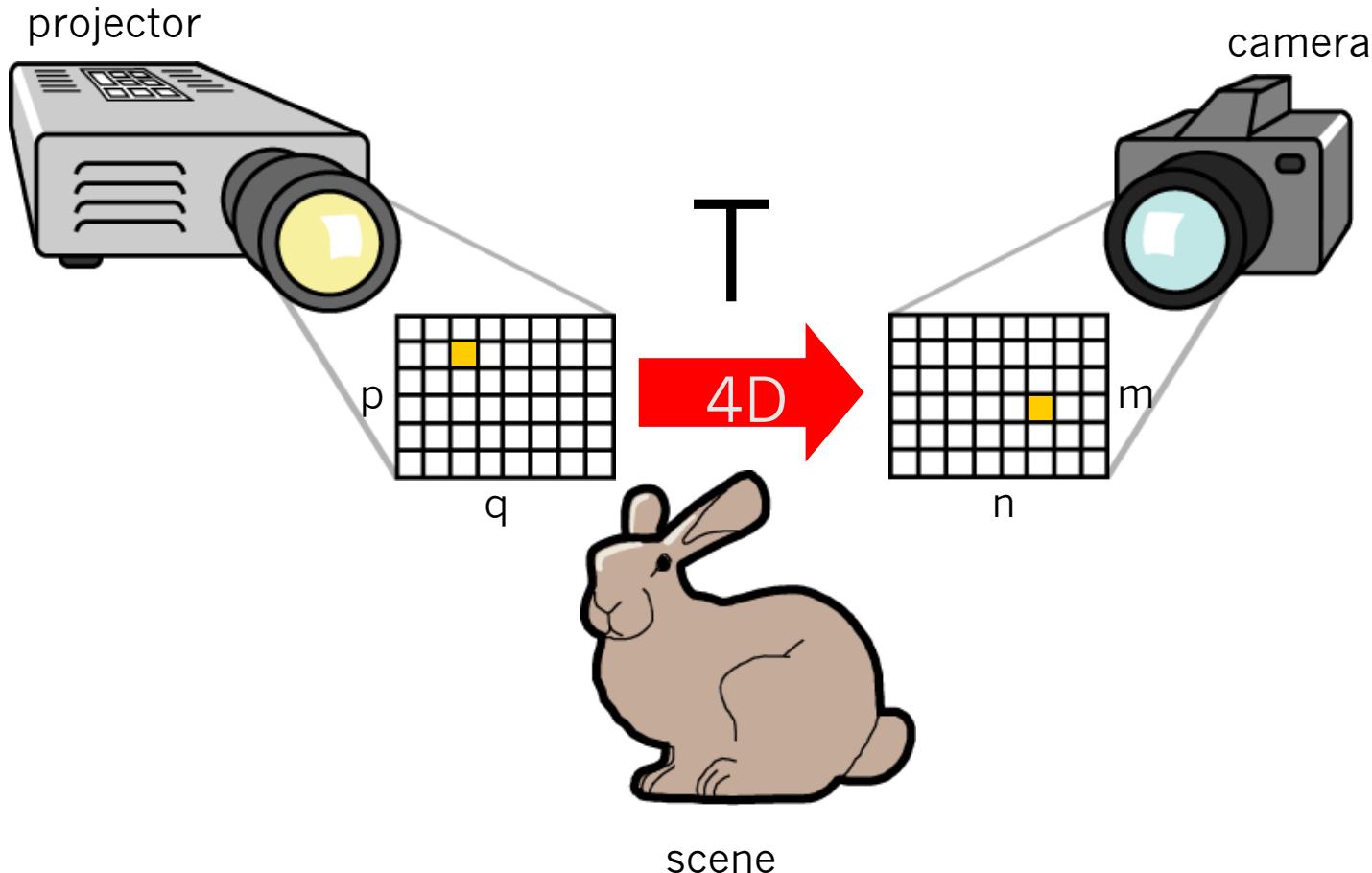
image-based rendering & relighting



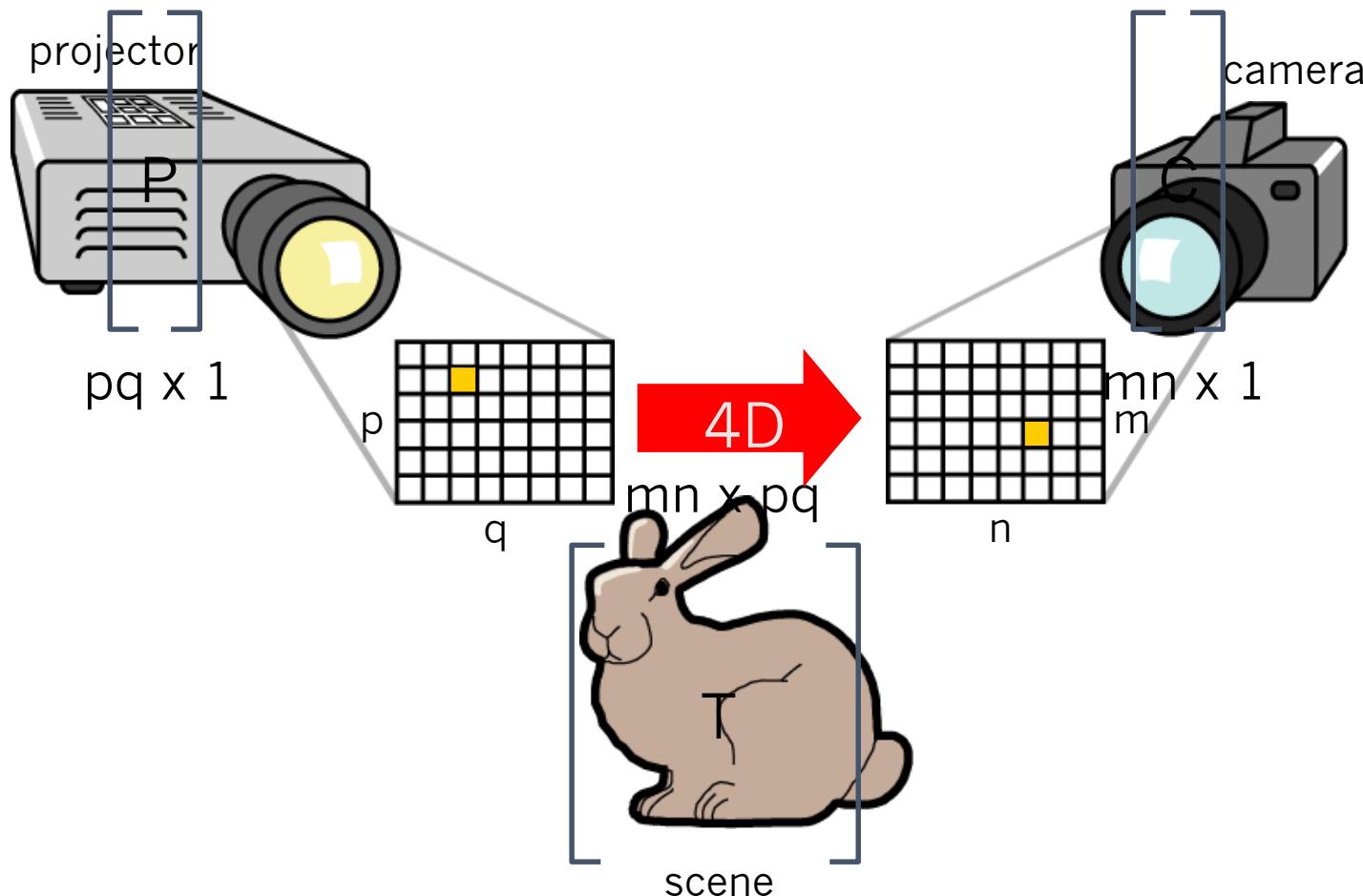
Reflectance field

Performance Relighting [Peers2007]

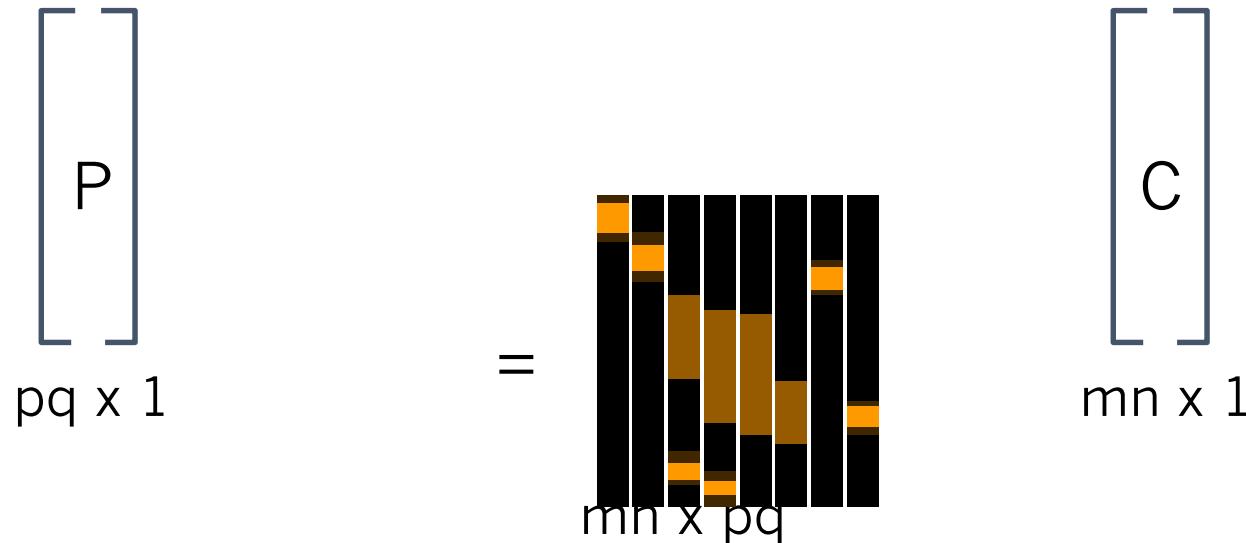
Pixel-to-Pixel Transport



Mathematical Notation



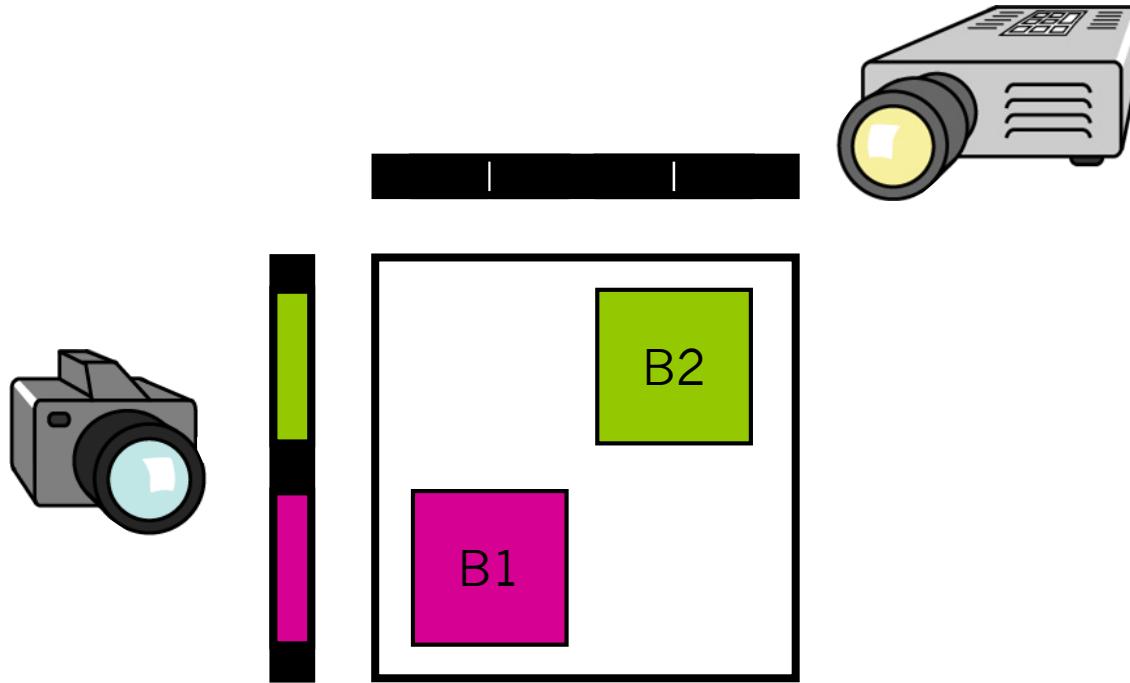
Mathematical Notation

$$\begin{bmatrix} P \\ \end{bmatrix}_{pq \times 1} = \begin{bmatrix} \text{[pq columns]} \\ \vdots \\ \text{[mn columns]} \end{bmatrix}_{mn \times pq} = \begin{bmatrix} C \\ \end{bmatrix}_{mn \times 1}$$


- Matrix properties
 - little interreflection $\leftrightarrow T$ rather sparse
 - many interreflections $\leftrightarrow T$ rather dense

Adaptive Parallel Acquisition

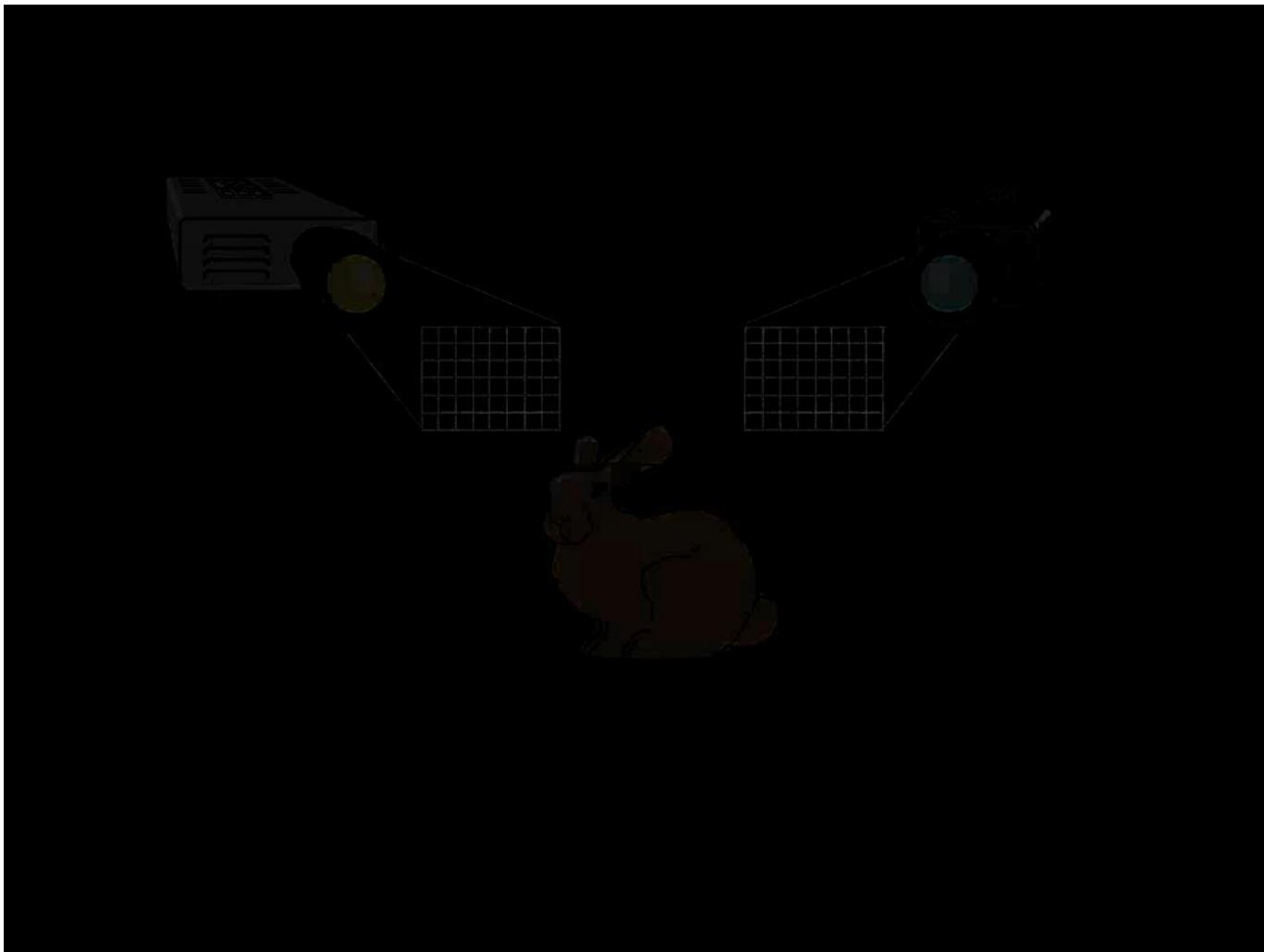
- assumption: sparse matrix
- radiometrically independent blocks can be sensed in parallel



[Sen et al., Dual Photography, SIGGRAPH 2005]

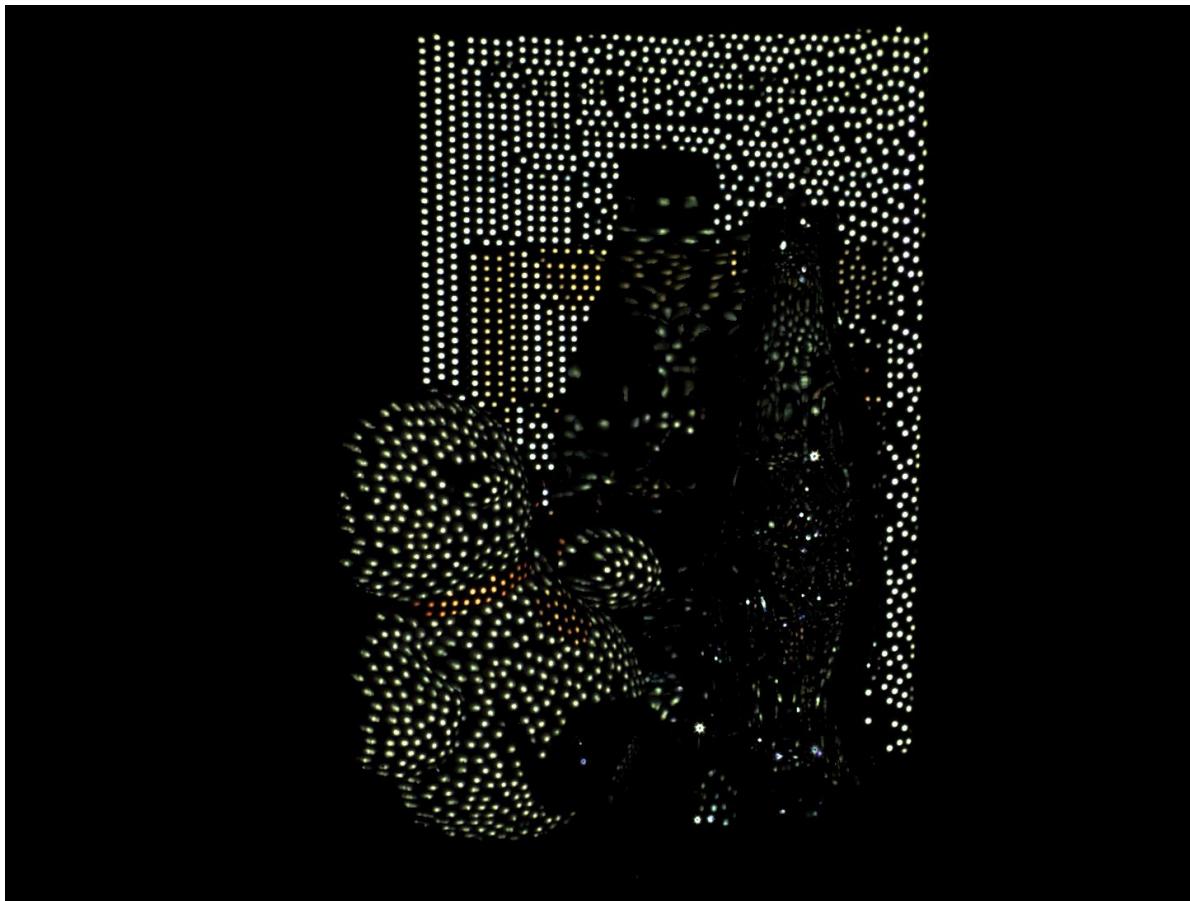
Adaptive Multiplexed Acquisition

- Parallel investigation possible if regions do not overlap in camera image



Adaptive Acquisition

- parallel investigation if regions do not overlap in camera image

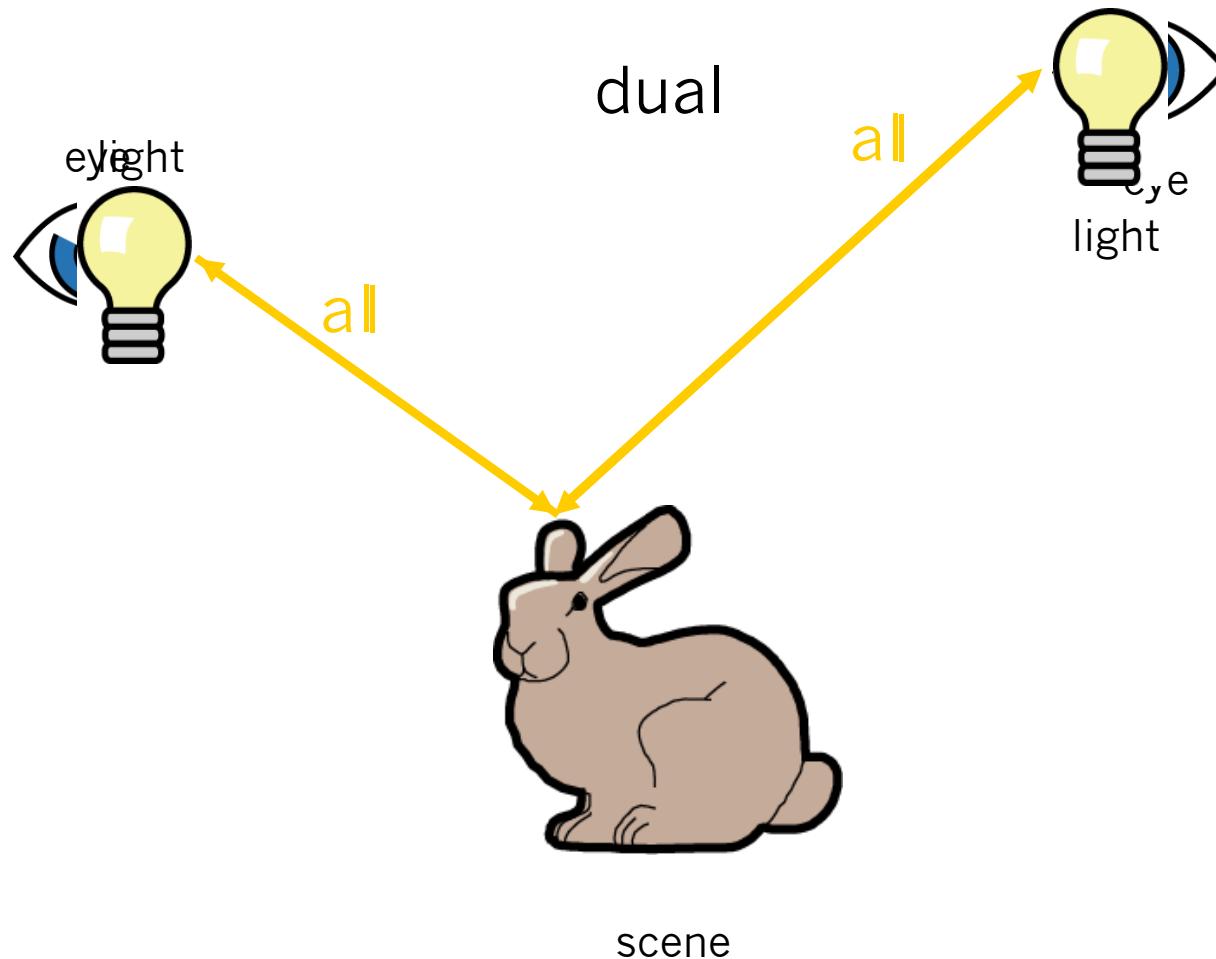


Relighting with 2D Patterns

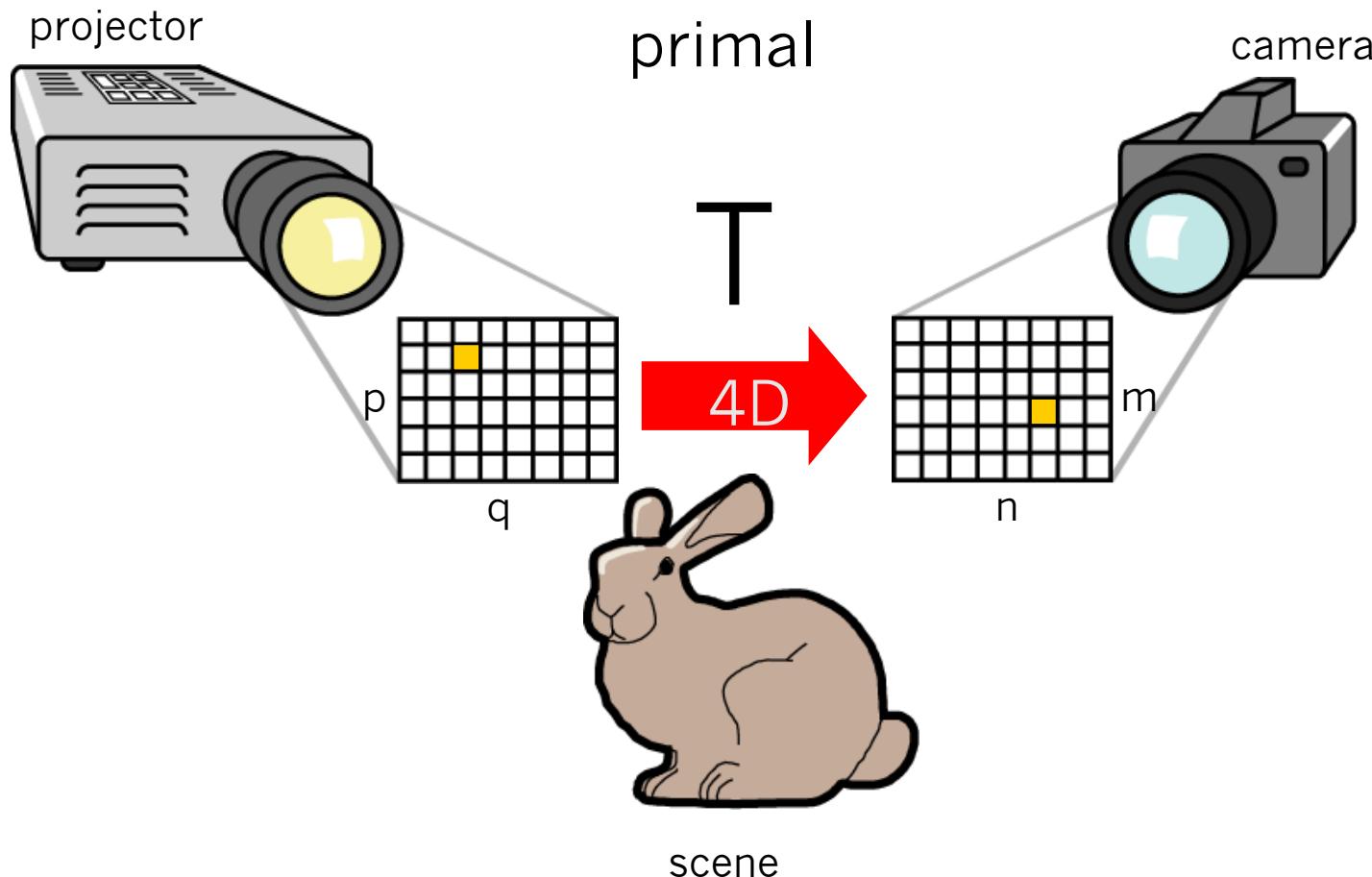
[Ser]



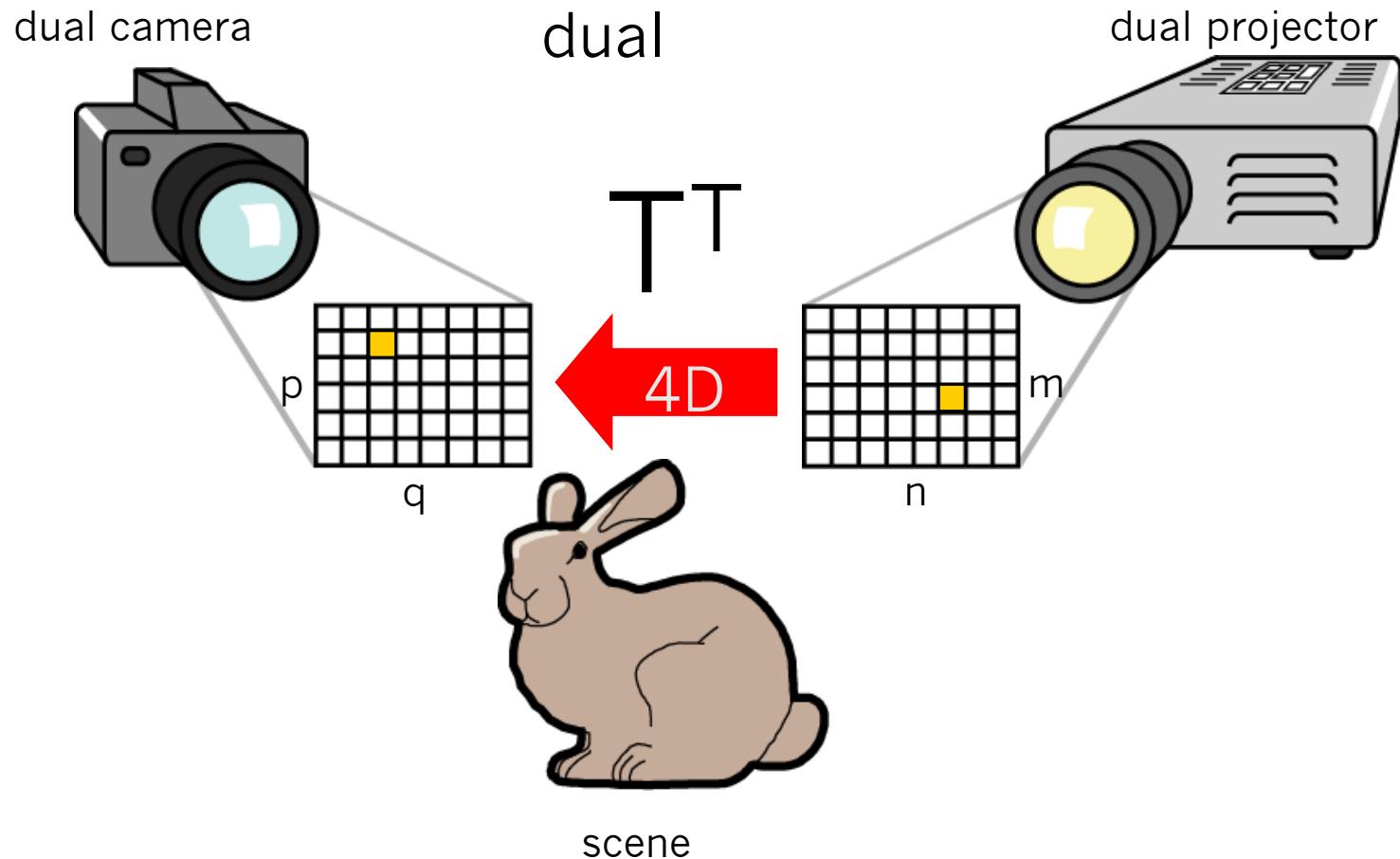
Helmholtz Reciprocity



Dual Photography for Relighting

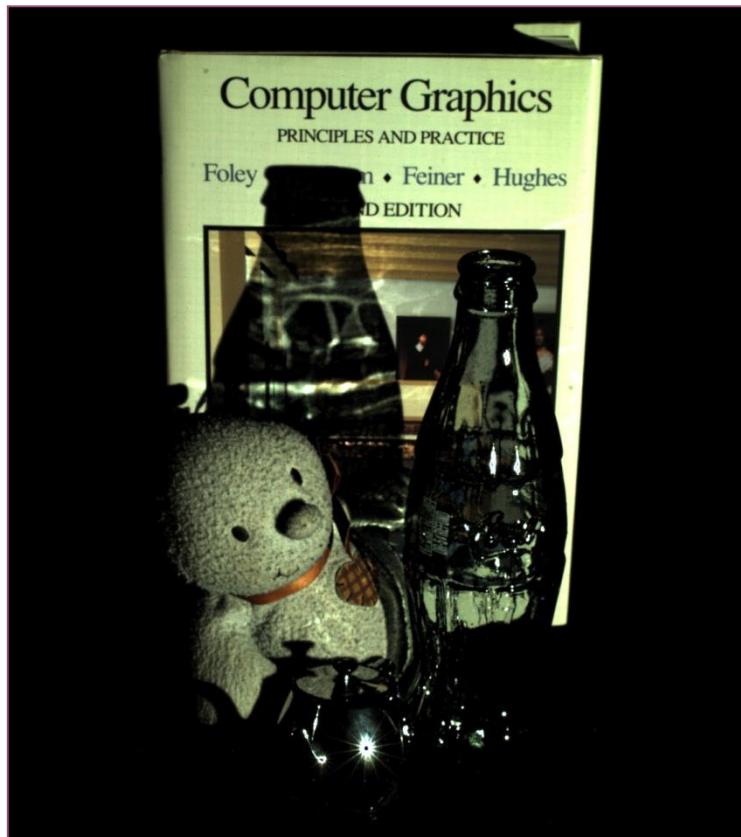


Dual Photography for Relighting

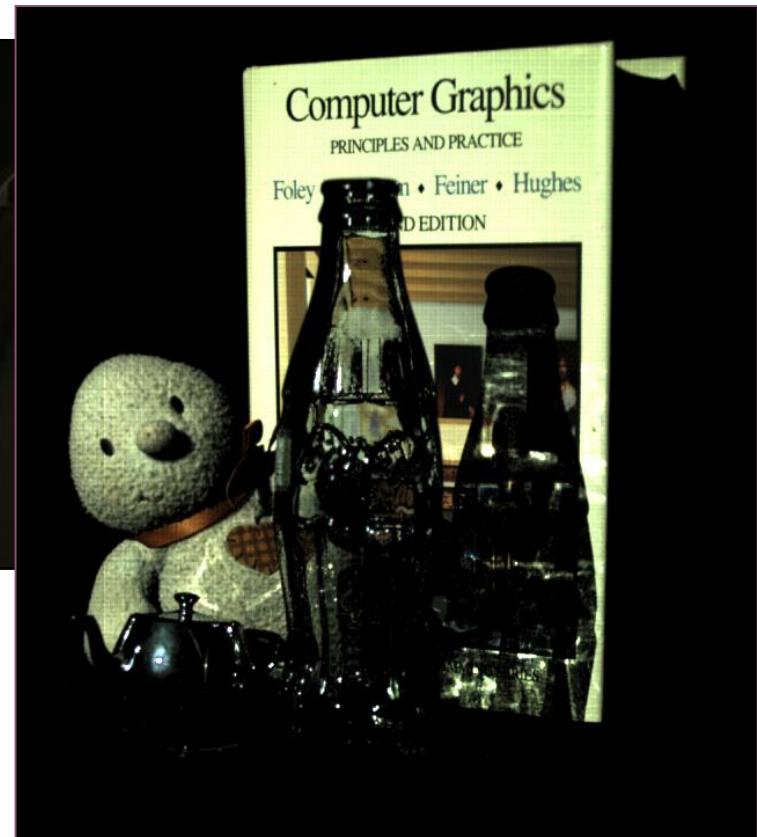


Using the Transposed T Matrix

standard photograph
from camera



dual photograph
from projector



Sample Results

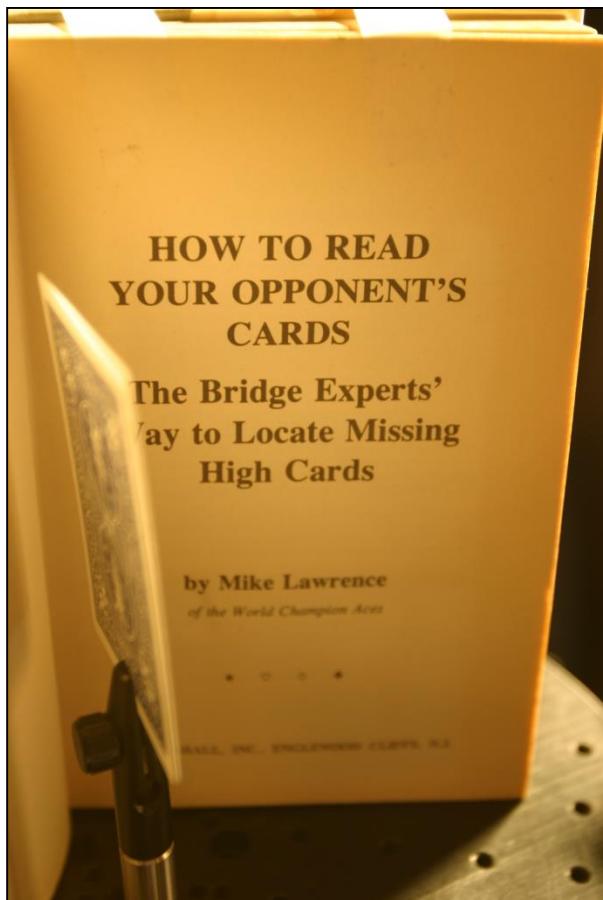


primal

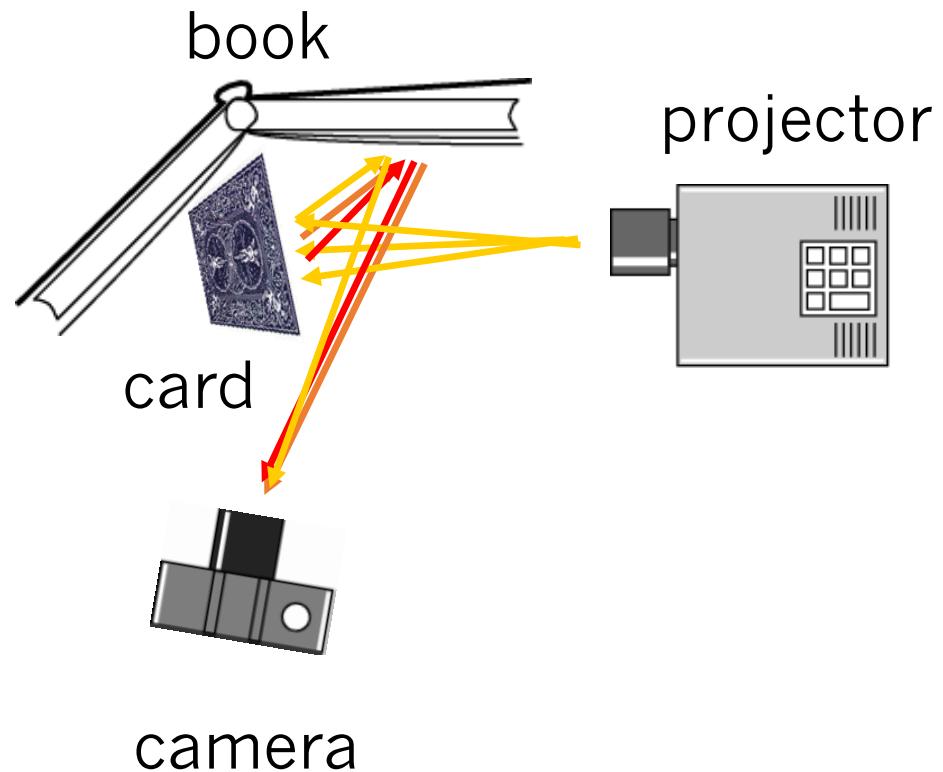


dual

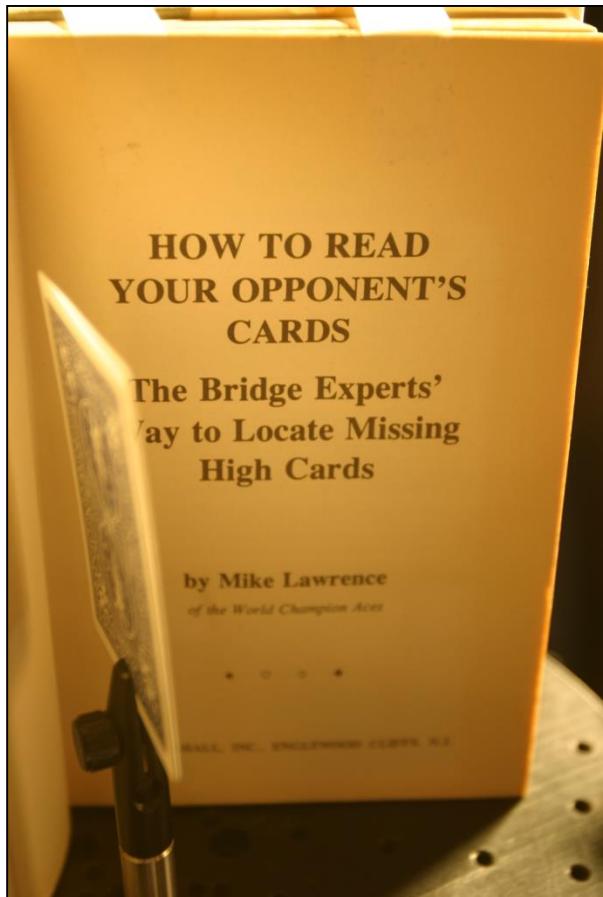
Card Experiment



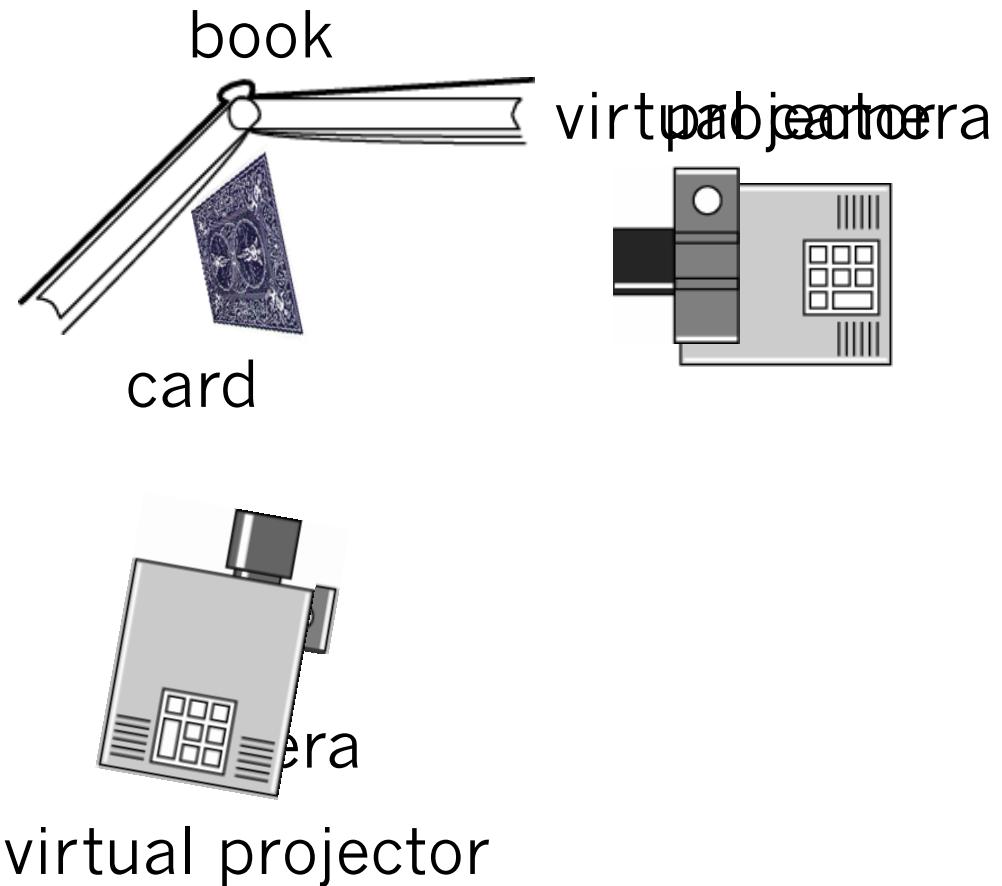
primal



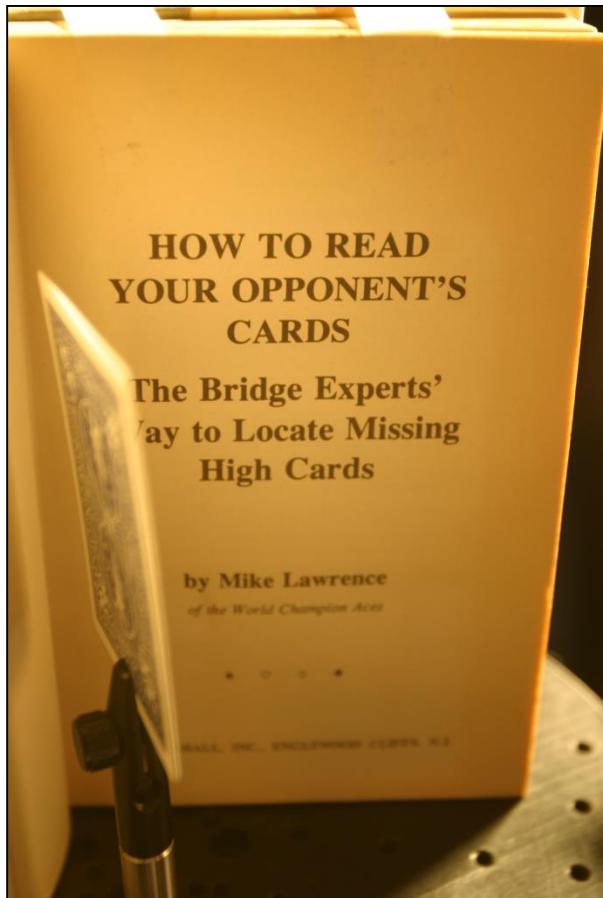
Card Experiment



primal



Card Experiment



primal



dual

Optical Computing [O'Toole 2010]

- Krylov subspace method for eigenspace computation
- Capture Tl, T^2l, T^3l, \dots (converges to T 's principal eigenvector)

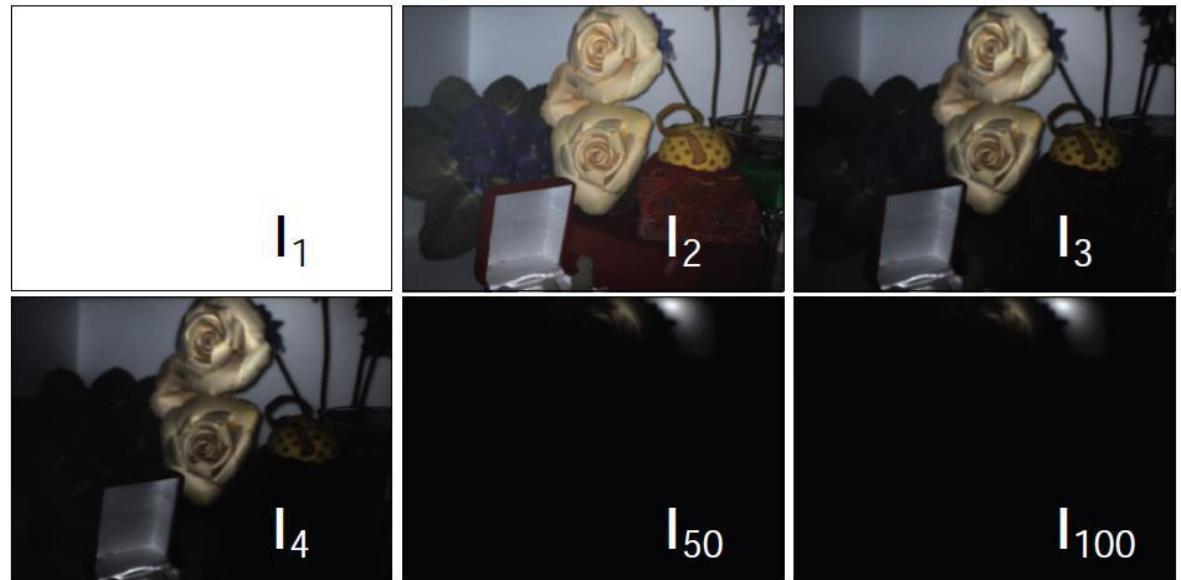
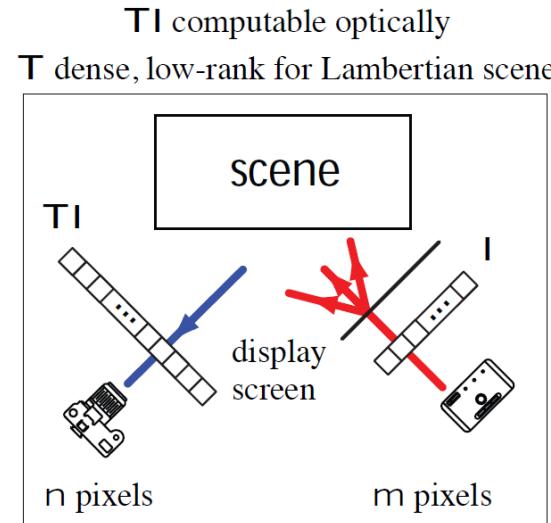
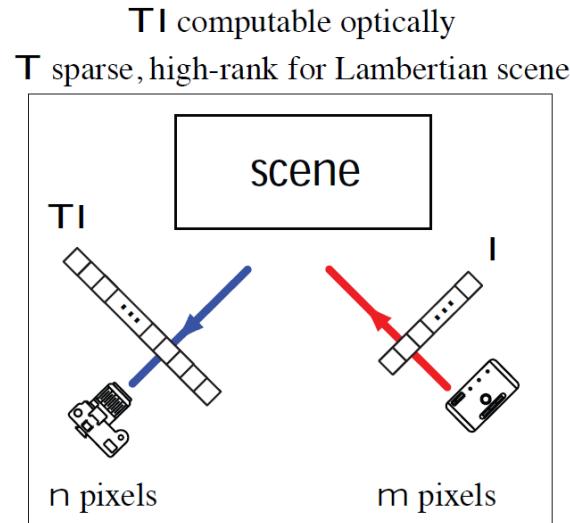


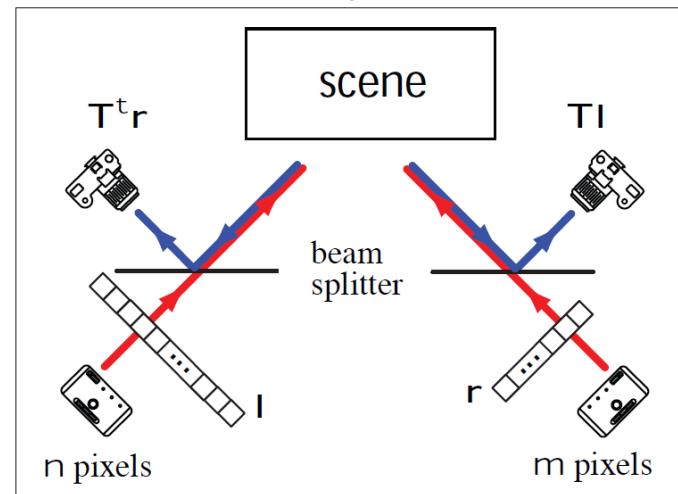
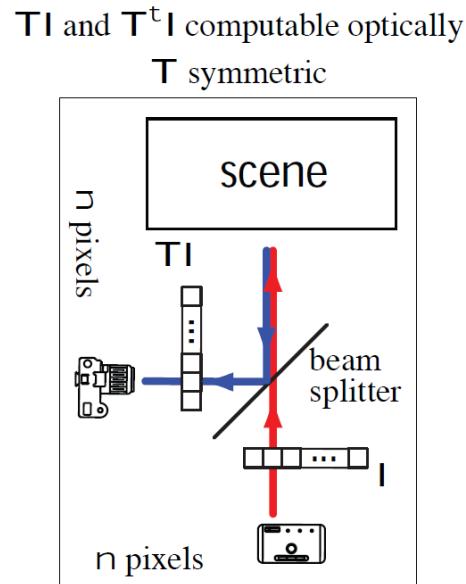
Figure 4: Optical power iteration in action. We used the coaxial arrangement of Figure 2(c) for this example, where a camera and a projector share the same viewpoint and \mathbf{T} is symmetric. We started with a constant illumination vector \mathbf{l}_1 , shown above, so the first photo of the scene was captured under constant illumination. That photo became the next illumination vector, \mathbf{l}_2 , also shown above. The illumination vectors change very little after about 50 captured photos, indicating that a good approximation of \mathbf{T} 's principal eigenvector has already been found.

Optical Computing



(a)

(b)



(c)

(d)

Optical Computing

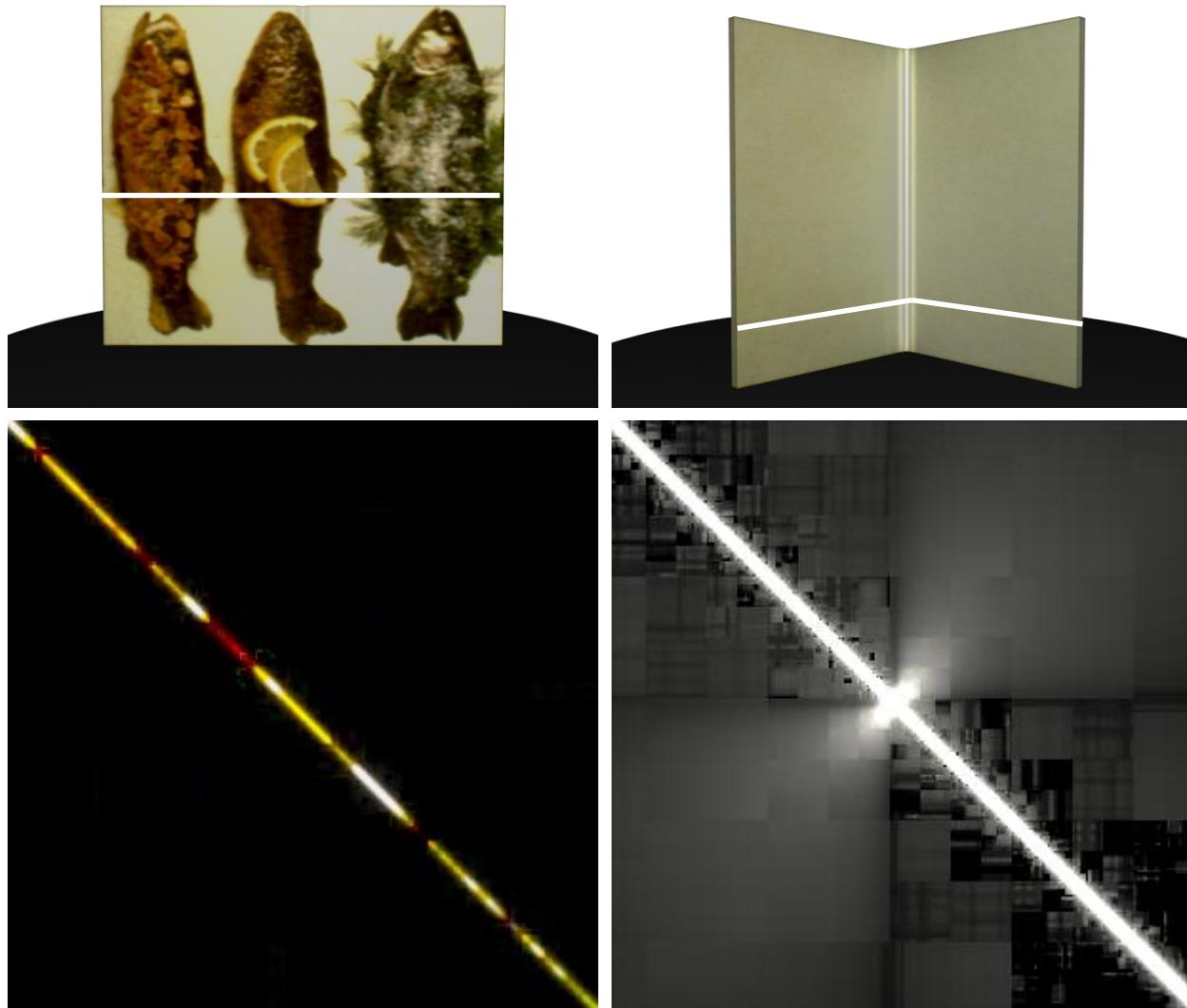


Exploiting light transport properties for photography

Light Transport Matrix

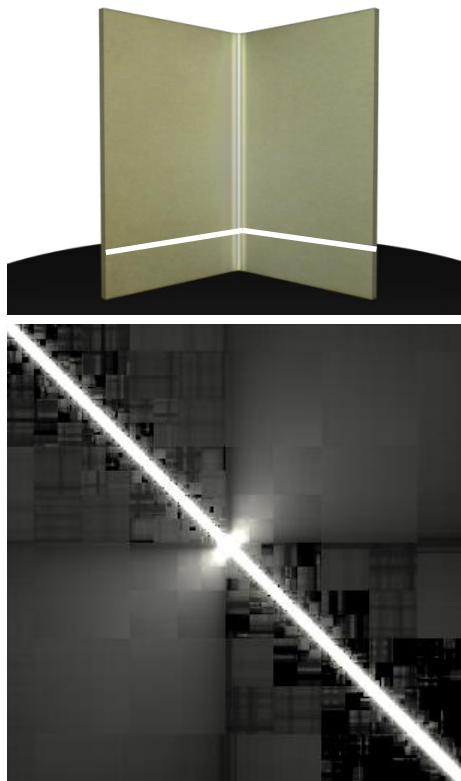


Local vs. Global Reflections



[Garg, Talvala, Levoy, Lisch - EGSR 2006]

Local vs. Global Reflections



- We are interested in the direct reflections.
- Indirect reflections correspond to smooth regions in the matrix.

Getting Rid of Global Effects



- Remove off-diagonal components
- Diagonal entries might still contain global components

Fast Separation of Direct and Global Effects

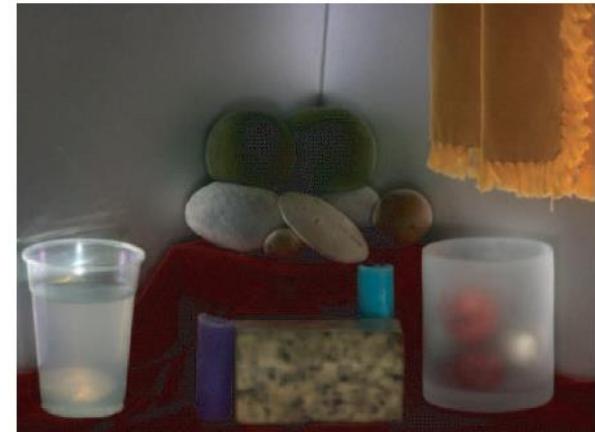
- Main idea: global illumination effects dampen high frequencies
 - illuminate with shifted high frequency patterns
 - only the local illumination will change
 - global illumination will be invariant to phase shifts



image



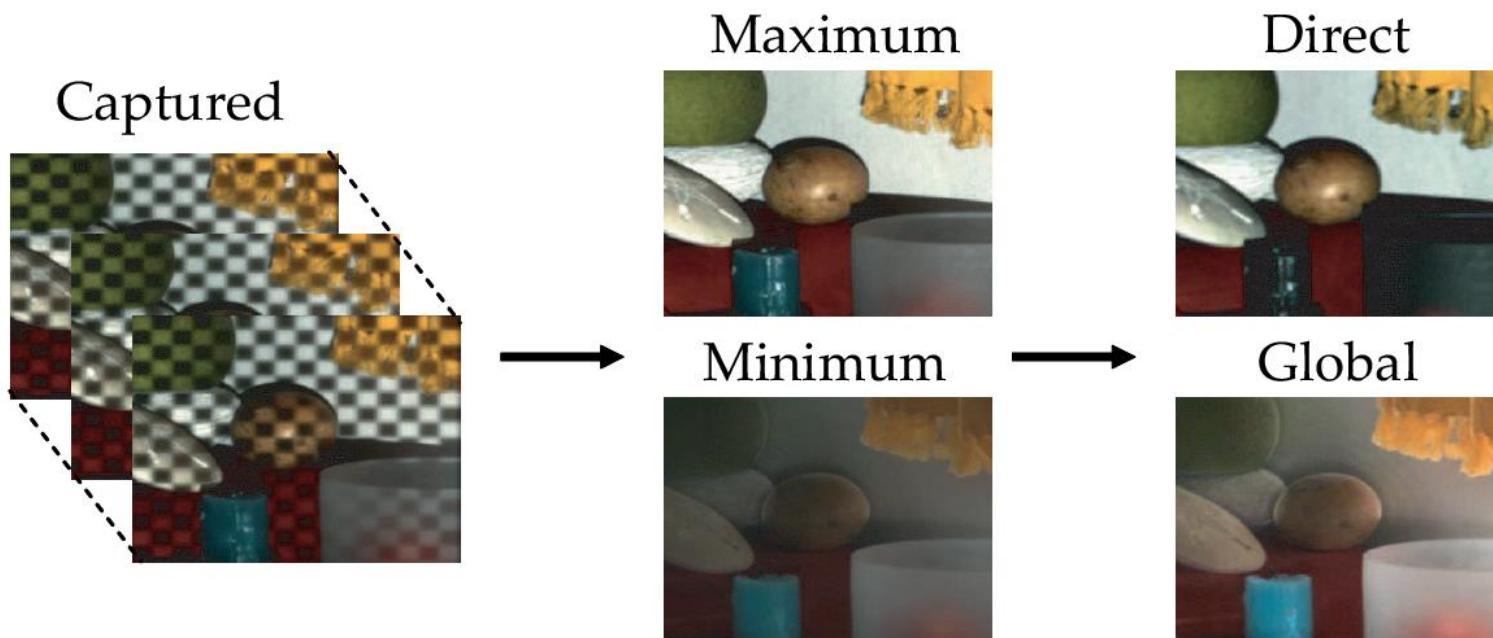
direct



global component

Fast Separation of Direct and Global Effects

- shifted periodic patterns (e.g. checker board)
- record per-pixel minimum and maximum
- approximation:
 - $L_g = L_{\min}, \quad L_d = L_{\max} - L_{\min}$



References

- Sloan et al., Precomputed Radiance Transfer for Real-Time Rendering in Dynamic, Low-Frequency Lighting Environments, SIGGRAPH 2002
- Ng et al., All-Frequency Shadows Using Non-linear Wavelet Lighting Approximation, SIGGRAPH 2003
- Seitz et al., A Theory of Inverse Light Transport, ICCV 2005
- Sen et al., Dual Photography, SIGGRAPH 2005
- Debevec et al., Capturing the Reflectance Field of a Human Face, SIGGRAPH 2000
- Peers et al., Post-Production Facial Performance Relighting using Reflectance Transfer, SIGGRAPH 2007
- O'Toole and Kutulakos, Optical Computing for Fast Light Transport Analysis, SIGGRAPH Asia 2010
- Garg et al., Symmetric Photography: Exploiting Data-sparseness in Reflectance Fields, EGSR 2006