

Eurographics2019



Analysis of Sample Correlations for Monte Carlo Rendering

Gurprit Singh



Cengiz Oztireli



Abdalla G. Ahmed



David Coeurjolly

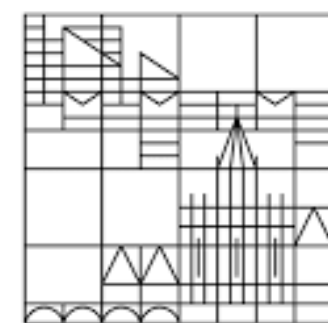


Kartic Subr



Oliver Deussen

Universität
Konstanz



Victor Ostromoukhov



Ravi Ramamoorthi



Wojciech Jarosz







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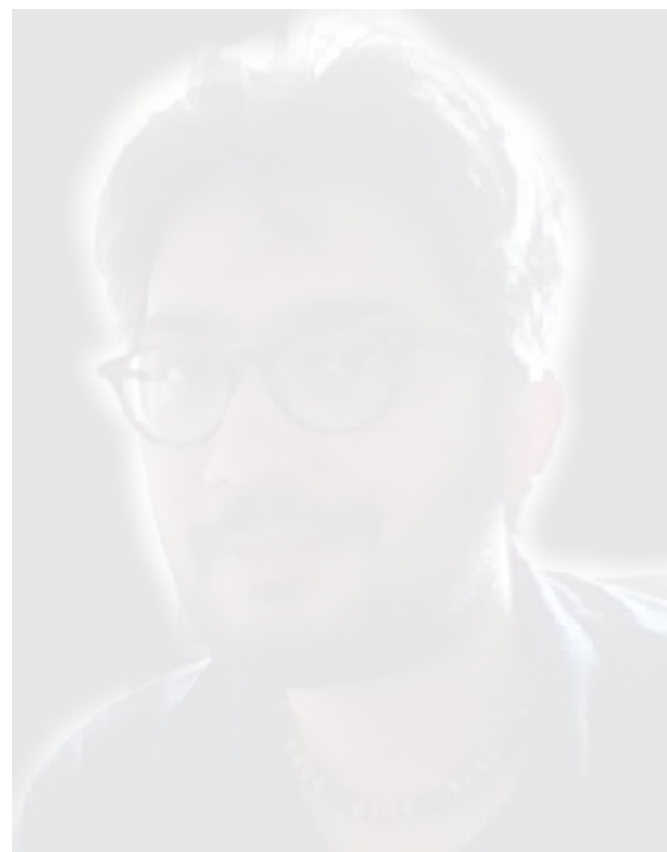
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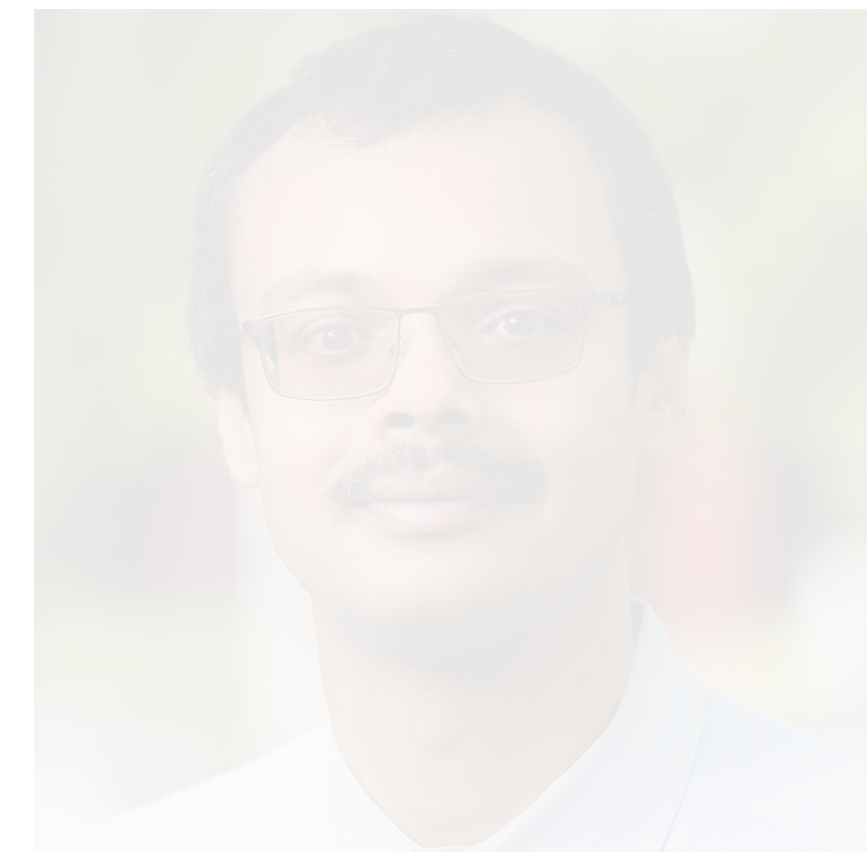
Kartic Subr



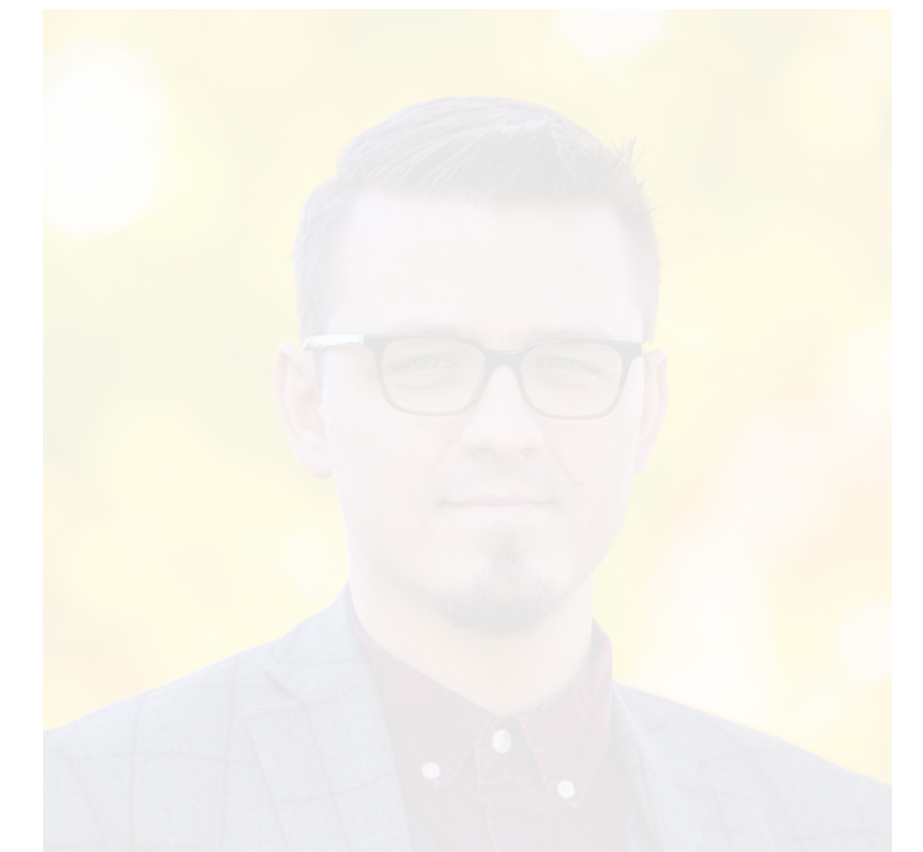
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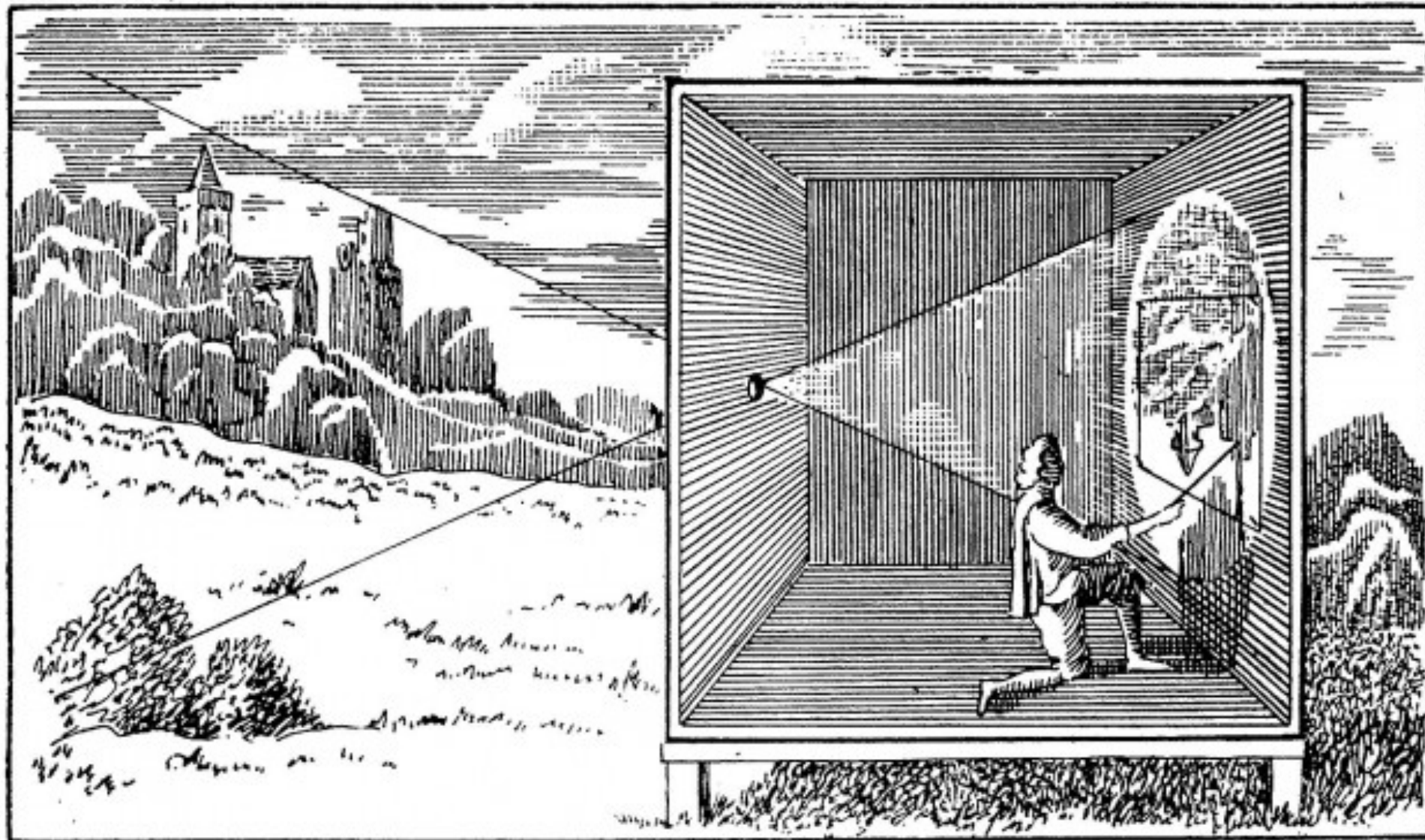


Wojciech Jarosz

Rendering = Geometry + Radiometry

Geometry / Projection

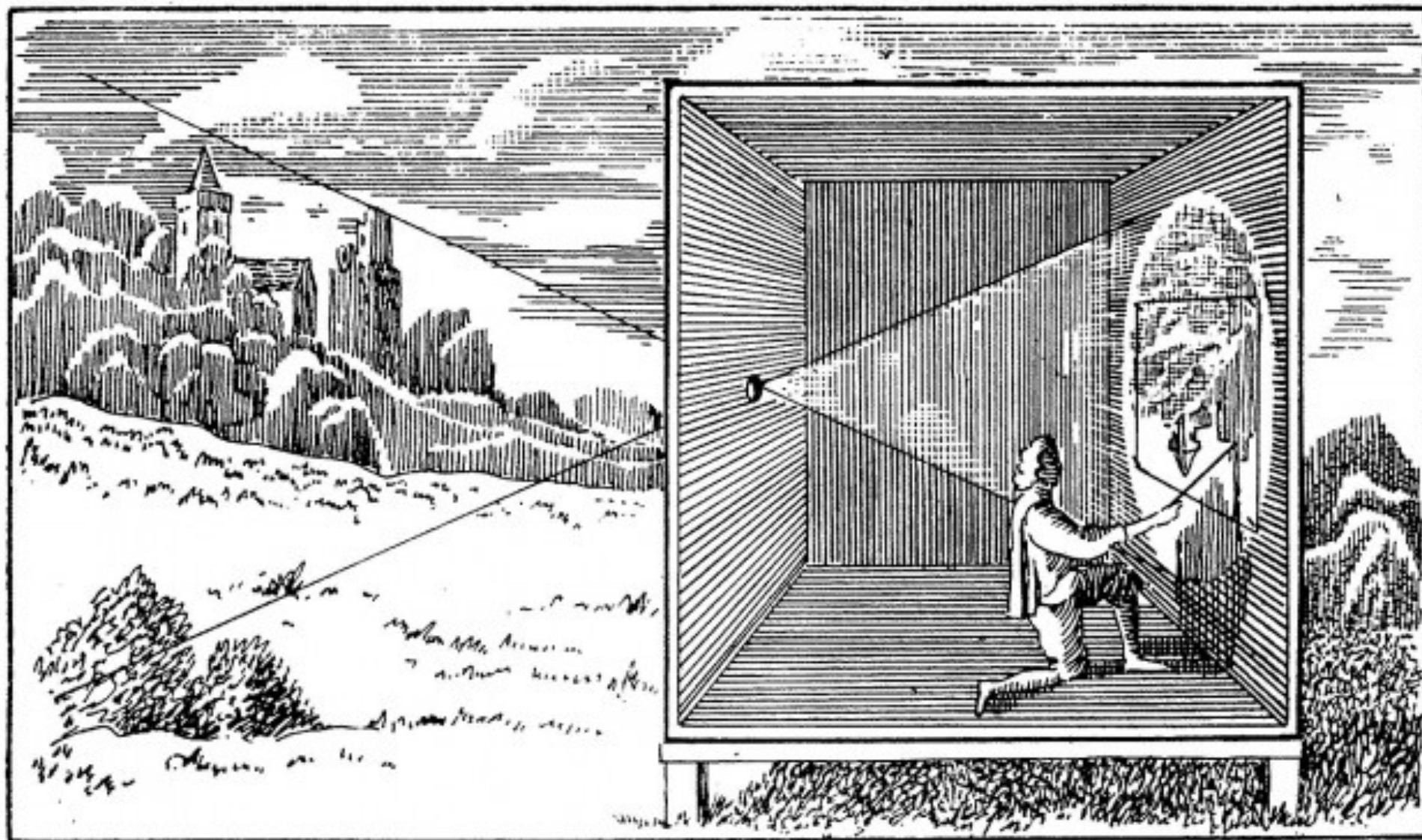
for pin-hole model is known since 400BC



Rendering = Geometry + Radiometry

Geometry / Projection

for pin-hole model is known since 400BC



Radiometrically accurate simulation

is importance of realism



Rendering = Geometry + Radiometry

Geometry / Projection

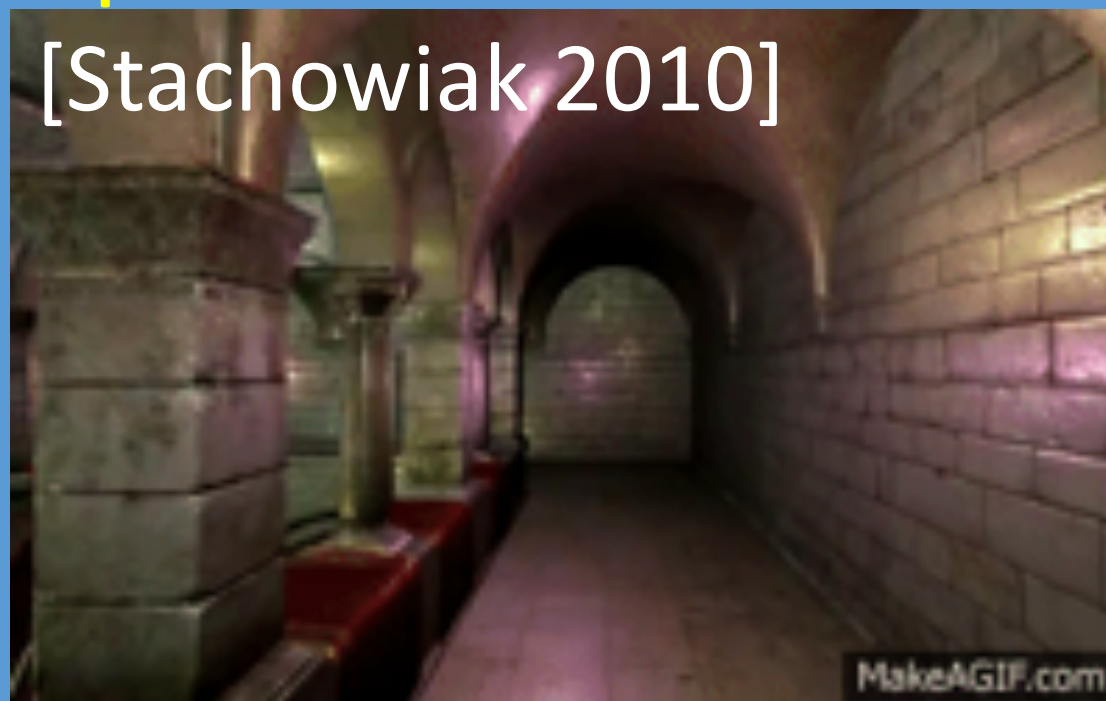
for pin-hole model is known since 400BC

Radiometrically accurate simulation

is importance of realism

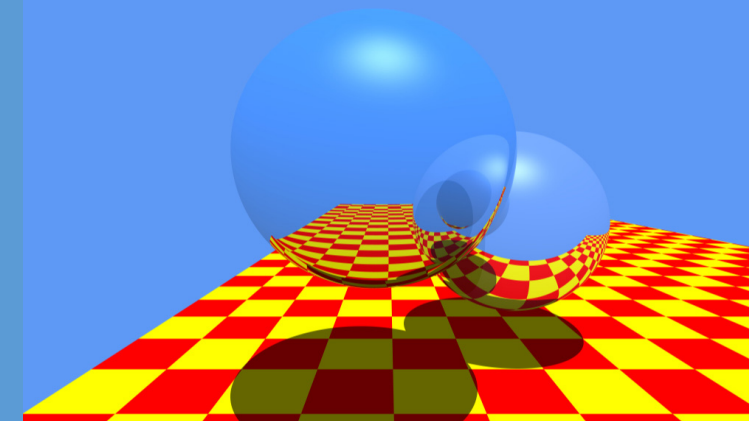
OpenGL

[Stachowiak 2010]



Raytracing

[Whitted 1980]



Radiometric fidelity improves photorealism

Papas et al. [2013]

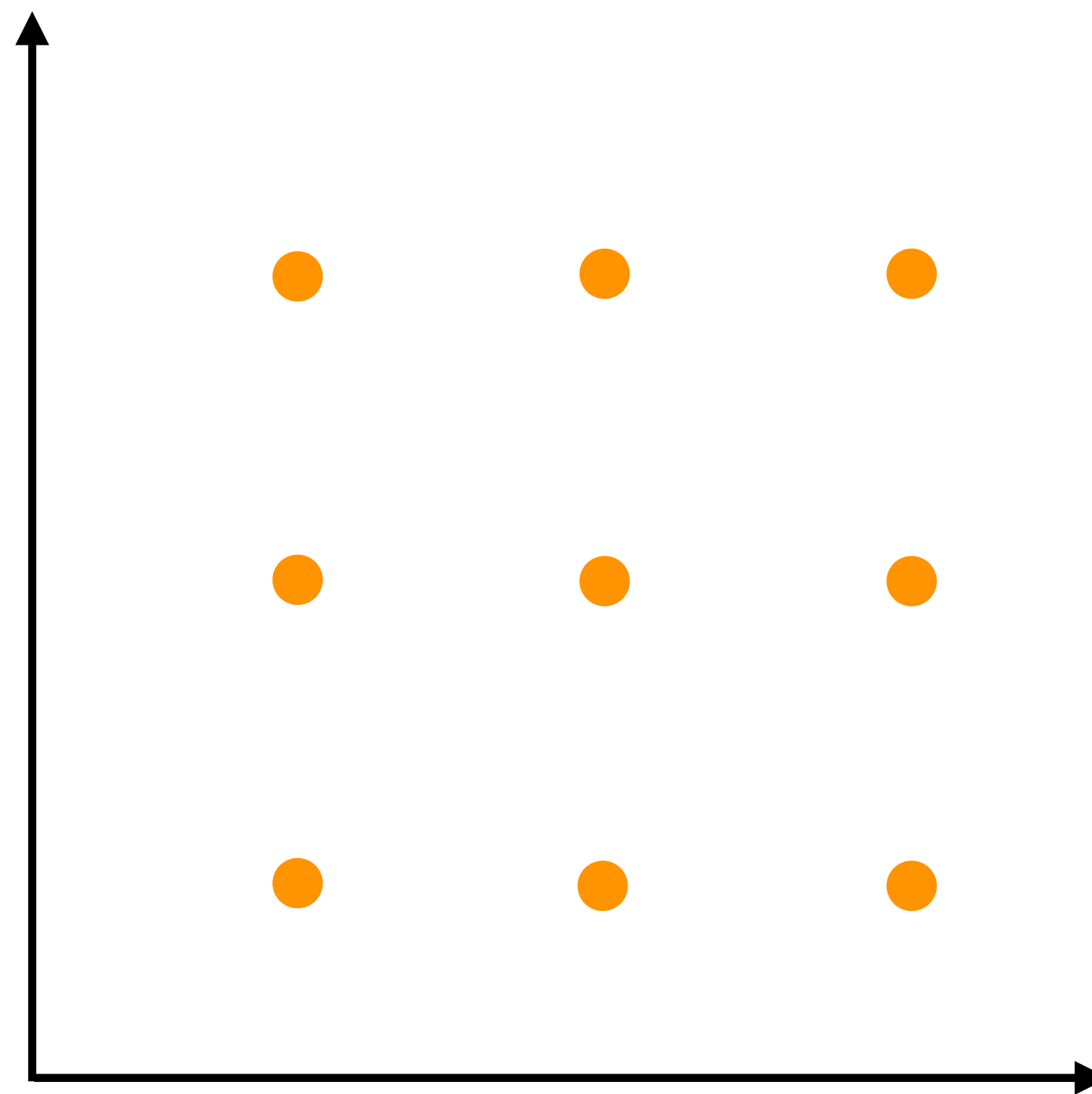
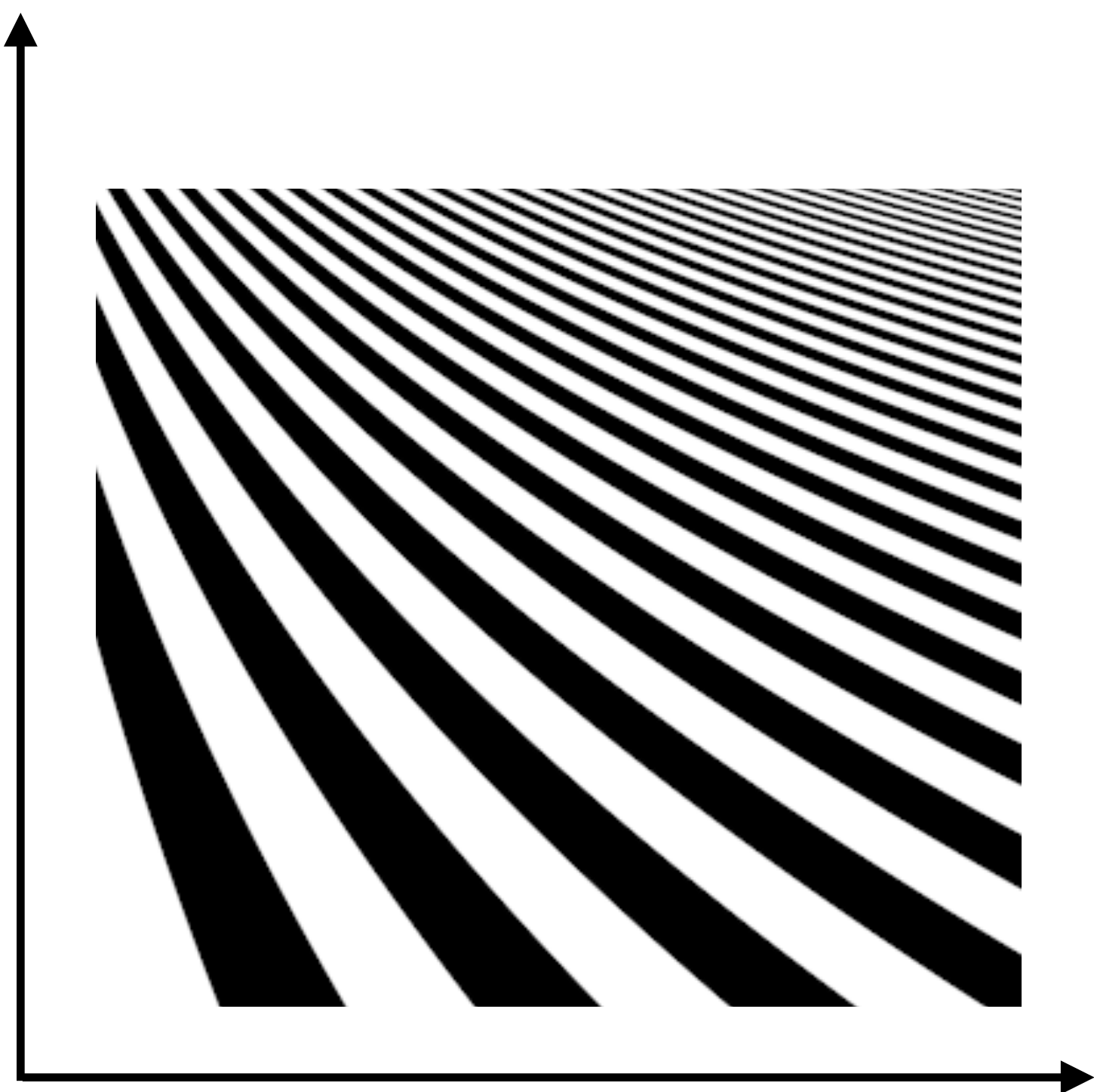


Radiometric fidelity improves photorealism

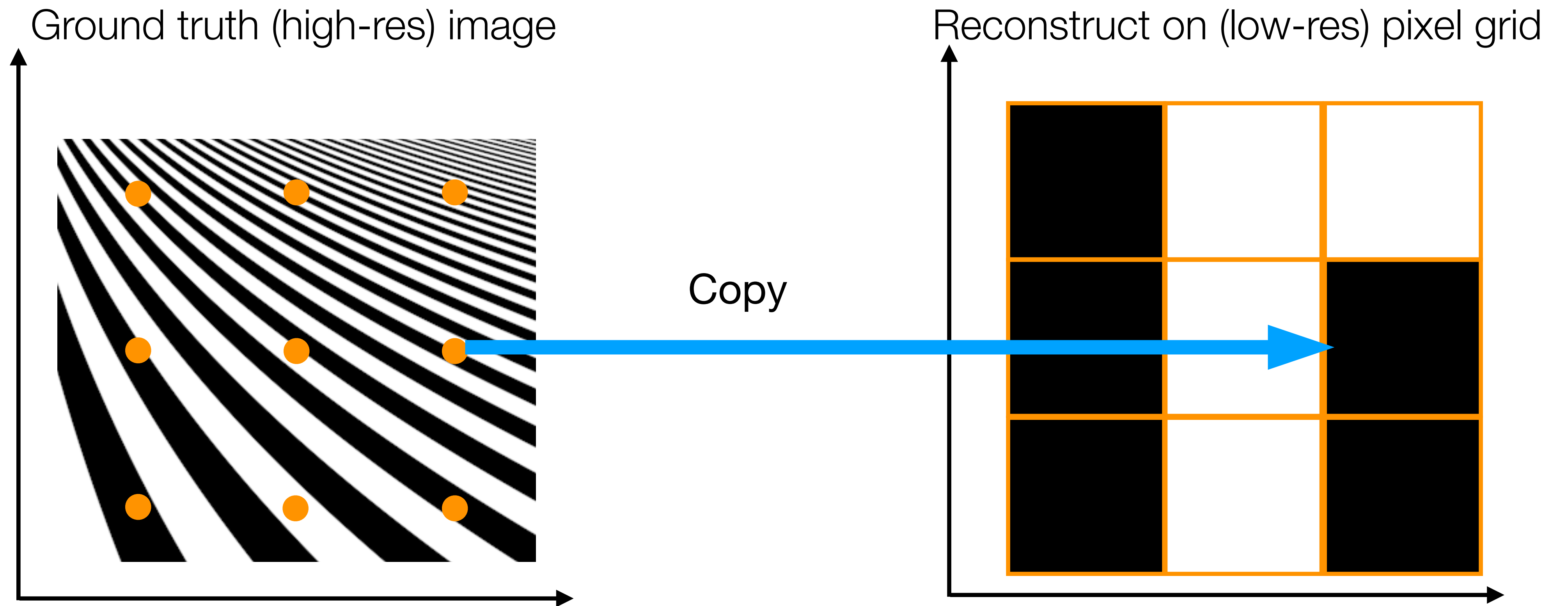
Krivanek et al. [2014]



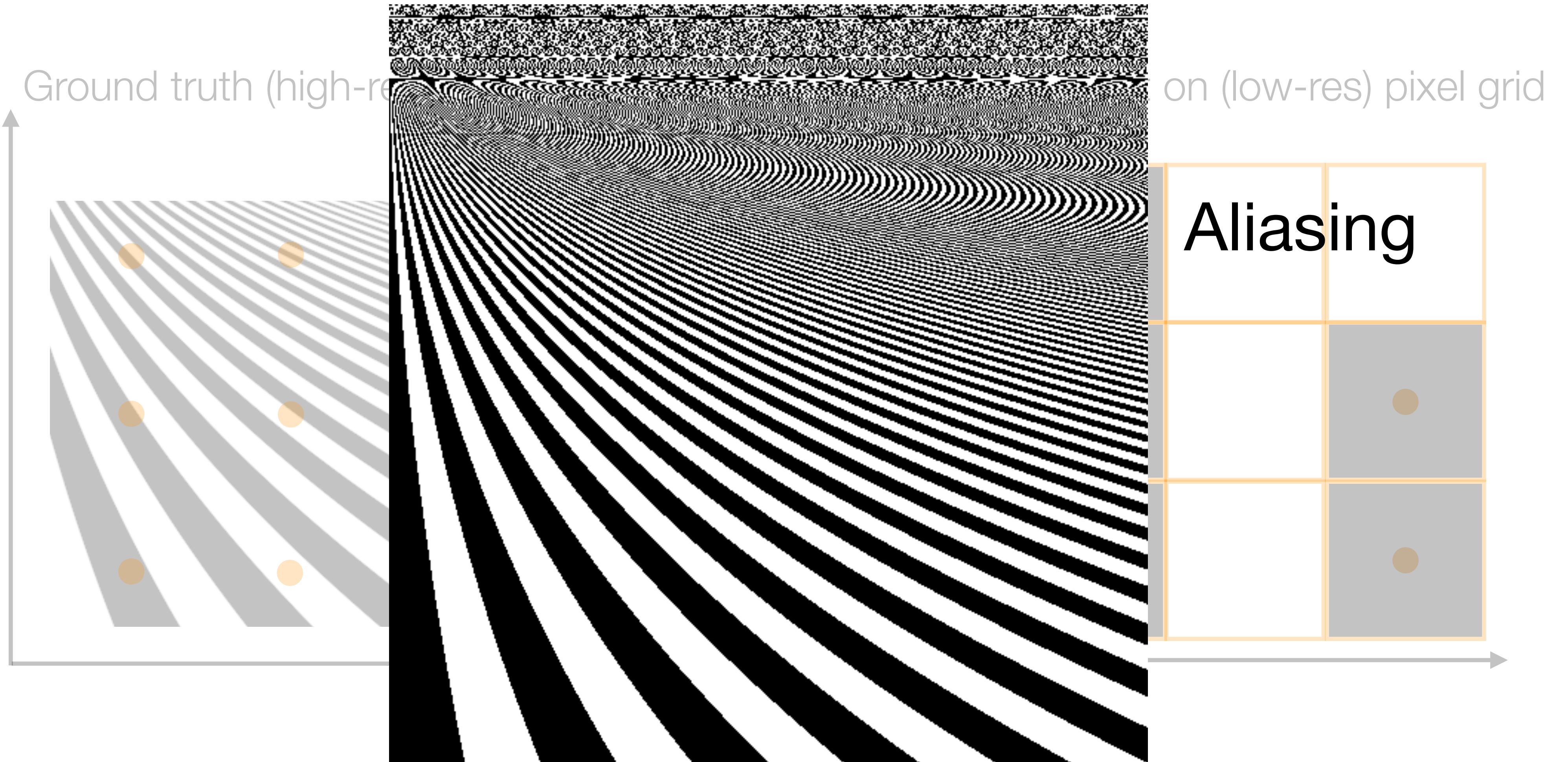
Reconstruction: Estimate image samples



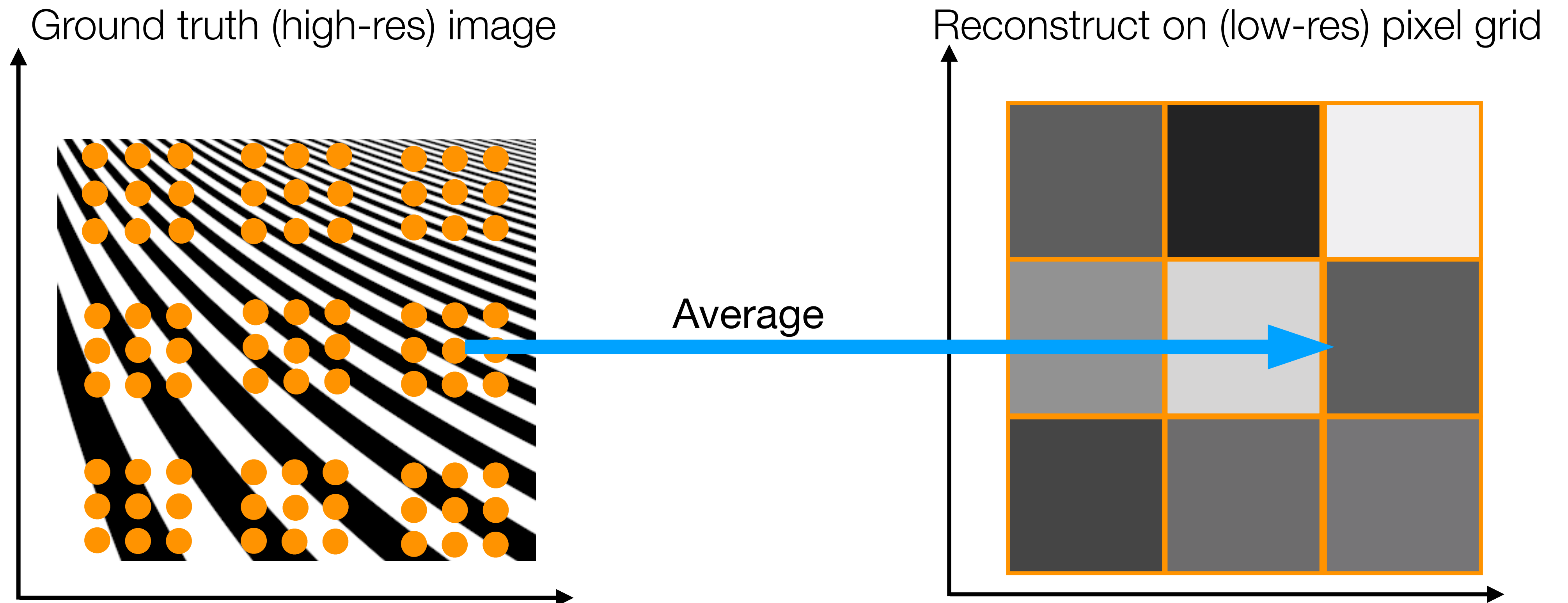
Naive method: sample image at grid locations



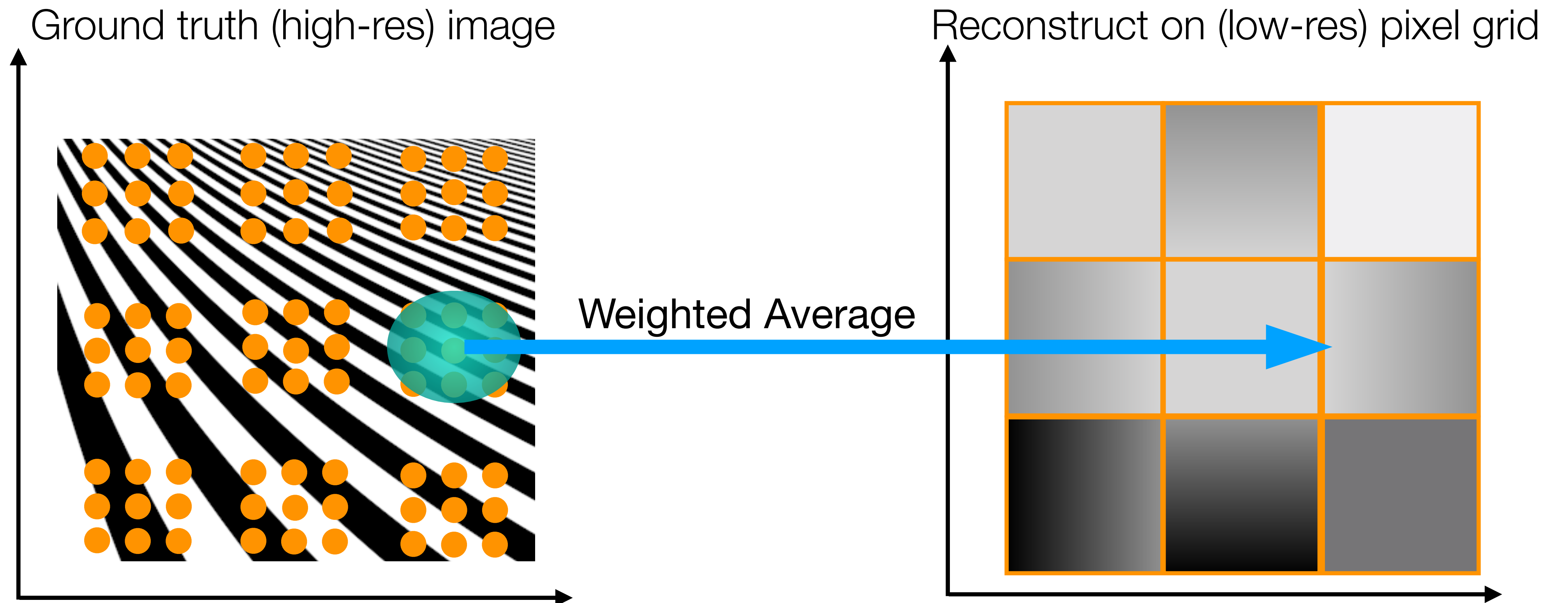
Naive method: sample image at grid locations



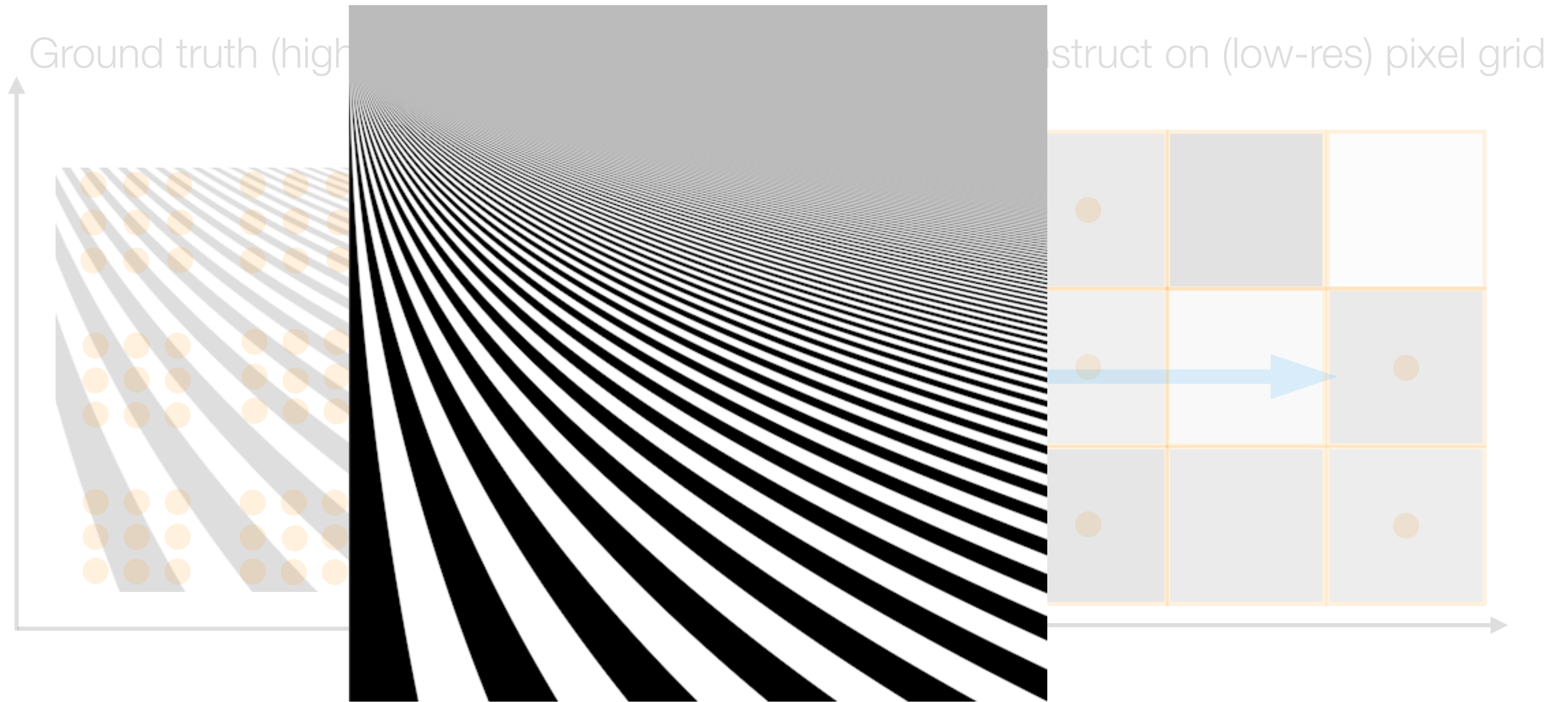
Naive method: sample image at grid locations



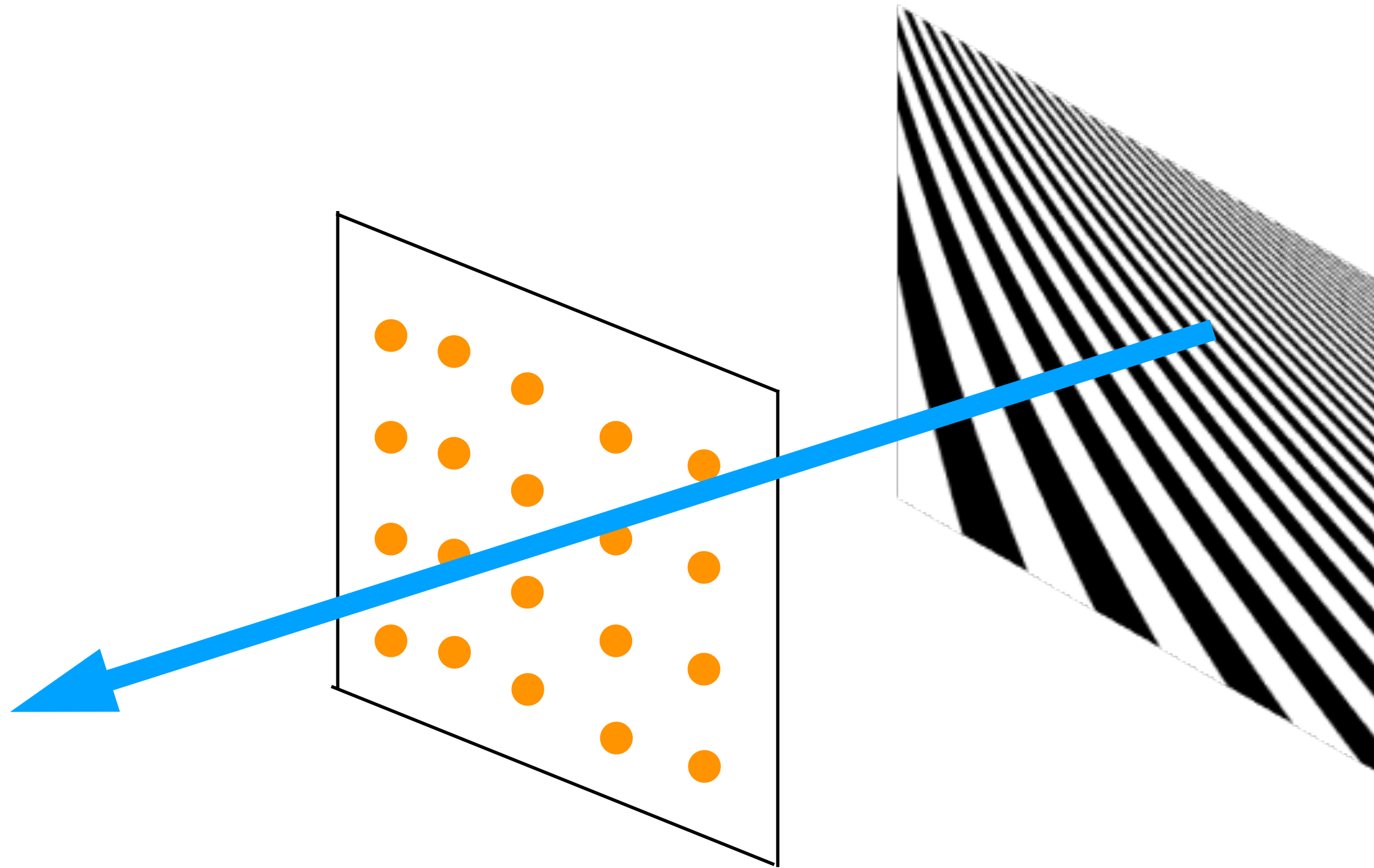
Antialiasing using general reconstruction filters



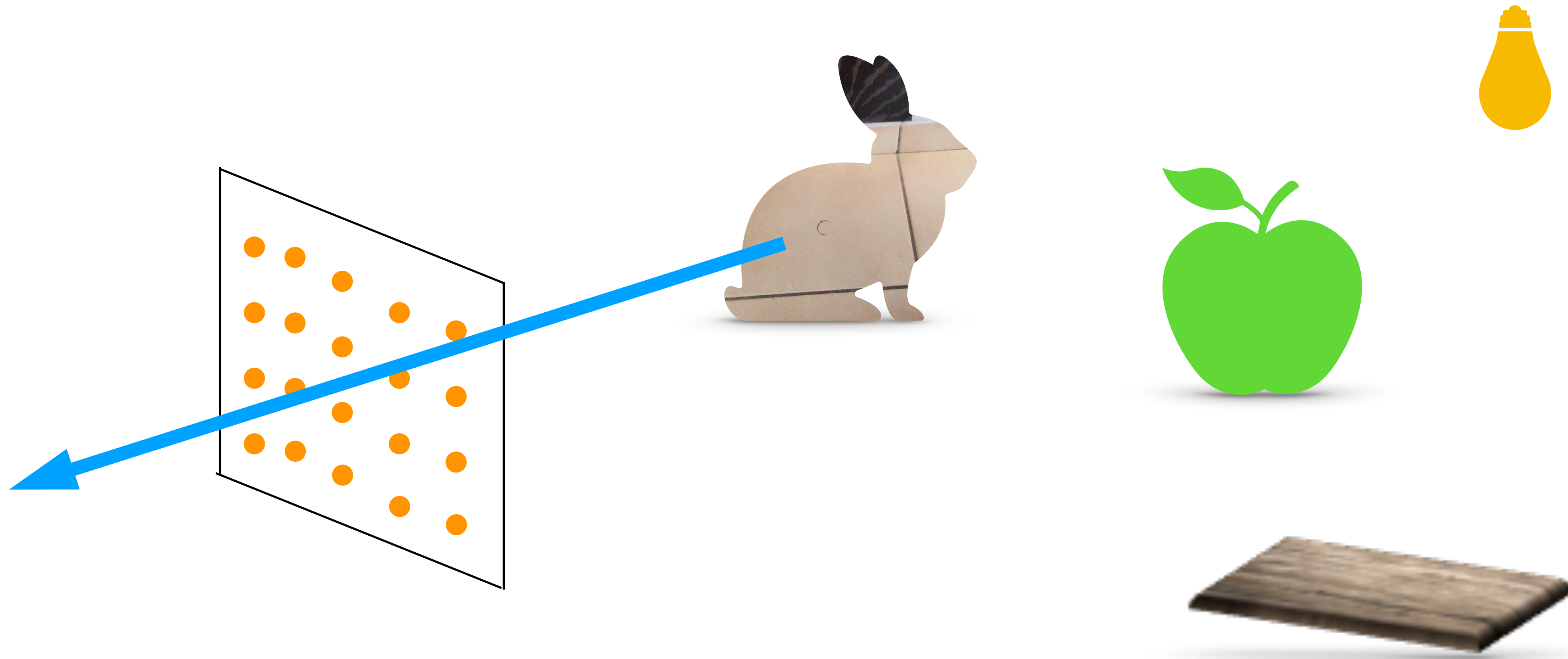
Naive method: sample image at grid locations



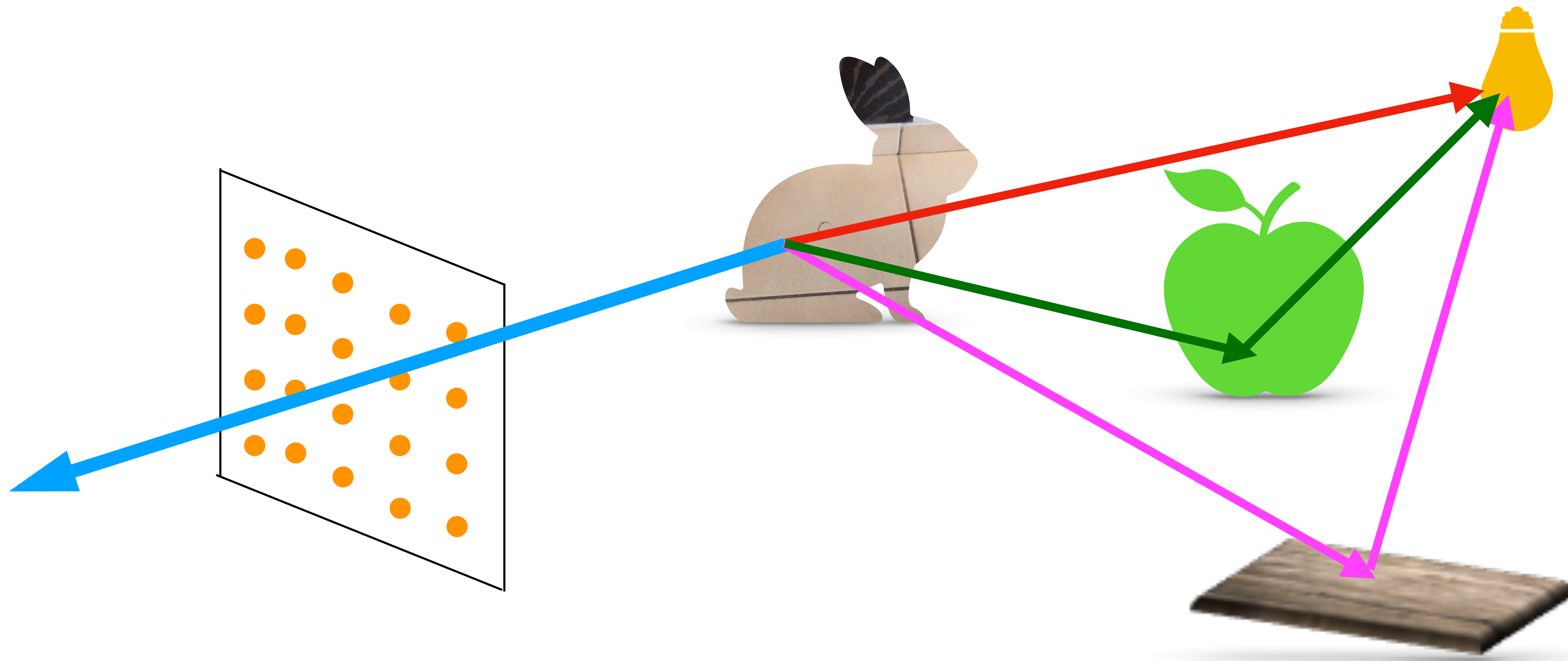
Rendering: reconstructing integrals



Rendering: reconstructing integrals

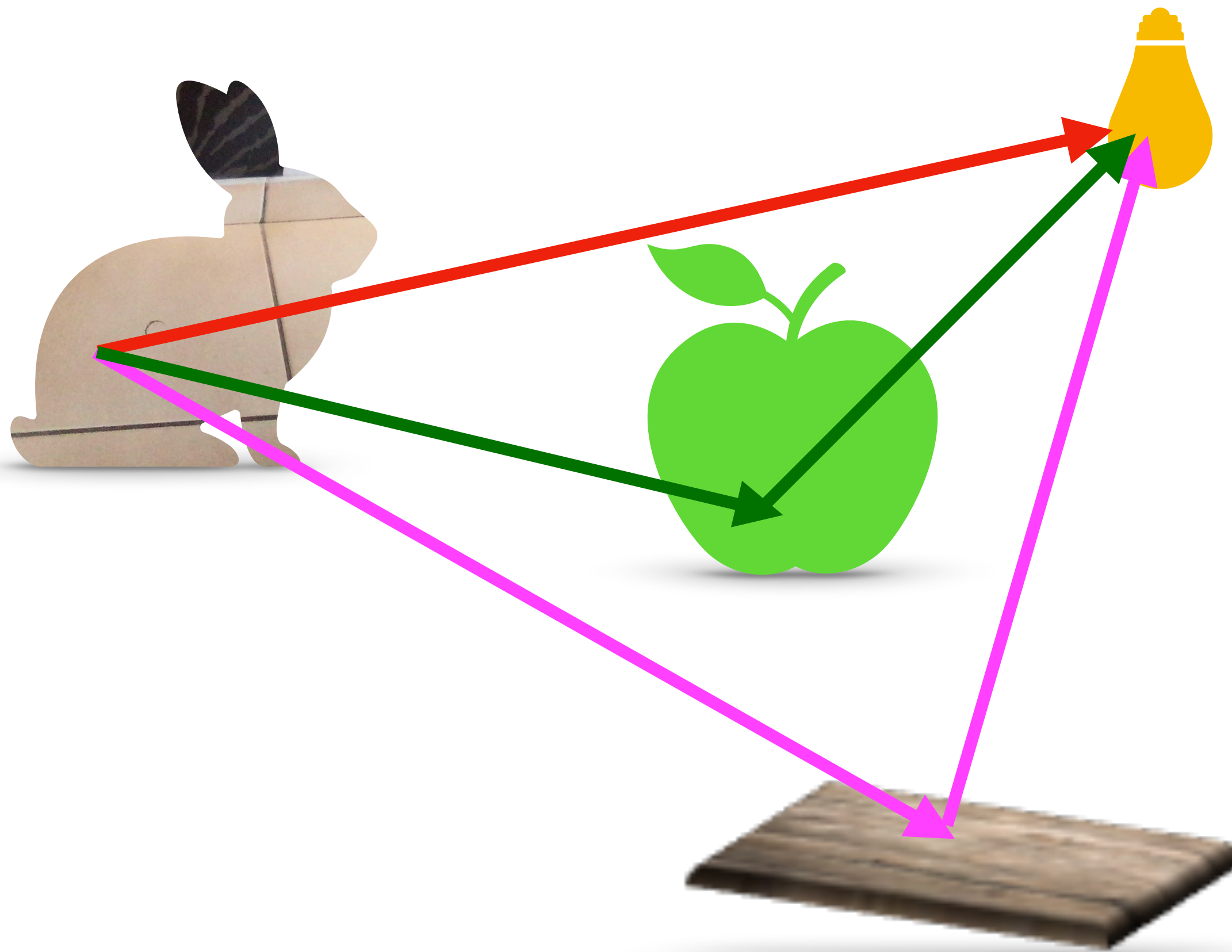
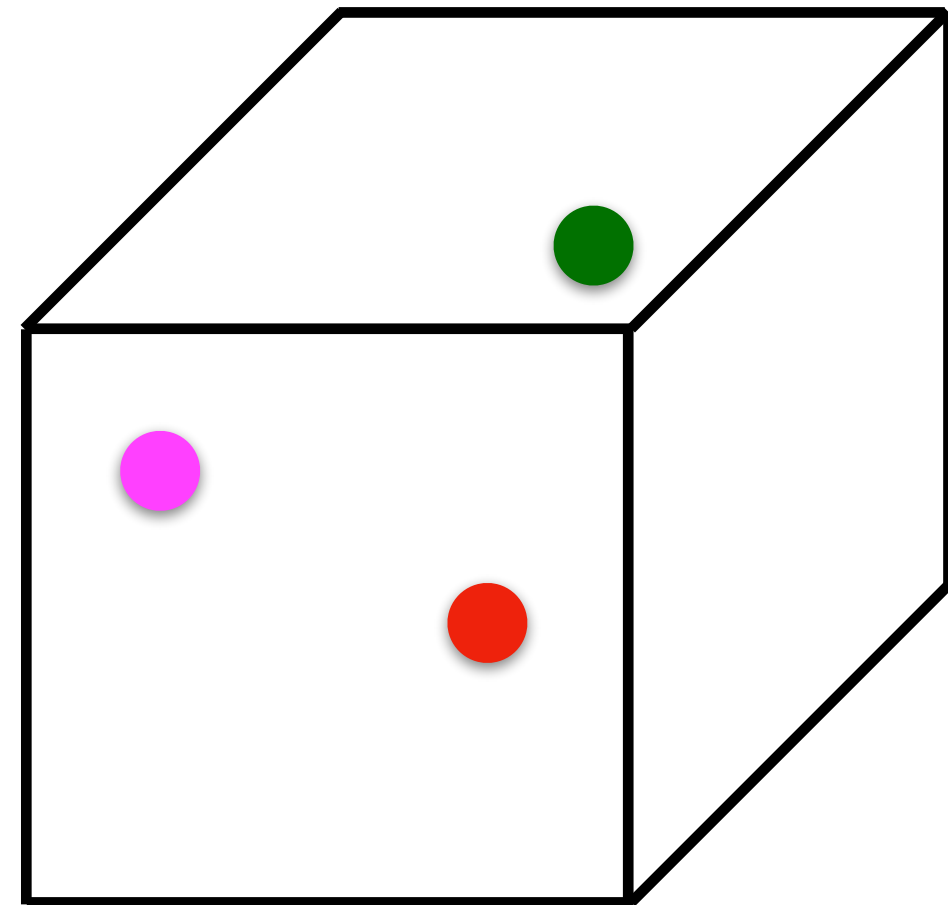


Rendering: reconstructing integrals



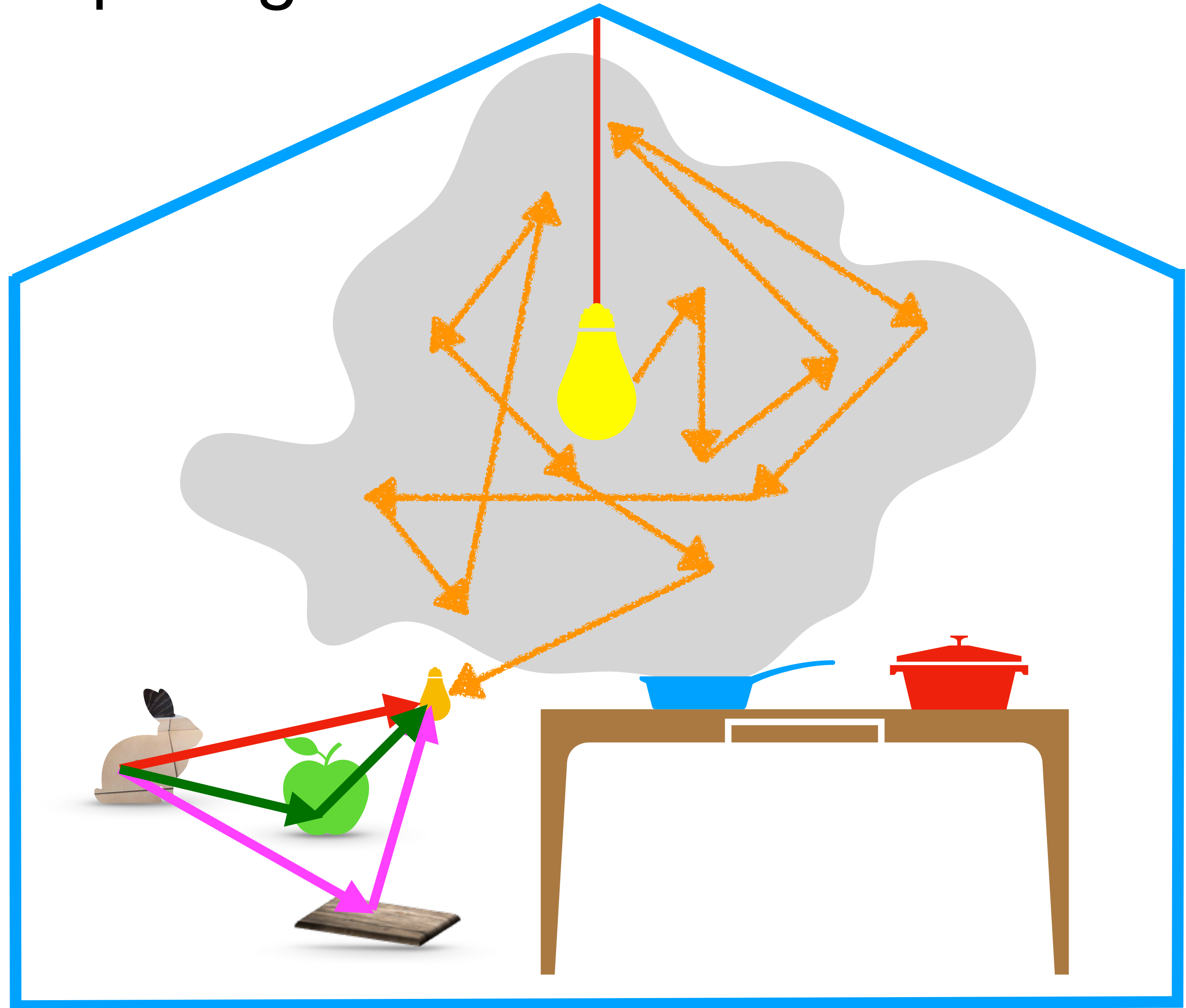
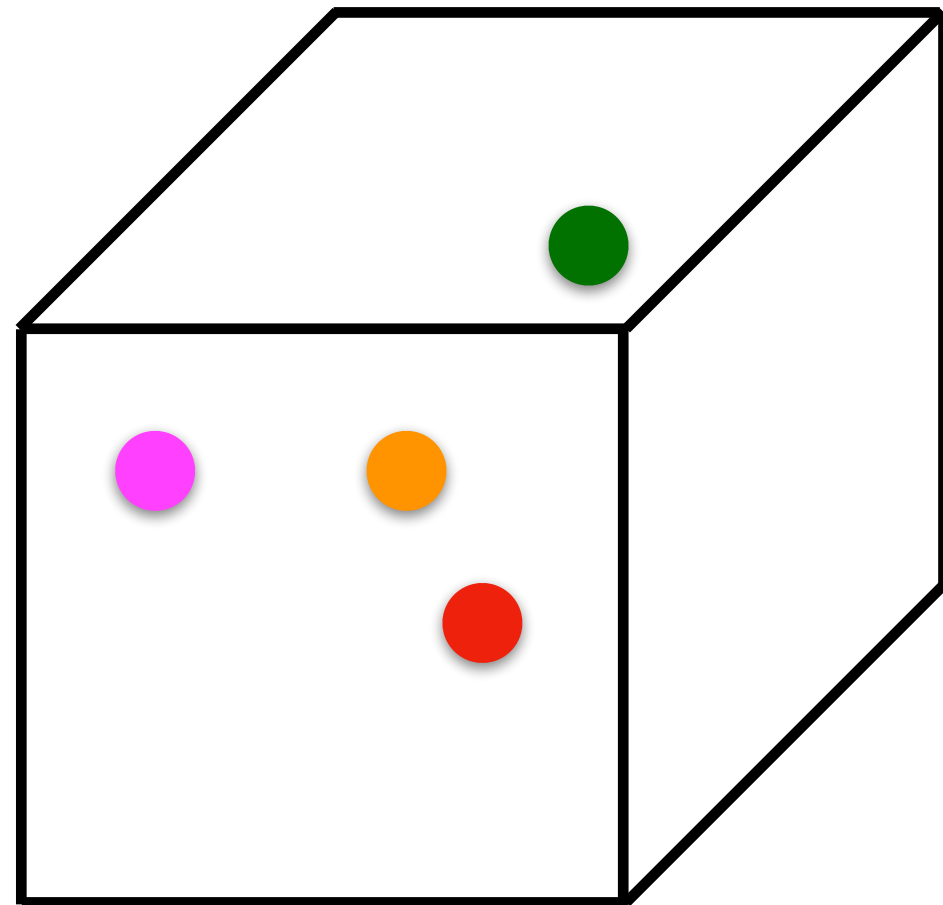
Rendering: reconstructing integrals

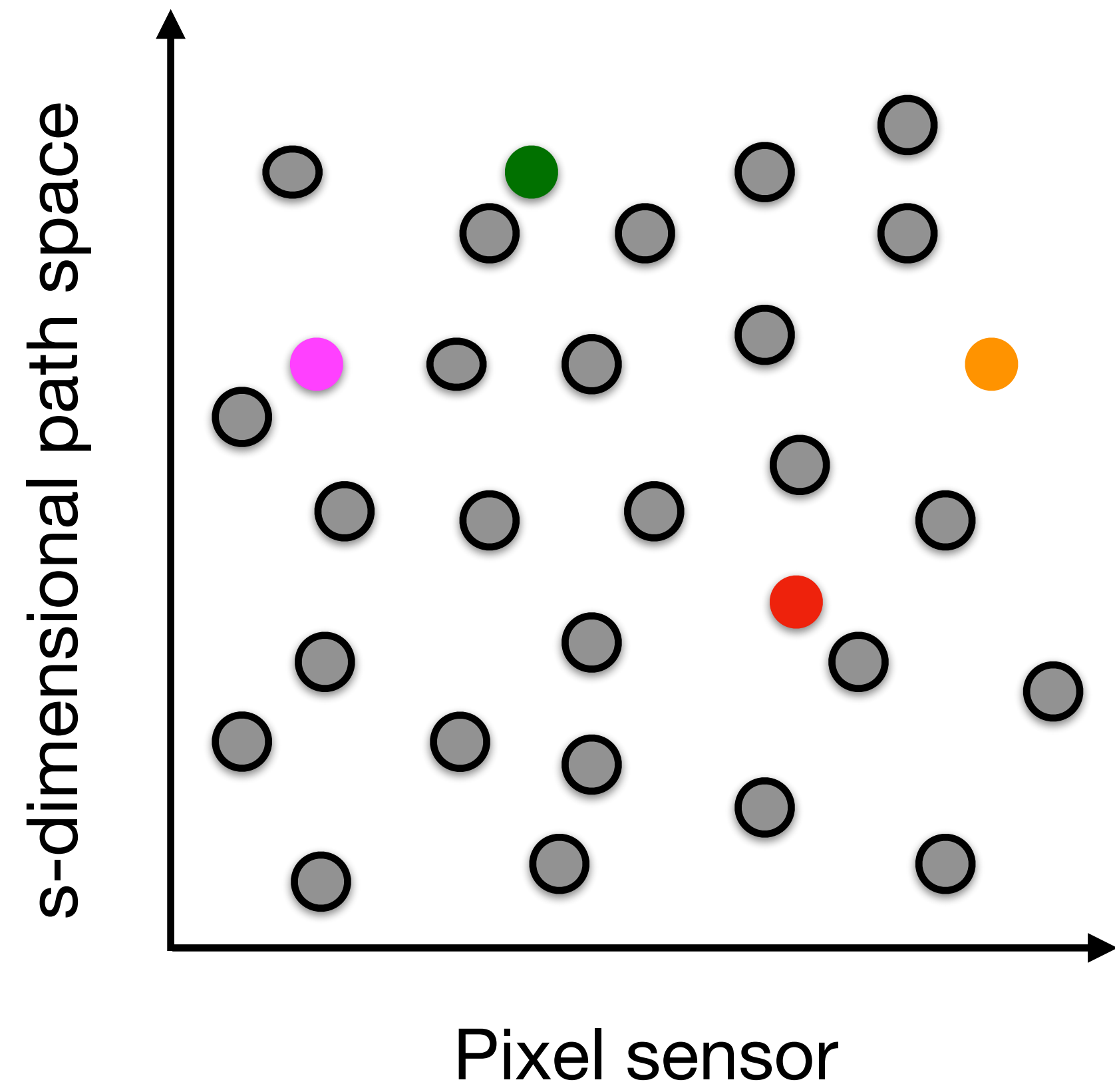
Each path has an associated radiance value

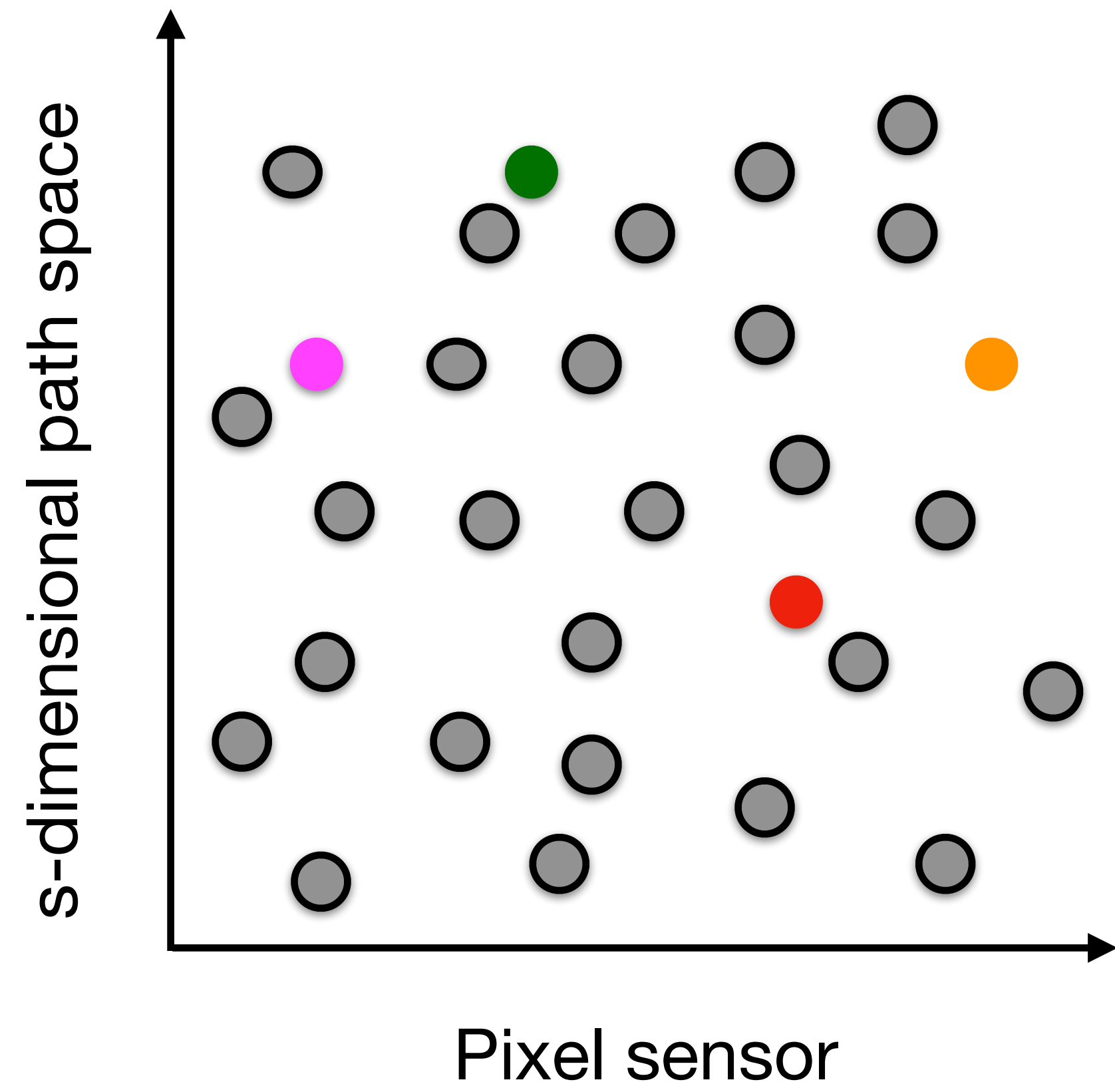


Global Illumination: Participating media

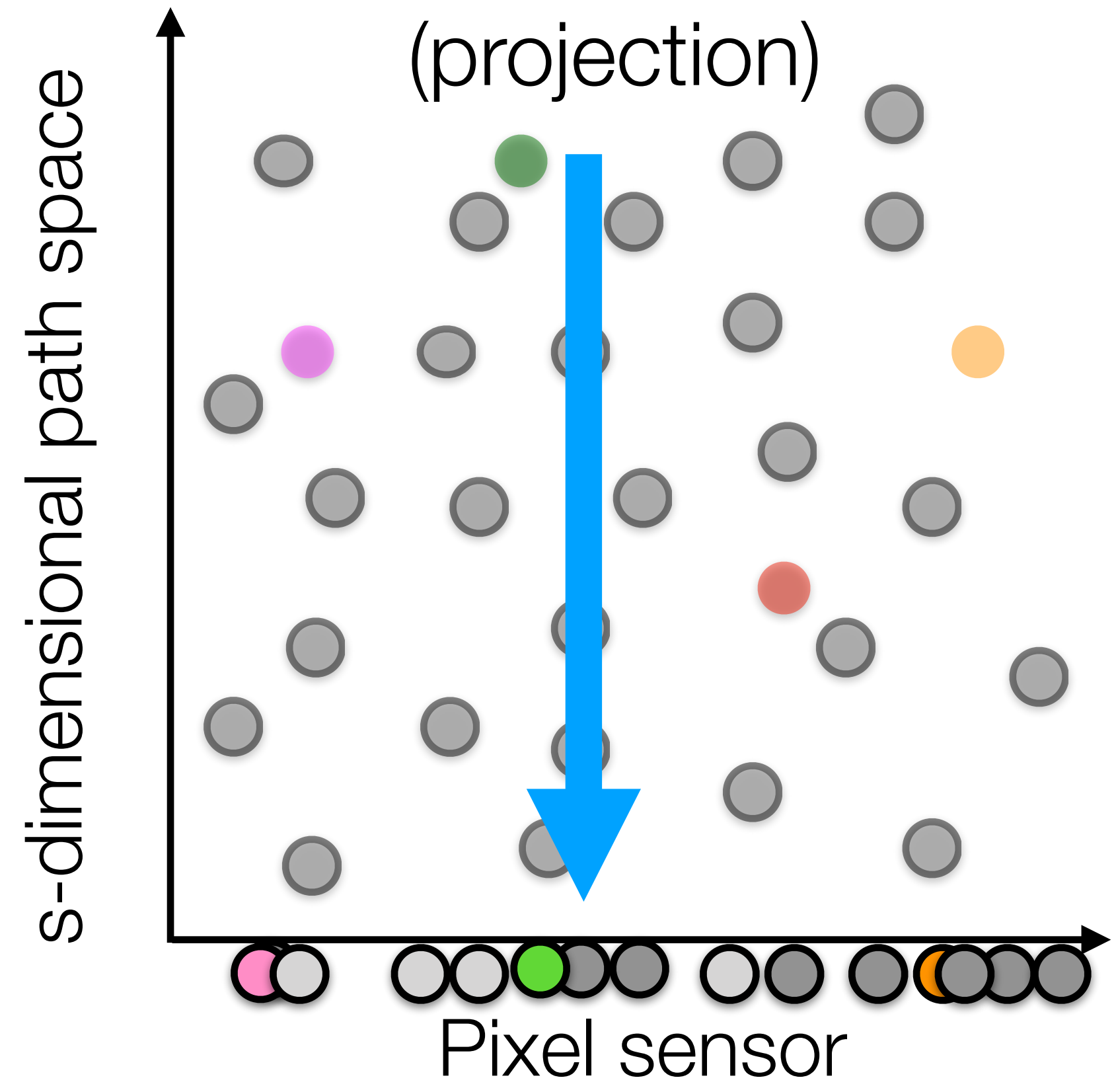
Each path has an associated radiance value





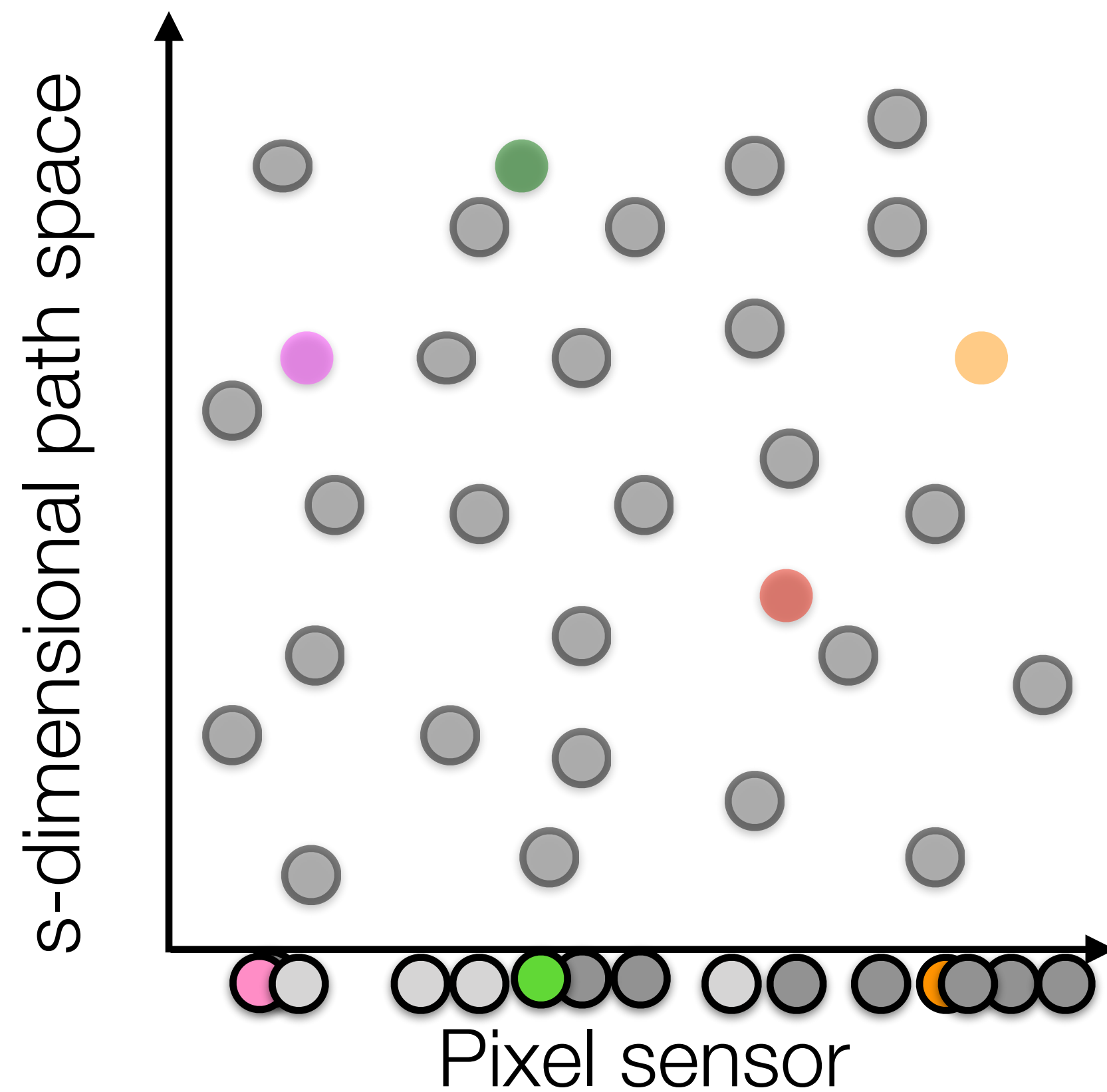


Path-space integration

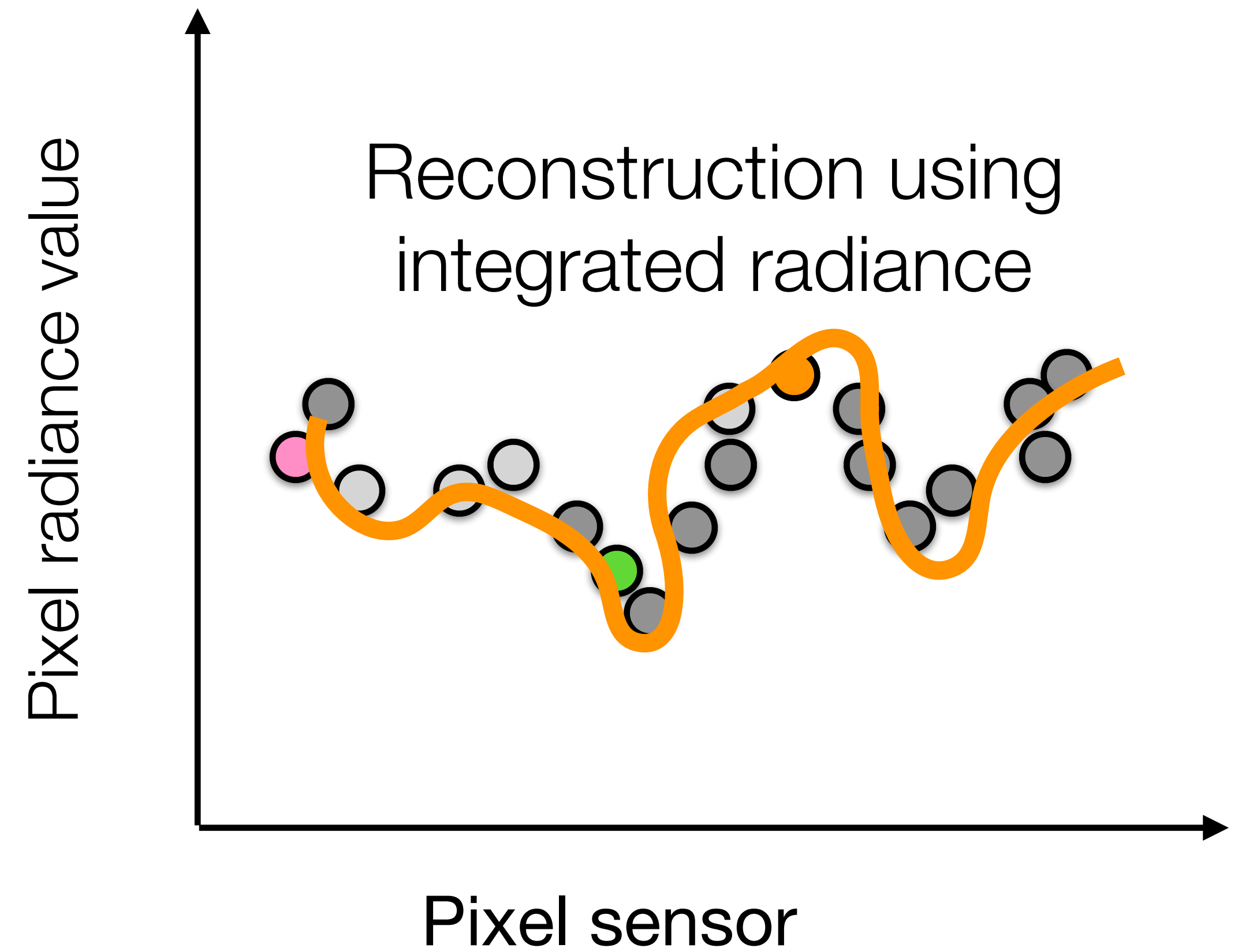


Rendering = integration + reconstruction

Path-space integration

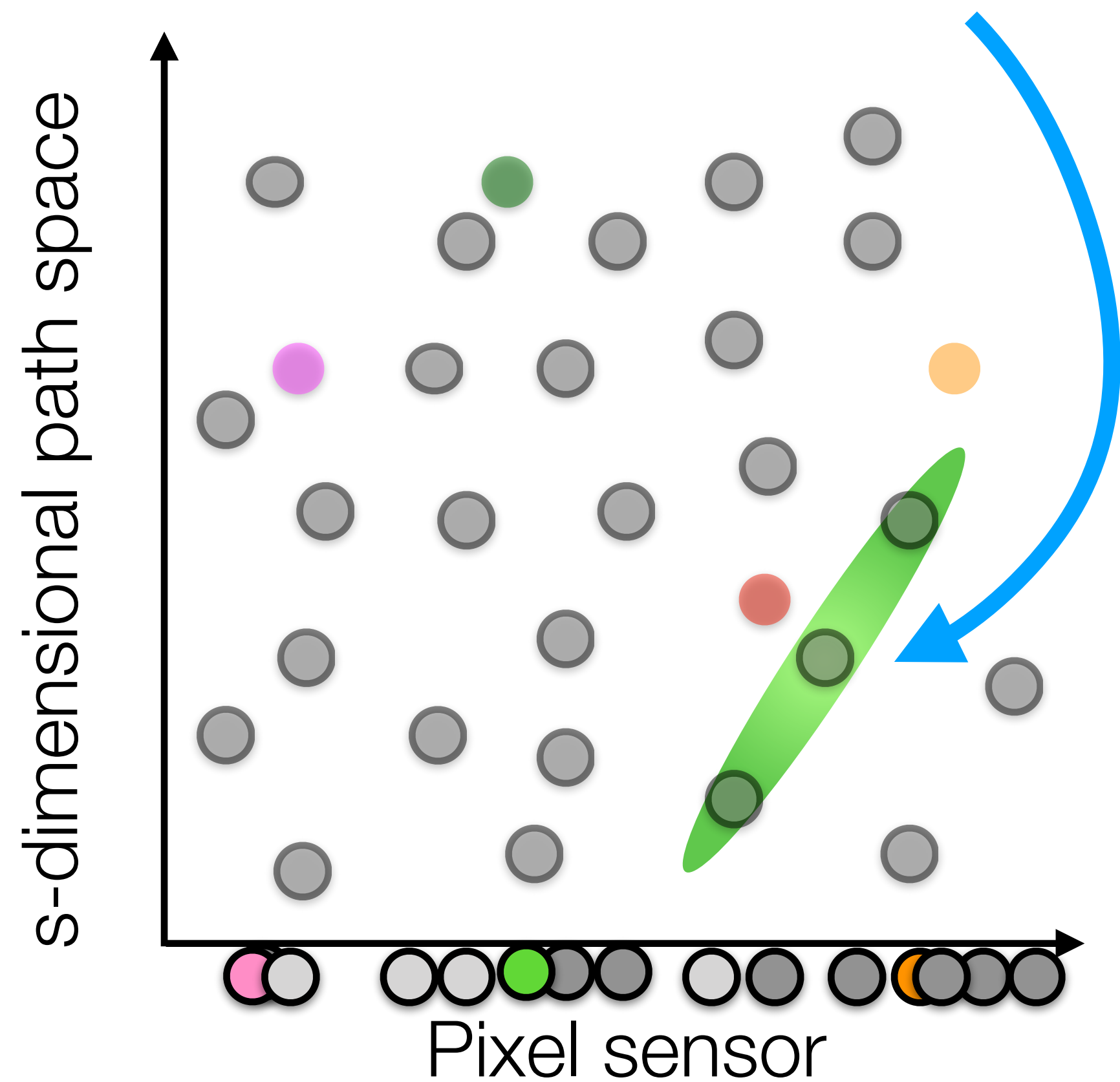


Reconstruction using integrated radiance

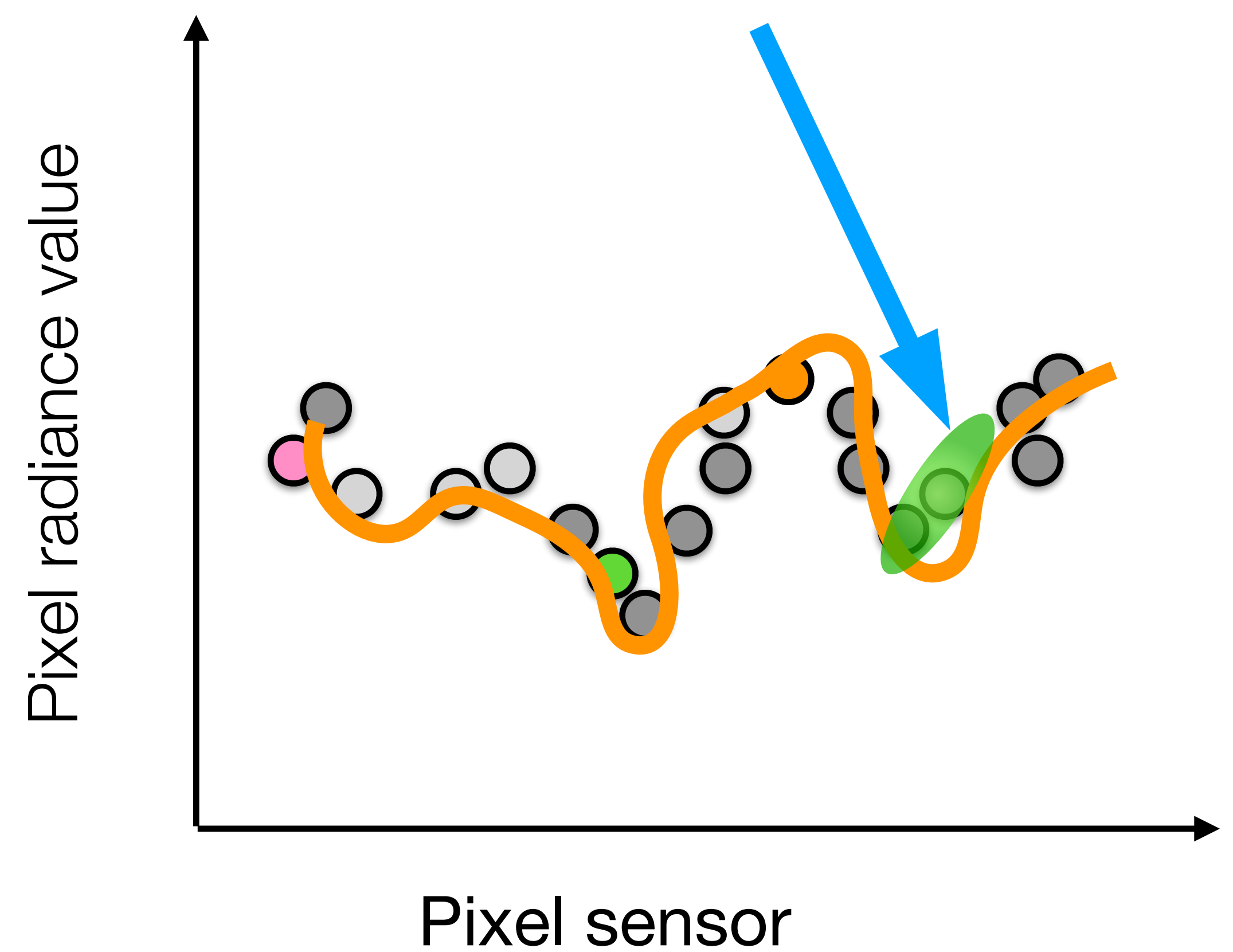


Frequency analysis of light fields in rendering

Local variation of the integrand



Reconstruction filter



A Frequency Analysis of Light Transport

Frédo Durand
MIT-CSAIL

Nicolas Holzschuch
ARTIS* GRAVIR/IMAG-INRIA

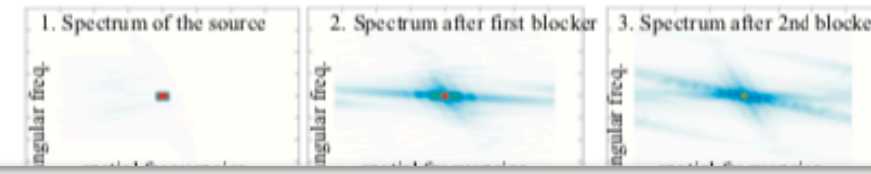
Cyril Soler

Eric Chan
MIT-CSAIL

François X. Sillion
ARTIS* GRAVIR/IMAG-INRIA

Abstract

We present a signal-processing framework for light transport. We study the frequency content of radiance and how it is altered by



To appear in the ACM SIGGRAPH conference proceedings

Frequency Analysis and Sheared Reconstruction for Rendering Motion Blur

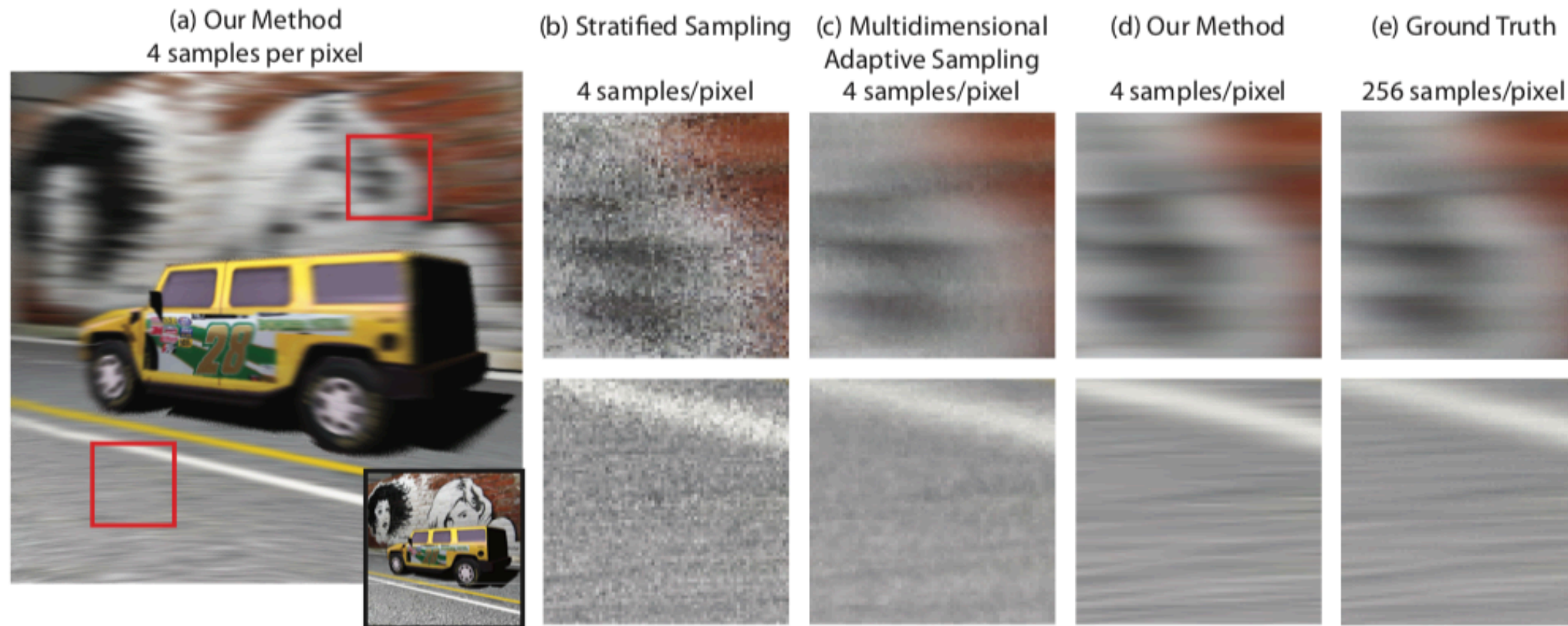
Kevin Egan *
Columbia University

Yu-Ting Tseng
Columbia University

Nicolas Holzschuch
INRIA — LJK

Frédo Durand
MIT CSAIL

Ravi Ramamoorthi
UC Berkeley



4D Frequency Analysis of Computational Cameras for Depth of Field Extension

Anat Levin^{1,2} Samuel W. Hasinoff¹ Paul Green¹ Frédo Durand¹ William T. Freeman¹
¹MIT CSAIL ²Weizmann Institute

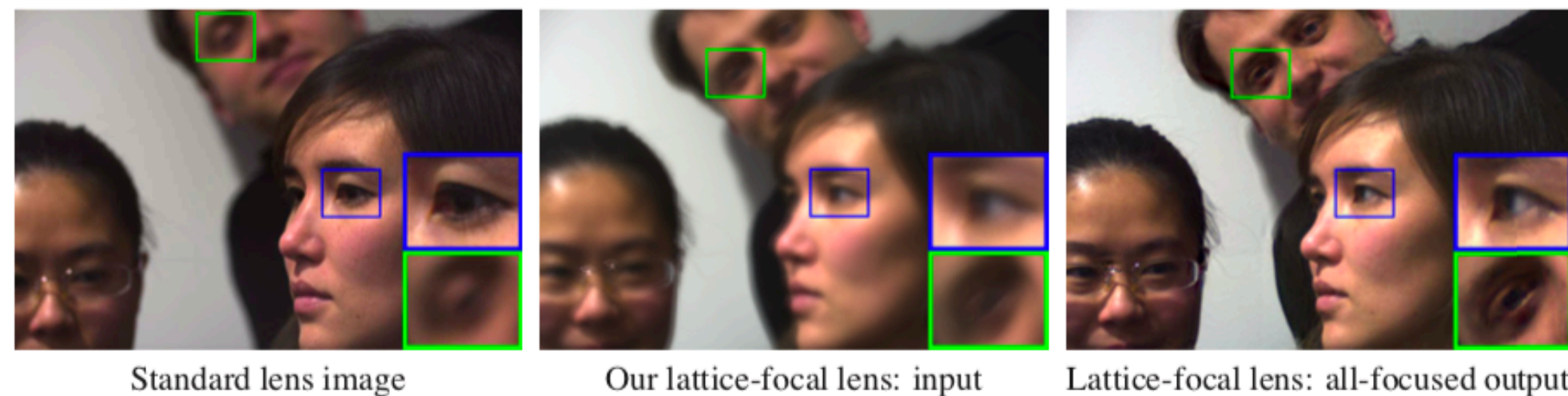


Figure 1: Left: Image from a standard lens showing limited depth of field, with only the rightmost subject in focus. Center: Input from our lattice-focal lens. The defocus kernel of this lens is designed to preserve high frequencies over a wide depth range. Right: An all-focused image processed from the lattice-focal lens input. Since the defocus kernel preserves high frequencies, we achieve a good restoration over the

5D Covariance Tracing for Efficient Defocus and Motion Blur

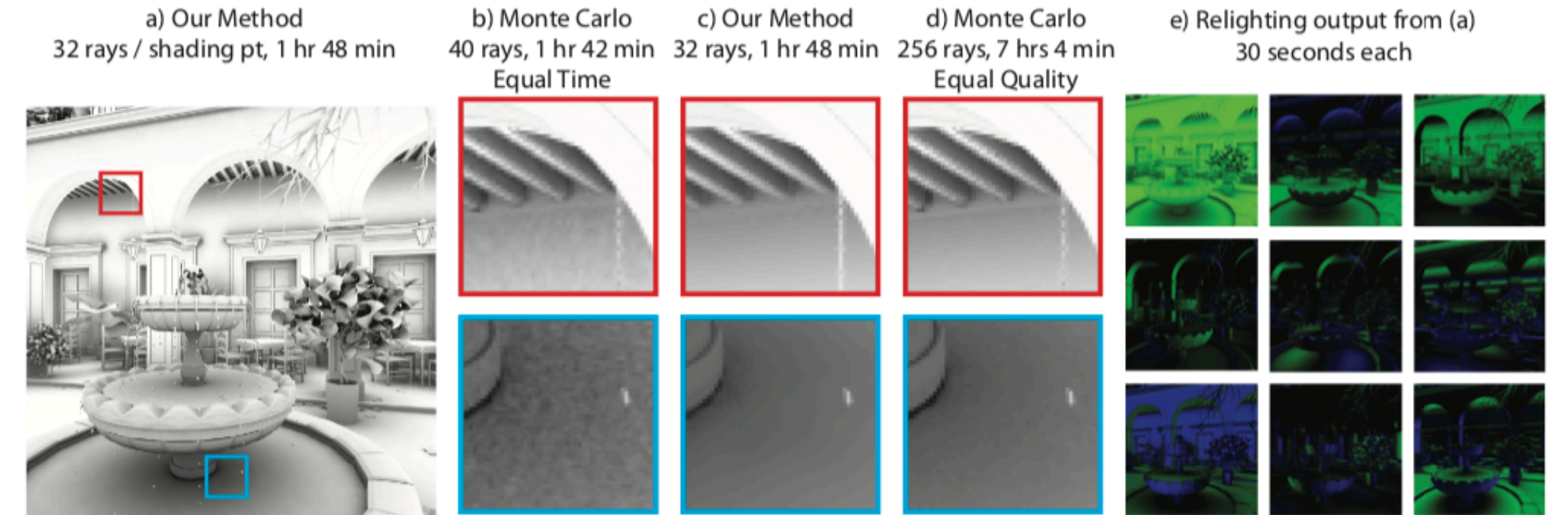
LAURENT BELCOUR¹, CYRIL SOLER², KARTIC SUBR³, NICOLAS HOLZSCHUCH², and FREDO DURAND⁴

Practical Filtering for Efficient Ray-Traced Directional Occlusion

Kevin Egan *
Columbia University

Frédo Durand
MIT CSAIL

Ravi Ramamoorthi
University of California, Berkeley



Temporal Light Field Reconstruction for Rendering Distribution Effects

Jaakko Lehtinen
NVIDIA Research

Timo Aila
NVIDIA Research

Jiawen Chen
MIT CSAIL

Samuli Laine
NVIDIA Research

Frédo Durand
MIT CSAIL



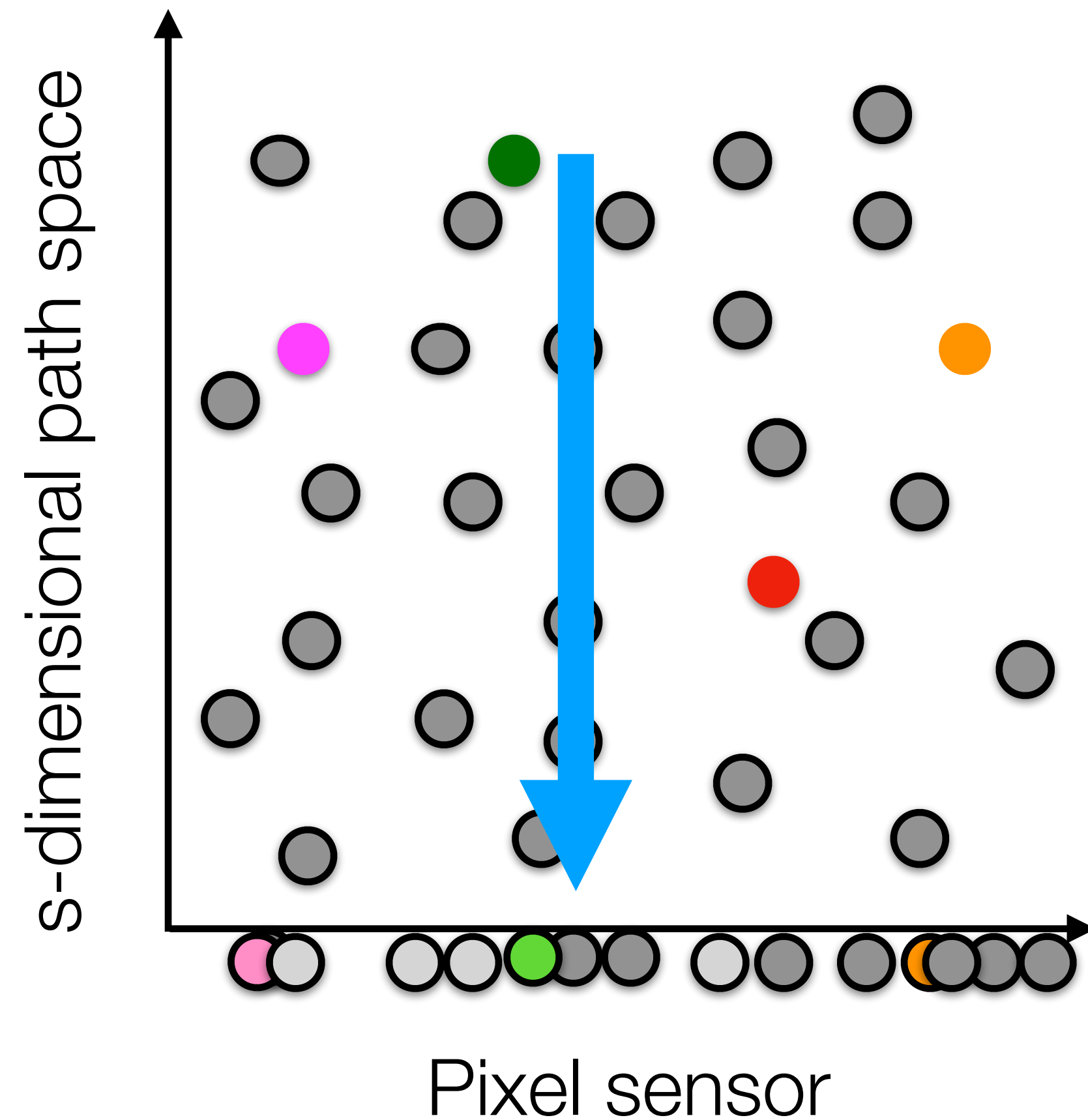
Figure 1: A scene with complex occlusion rendered with depth of field. Left: Images rendered by PBRT [Pharr and Humphreys 2010] using 16 and 256 low-discrepancy samples per pixel (spp) and traditional axis-aligned filtering. Right: Image reconstructed by our algorithm in 10 seconds from the same 16 samples per pixel. We obtain defocus quality similar to the 256 spp result in approximately 1/16th of the time.

Abstract

Traditionally, effects that require evaluating multidimensional integrals for each pixel, such as motion blur, depth of field and

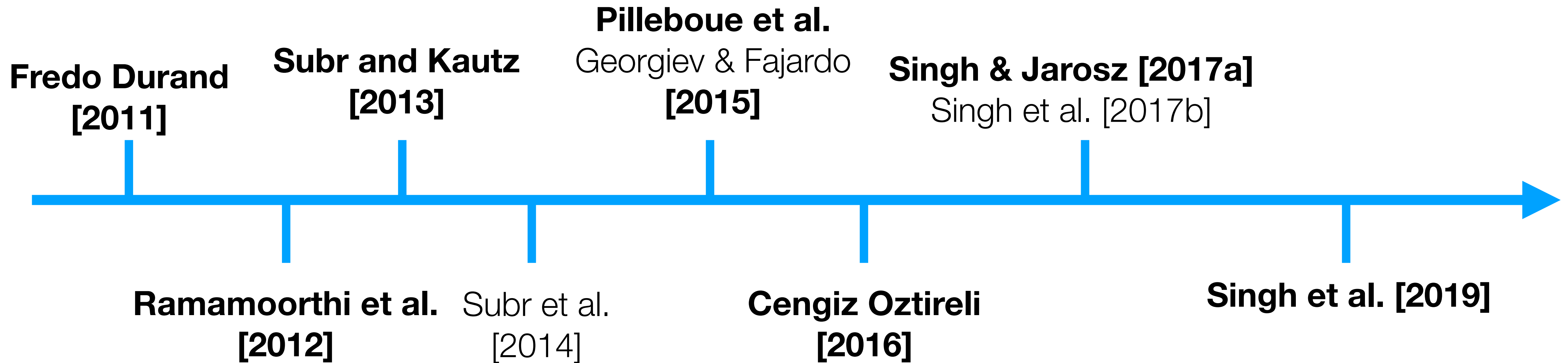
dramatic reductions in sampling rate, they rely on fairly simple reconstruction that suffers from a number of limitations. First, because they use linear reconstruction kernels and a simple model of local spectrum, they fail near object boundaries, and need to re

This STAR: Analyze sample correlations for MC sampling



Assessing MSE, bias, variance and convergence of Monte Carlo estimators using spatial and spectral tools

This STAR: Analyze sample correlations for MC sampling

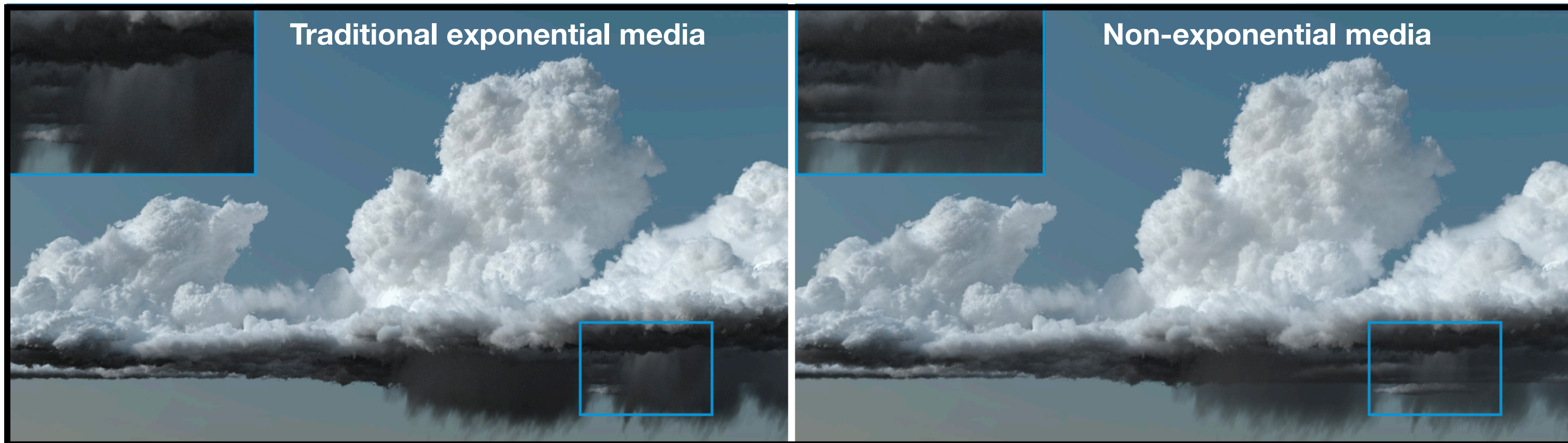


Sample correlations affect light transport / appearance

Jarabo et al. [2018]



Guo et al. [2019]



Bitterli et al. [2018]

Theoretical Tools

Point Processes

Fourier transform / Series

Samples Quality Assessment

Pair Correlation Function

Fourier Transform / Series

Error Formulations

Spatial Domain Formulations

Fourier Domain Formulations

Error Analysis

Stratification Strategies

Low Discrepancy Samplers

Stochastic Samplers