

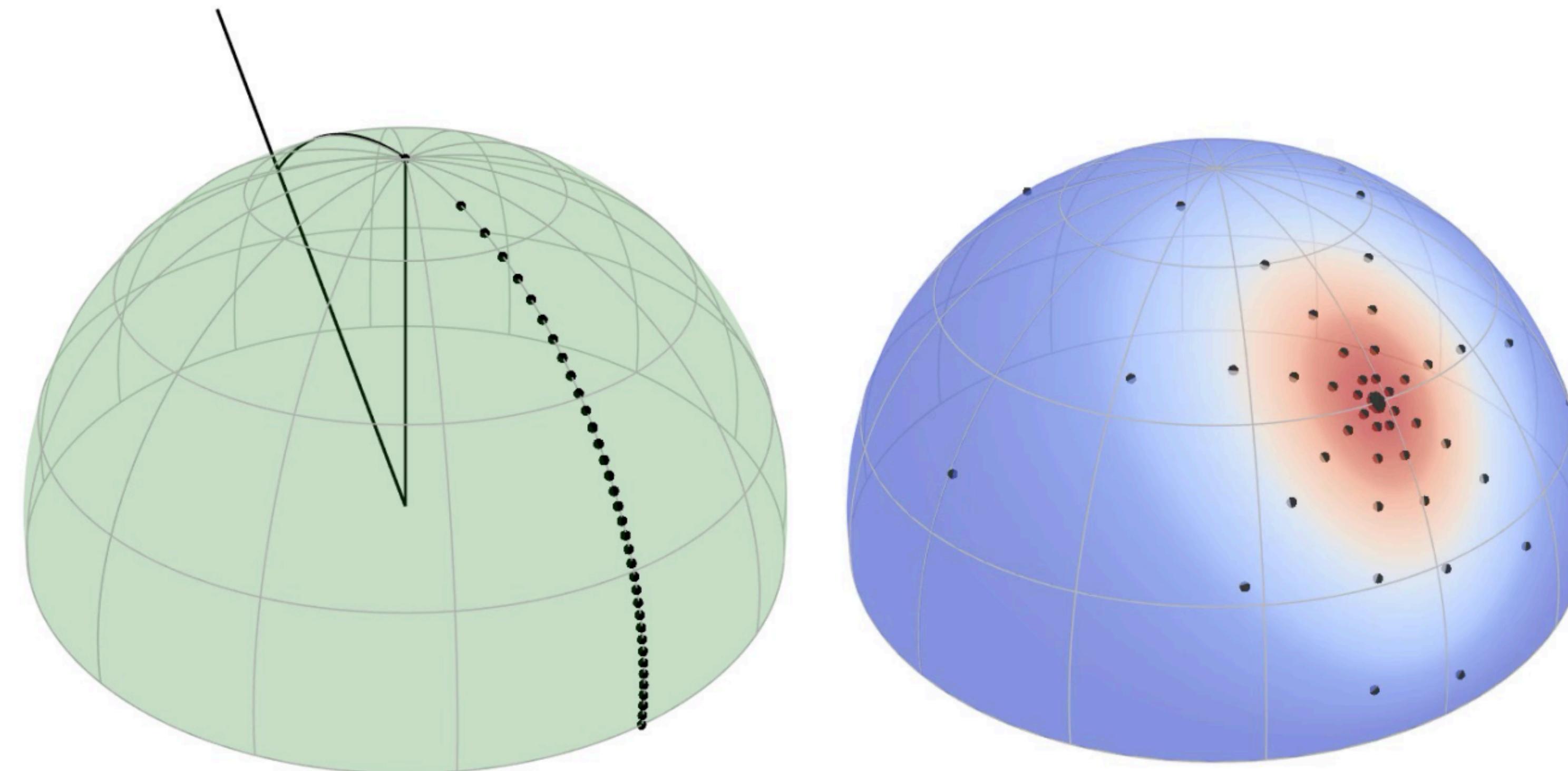
# Summary of the course and weird tricks for your renderers

UCSD CSE 272  
Advanced Image Synthesis  
Tzu-Mao Li

# What have we covered so far?

# What have we covered so far?

- BSDF measurement & microfacet theory, Schlick approximation



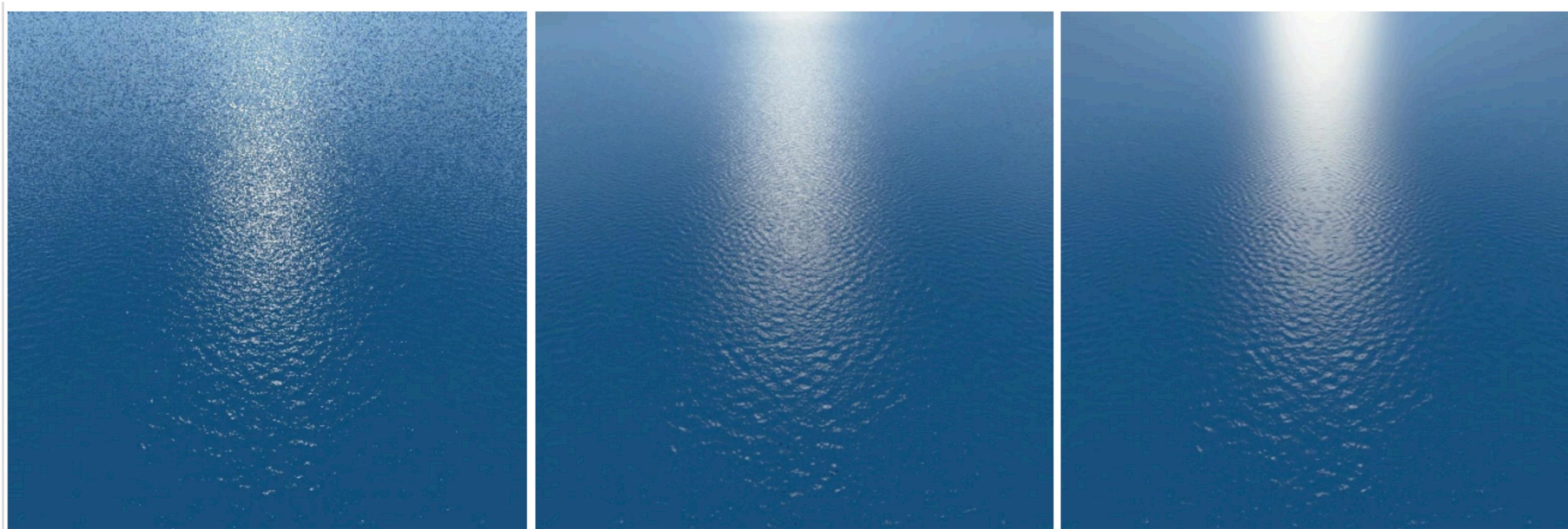
# What have we covered so far?

- Uber BSDF



# What have we covered so far?

- normal map filtering



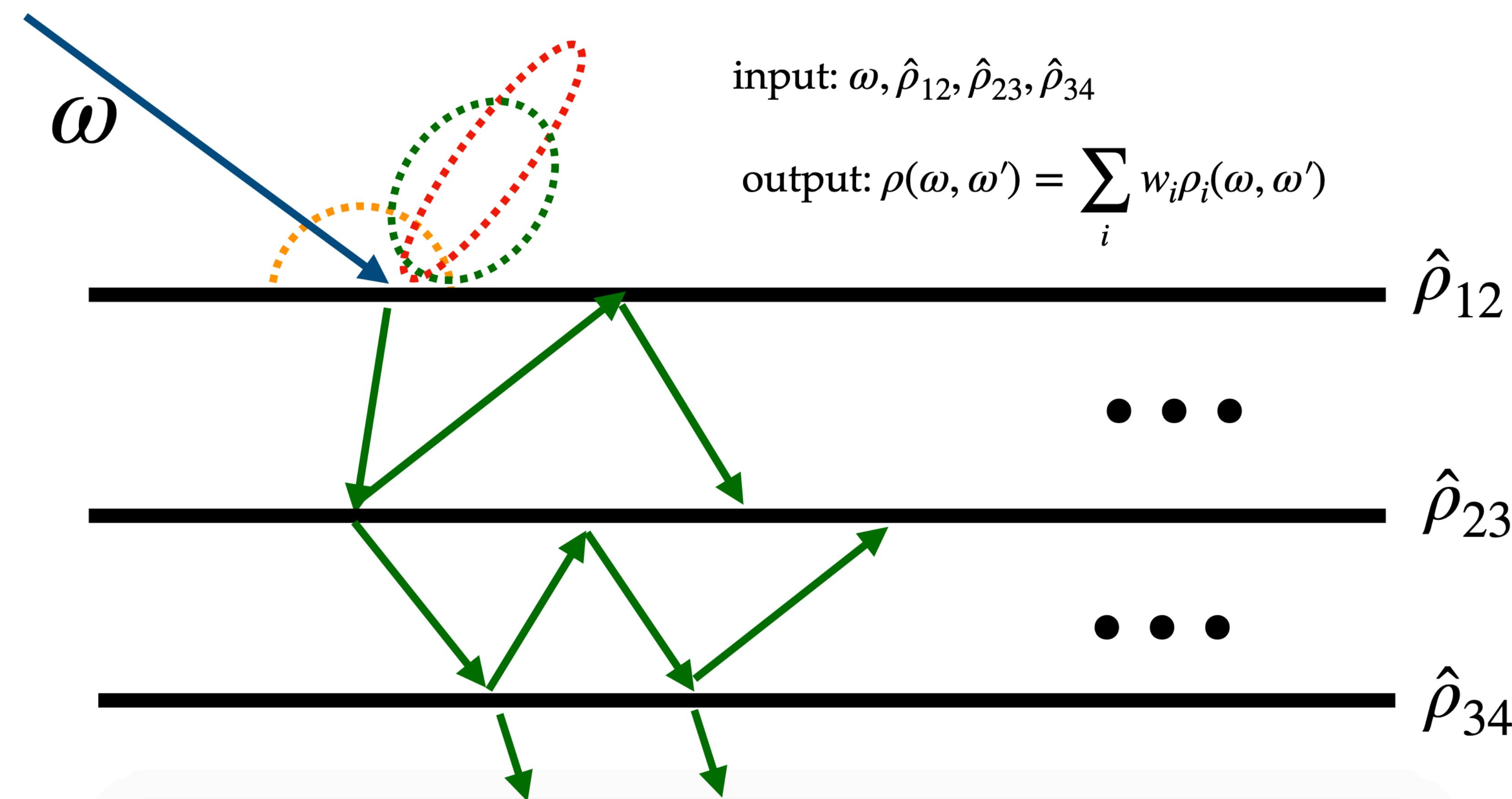
undersampled rendering

ground truth  
**(slow)**

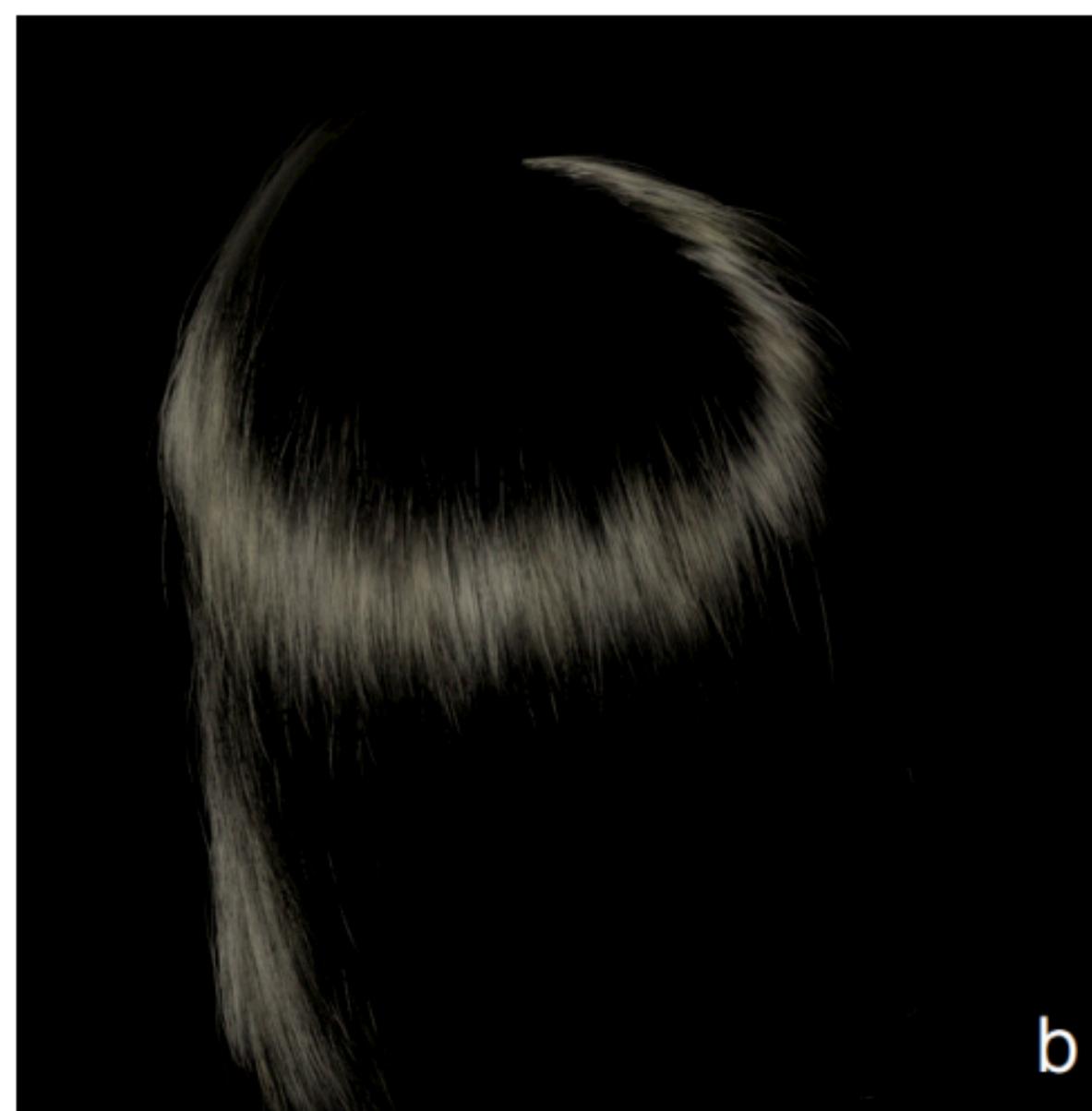
LEAN mapping  
+  
microfacet BRDF  
**(real time)**

# What have we covered so far?

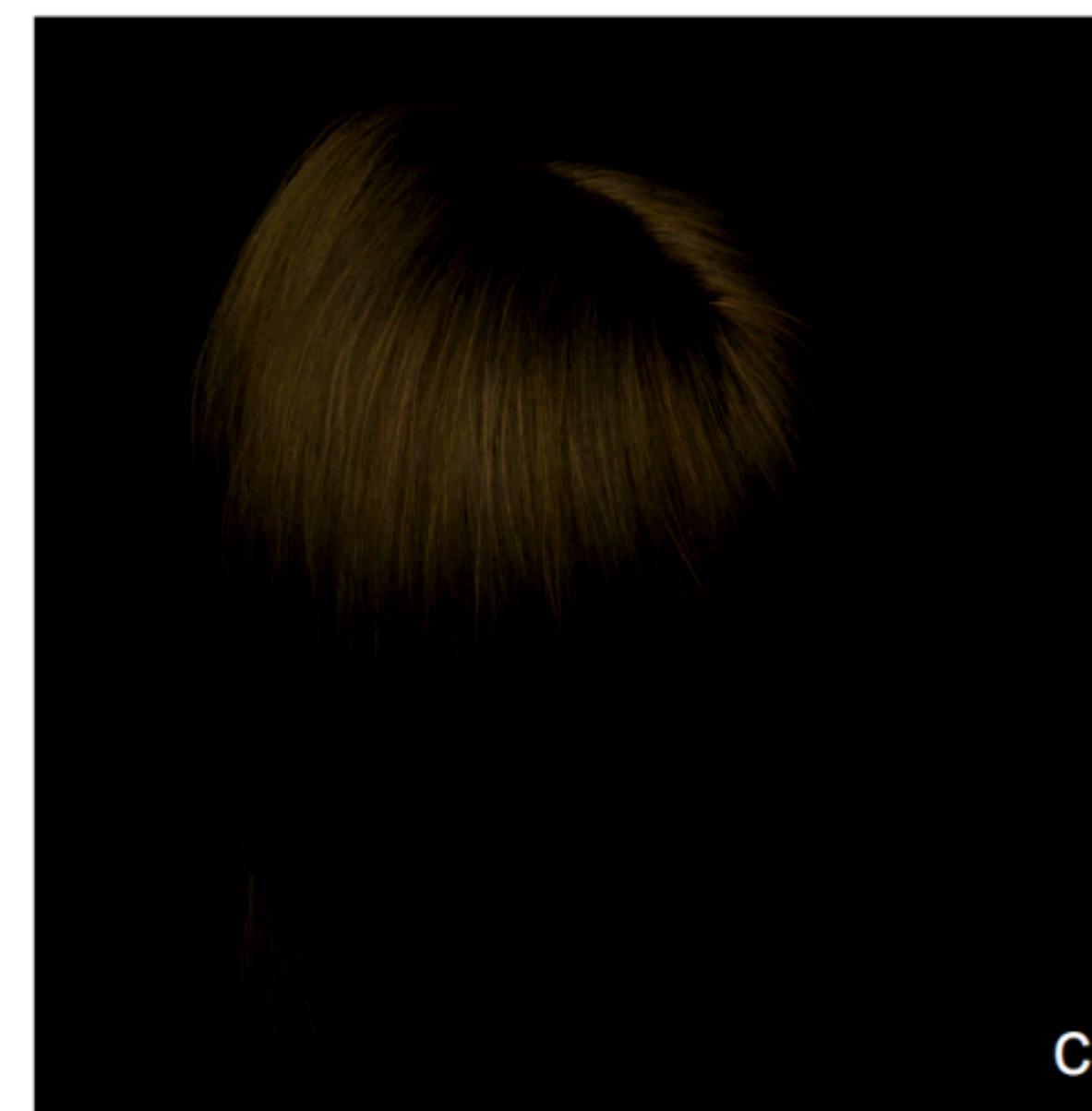
- layered BSDFs



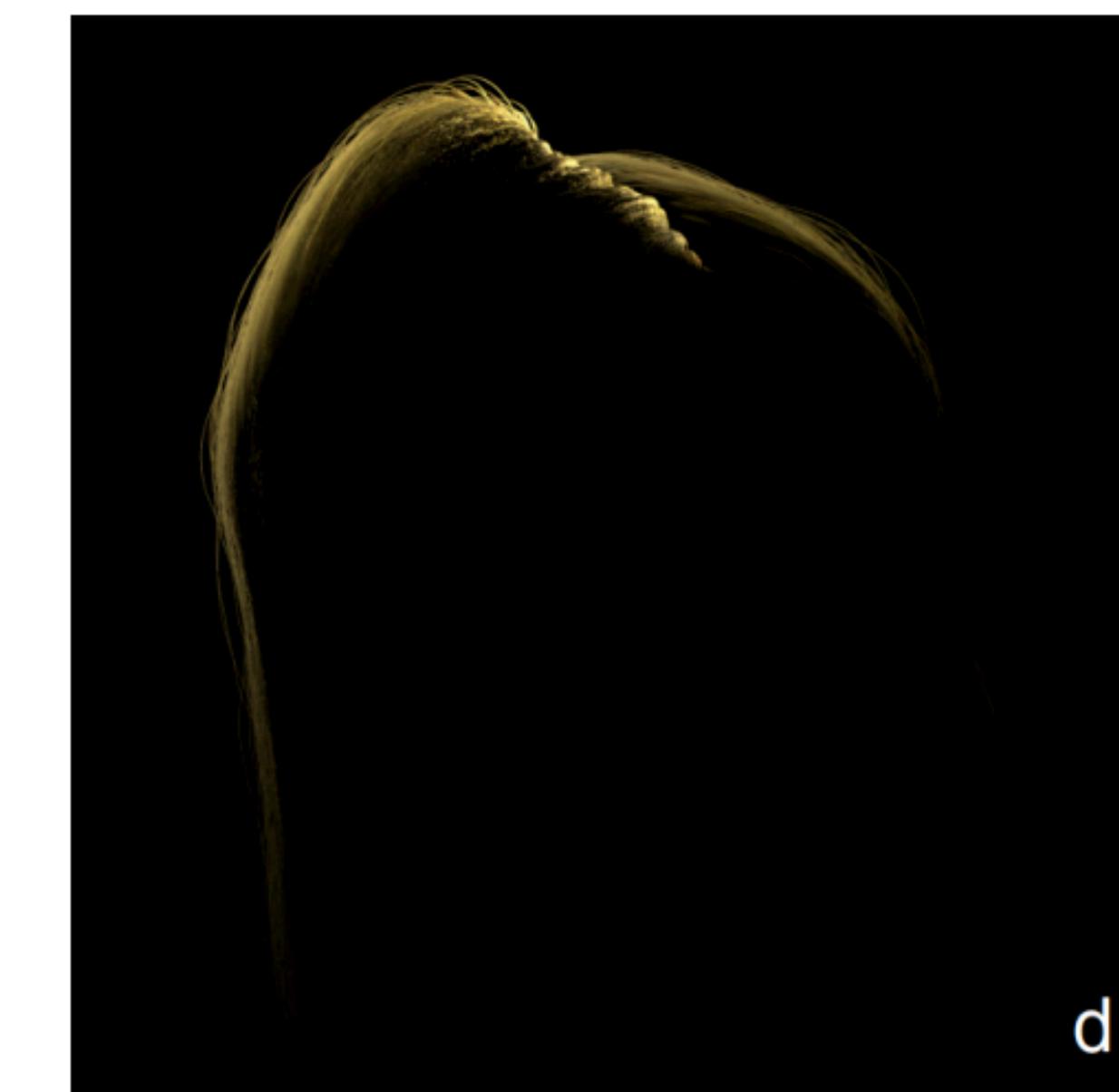
# Hair and cloth rendering



$R$

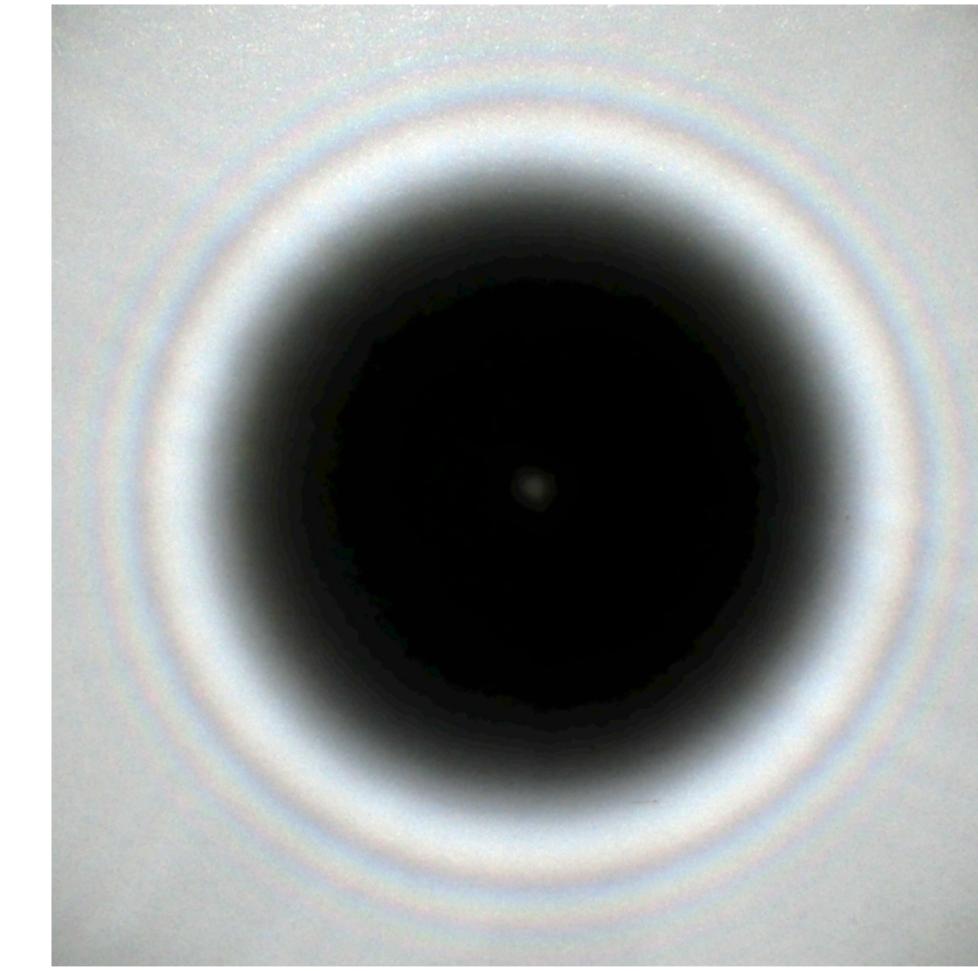
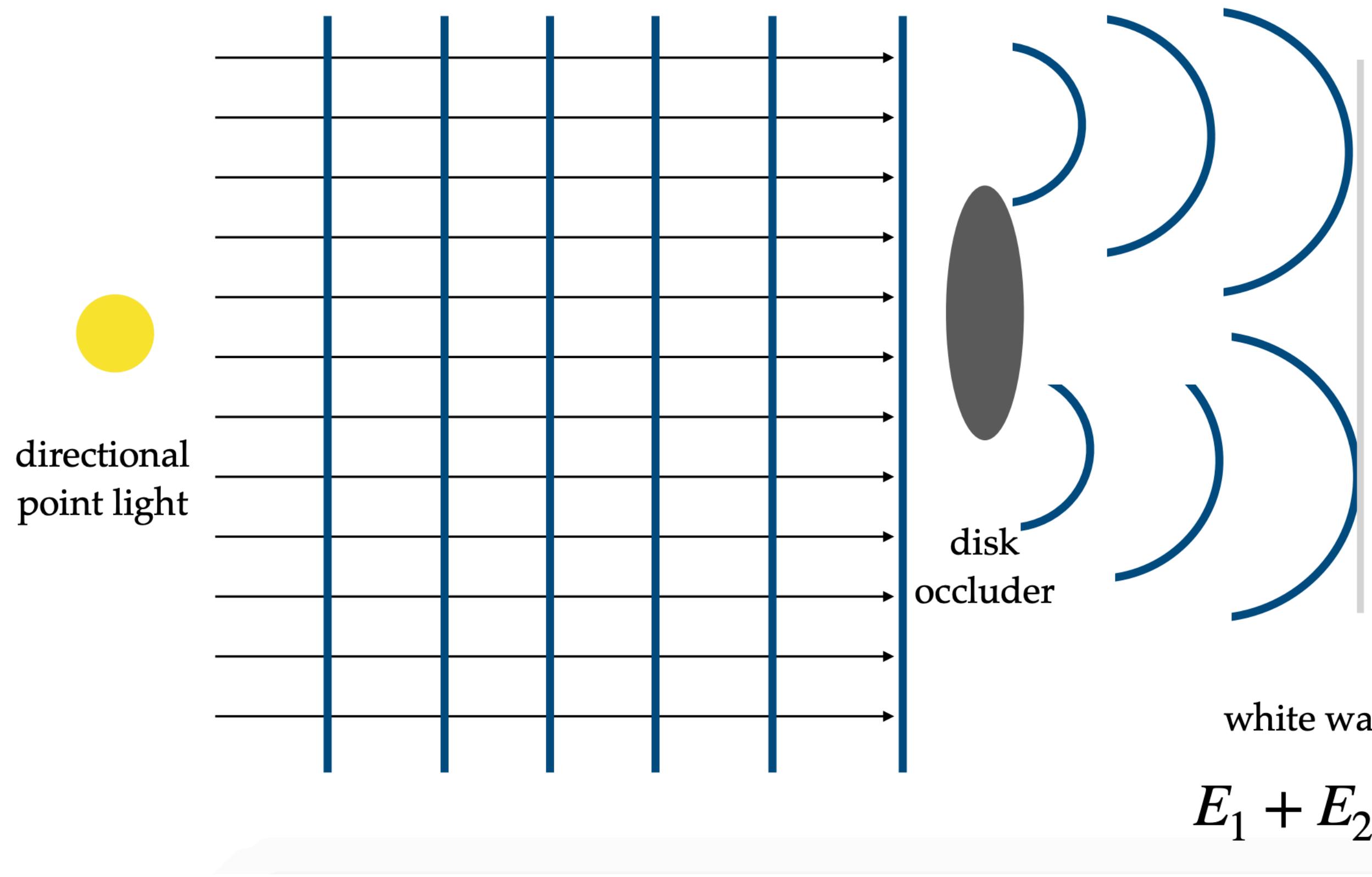


$TRT$



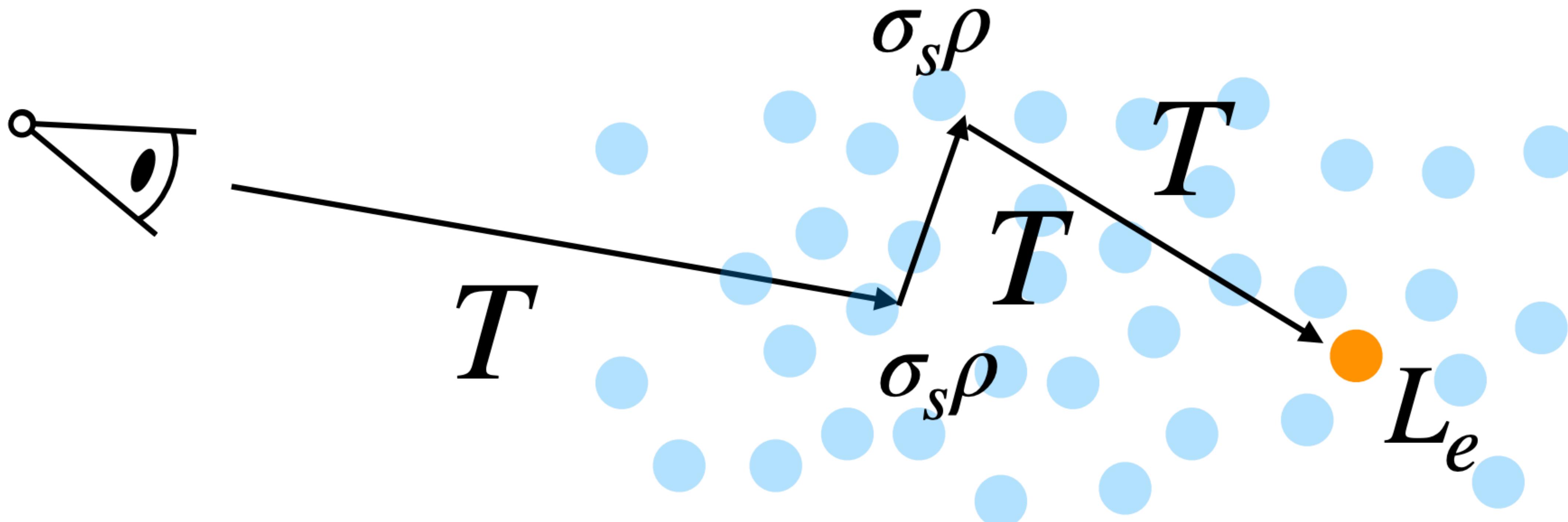
$TT$

# Wave optics

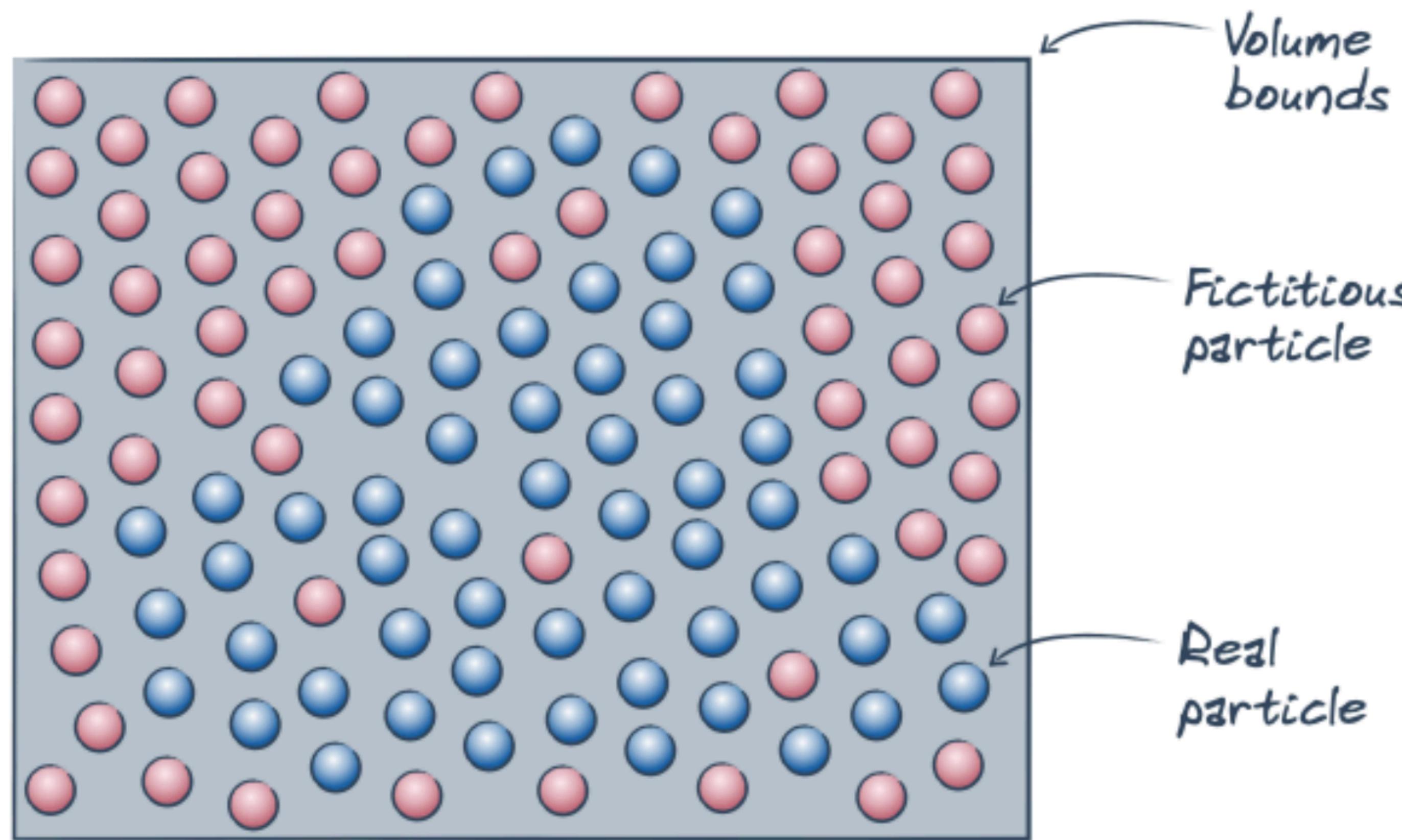


$$E_1 + E_2 = \Re \left( (A_1 e^{i\phi_1} + A_2 e^{i\phi_2}) e^{i\omega t} \right)$$

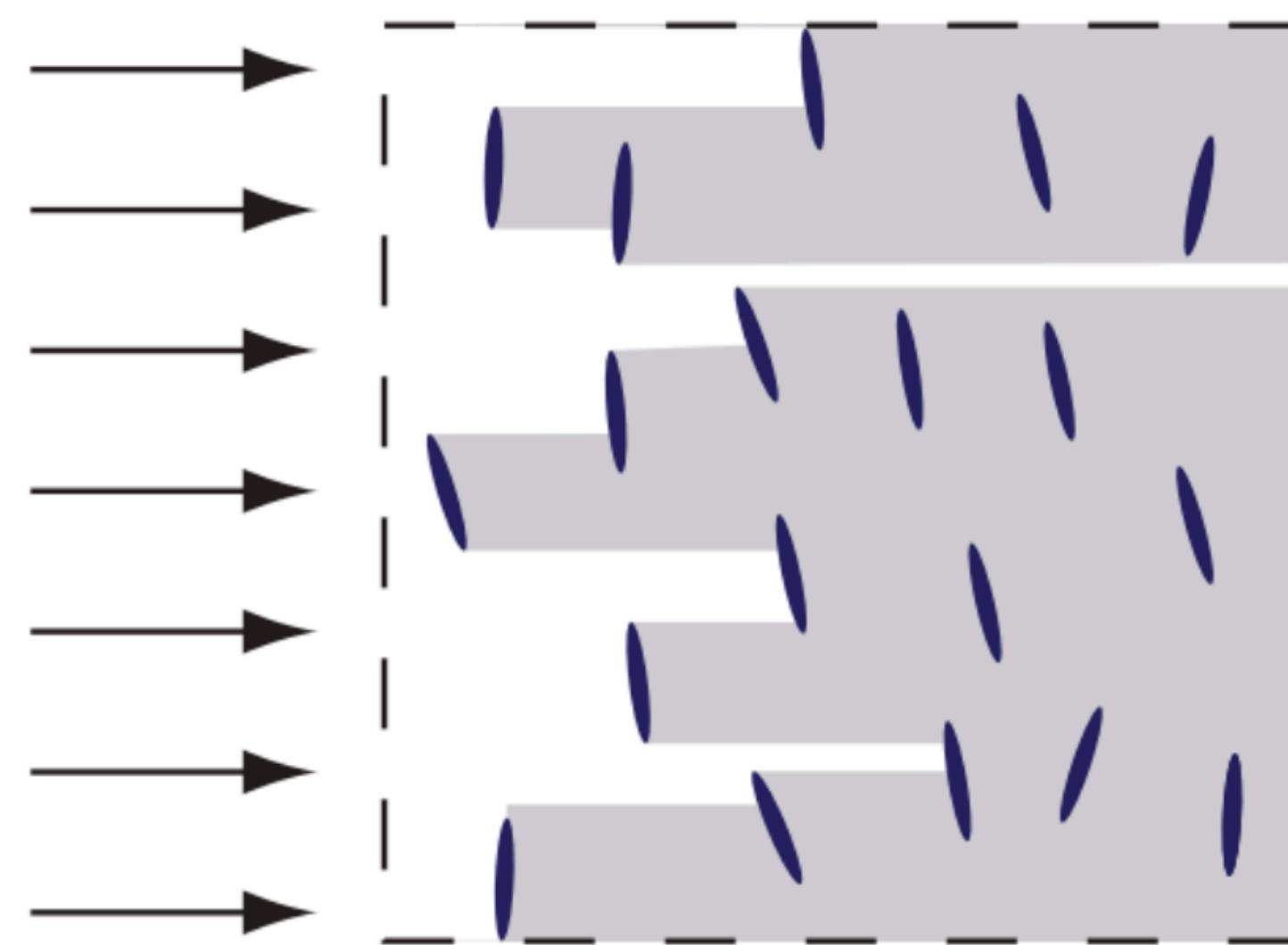
# Radiative transfer equation



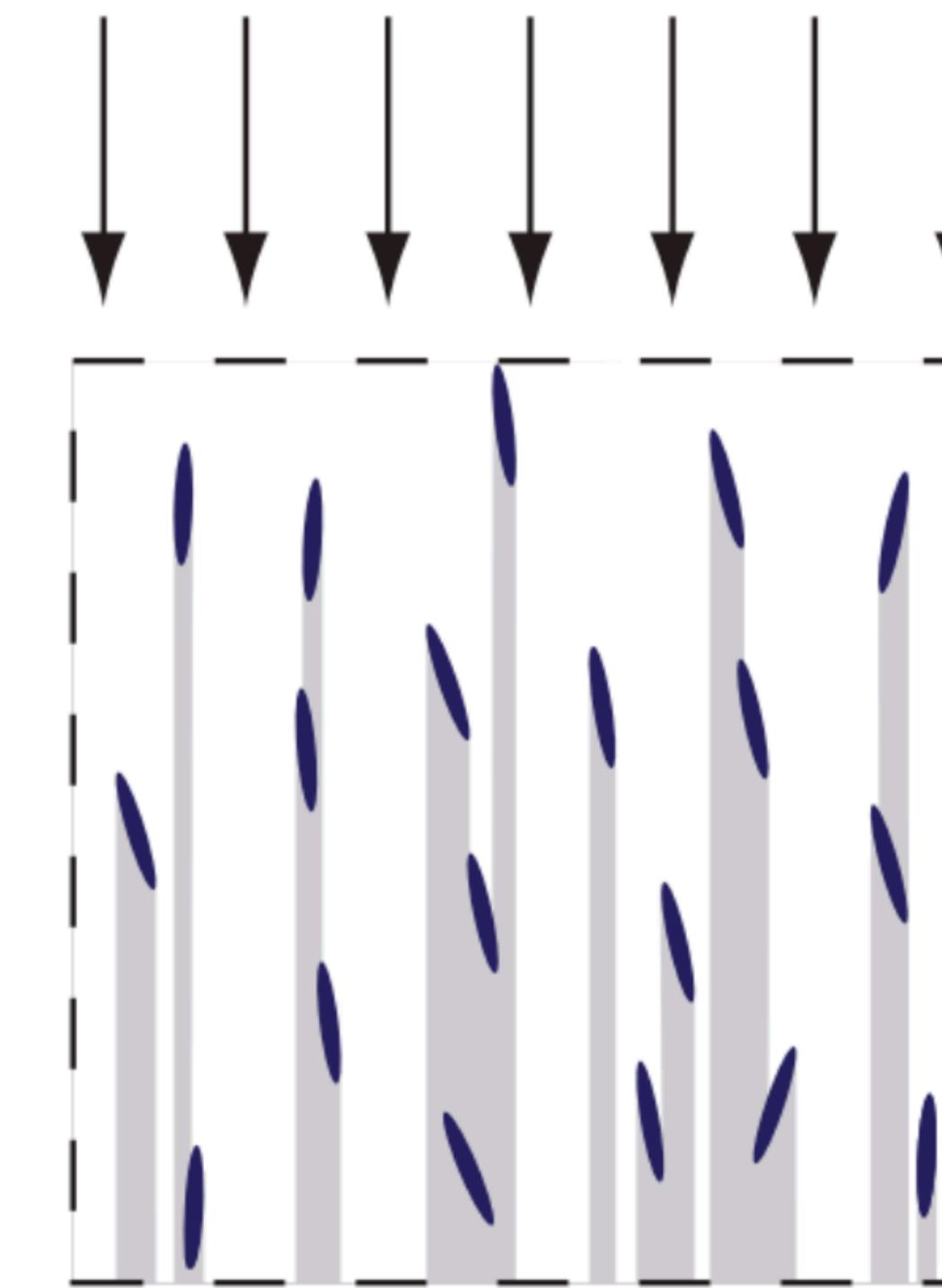
# Transmittance estimation & null scattering



# Microflake theory & SGGX



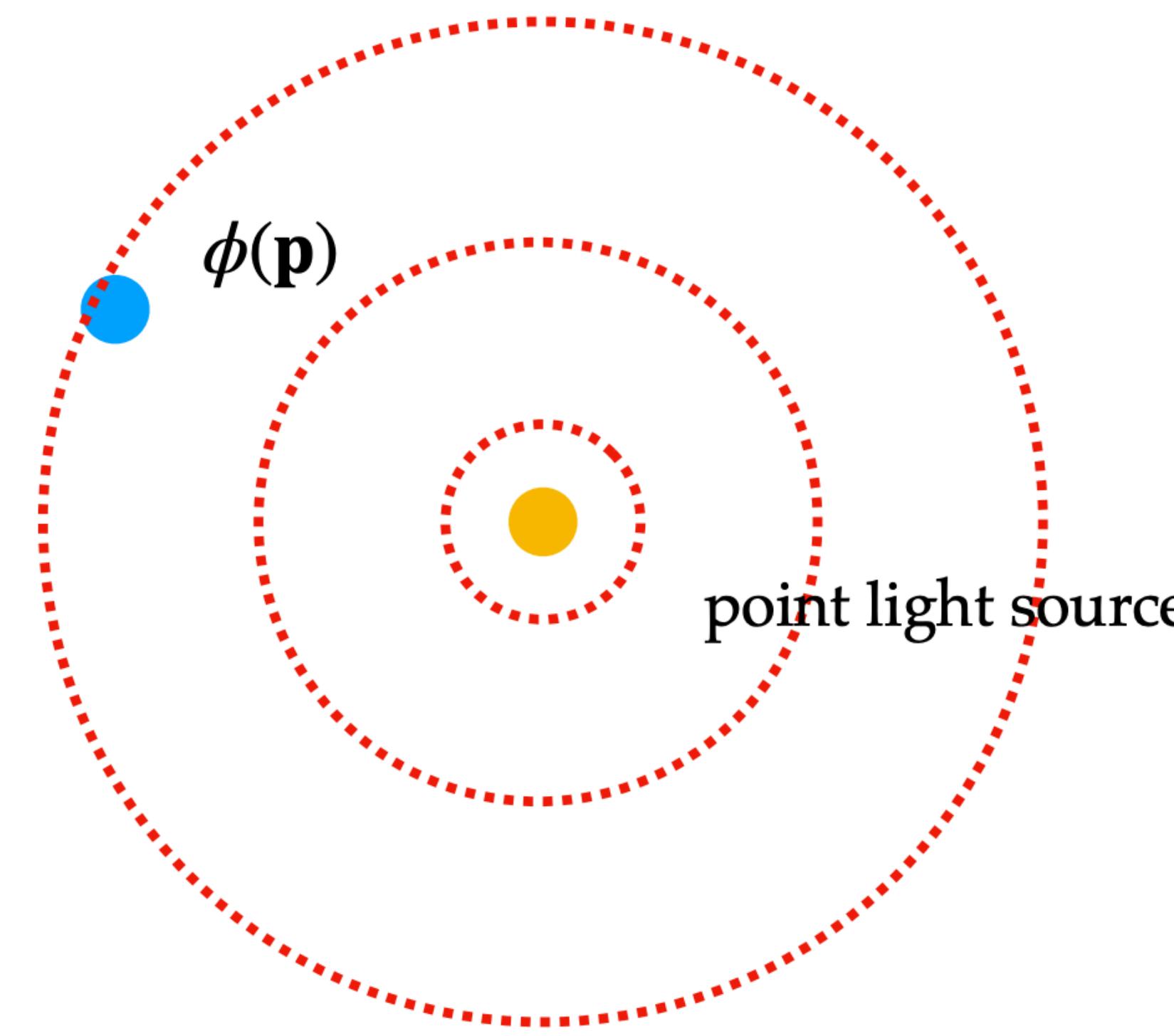
high  $\sigma_t$



low  $\sigma_t$

# Diffusion approximation

$$\frac{1}{3\sigma_t} \Delta \phi(\mathbf{p}) = \sigma_a \phi(\mathbf{p}) - Q_0(\mathbf{p}) + \frac{1}{\sigma_t} \nabla \cdot Q_1(\mathbf{p})$$



$$Q_0 = \delta(\mathbf{p})$$

$$Q_1 = 0$$

# Differentiable rendering: edge sampling

$$\frac{\partial}{\partial p} \iint \text{triangle} = \iint \frac{\partial}{\partial p} \text{triangle} + \int \text{boundary}$$

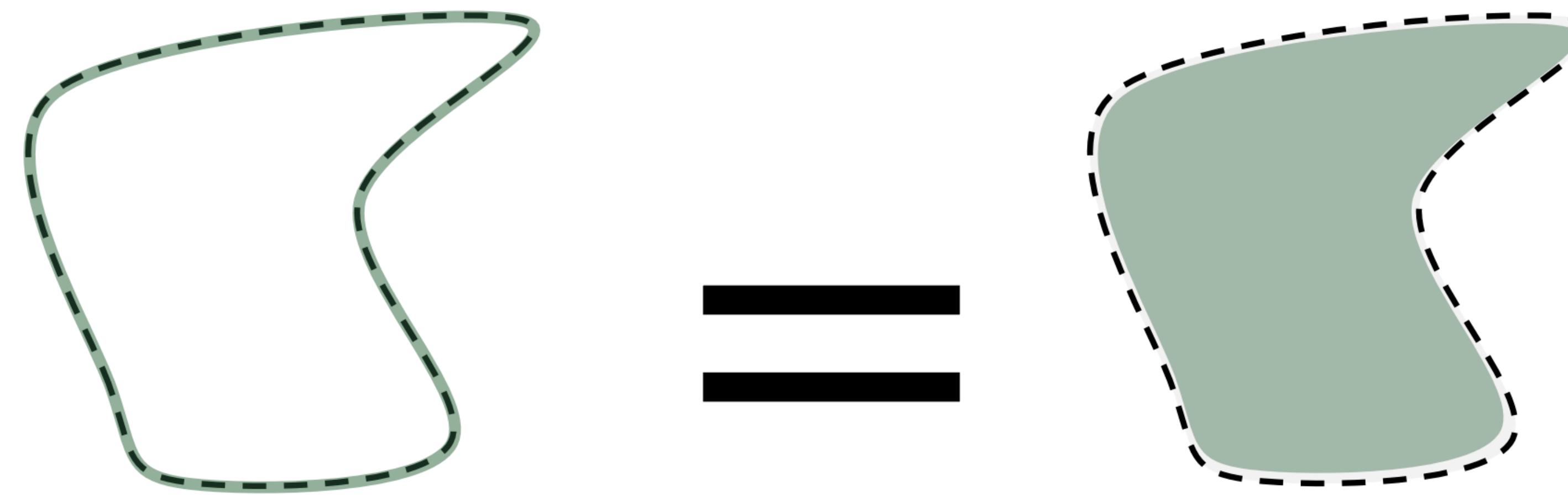
interior derivative

boundary derivative

Reynolds transport theorem  
[Reynolds 1903]

The diagram illustrates the Reynolds transport theorem for differentiable rendering. It shows a triangle being divided into interior and boundary regions. The interior region is labeled "interior derivative" and the boundary region is labeled "boundary derivative". The boundary is shown as a dashed line with points on it. The equation shows the total derivative being equal to the sum of the interior derivative and the boundary derivative.

# Differentiable rendering: warped-area sampling

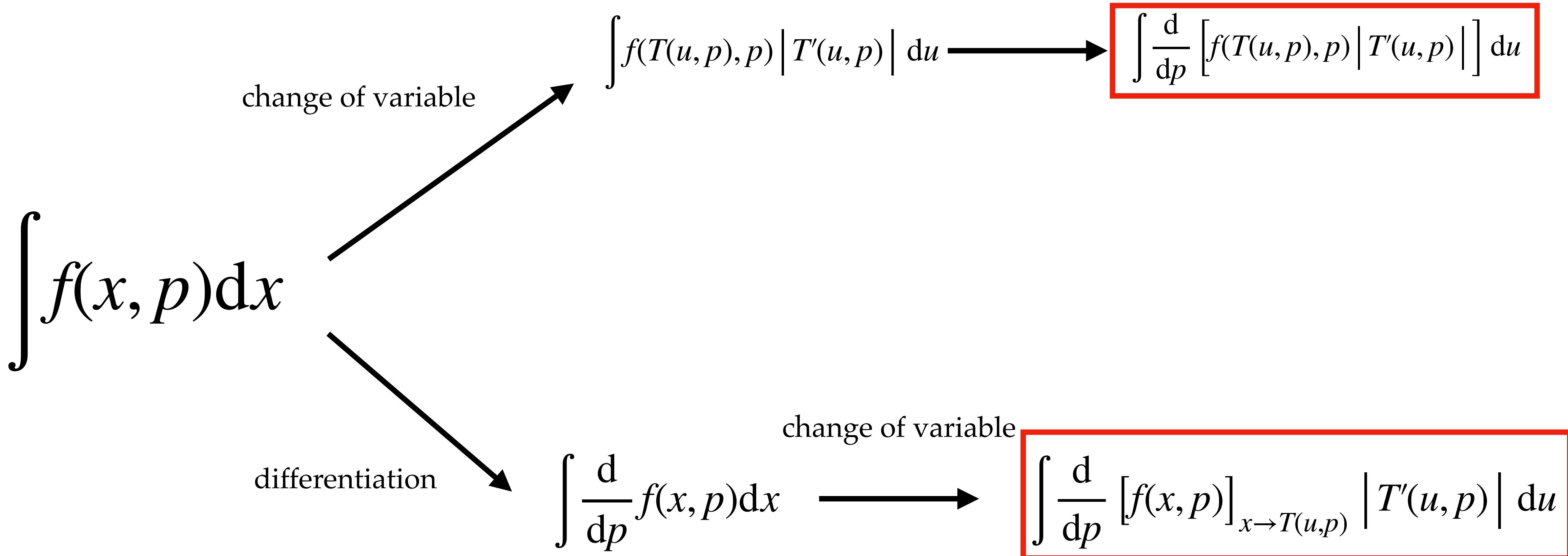


$$\int_{\partial D} \vec{f} \cdot \vec{n}$$

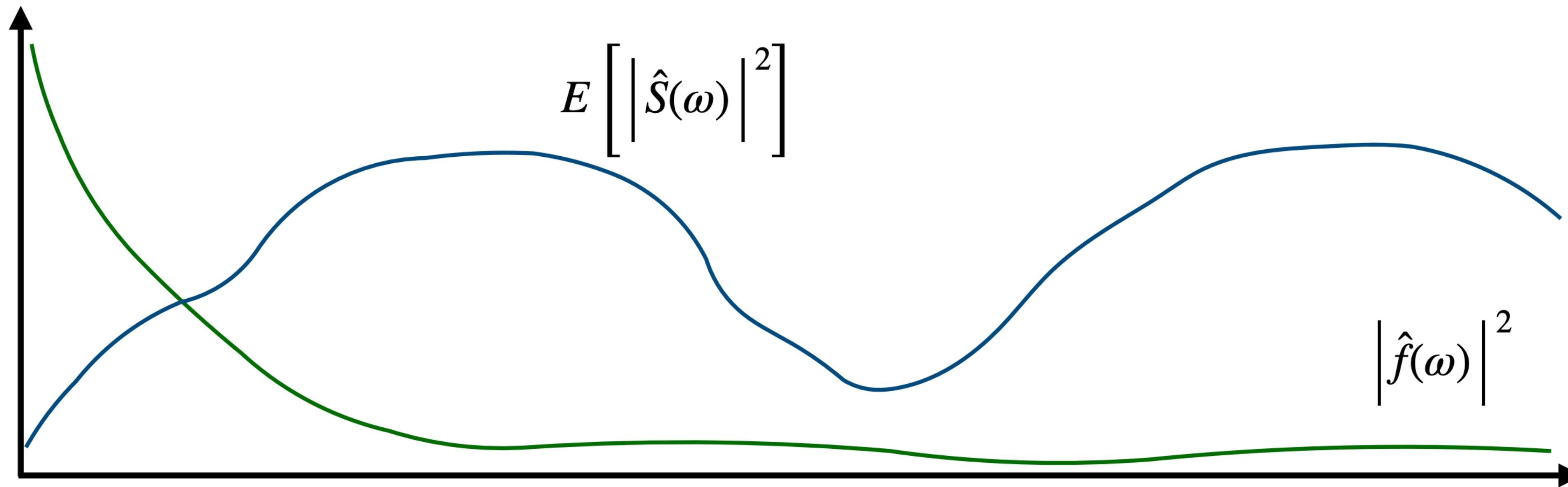
$$\int_D \nabla \cdot \vec{f}$$

# Differentiable rendering: importance sampling

differentiation

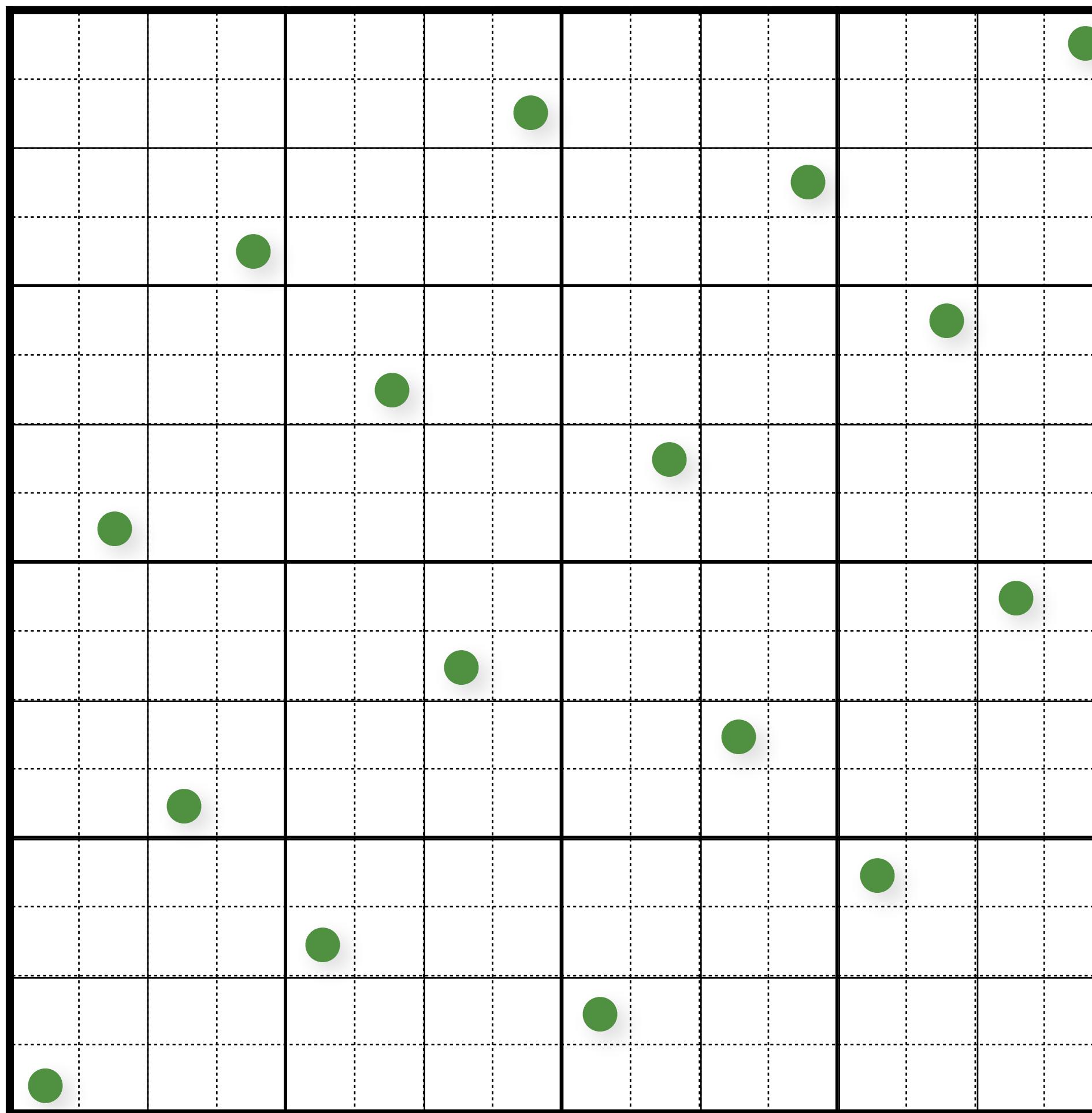


# Stratification

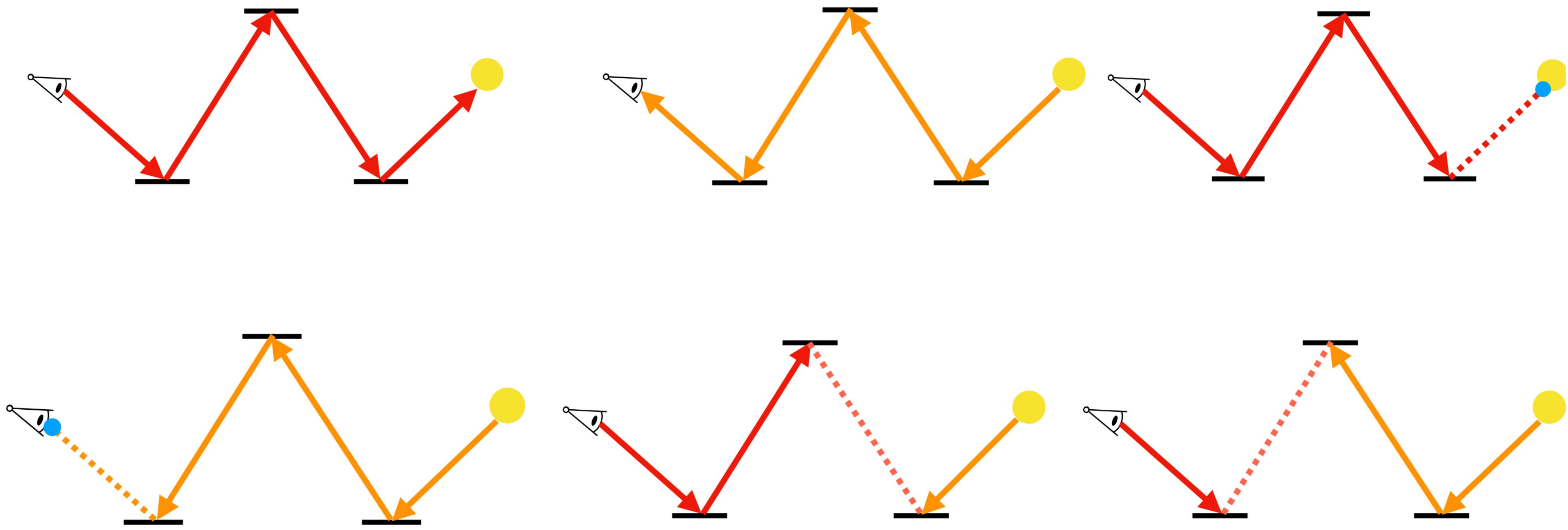


$$\text{variance} = \int |\hat{f}(\omega)|^2 E[|\hat{S}(\omega)|^2] d\omega$$

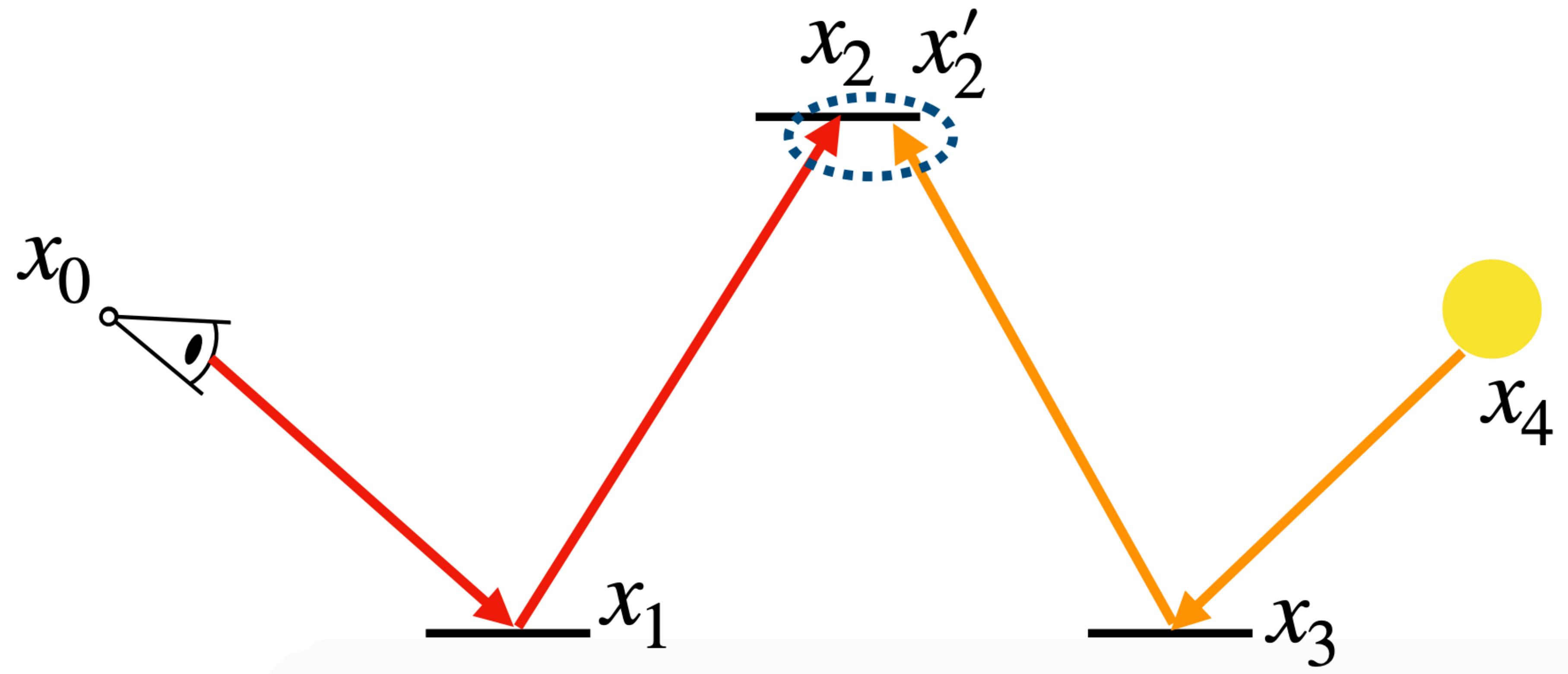
# Low-discrepancy sequences



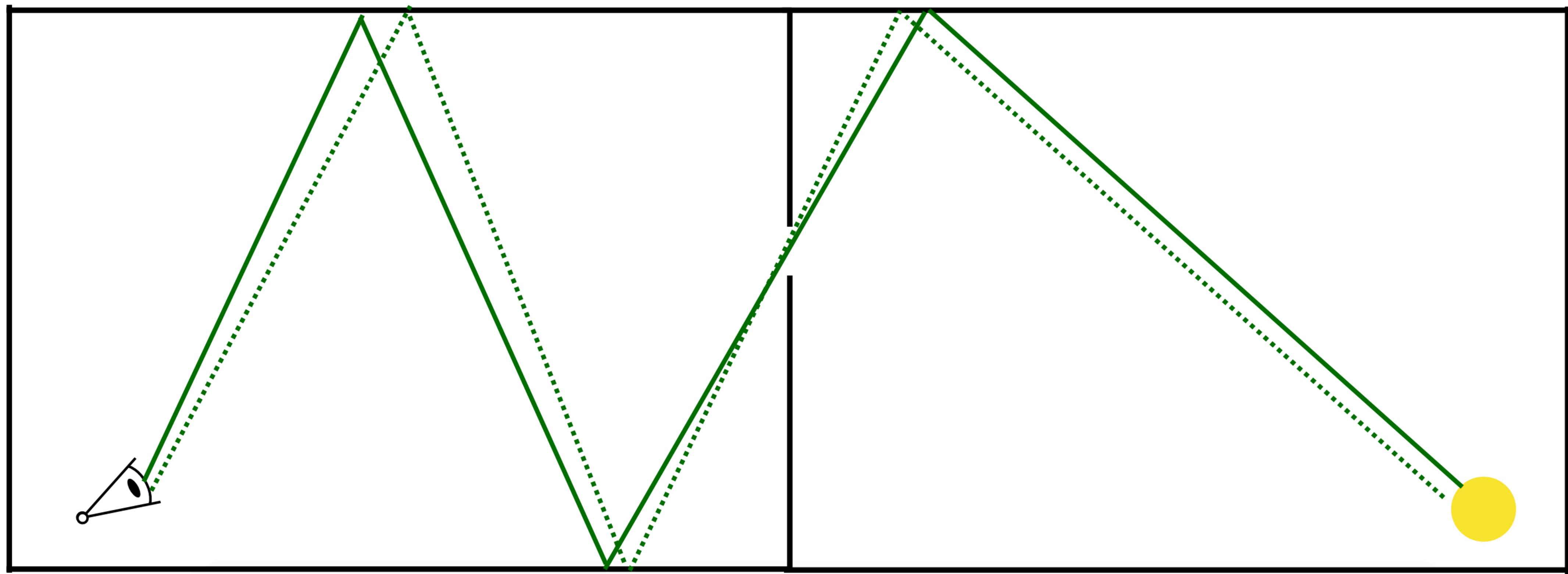
# Path-space



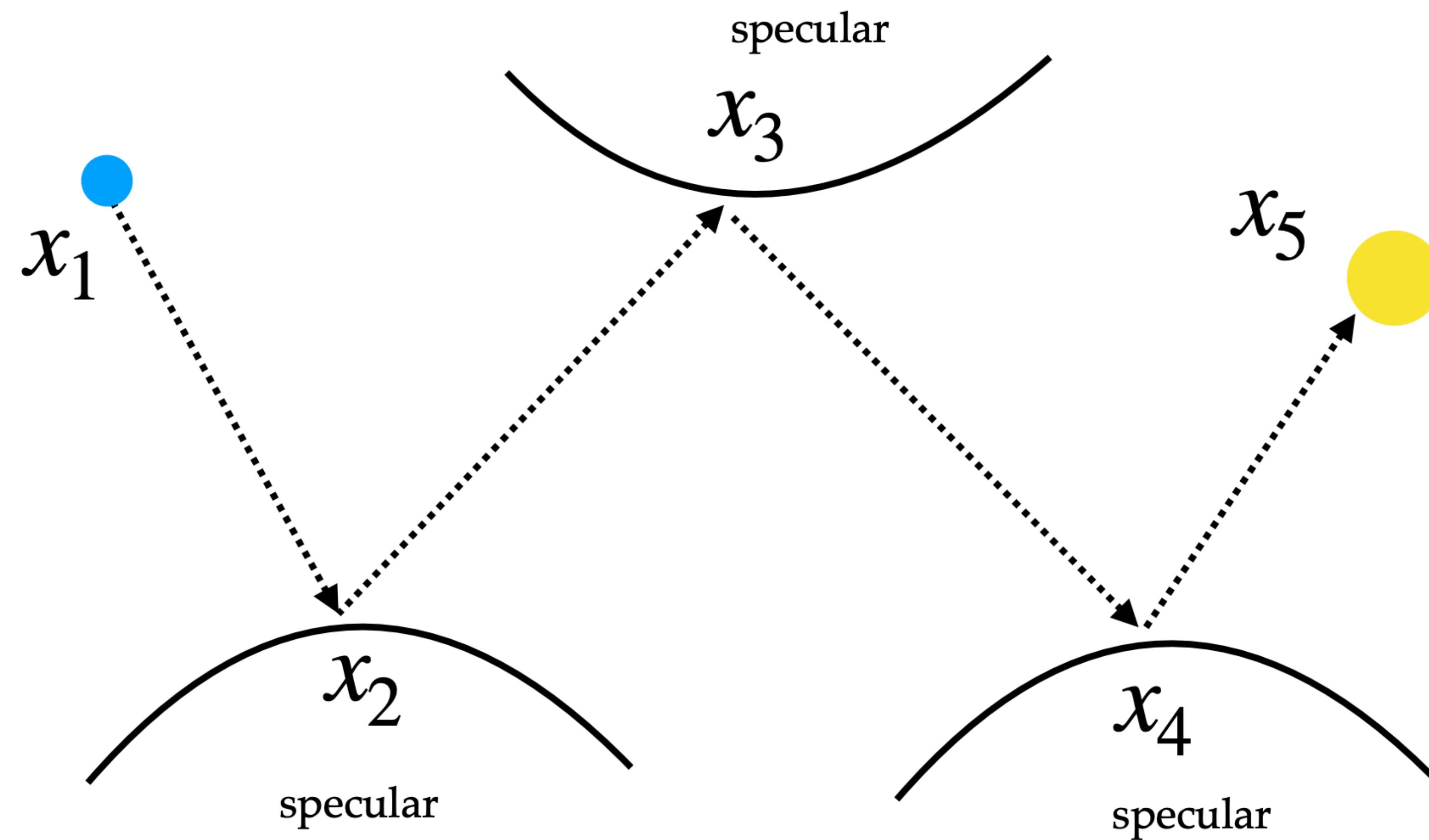
# Photon mapping



# Metropolis light transport



# Specular light path rendering



# Multiple importance sampling++

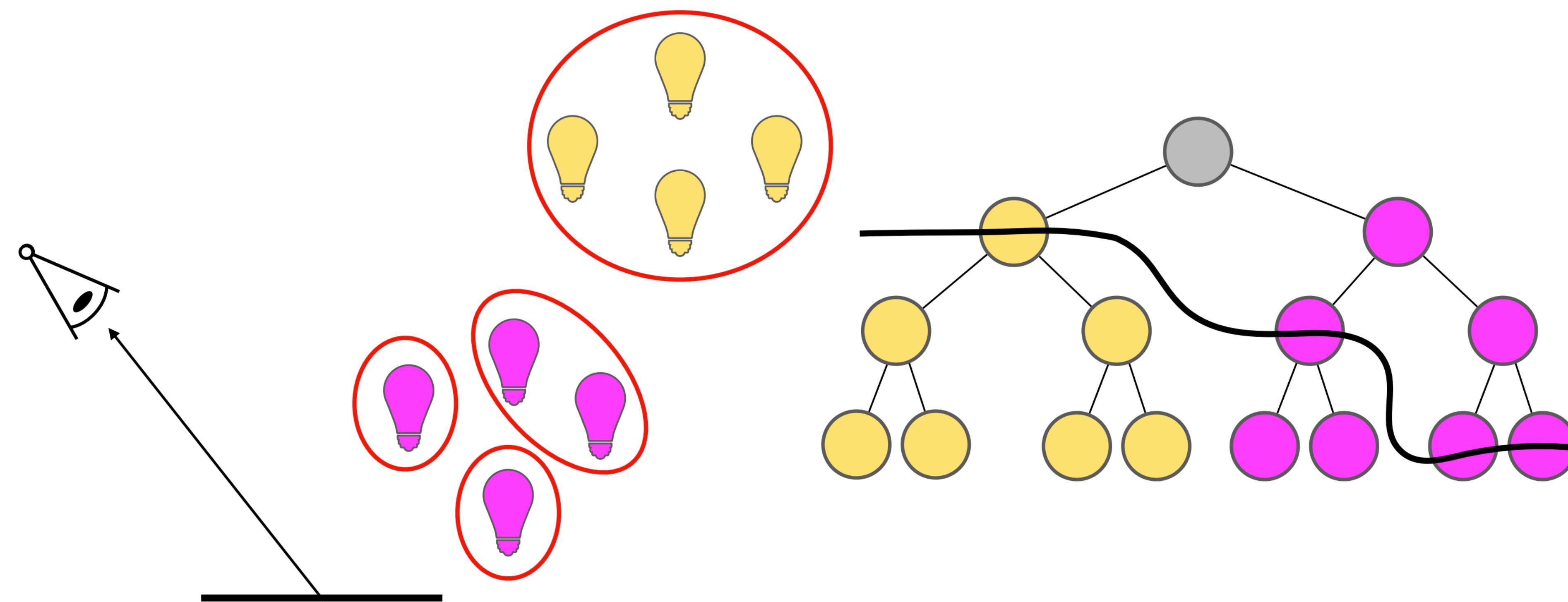
$$w_i = \frac{p_i}{f} a_i - p_i \cdot \frac{\sum_j \frac{p_j}{f} a_j - 1}{\sum_j p_j}$$

$$\begin{bmatrix} A_{00} & A_{01} \\ A_{10} & A_{11} \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \end{bmatrix} = \begin{bmatrix} b_0 \\ b_1 \end{bmatrix}$$

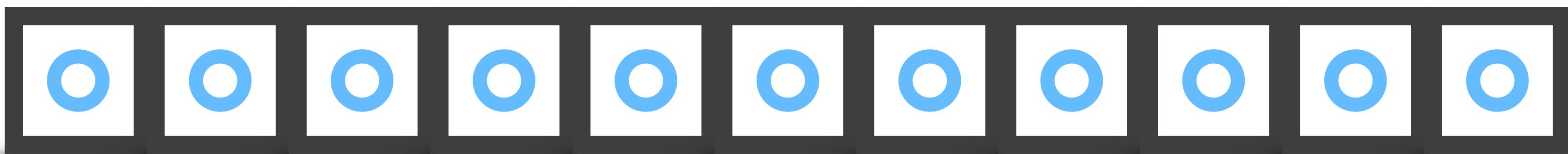
$$A_{ij} = \int \frac{p_i p_j}{\sum_k p_k}$$

$$b_i = \int \frac{fp_i}{\sum_k p_k}$$

# Many-lights rendering



# ReSTIR



# History of Computer Animation



# Production rendering

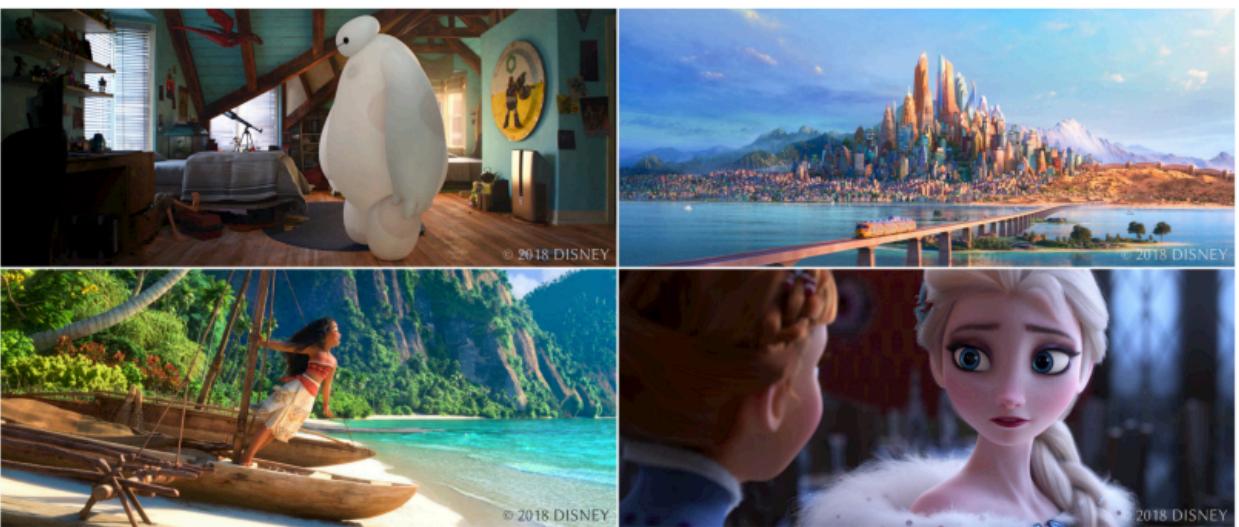
## RenderMan: An Advanced Path Tracing Architecture for Movie Rendering

PER CHRISTENSEN, JULIAN FONG, JONATHAN SHADE, WAYNE WOOTEN, BRENDEN SCHUBERT, ANDREW KENSLER, STEPHEN FRIEDMAN, CHARLIE KILPATRICK, CLIFF RAMSHAW, MARC BANNERISTER, BRENTON RAYNER, JONATHAN BROUILLAT, and MAX LIANI, Pixar Animation Studios



## The Design and Evolution of Disney's Hyperion Renderer

BRENT BURLEY, DAVID ADLER, MATT JEN-YUAN CHIANG, HANK DRISKILL, RALF HABEL, PATRICK KELLY, PETER KUTZ, YINING KARL LI, and DANIEL TEECE, Walt Disney Animation Studios



## Vectorized Production Path Tracing

Mark Lee  
DreamWorks Animation

Feng Xie  
DreamWorks Animation

Brian Green  
DreamWorks Animation

Eric Tabellion  
DreamWorks Animation



## Sony Pictures Imageworks Arnold

CHRISTOPHER KULLA, Sony Pictures Imageworks  
ALEJANDRO CONTY, Sony Pictures Imageworks  
CLIFFORD STEIN, Sony Pictures Imageworks  
LARRY GRITZ, Sony Pictures Imageworks



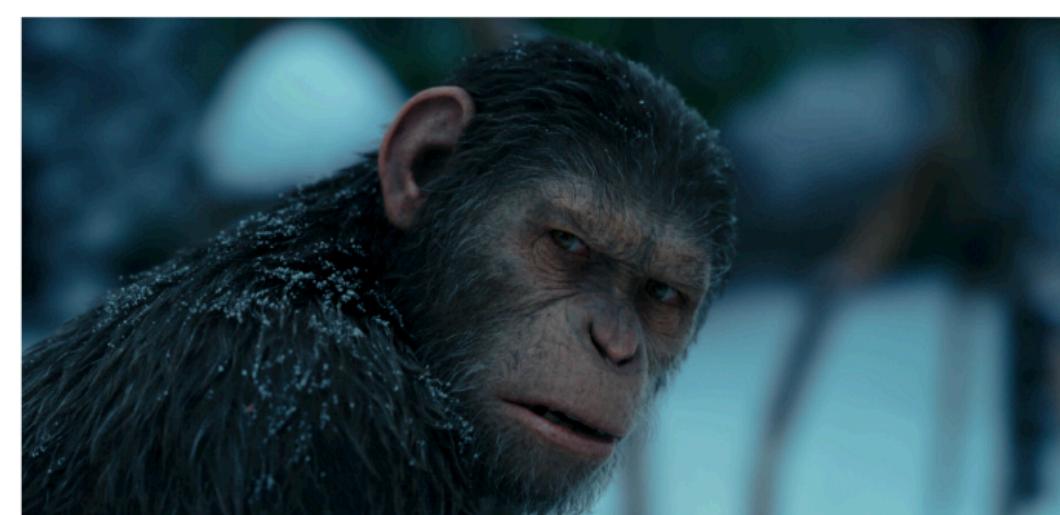
## Arnold: A Brute-Force Production Path Tracer

ILYAN GEORGIEV, THIAGO IZE, MIKE FARNSWORTH, RAMÓN MONTOYA-VOZMEDIANO, ALAN KING, BRECHT VAN LOMMEL, ANGEL JIMENEZ, OSCAR ANSON, SHINJI OGAKI, ERIC JOHNSTON, ADRIEN HERUBEL, DECLAN RUSSELL, FRÉDÉRIC SERVANT, and MARCOS FAJARDO, Solid Angle

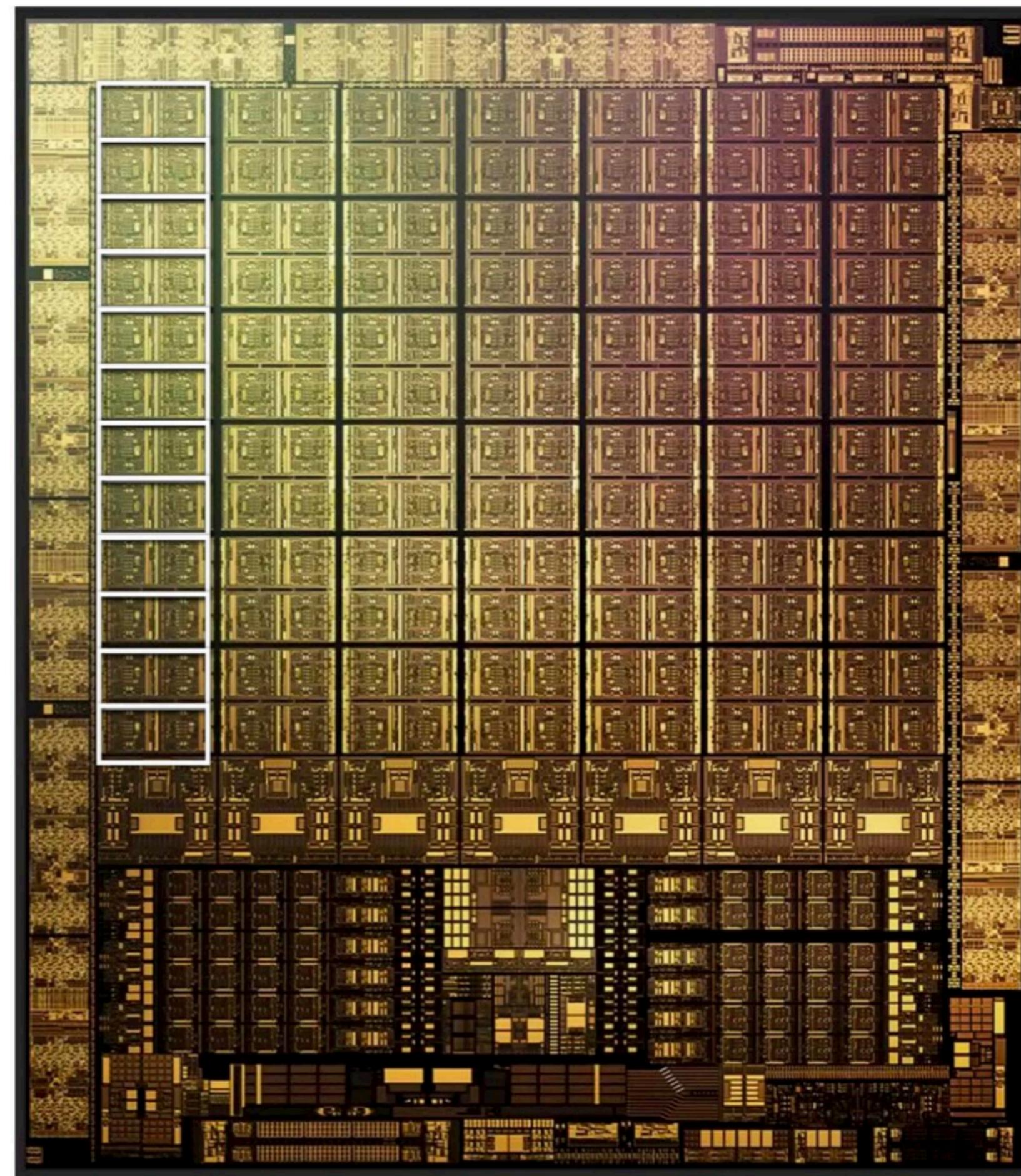


## Manuka: A batch-shading architecture for spectral path tracing in movie production

LUCA FASCIONE, JOHANNES HANIKA, MARK LEONE, MARC DROSKE, JORGE SCHWARZHaupt, TOMÁŠ DAVIDOVÍC, ANDREA WEIDLICH, and JOHANNES MENG, Weta Digital



# GPU architectures



# Nanite



# What haven't we covered?

- closed-form / analytical methods
- transient rendering
- speckle rendering
- refractive radiative transfer equation
- light baking / precomputed radiance transfer (in CSE 168)
- Monte Carlo denoising (in CSE 168)
- adaptive importance sampling (path guiding) (in CSE 168)



# Fill the teaching evaluation!

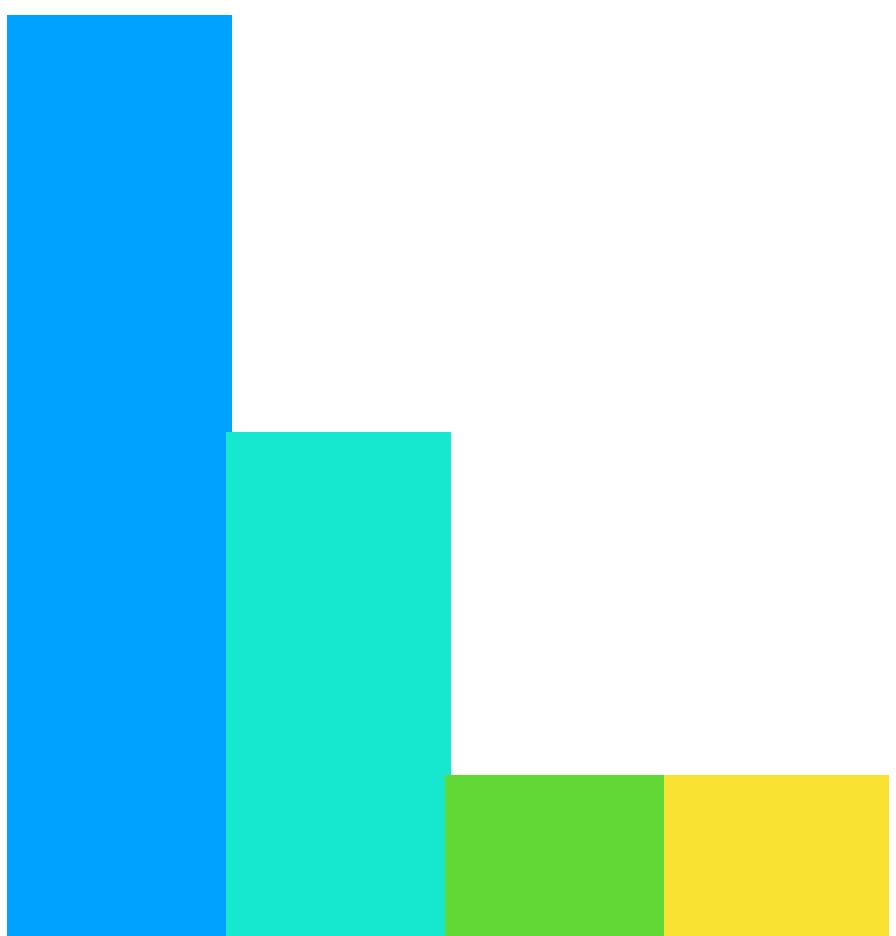
- <https://academicaffairs.ucsd.edu/Modules/Evals/>
- important for future students and my career : >



# Random tricks and stuff

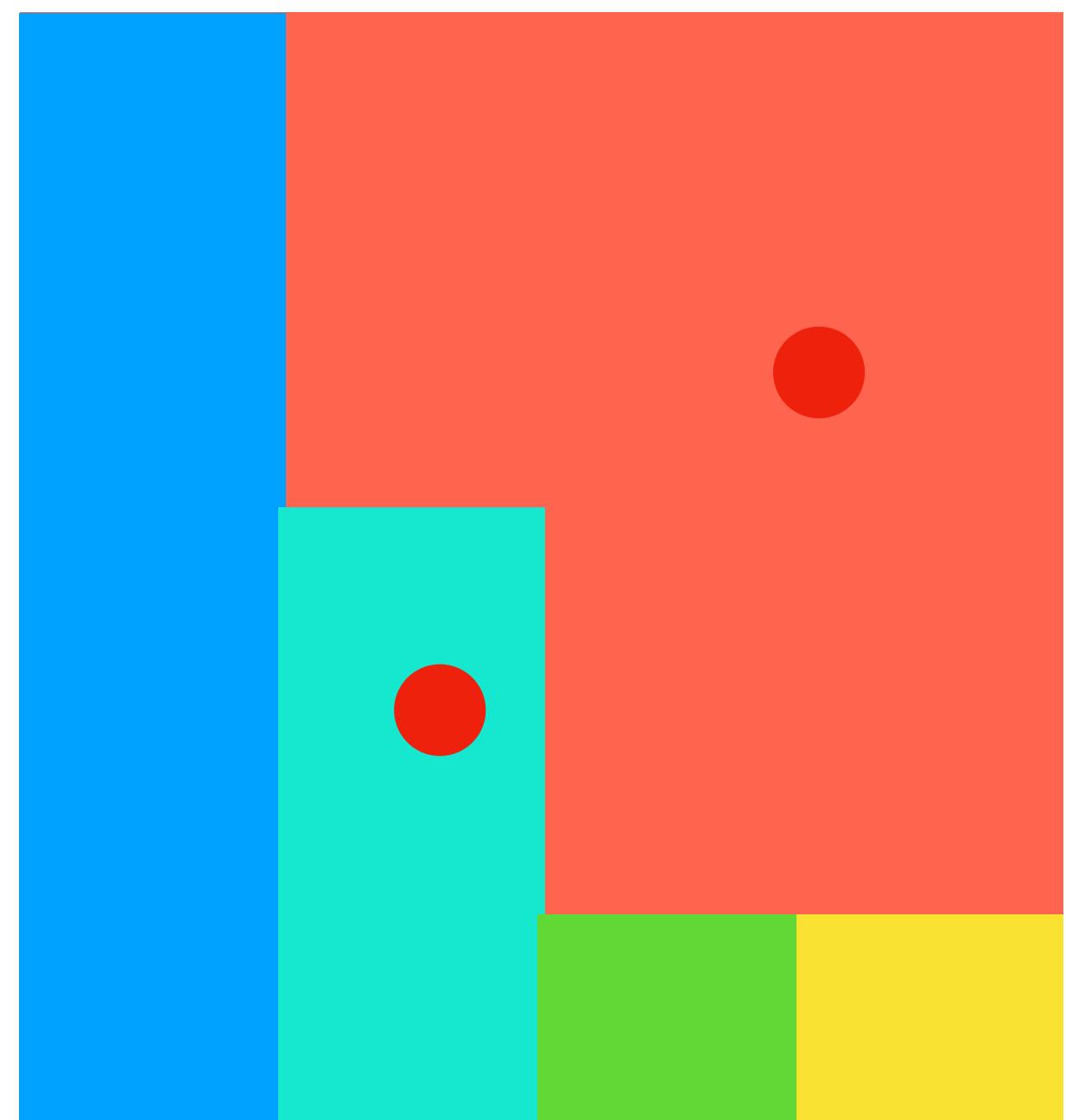
# Alias method

- goal: sampling from a discrete probability distribution
- lajolla's current implementation takes  $O(n)$  time to construct the CDF,  $O(\log(n))$  time to query
- alias method takes  $O(n)$  time to construct an “alias table”, and takes  $O(1)$  time to query



# Idea: sample from this rectangle

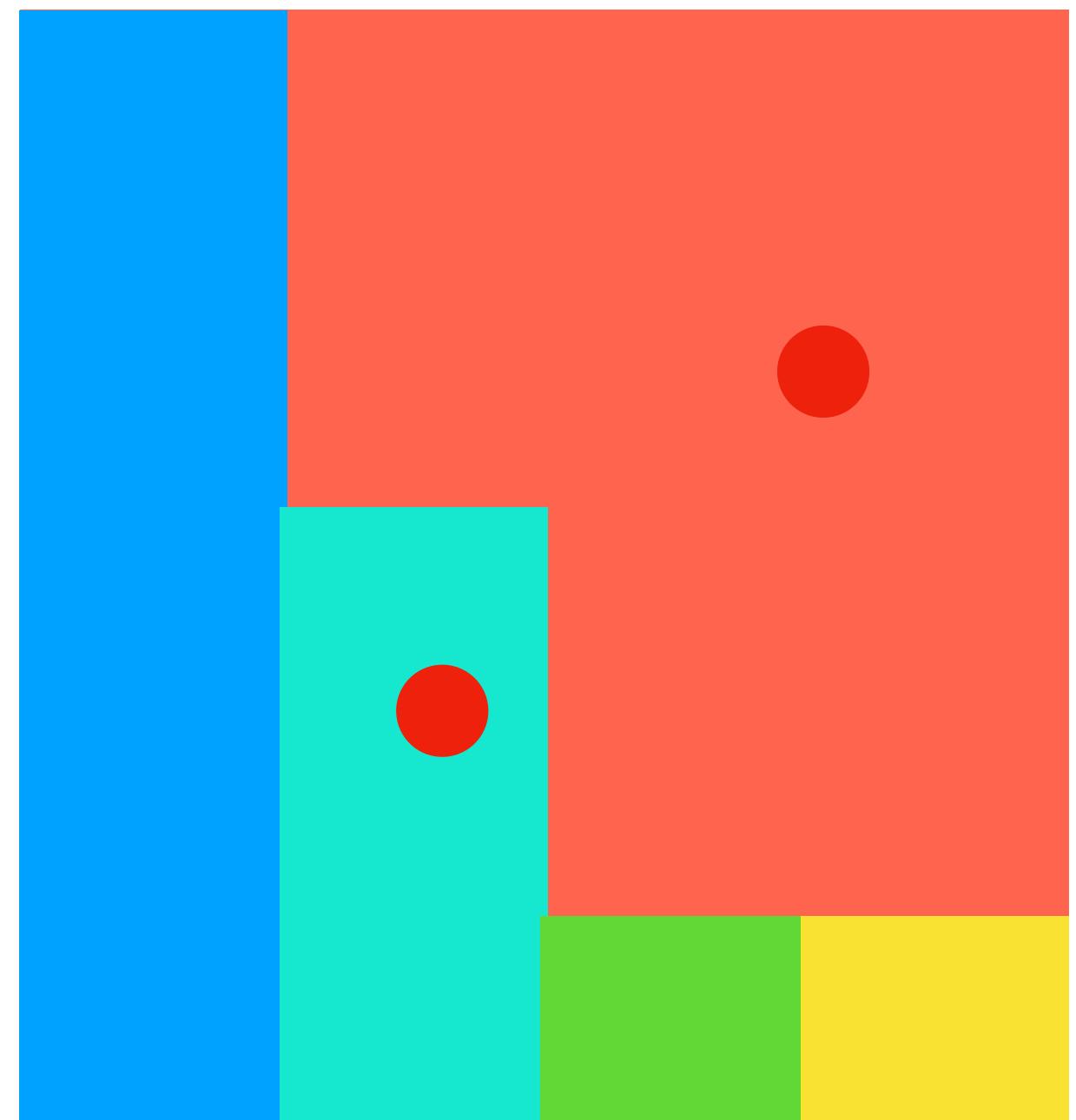
1. pick a point horizontally
2. accept/reject the sample by picking a point vertically
3. repeat until accept



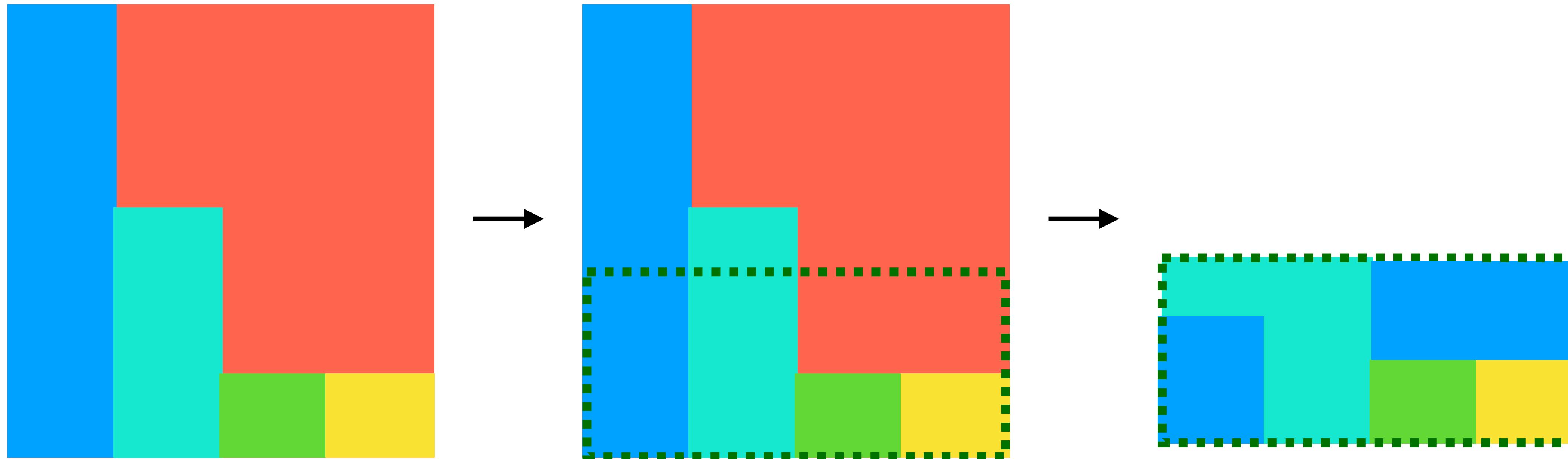
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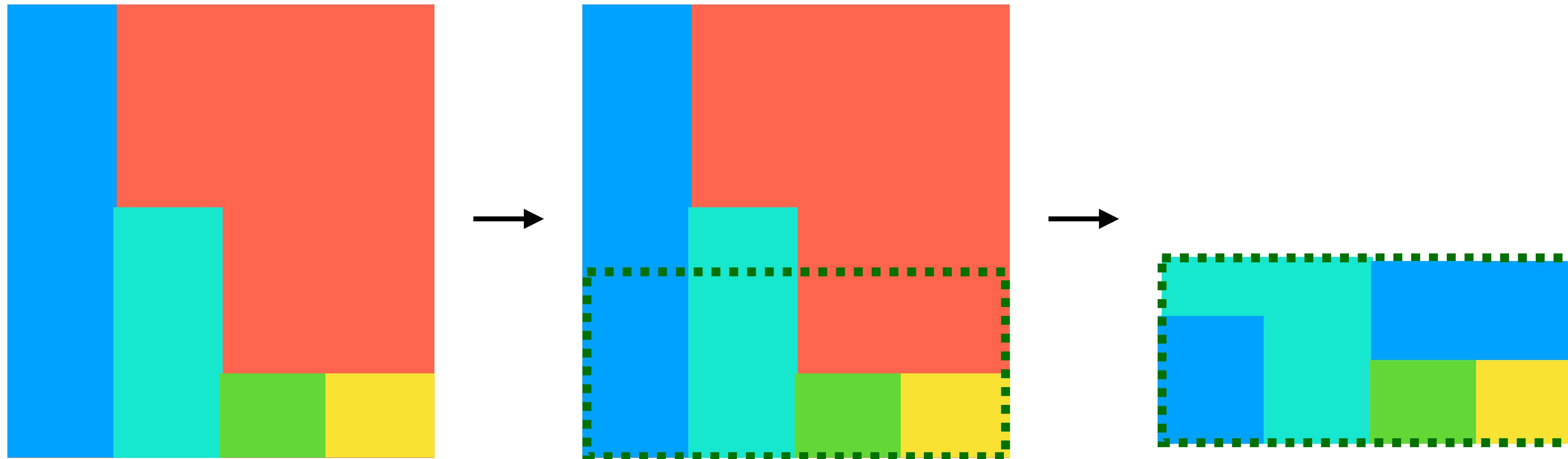
inefficient — can we improve this?



Idea: cut the rectangle and redistribute



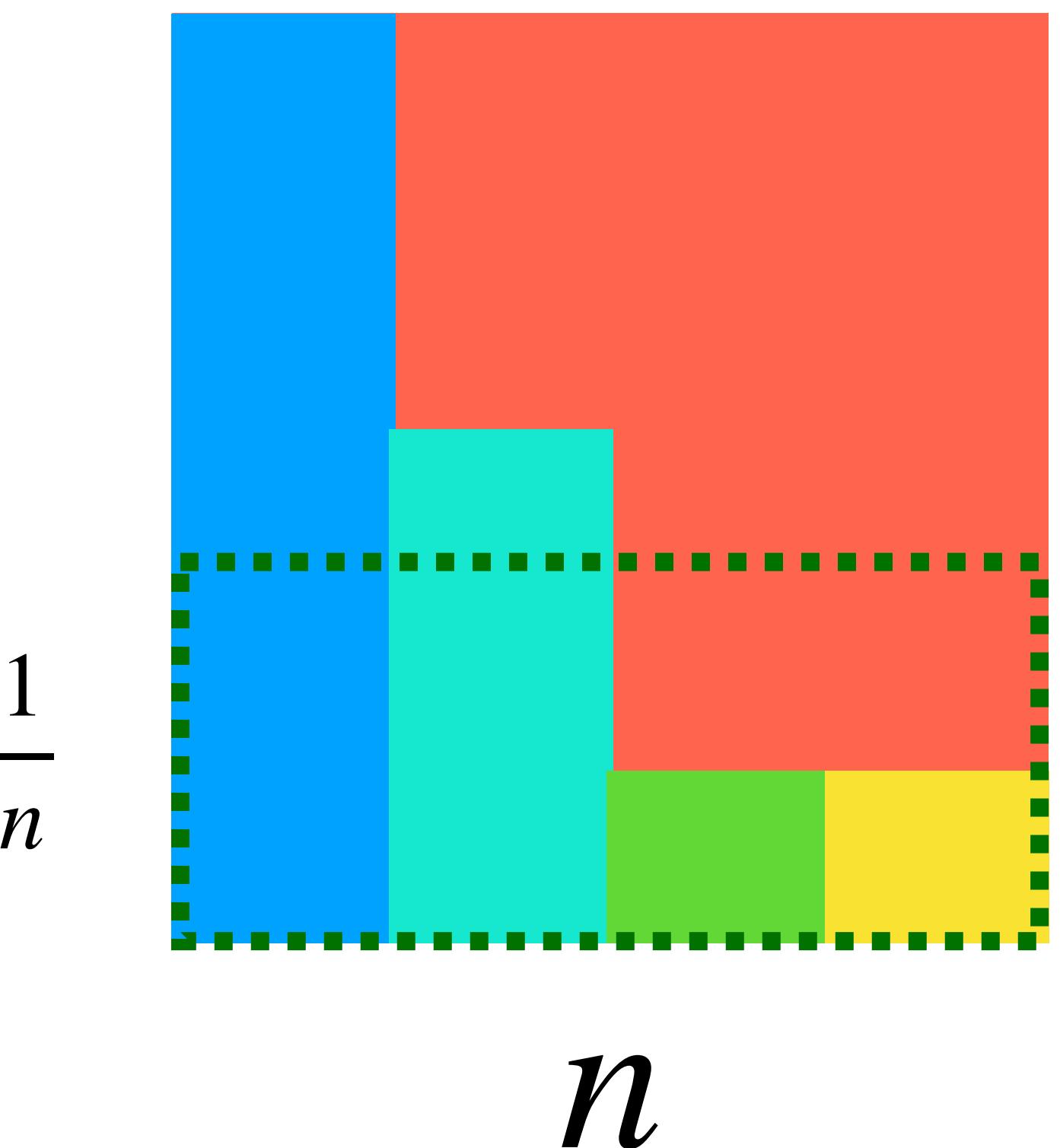
# Idea: cut the rectangle and redistribute



- how large should the cut be?
- how do we redistribute?

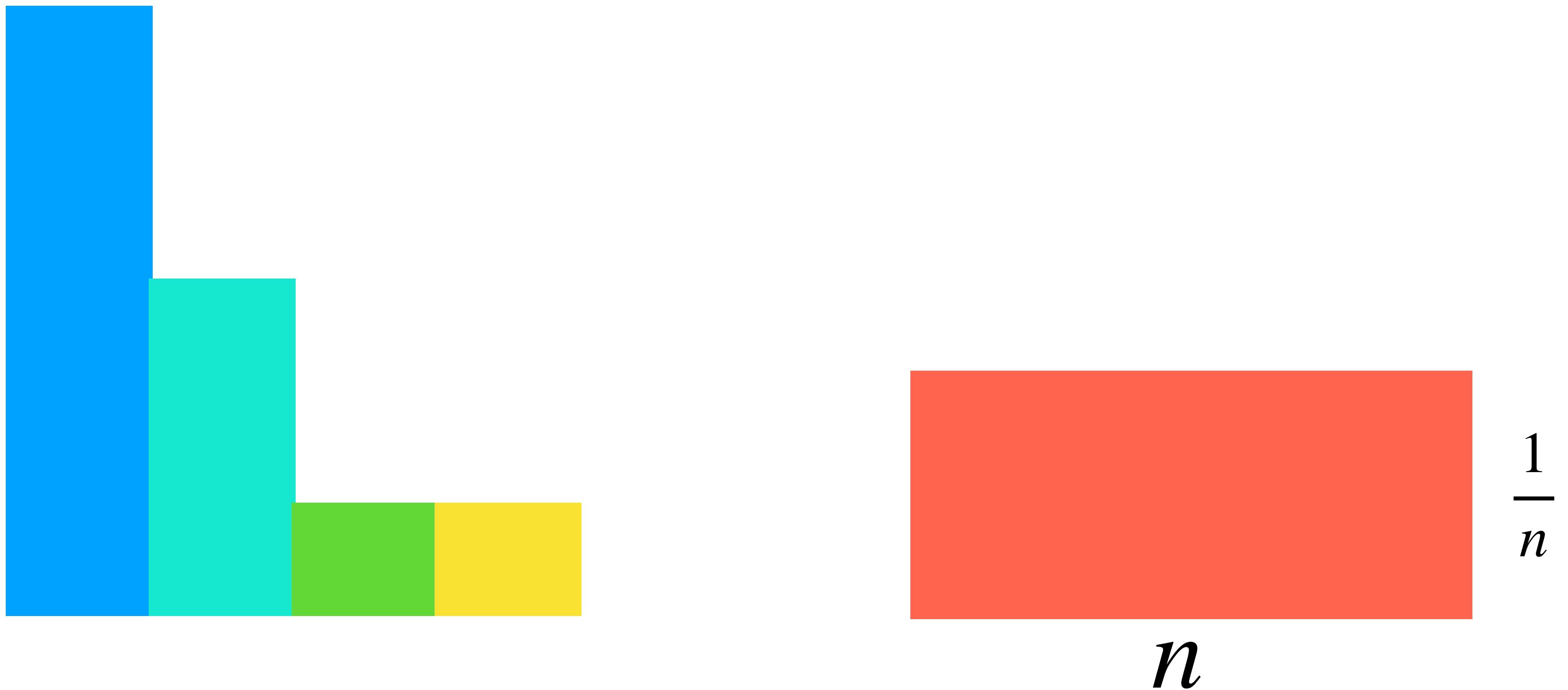
# How large should the cut be?

- we know the total probabilities sum to 1
- this means that to not waste space, if the width of the rectangle is  $n$ , the height must be  $\frac{1}{n}$



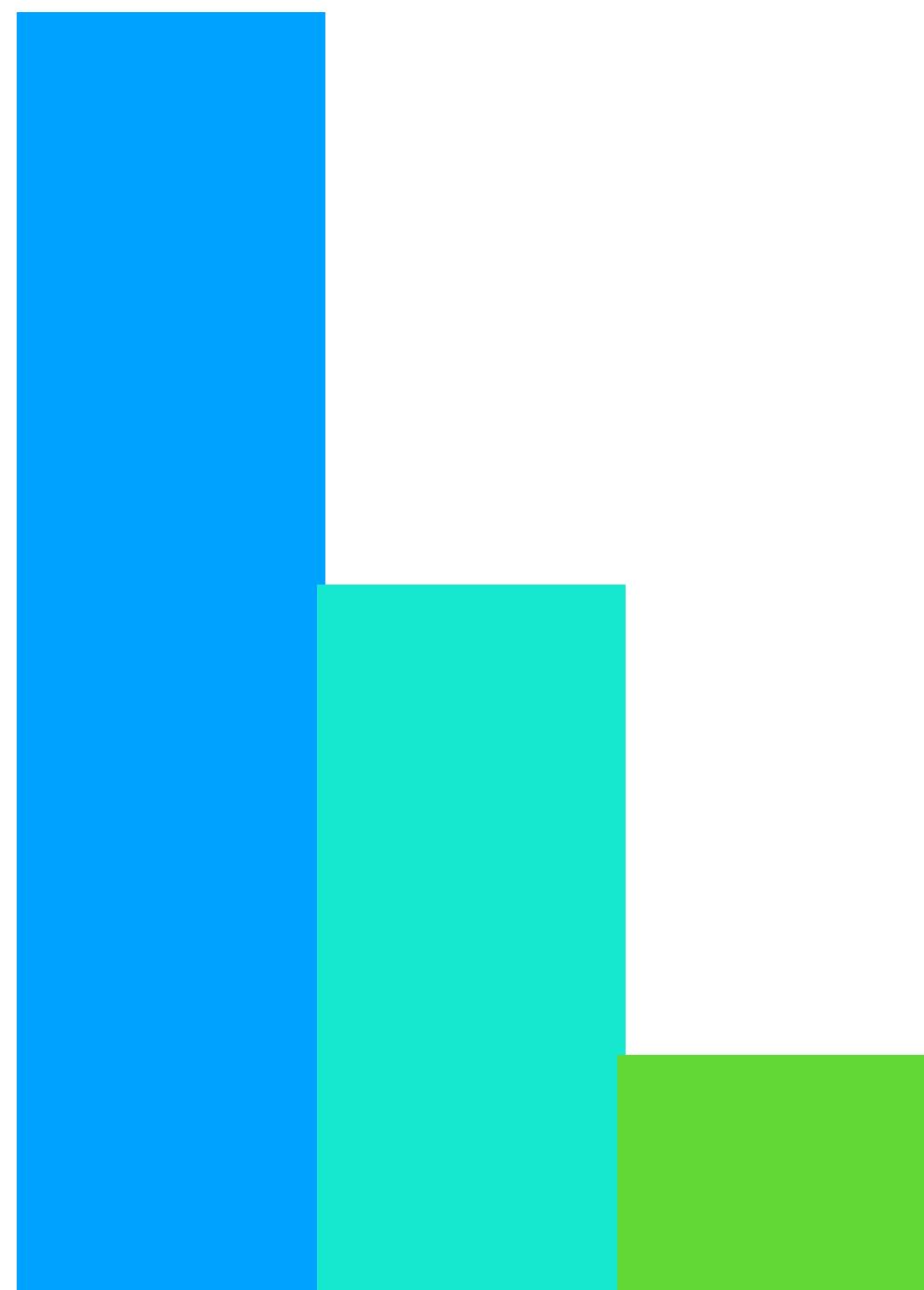
# How do we redistribute?

- start from an empty rectangle



# How do we redistribute?

- find a probability  $< \frac{1}{n}$ , put it on the rectangle



# How do we redistribute?

- find a probability  $\leq \frac{1}{n}$ , put it on the rectangle
- find a probability  $> \frac{1}{n}$ , cut it an put it on the rectangle

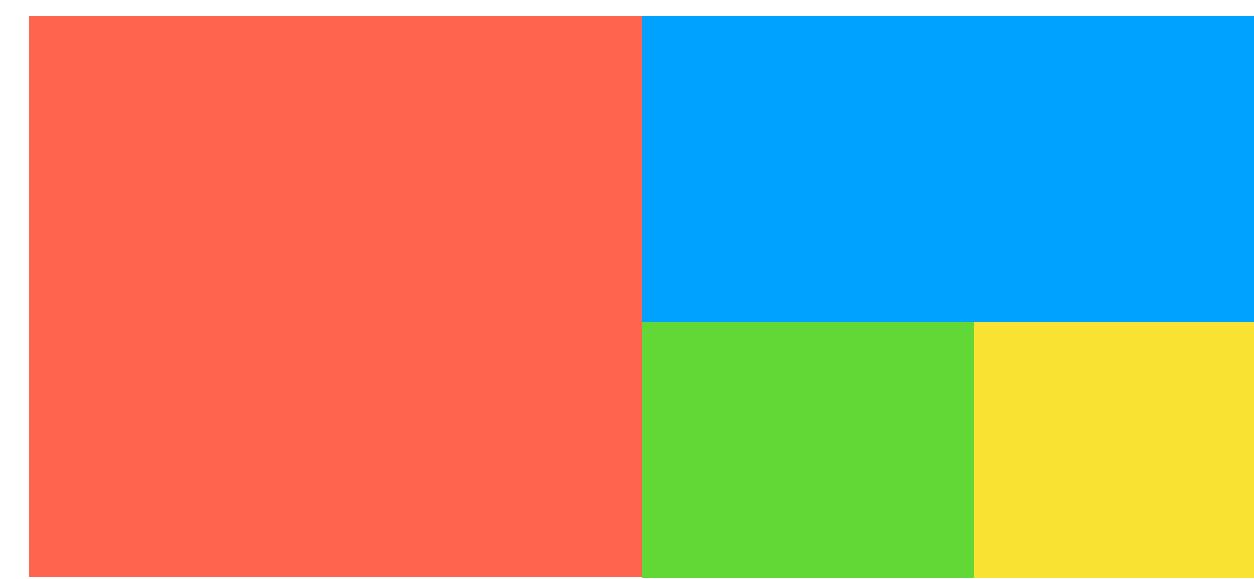
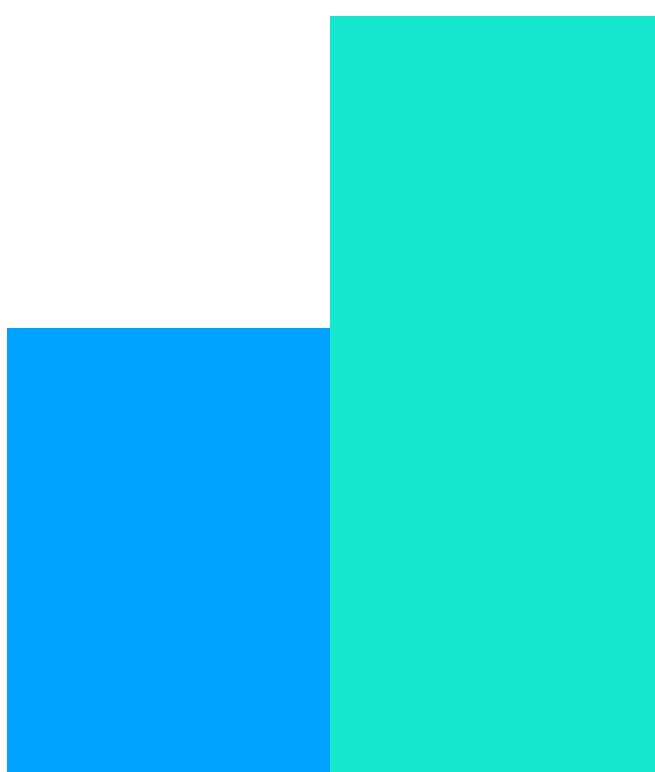


$n$

$$\frac{1}{n}$$

# How do we redistribute?

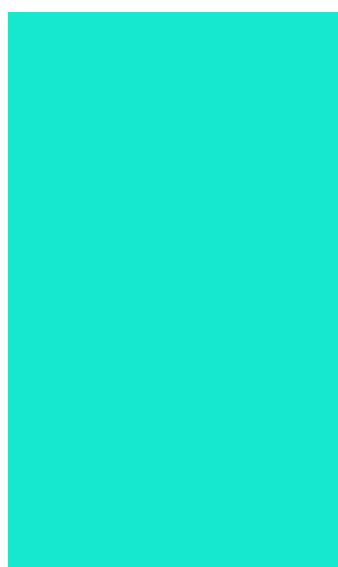
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- repeat



$$\frac{1}{n}$$

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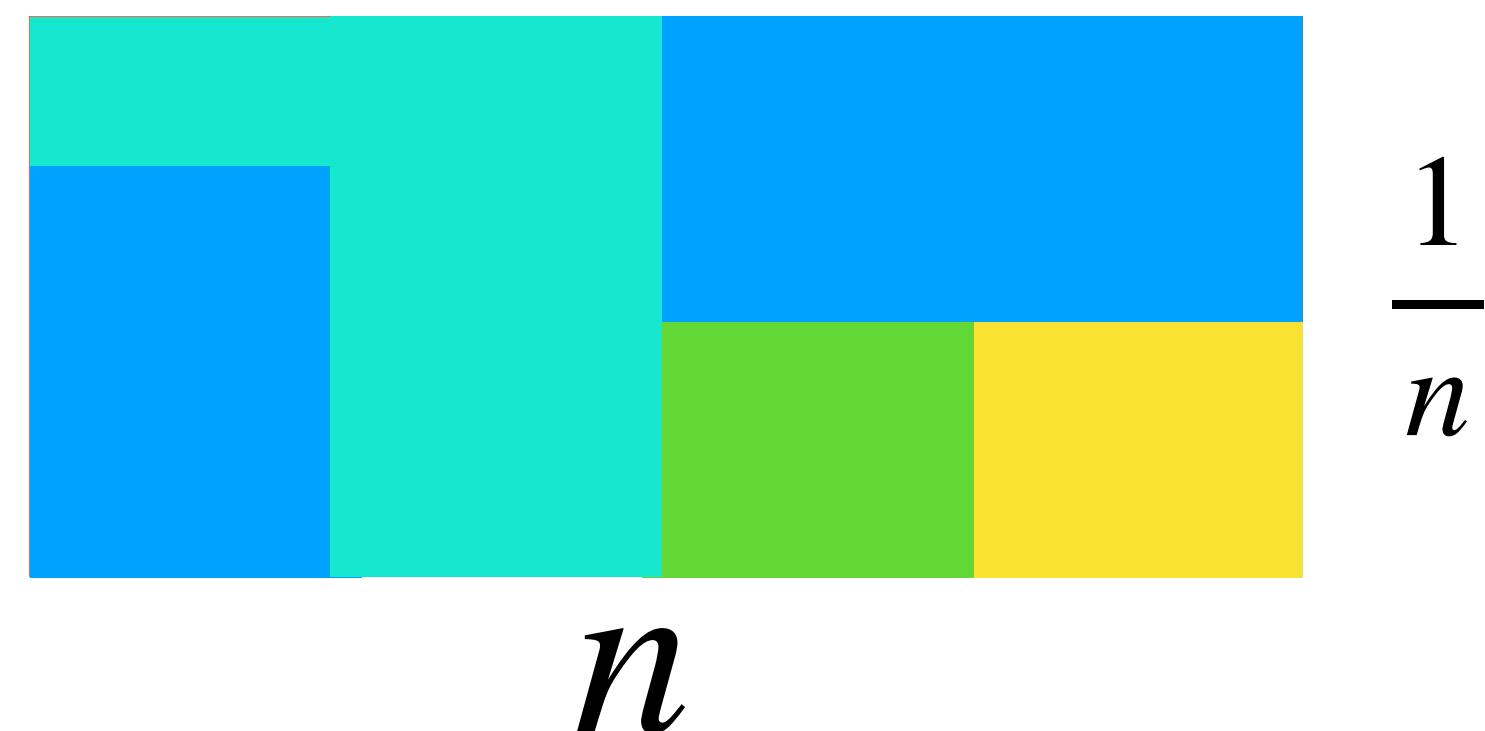


$n$

$$\frac{1}{n}$$

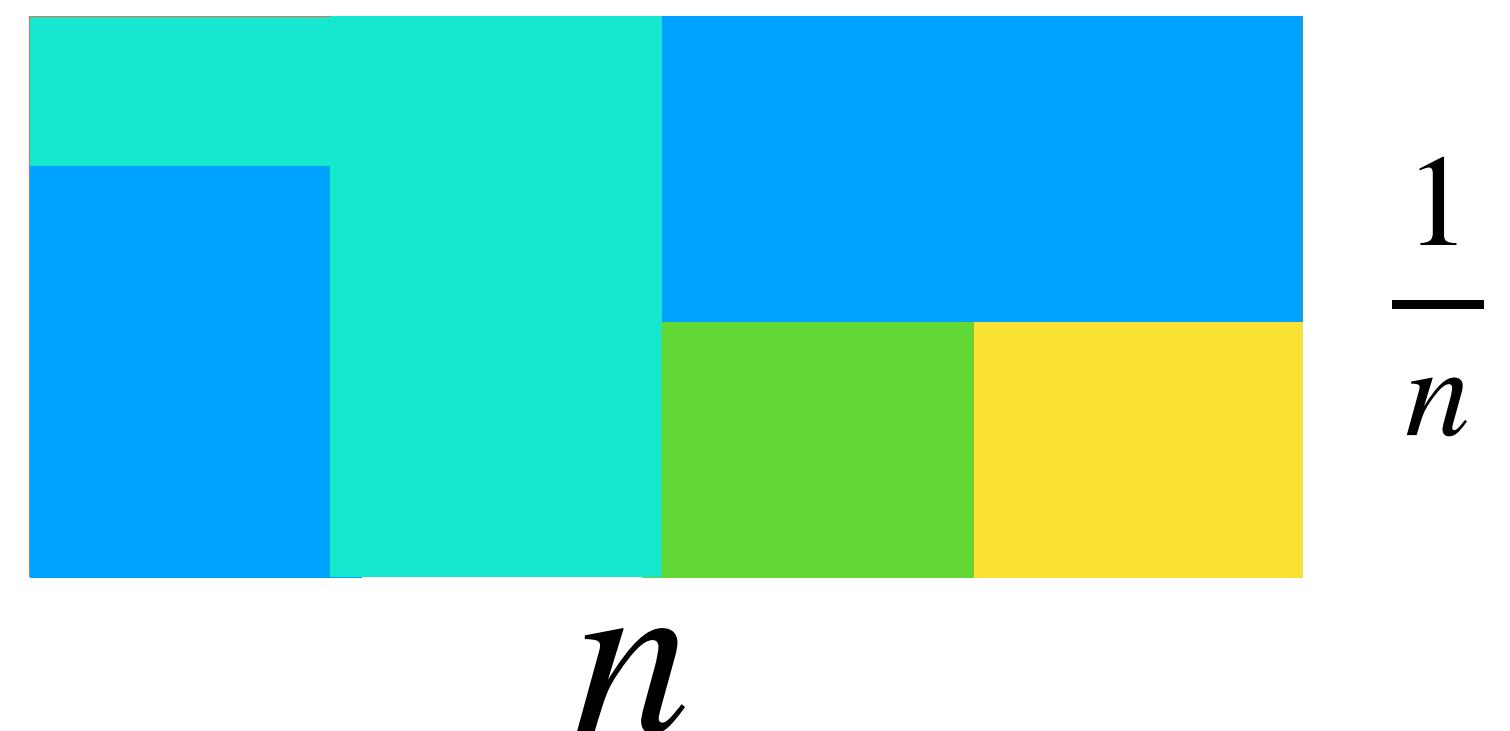
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- repeat



# How do we redistribute?

- find a probability  $\leq \frac{1}{n}$ , put it on the rectangle
- find a probability  $> \frac{1}{n}$ , cut it an put it on the rectangle
- repeat
- can be done in  $O(n)$  time if we keep track of which entry is  $\leq \frac{1}{n}$  and which is not



# Alias method: pros and cons

- pro(s):
- con(s):



# Alias method: pros and cons

- pro(s): fast
- con(s): stratification



# Russian roulette debiasing

- goal: turn any consistent estimator into an unbiased estimator

$$\lim_{i \rightarrow \infty} A_i = A$$

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idea: rewrite  $A$  as a **telescoping sum**

$$A = A_0 + (A_1 - A_0) + (A_2 - A_1) + \cdots = A_0 + \sum_{i=1}^{\infty} (A_i - A_{i-1})$$

# Russian roulette debiasing

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idea: rewrite  $A$  as a **telescoping sum**

$$A = A_0 + (A_1 - A_0) + (A_2 - A_1) + \cdots = A_0 + \sum_{i=1}^{\infty} (A_i - A_{i-1})$$

sample  $N$  with probability  $p(N)$ , estimate  $A$  as  $\frac{A_0 + \sum_{i=1}^N (A_i - A_{i-1})}{p(N)} = \frac{A_N}{p(N)}$

# Russian roulette debiasing

- special case: a non-linear transformation of an integral

$$g \left( \int f \right)$$

# Russian roulette debiasing

- special case: a non-linear transformation of an integral

can be estimated using Taylor expansion at  $F = \int f$

$$g \left( \int f \right)$$

$$g(x) = g(F) + g'(F)(x - F) + g''(F) \frac{(x - F)^2}{2!} + \dots$$

# Russian roulette debiasing: applications

- unbiased estimation for non-linear transformation of integrals

$$\frac{1}{\int f} \exp \left( \int f \right) \quad \left| \int f \right| \quad \max \left( \int f, 0 \right)$$

# Russian roulette debiasing: applications

read this paper for more detail!

Unbiased and consistent rendering using biased estimators

ZACKARY MISSO, Dartmouth College, USA

BENEDIKT BITTERLI, Dartmouth College, USA and NVIDIA, USA

ILIYAN GEORGIEV, Autodesk, United Kingdom

WOJCIECH JAROSZ, Dartmouth College, USA

# Tagged pointer

- in lajolla, we used `std::variant` for dynamic polymorphism

```
struct Foo {  
    int type;  
    union {  
        TypeA a;  
        TypeB b;  
        ...  
    };  
};
```

this is somewhat memory consuming —  
can we save some memory?

# Tagged pointer

- hack: on x86 machines, only 48-bits out of 64-bits of a pointer are used
- use the 16-bits to store the type information!

```
struct TaggedPointer {  
    uint32_t tag() const { return ((bits & tag_mask) >> tag_shift); }  
    void* ptr() const { return reinterpret_cast<void*>(bits & ptr_mask); }  
  
    uint64_t bits;  
    static constexpr int tag_shift = 48;  
    static constexpr int tag_bits = 64 - tagShift;  
    static constexpr uint64_t tag_mask = ((1ull << tag_bits) - 1) << tag_shift;  
    static constexpr uint64_t ptr_mask = ~tag_mask;  
};
```

# Numerically stable cross product

```
// a * b - c * d
difference_of_products(a, b, c, d) {
    cd = c * d
    err = fma(-c, d, cd) // -c*d + cd
    dop = fma(a, b, -cd) // a*b - cd
    return dop + err
}
```

# Numerically stable quadratic solve

$$ax^2 + bx + c = 0$$

$$x_1 = \frac{-b - \sqrt{b^2 - 4ac}}{2a}$$

$$x_2 = \frac{2c}{-b - \sqrt{b^2 - 4ac}} \quad b > 0$$

$$x_1 = \frac{2c}{-b + \sqrt{b^2 - 4ac}}$$

$$x_2 = \frac{-b + \sqrt{b^2 - 4ac}}{2a} \quad b \leq 0$$

# Cosine-weighted hemisphere sampling in 4 lines without building a frame

```
vec3 lambertNoTangent(in vec3 normal, in vec2 uv) {  
    float theta = 6.283185 * uv.x;  
    uv.y = 2.0 * uv.y - 1.0;  
    vec3 spherePoint = vec3(sqrt(1.0 - uv.y * uv.y) * vec2(cos(theta), sin(theta)), uv.y);  
    return normalize(normal + spherePoint);  
}
```

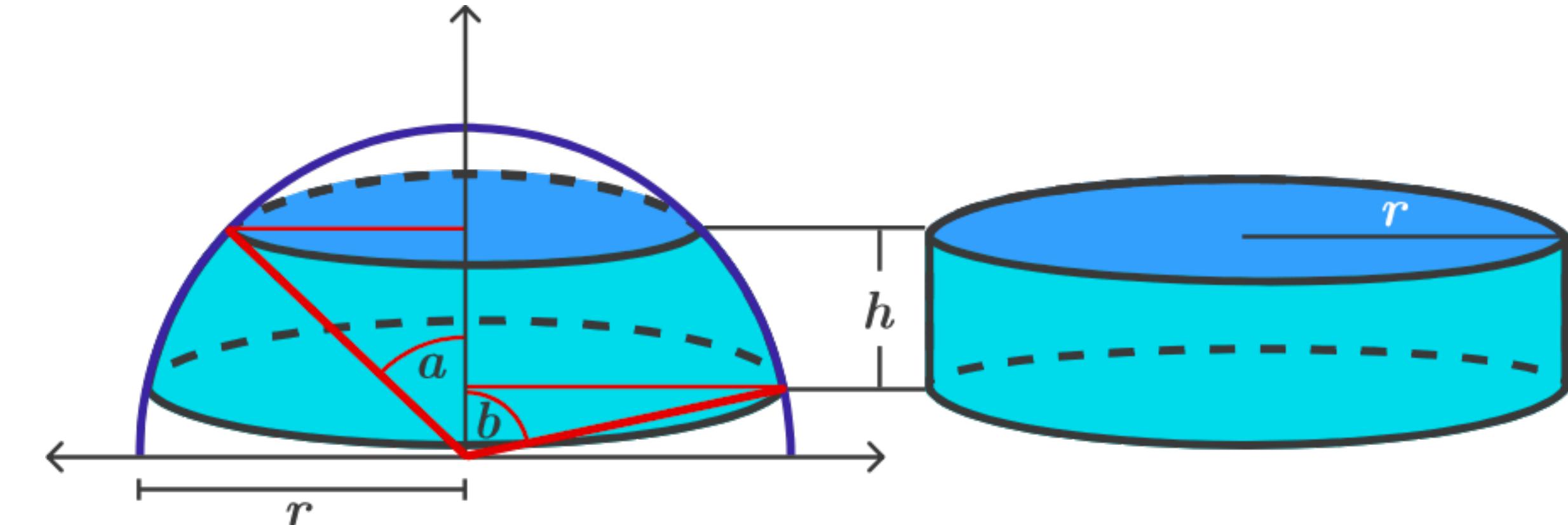
credit: Edd Biddulph

<https://web.archive.org/web/20170610002747/http://amietia.com/lambertnotangent.html>

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```

Archimedes' Hat-Box Theorem:  
the area of a spherical section  
= the area of a cylinder with the same radius



<https://brilliant.org/wiki/surface-area-sphere/#archimedes-hat-box-theorem>

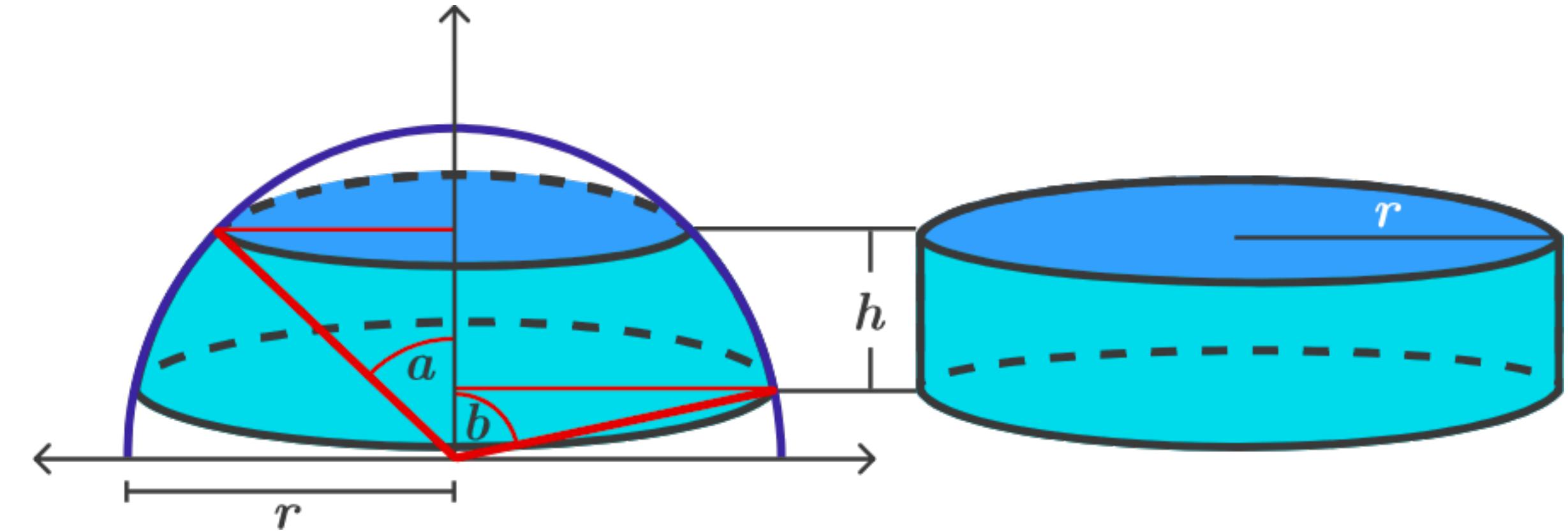
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Archimedes' Hat-Box Theorem:  
the area of a spherical section  
= the area of a cylinder with the same radius



<https://brilliant.org/wiki/surface-area-sphere/#archimedes-hat-box-theorem>

uniformly sampling a sphere  
= uniformly sampling on concentric rings of a disk

credit: Edd Biddulph

<https://web.archive.org/web/20170610002747/http://amietia.com/lambertnotangent.html>

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    return normalize(normal + spherePoint);  
}
```

algorithm:

- sampling on a unit sphere uniformly
- project onto a unit disk
- scale their distance to the origin to make it a uniform sampling of a disk
- project back onto the hemisphere (Malley's method)

credit: Edd Biddulph

<https://web.archive.org/web/20170610002747/http://amietia.com/lambertnotangent.html>

# What next?