

Analytical Motion Blur Rasterization with Compression

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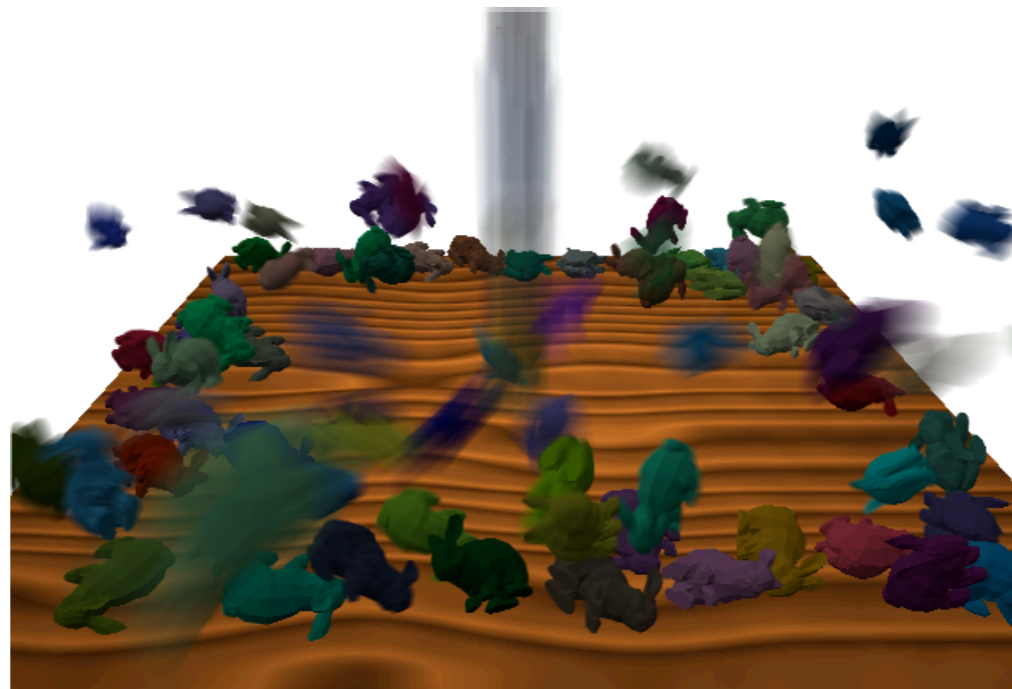
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Motivation



Motivation

- Human visual system designed to detect motion
- This fails when presented too few images, or too much motion per image
 - motion gets jumpy
- Motion blur aids the motion detection of the visual system



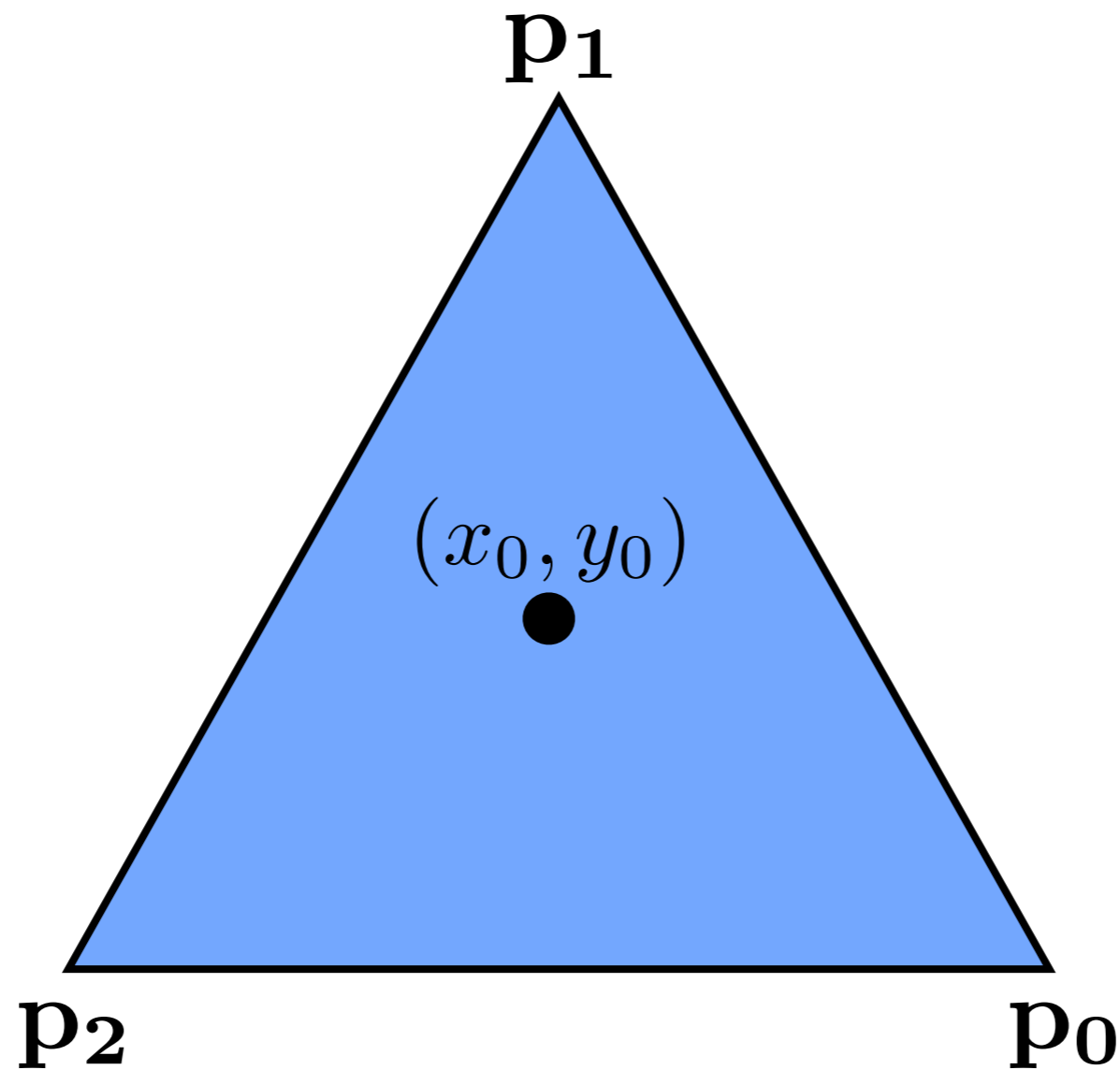
Talk outline

- Edge Equations and exact exposure intervals
 - analytic inside-test
 - visibility management
- Computing the time integral
 - opaque/translucent resolve
- Compression

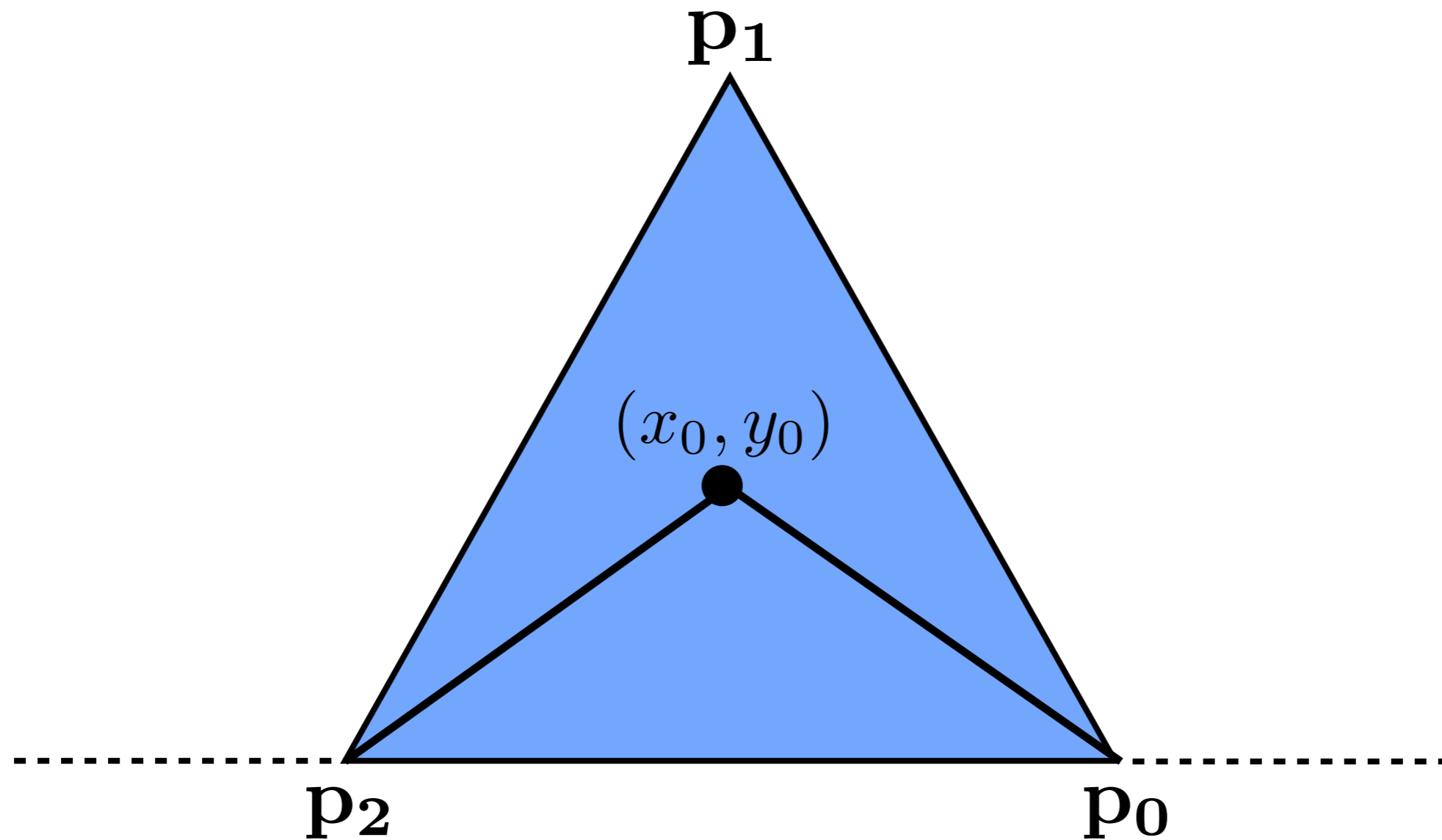
Previous work

- Accumulation buffer [HA90][WGER05][DWS88]
 - expensive
 - strobing artifacts
- Stochastic [CCC87] [AMMH07][McGuire et al. HPG2010]
 - correct solution but slow convergence
- Single sample analytical visibility [KB83]
Extended with stochastic shading by Sung et al. [SPW02]
 - details missing for visibility computations

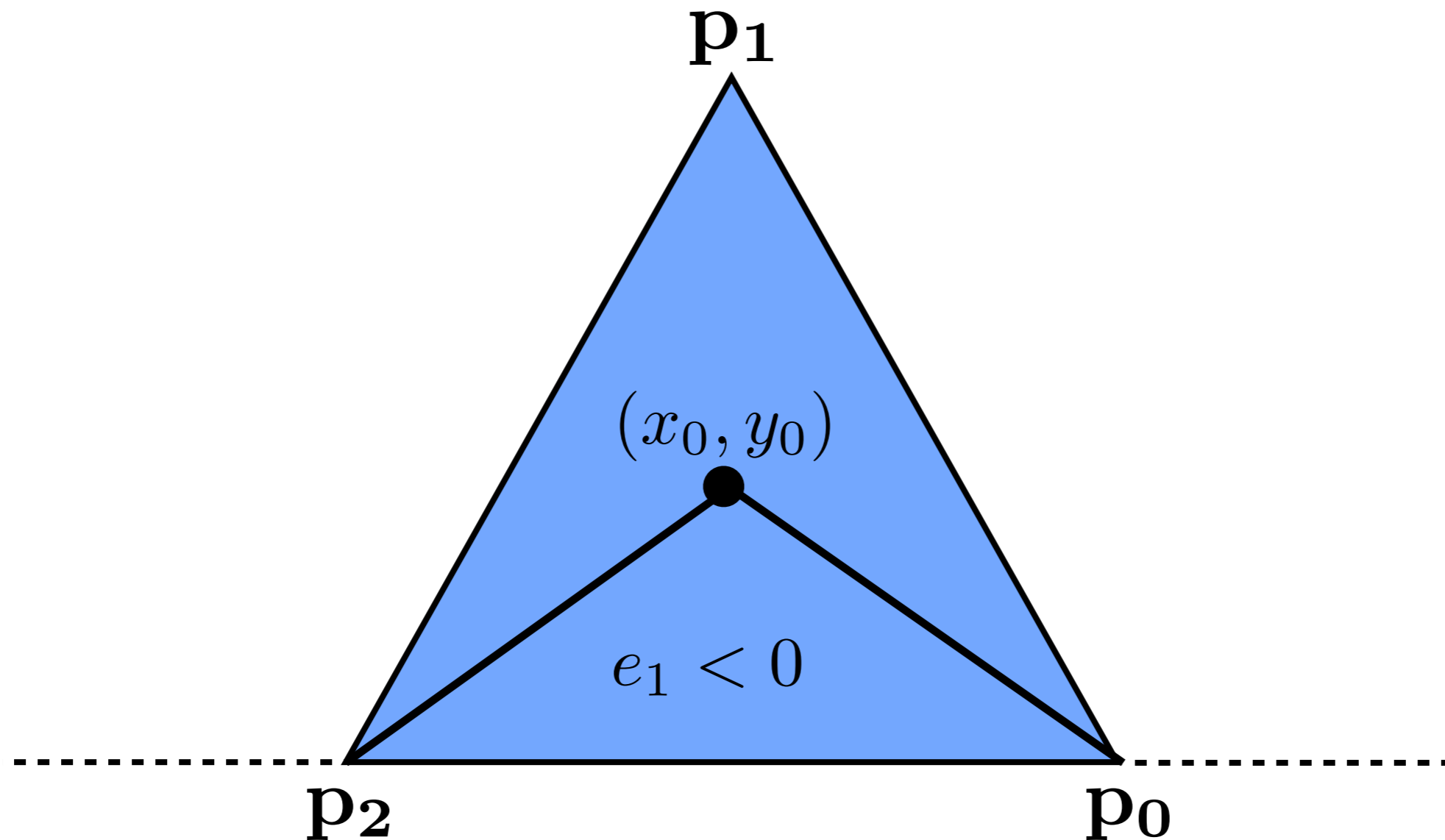
Edge Equation primer for static triangle



Edge Equation primer for static triangle



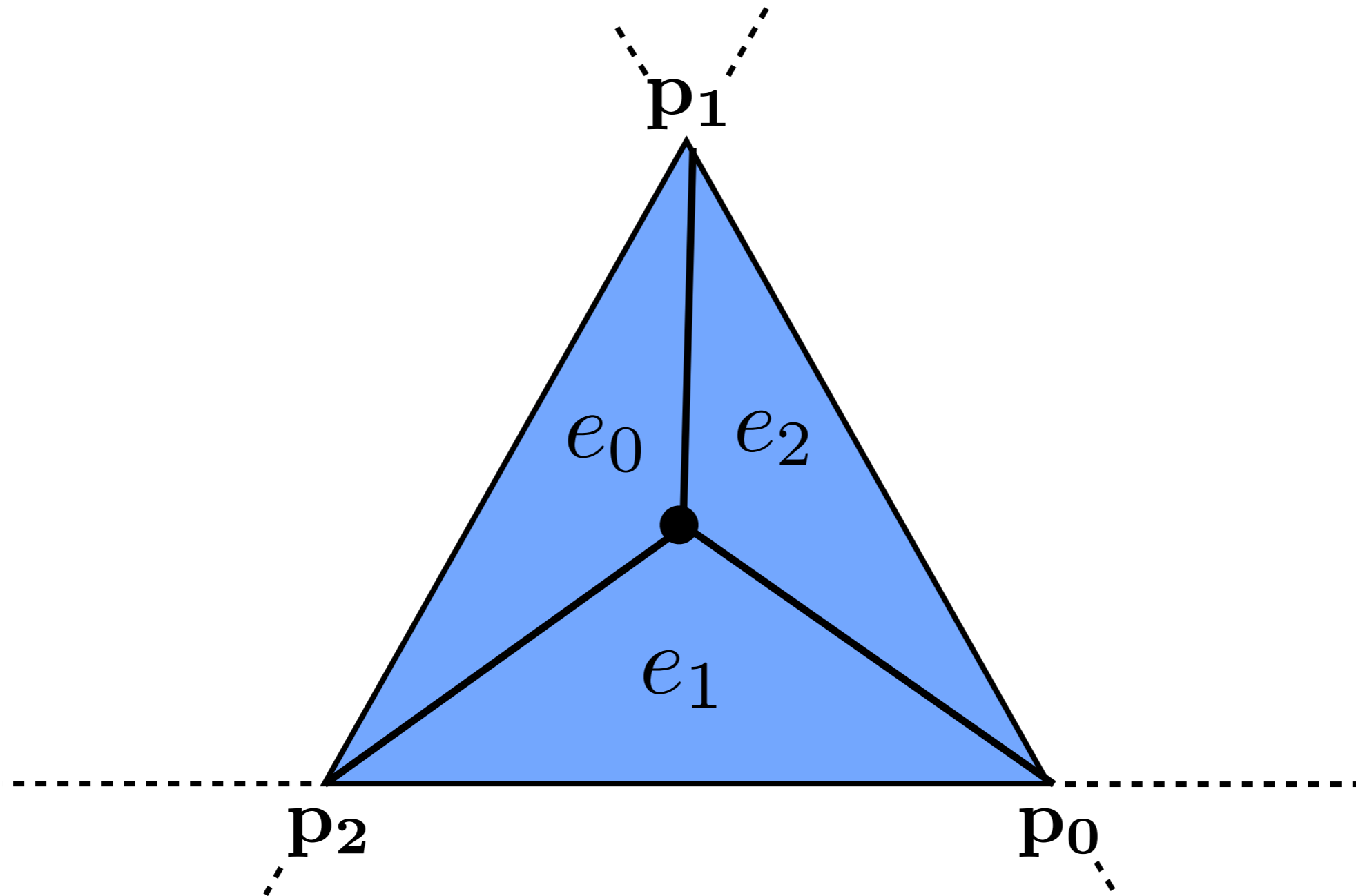
Edge Equation primer for static triangle



3D homogeneous space:

$$e_1 = (\mathbf{p}_2 \times \mathbf{p}_0) \cdot (x_0, y_0, 1)$$

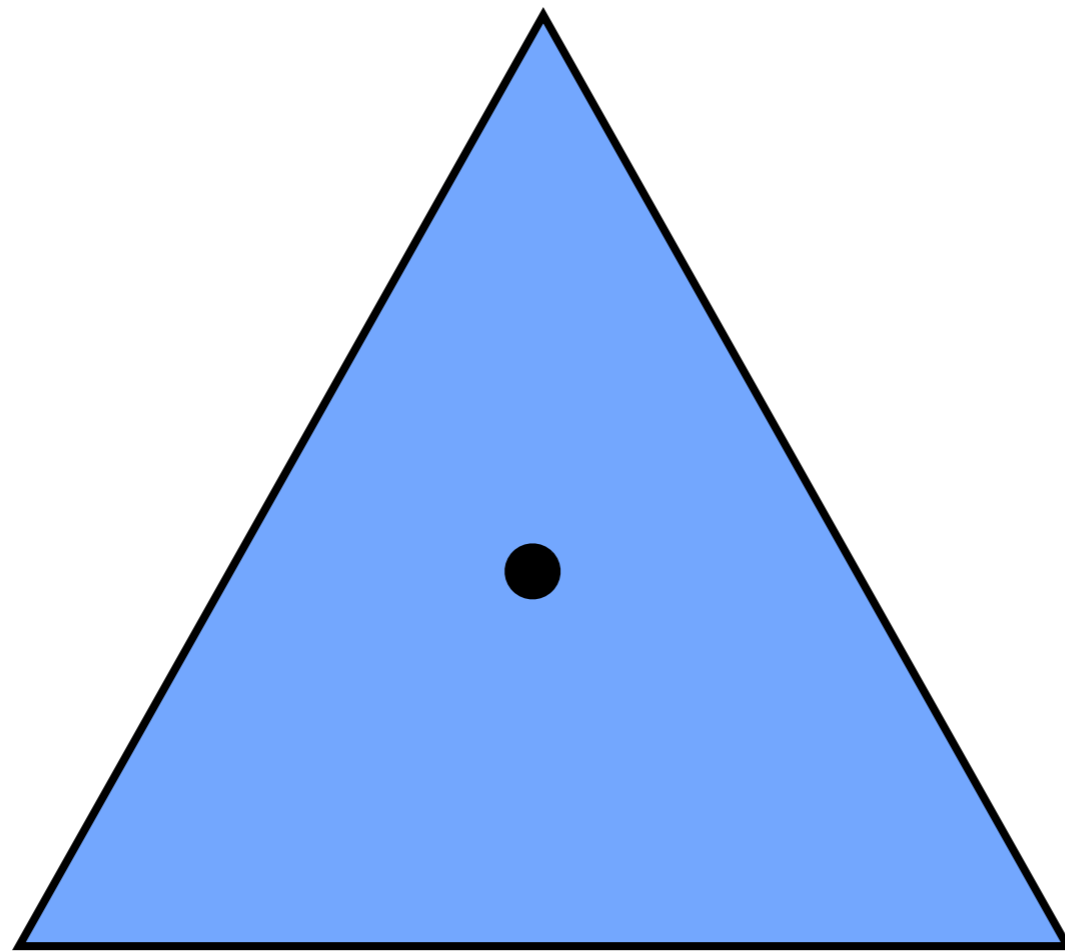
Edge Equation primer for static triangle



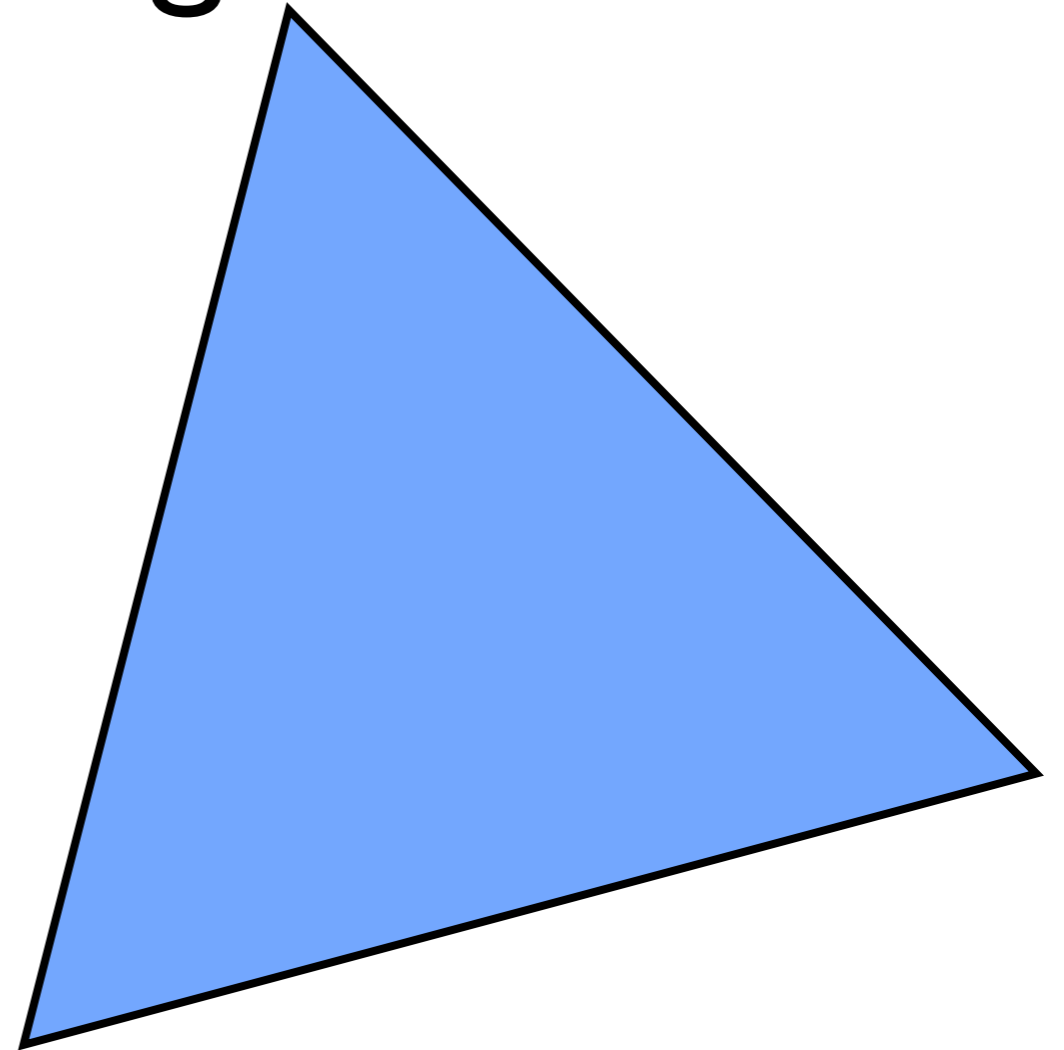
3D homogeneous space:

$$e_1 = (\mathbf{p}_2 \times \mathbf{p}_0) \cdot (x_0, y_0, 1)$$

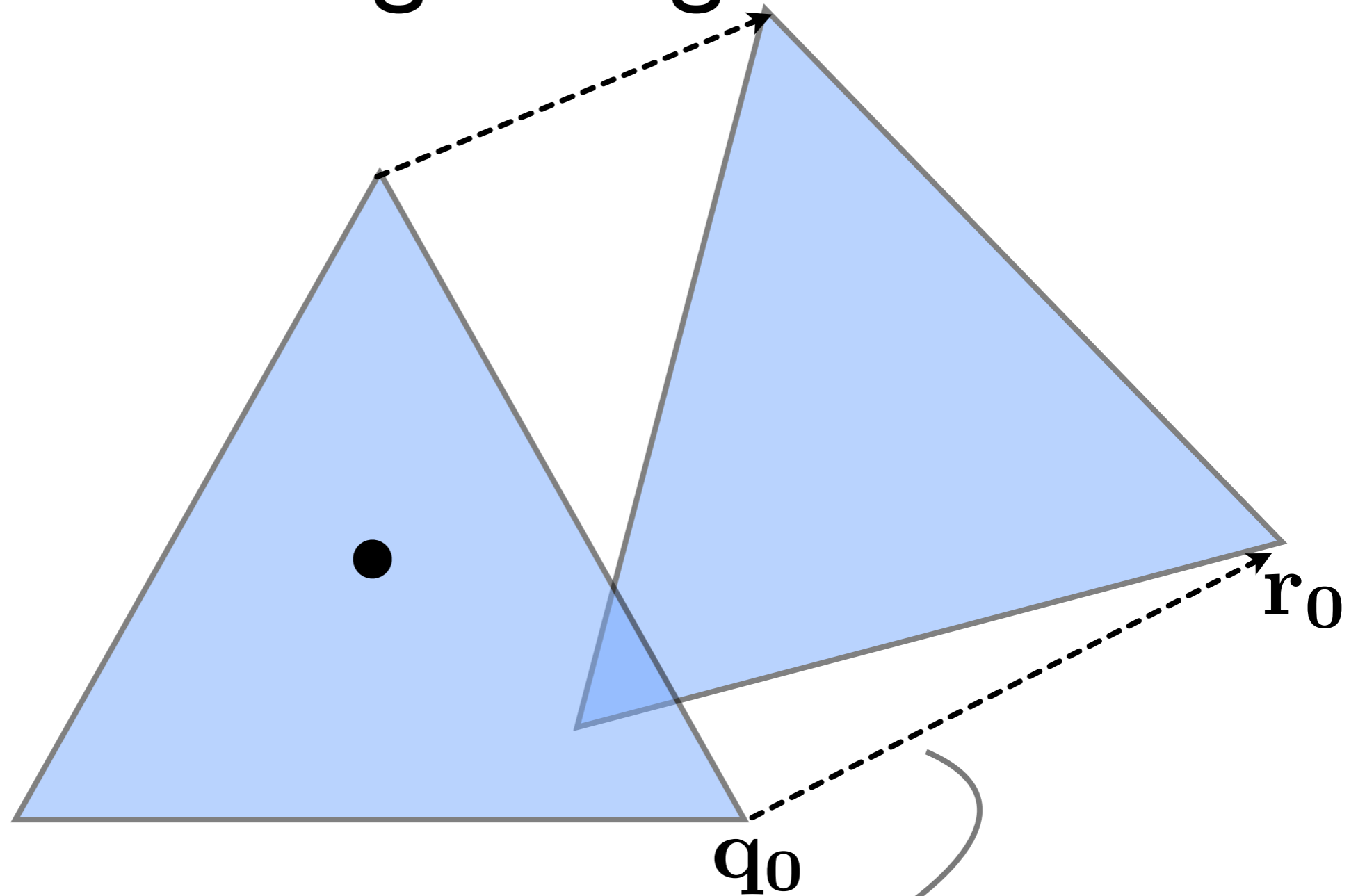
Edge Equation primer for moving triangle



Edge Equation primer for moving triangle

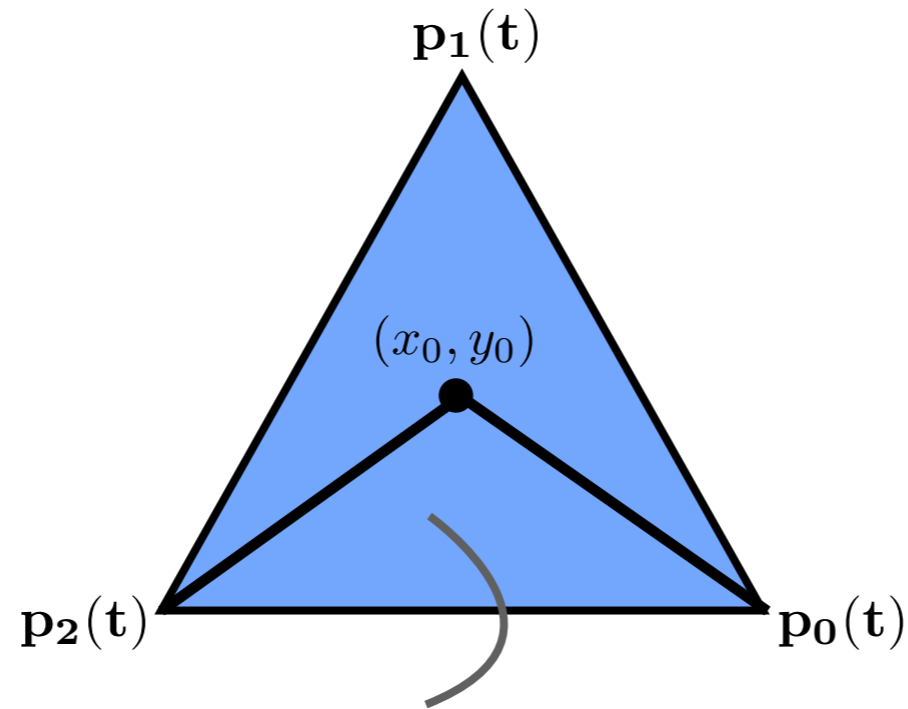


Edge Equation primer for moving triangle



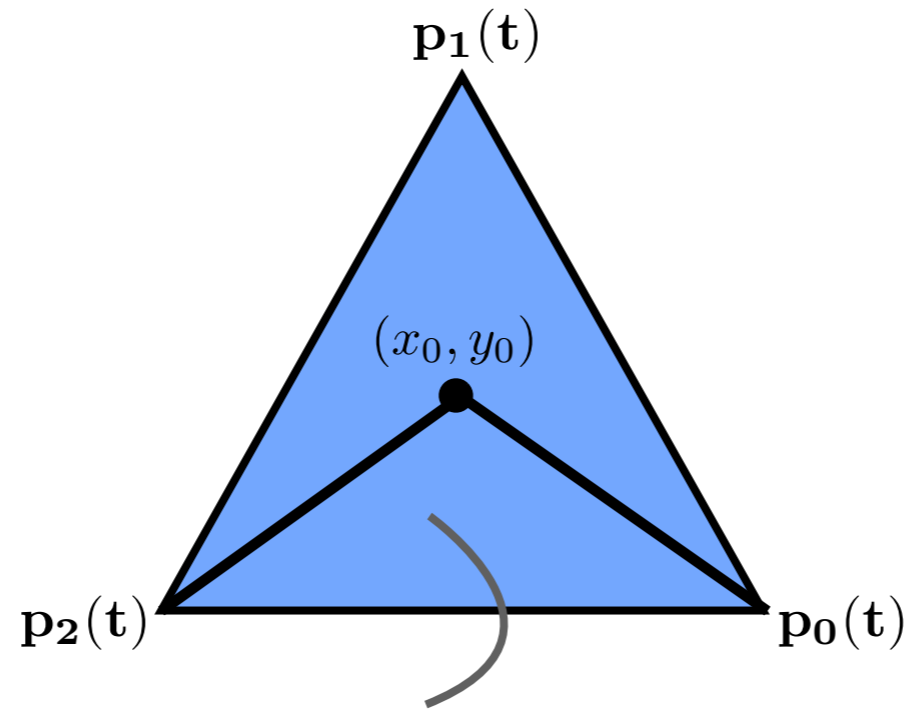
$$p_0(t) = (1 - t)q_0 + tr_0$$
$$t \in [0, 1]$$

Edge Equations



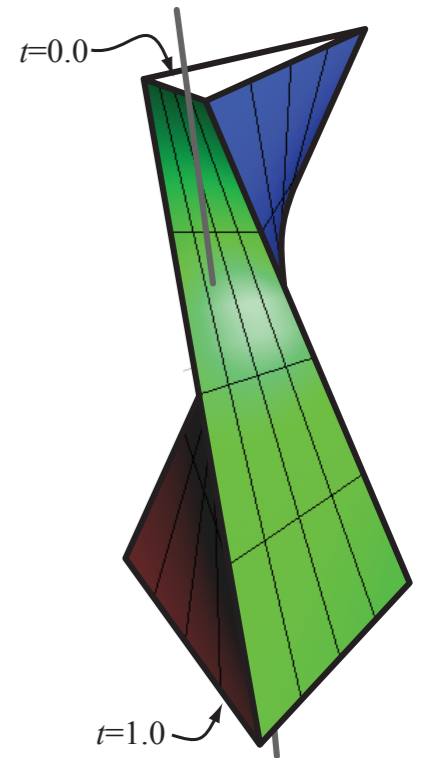
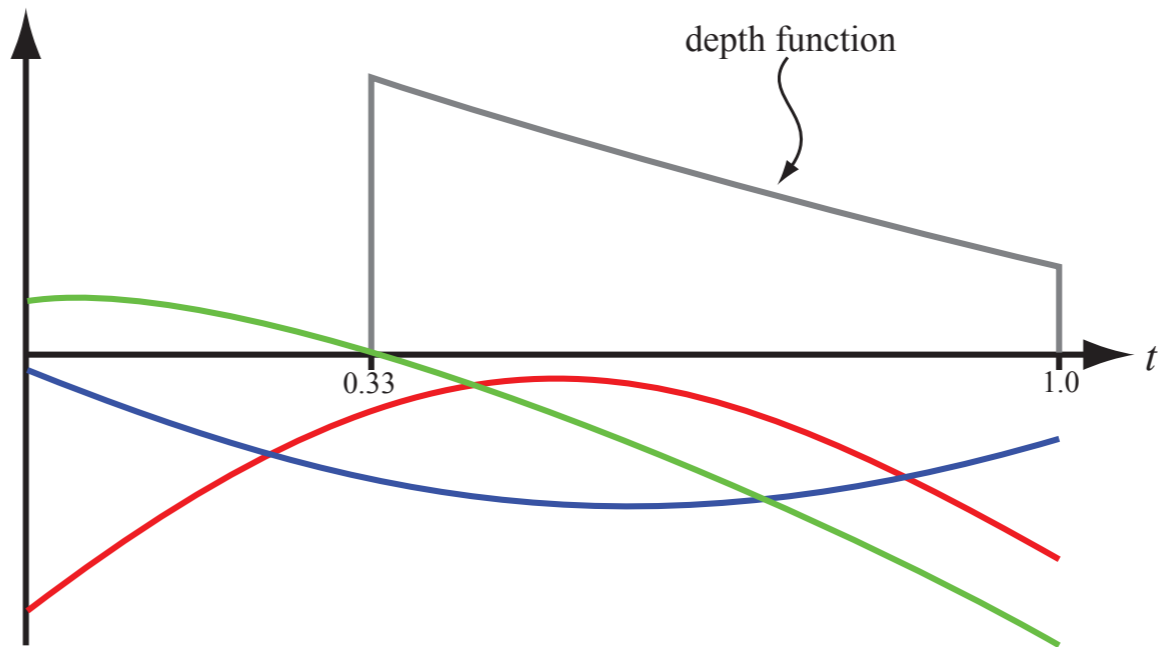
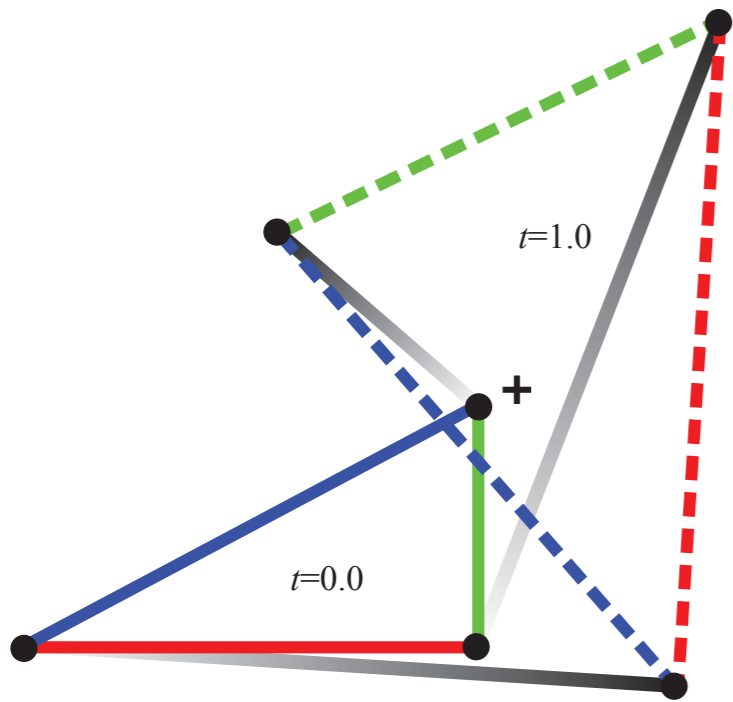
$$\begin{aligned} e_1(t) &= (\mathbf{p}_2(\mathbf{t}) \times \mathbf{p}_0(\mathbf{t})) \cdot (x_0, y_0, 1) \\ &= (((1-t)\mathbf{q}_2 + t\mathbf{r}_2) \times ((1-t)\mathbf{q}_0 + t\mathbf{r}_0)) \cdot (x_0, y_0, 1) \\ &= \boxed{(\mathbf{f}t^2 + \mathbf{g}t + \mathbf{h}) \cdot (x_0, y_0, 1)} \end{aligned}$$

Edge Equations



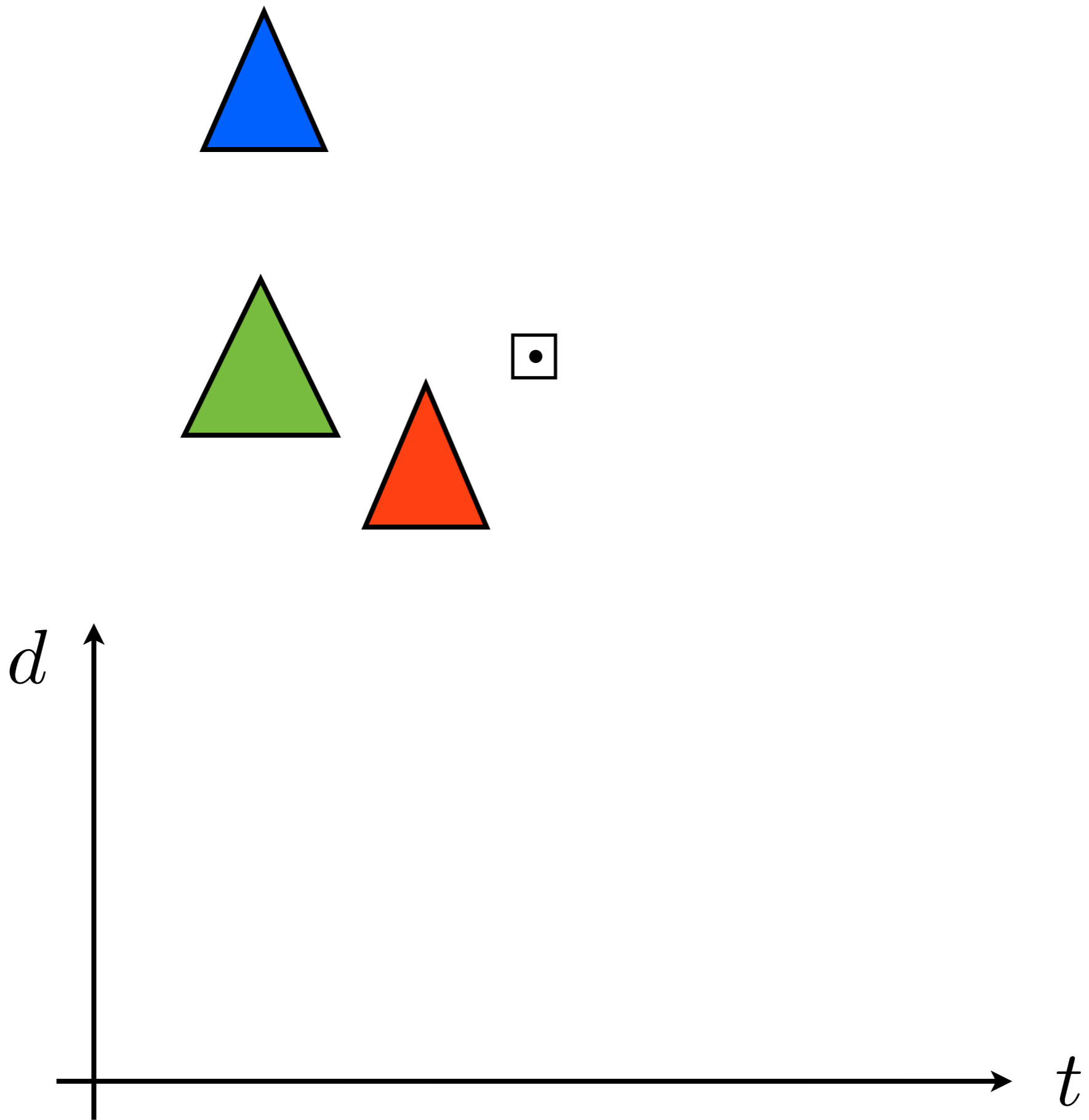
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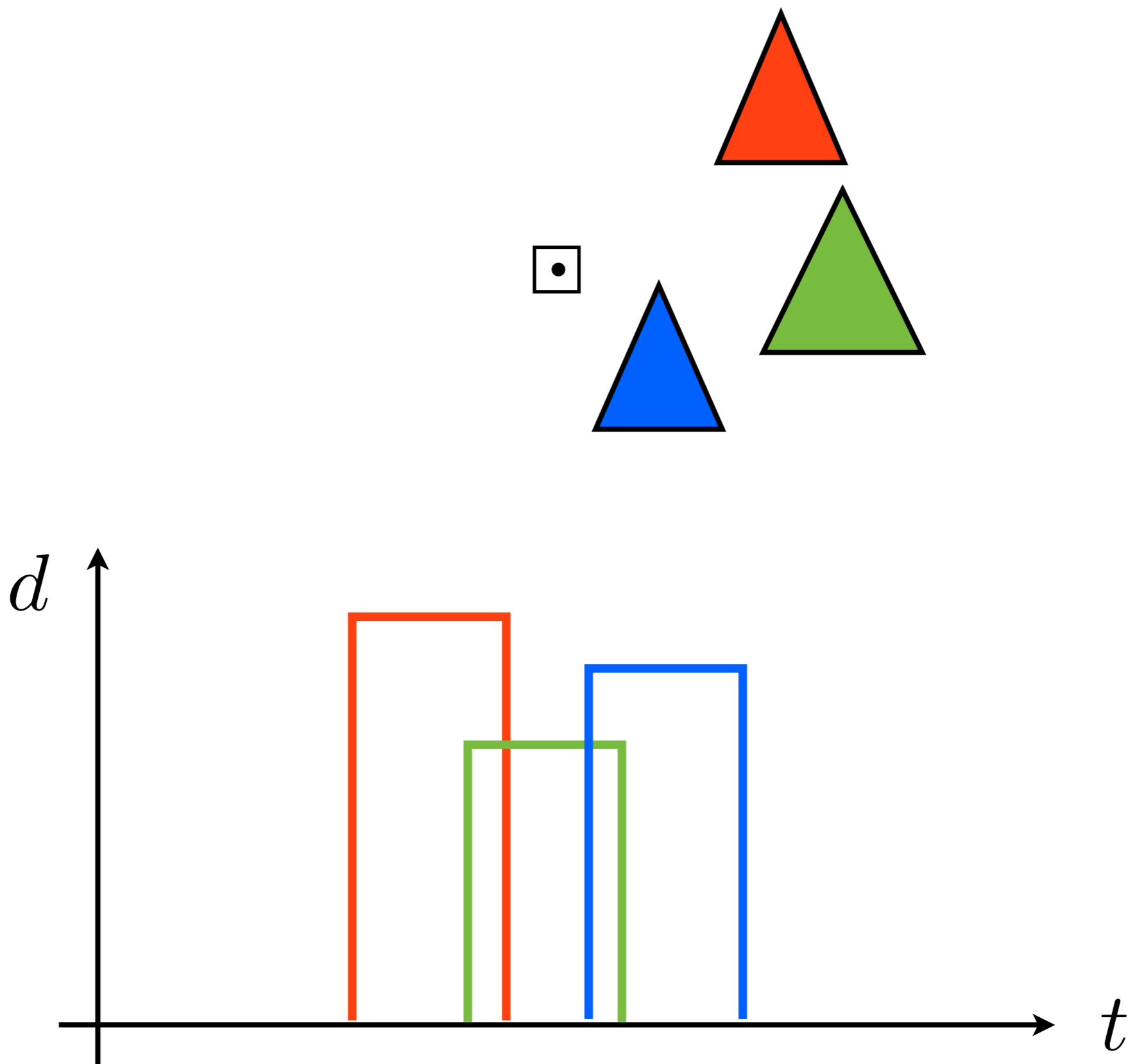
We also derive the analytical depth function $d(t)$...

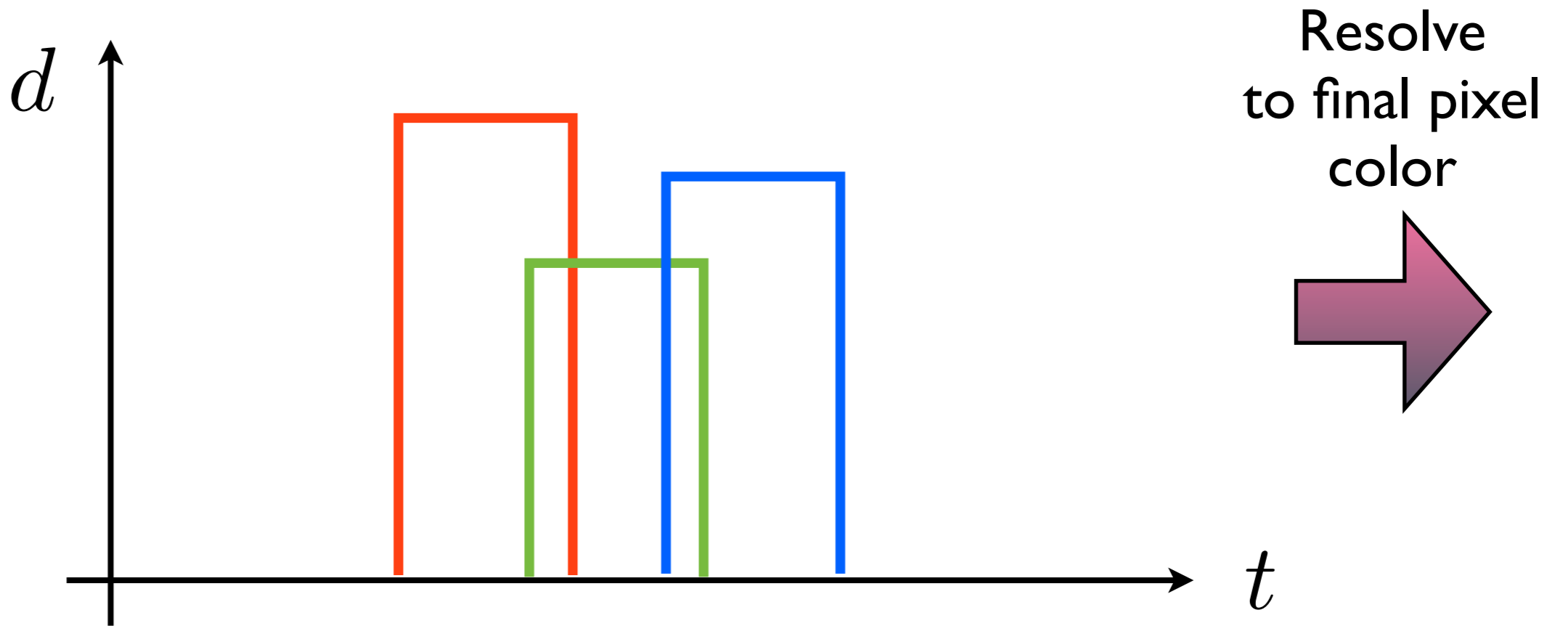
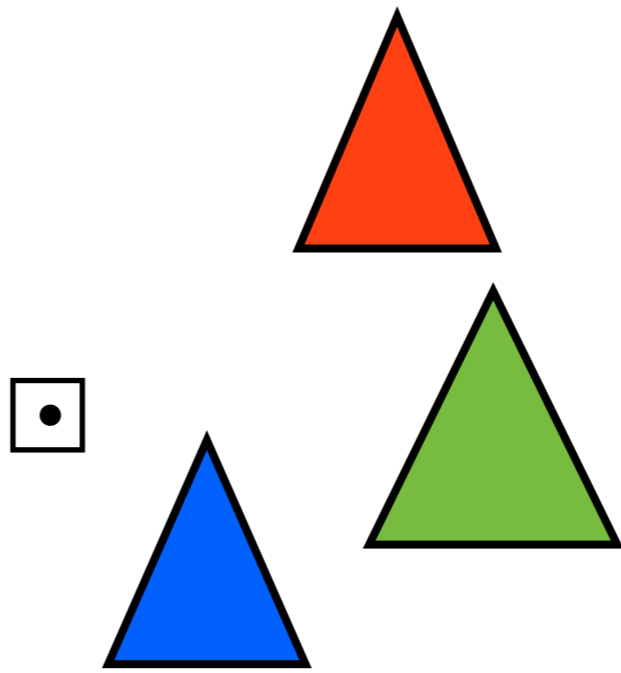


Rasterization

- For each pixel of bounding box, compute 2nd degree solutions
- Insert intervals into per-pixel list

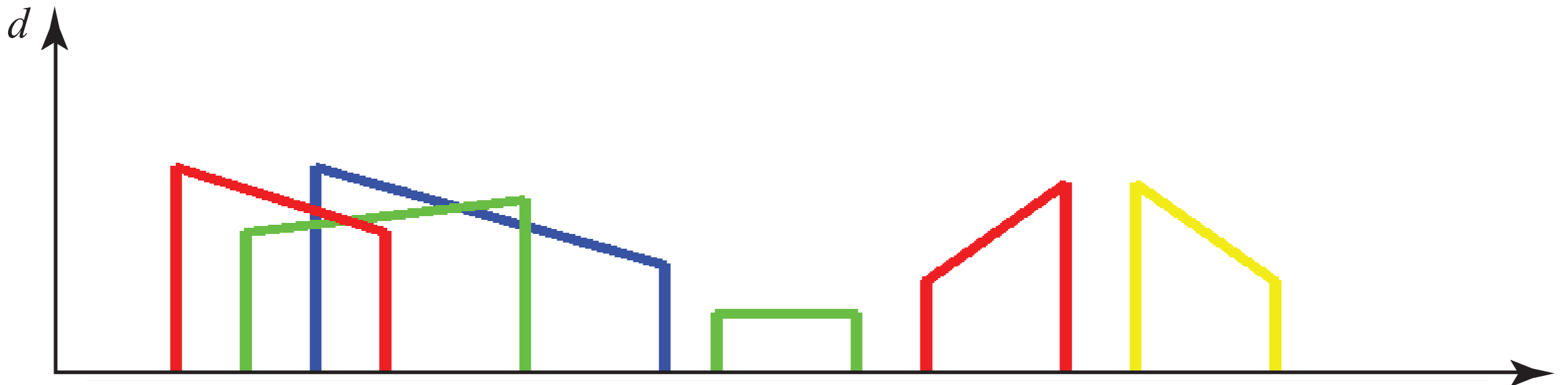






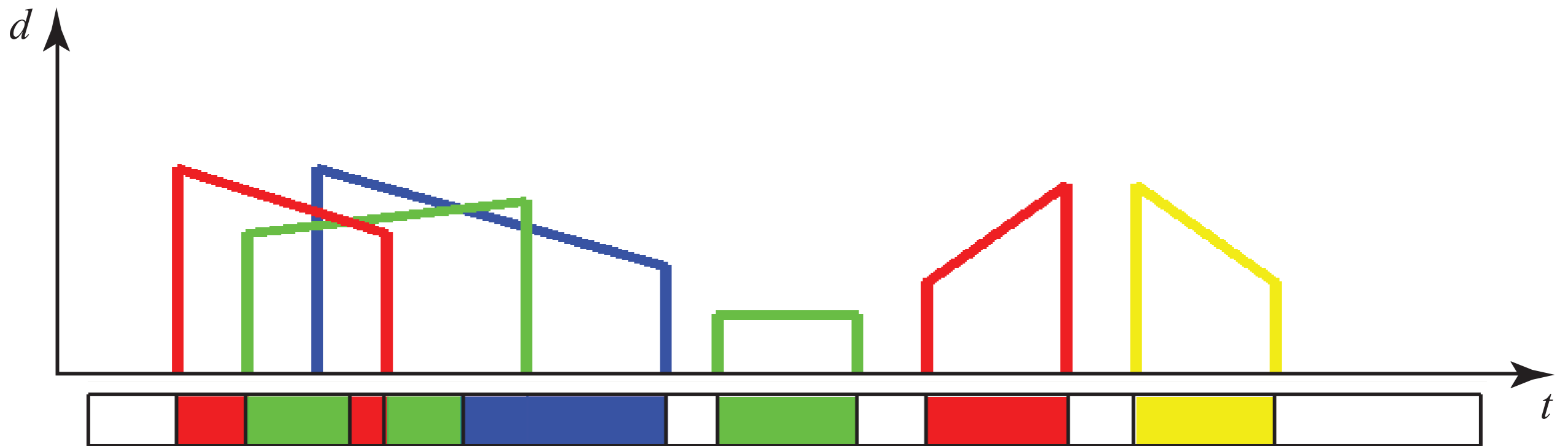
Opaque resolve

- Sweep over time and keep track of overlapping intervals
- Pick colors of the closest intervals
- Blend together for final color



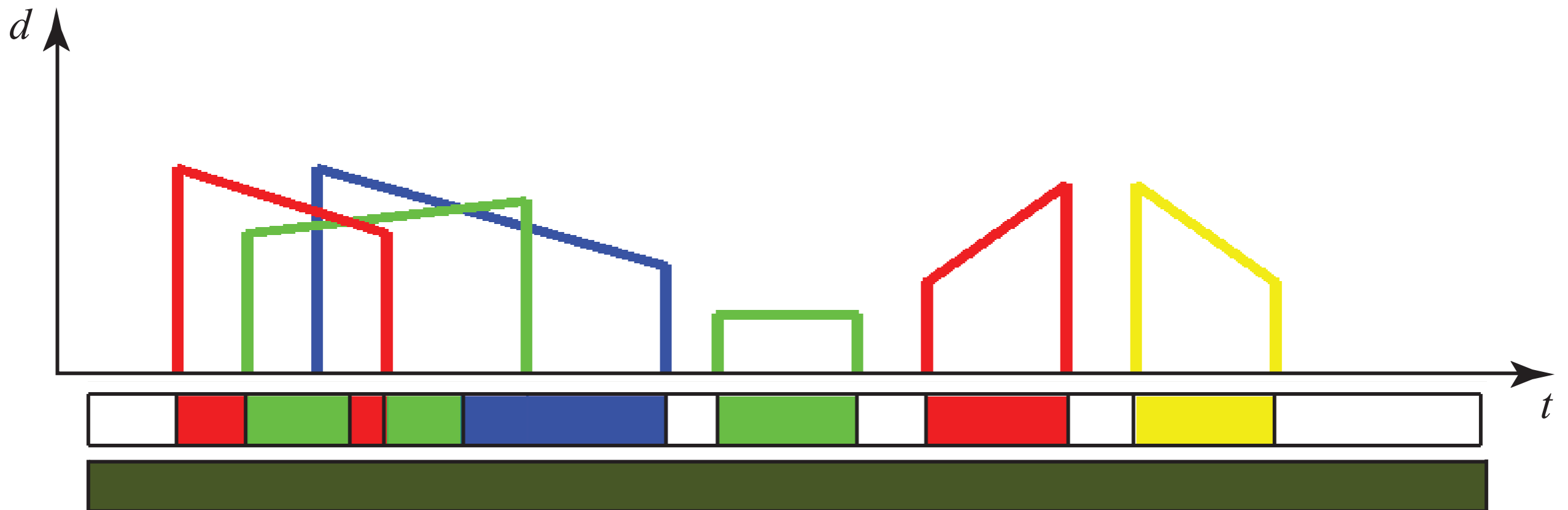
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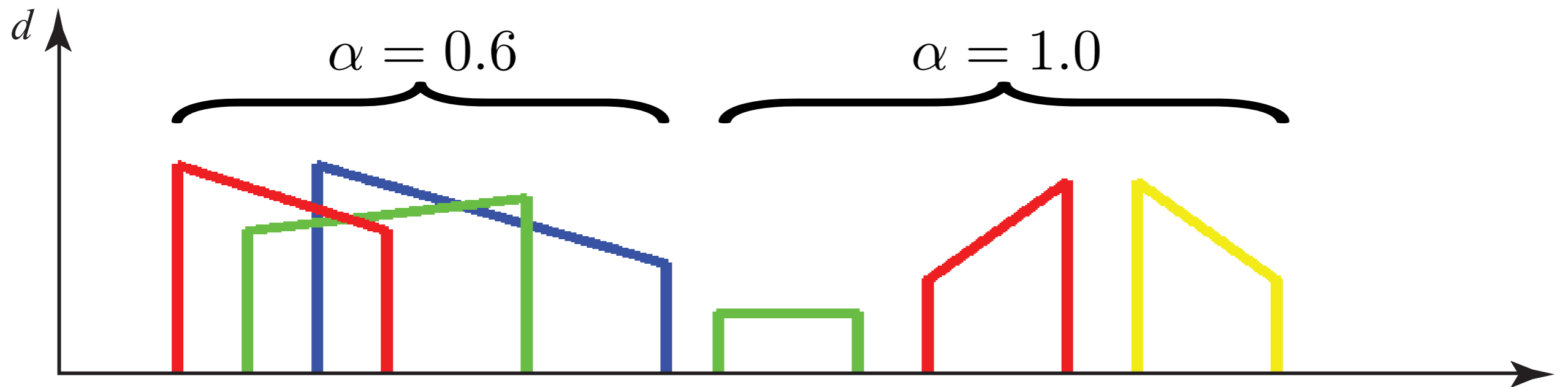
Opaque resolve

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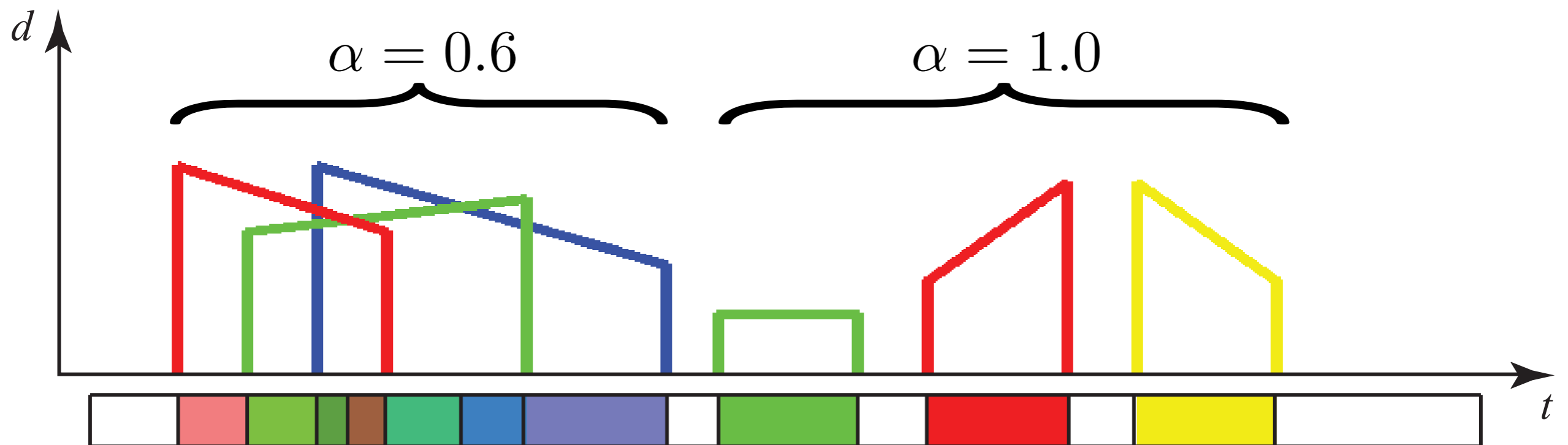
Translucent resolve

- Sweep over time and keep track of overlapping intervals
- **Alpha-blend each subsection**
- Blend together for final color



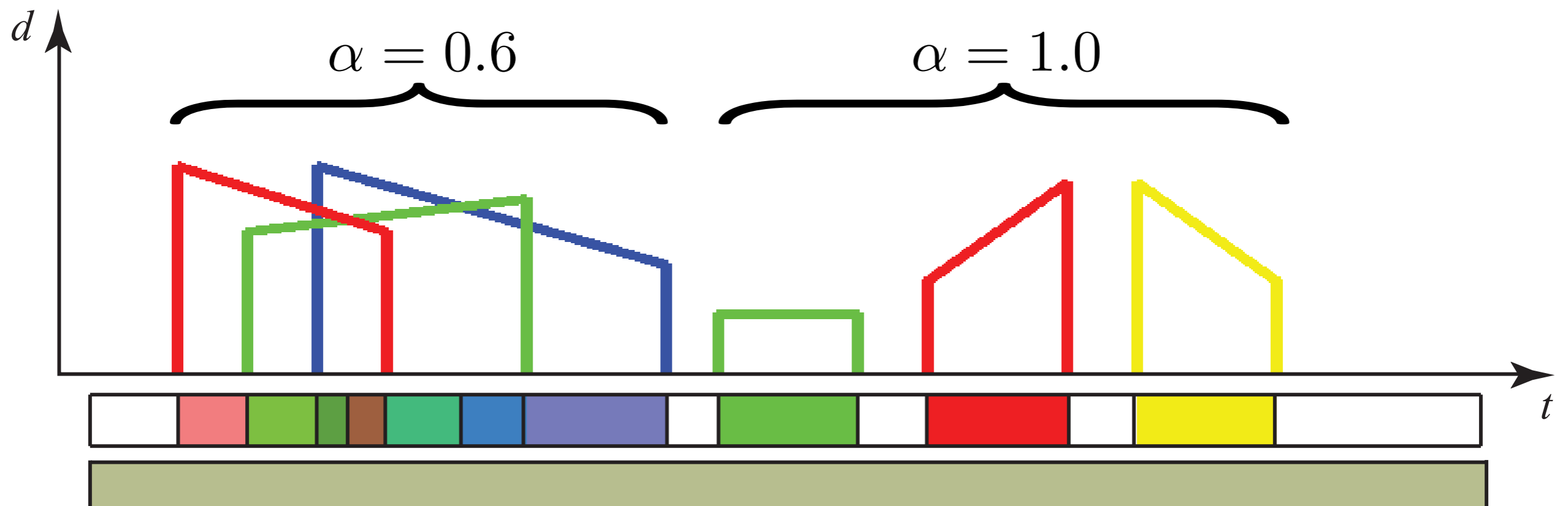
Translucent resolve

- Sweep over time and keep track of overlapping intervals
- **Alpha-blend each subsection**
- Blend together for final color



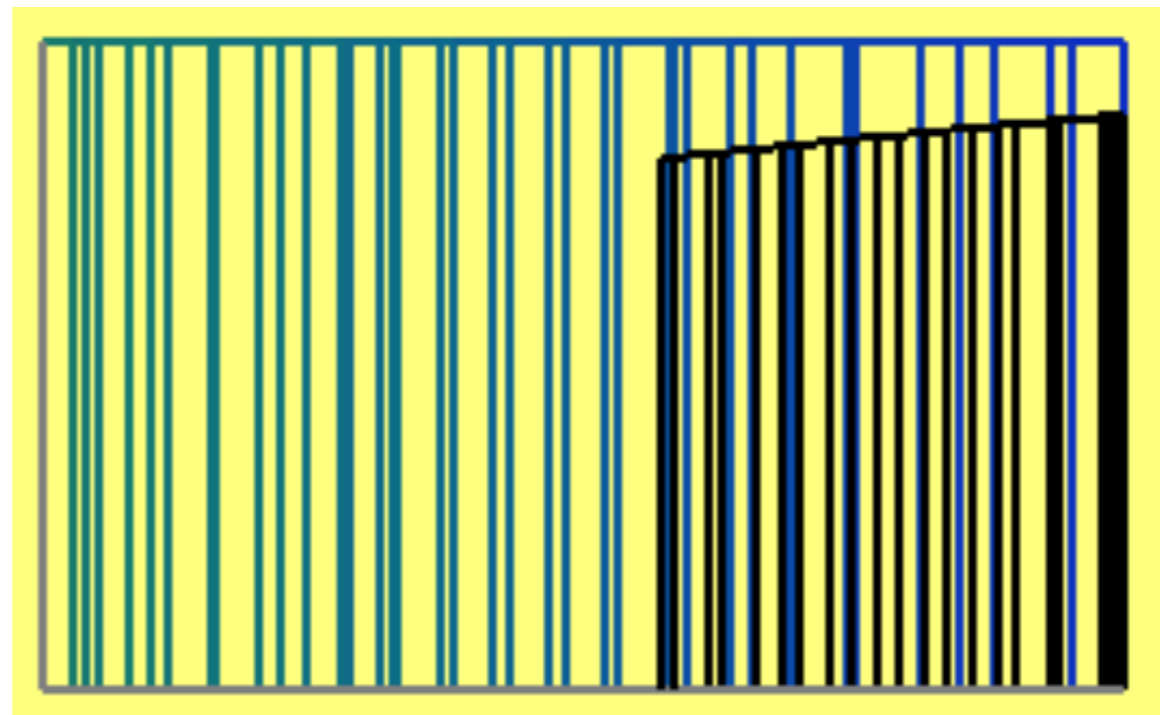
Translucent resolve

- Sweep over time and keep track of overlapping intervals
- **Alpha-blend each subsection**
- Blend together for final color



Compression

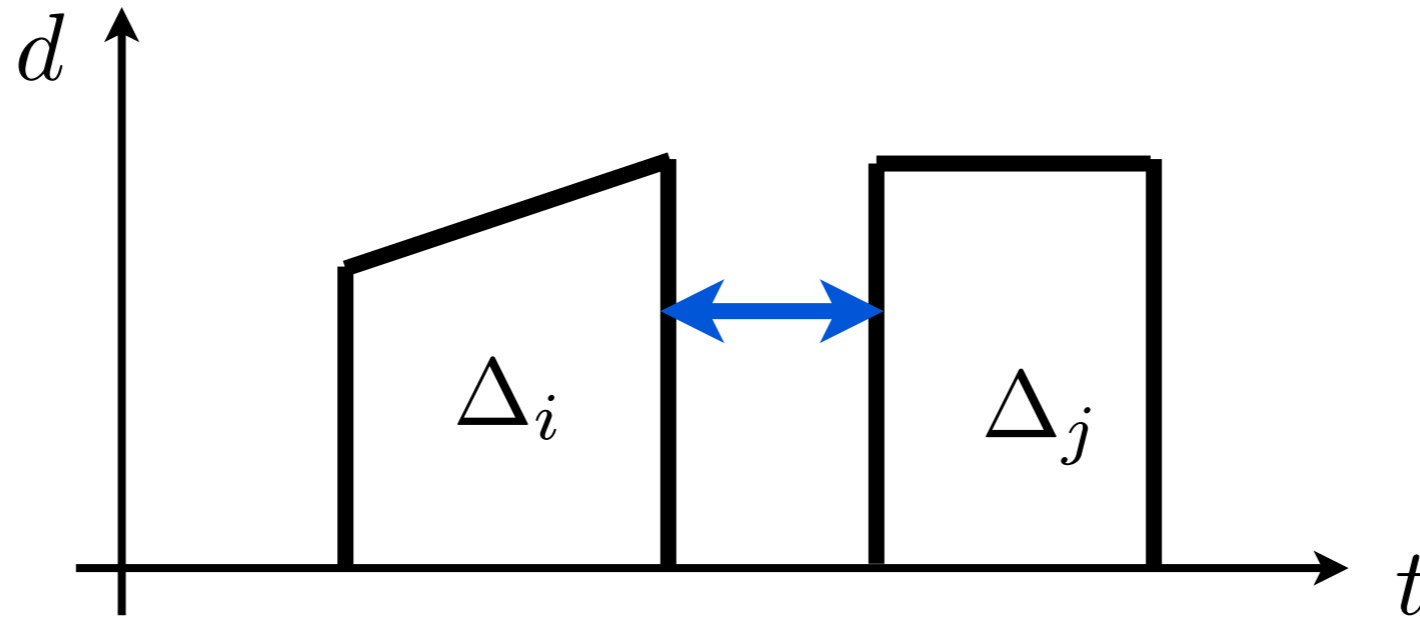
- Will end up with lots of intervals
- But, can expect spatial & temporal coherency
- Proposal: merge similar intervals



Compression: Oracle function

$$O(\Delta_i, \Delta_j)$$

- Metric for interval similarity
- Merge pairs of intervals deemed most similar



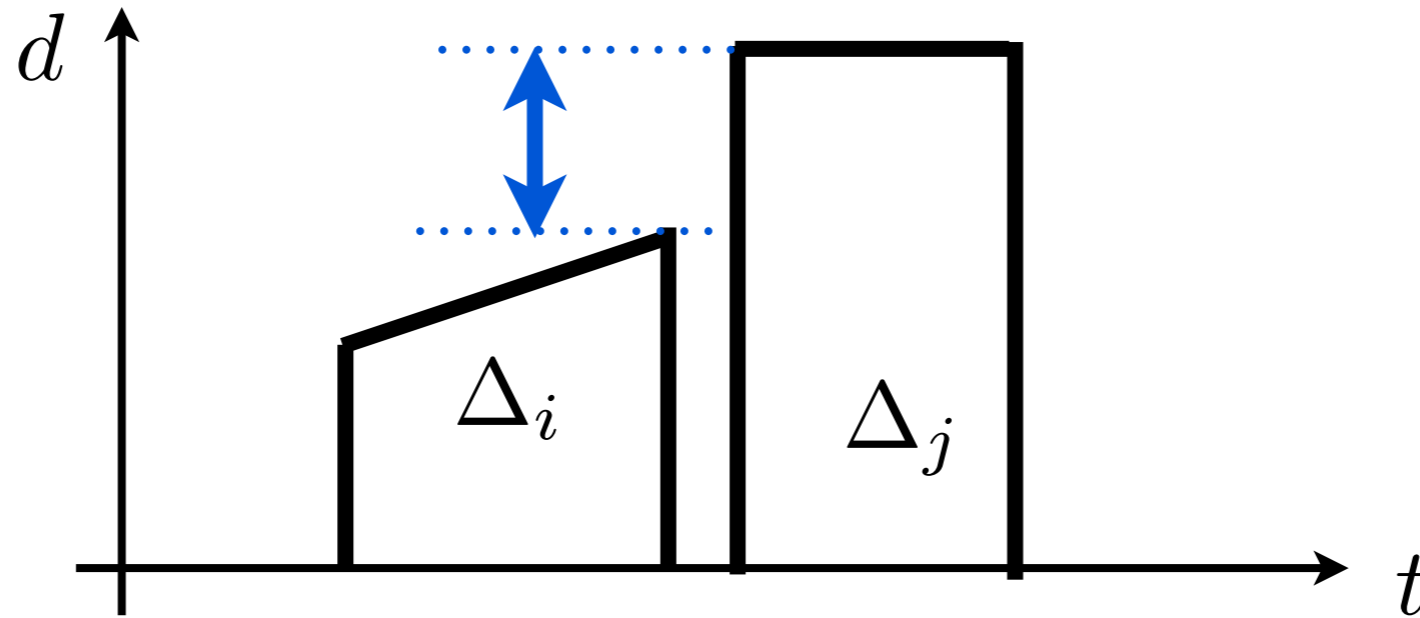
$$O(\Delta_i, \Delta_j) = \boxed{h_1 \max(t_j^s - t_i^e, 0) +} \quad \text{temporal proximity}$$

$$h_2 |\bar{z}_i - z_j| +$$

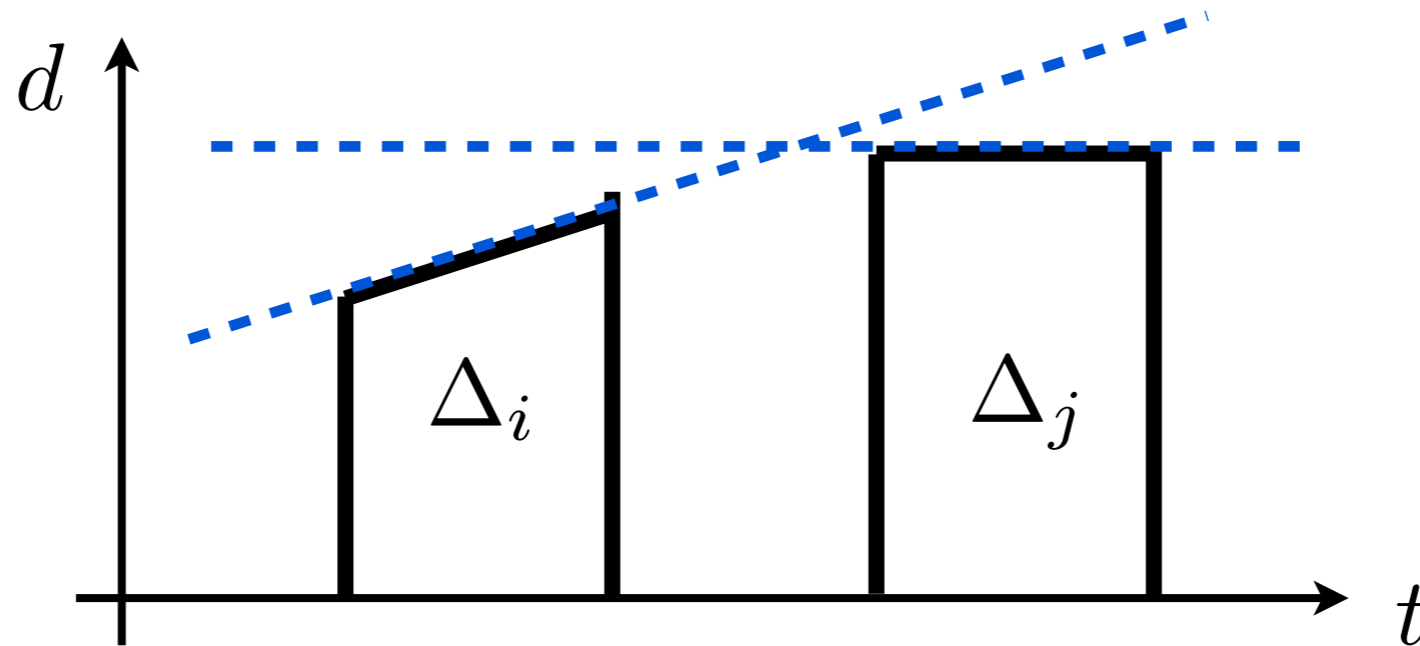
$$h_3 |k_i - k_j| +$$

$$h_4 (t_i^e - t_i^s + t_j^e - t_j^s) +$$

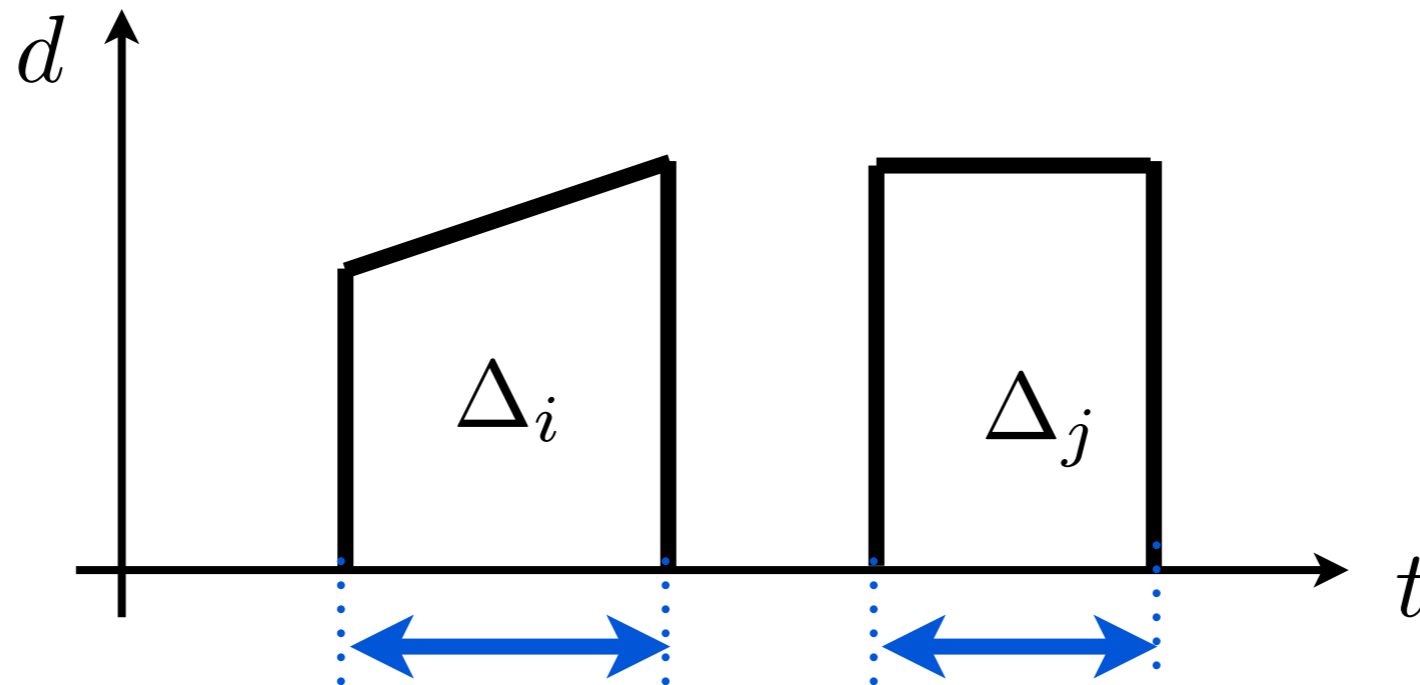
$$h_5 (|c_i - c_j|)$$



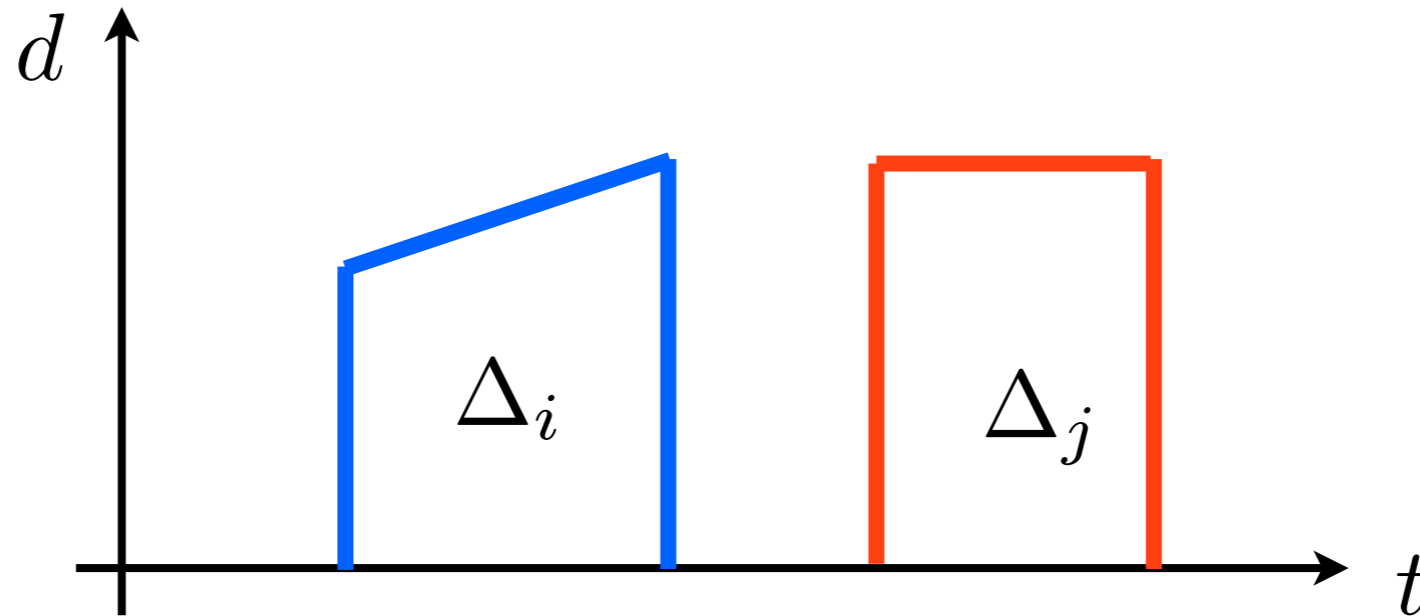
$$\begin{aligned}
 O(\Delta_i, \Delta_j) = & h_1 \max(t_j^s - t_i^e, 0) + && \text{temporal proximity} \\
 & h_2 |\bar{z}_i - z_j| + && \text{depths} \\
 & h_3 |k_i - k_j| + \\
 & h_4 (t_i^e - t_i^s + t_j^e - t_j^s) + \\
 & h_5 (|c_i - c_j|)
 \end{aligned}$$



$$\begin{aligned}
 O(\Delta_i, \Delta_j) = & h_1 \max(t_j^s - t_i^e, 0) + && \text{temporal proximity} \\
 & h_2 |\bar{z}_i - z_j| + && \text{depths} \\
 & h_3 |k_i - k_j| + && \text{slope} \\
 & h_4 (t_i^e - t_i^s + t_j^e - t_j^s) + \\
 & h_5 (|c_i - c_j|)
 \end{aligned}$$



$$\begin{aligned}
 O(\Delta_i, \Delta_j) = & h_1 \max(t_j^s - t_i^e, 0) + && \text{temporal proximity} \\
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 & h_3 |k_i - k_j| + && \text{slope} \\
 & h_4 (t_i^e - t_i^s + t_j^e - t_j^s) + && \text{temporal coverage} \\
 & h_5 (|c_i - c_j|)
 \end{aligned}$$



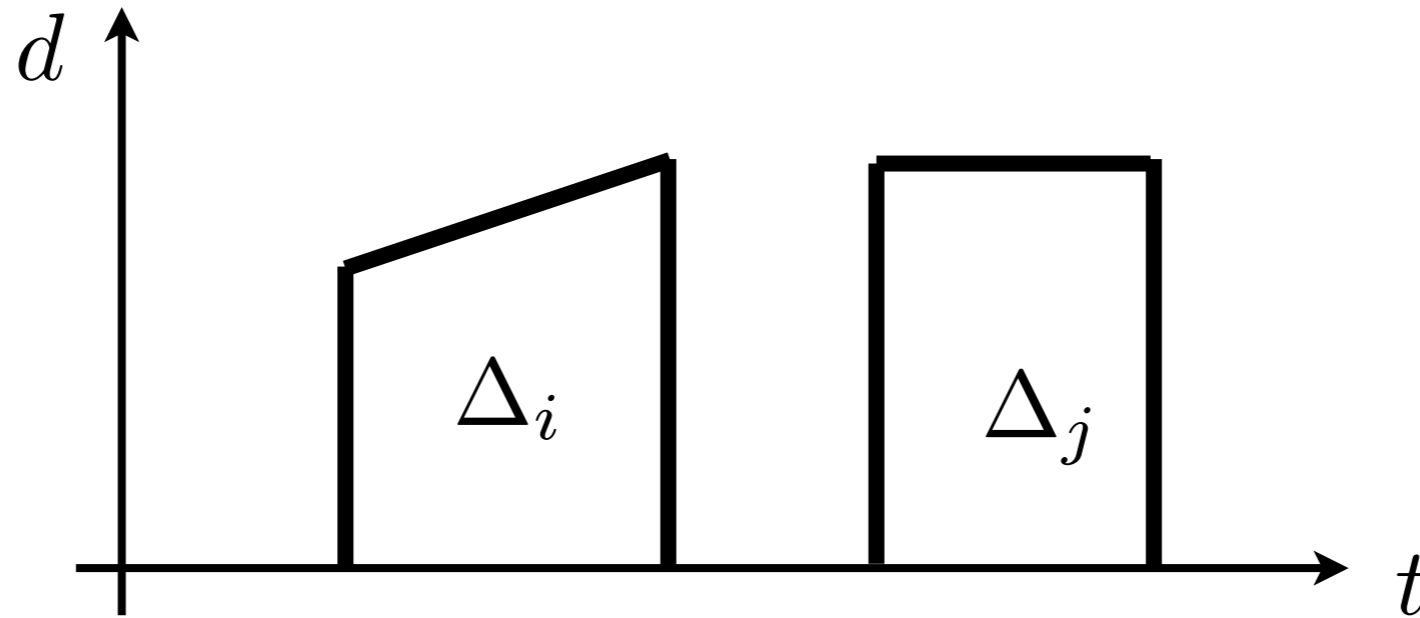
$$O(\Delta_i, \Delta_j) = h_1 \max(t_j^s - t_i^e, 0) + \text{temporal proximity}$$

$$h_2 |\bar{z}_i - z_j| + \text{depths}$$

$$h_3 |k_i - k_j| + \text{slope}$$

$$h_4 (t_i^e - t_i^s + t_j^e - t_j^s) + \text{temporal coverage}$$

$$\boxed{h_5 (|c_i - c_j|)} \quad \text{color norm}$$



$$O(\Delta_i, \Delta_j) = h_1 \max(t_j^s - t_i^e, 0) + \text{temporal proximity}$$

$$h_2 |\bar{z}_i - z_j| + \text{depths}$$

$$h_3 |k_i - k_j| + \text{slope}$$

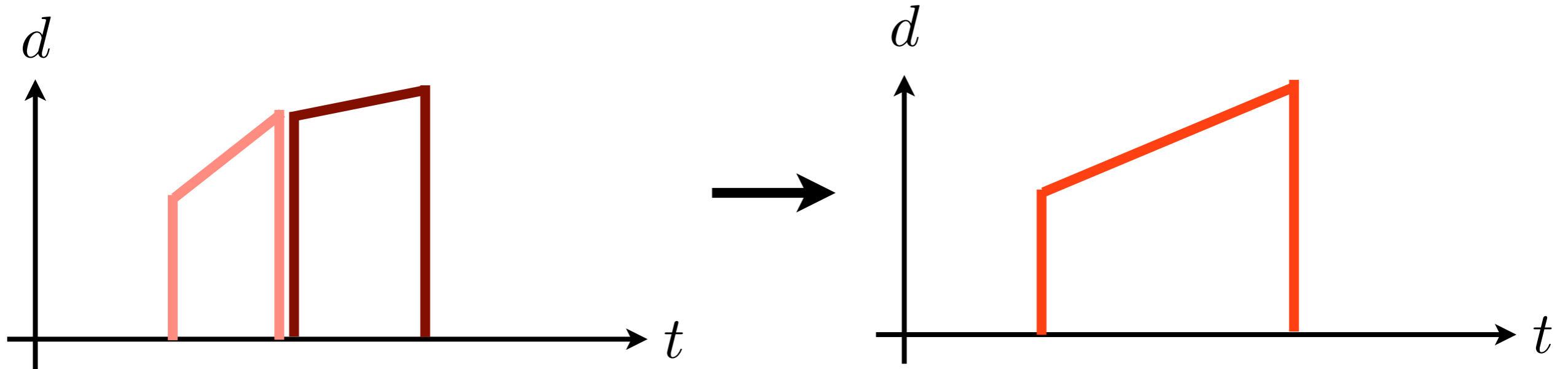
$$h_4 (t_i^e - t_i^s + t_j^e - t_j^s) + \text{temporal coverage}$$

$$h_5 (|c_i - c_j|) \text{ color norm}$$

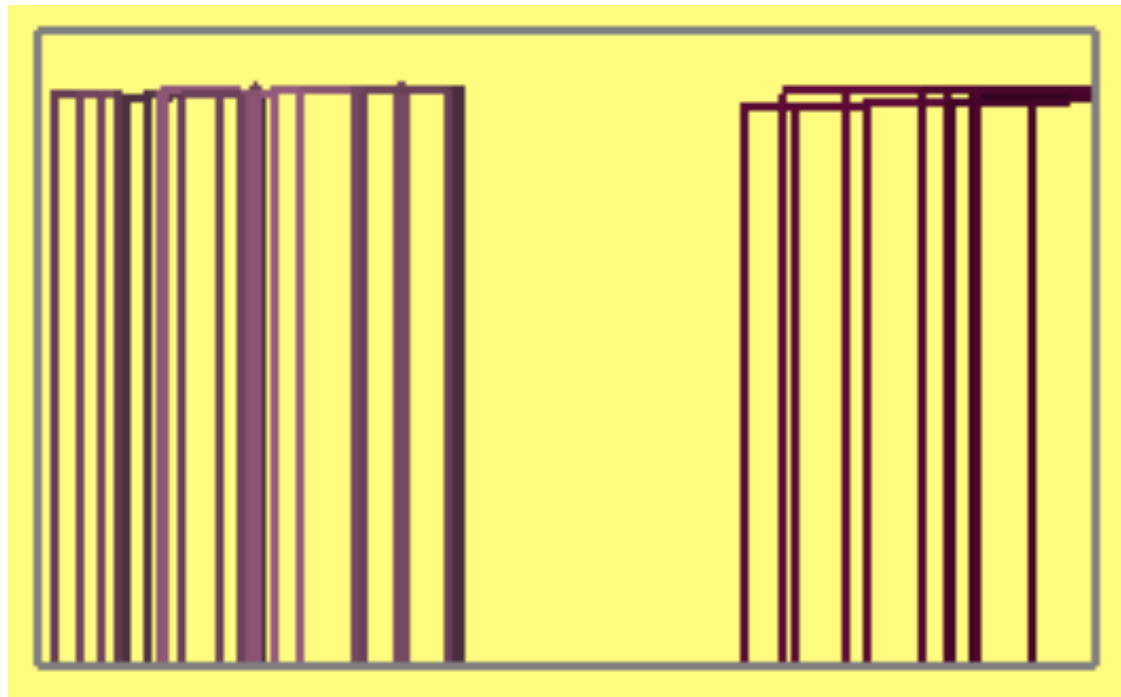
weights

Interval merging

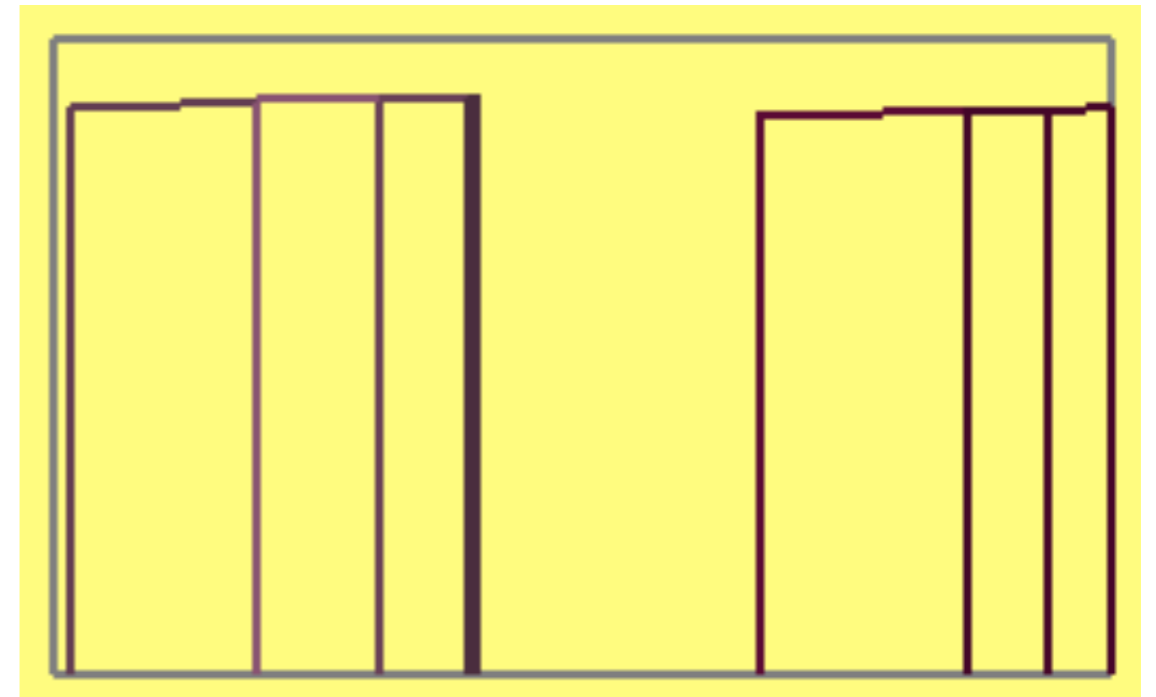
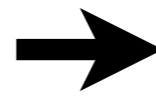
- Merge interval pair with lowest Oracle value
- Combine temporal coverage
- Blend attributes



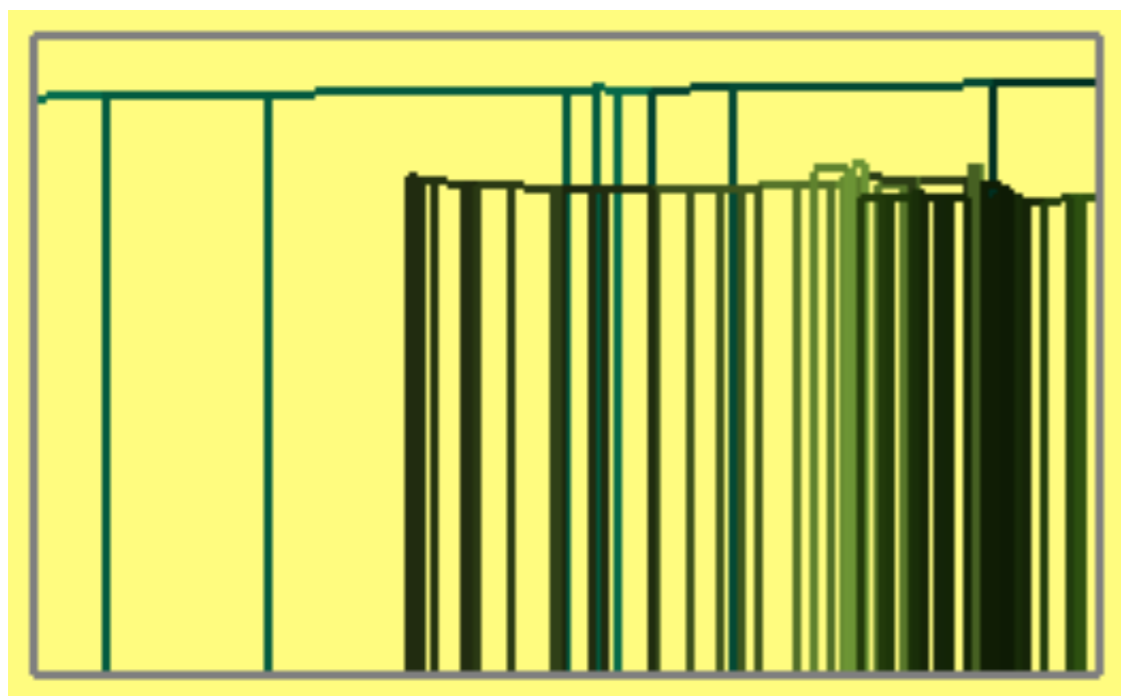
Compression with occlusion



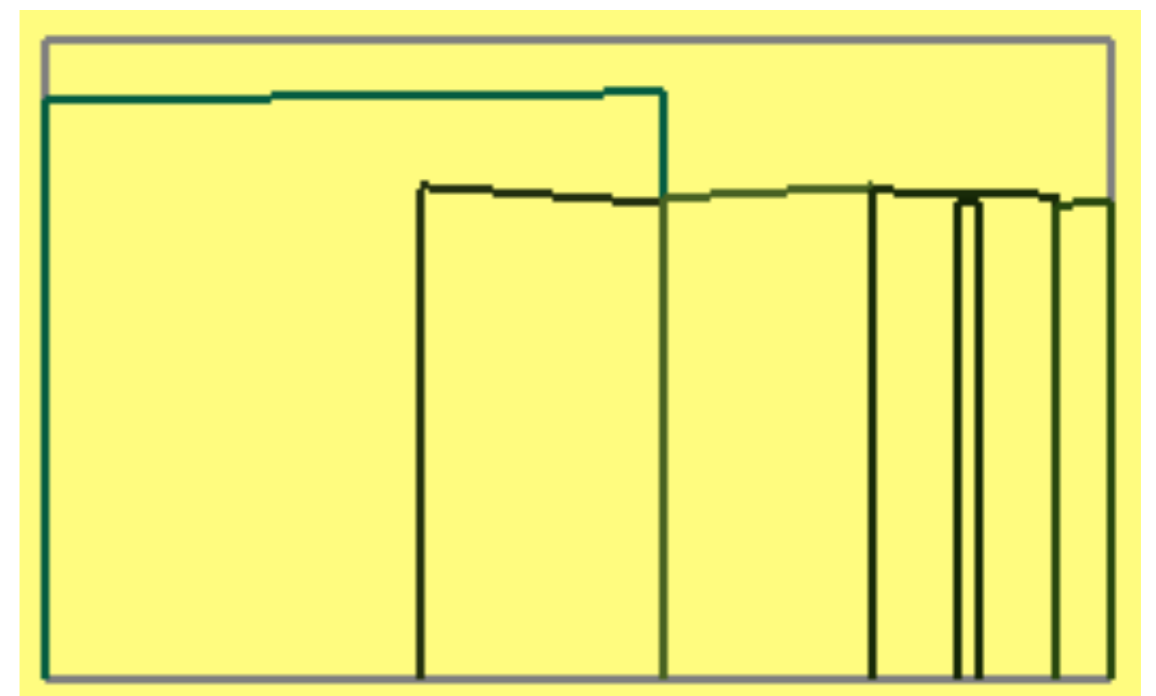
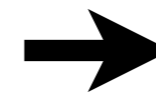
39 intervals



8 intervals

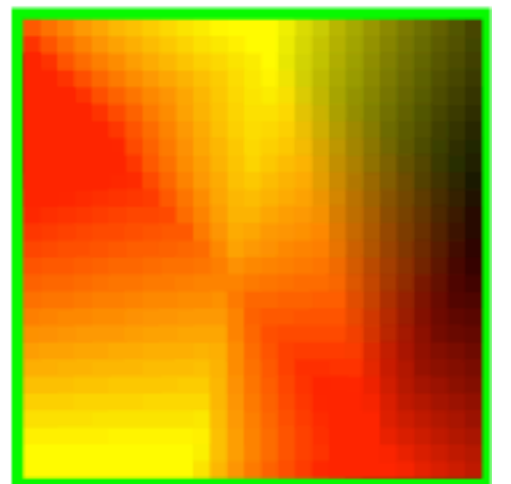
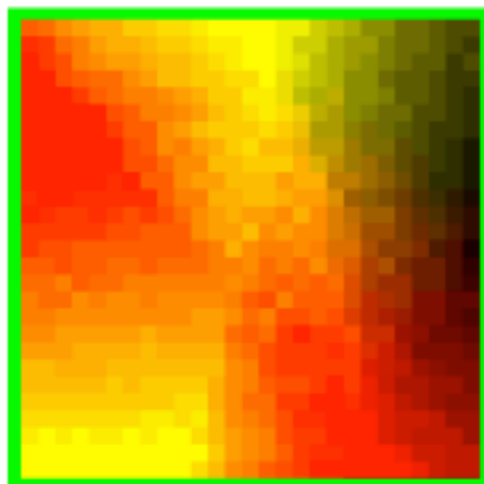
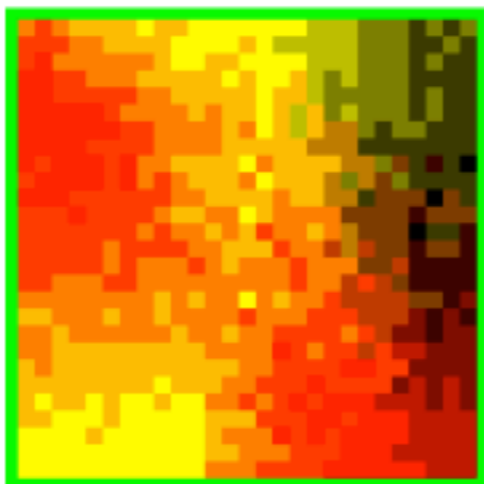
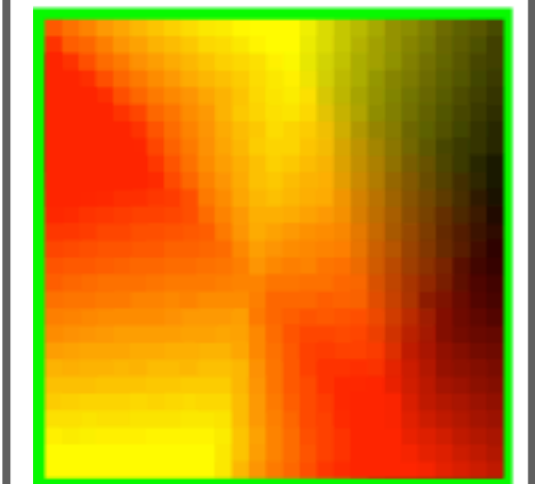
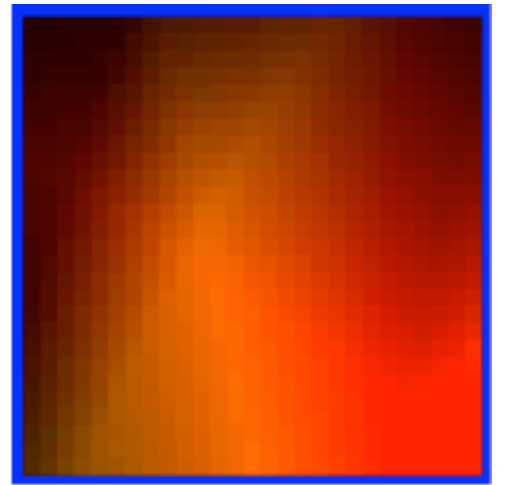
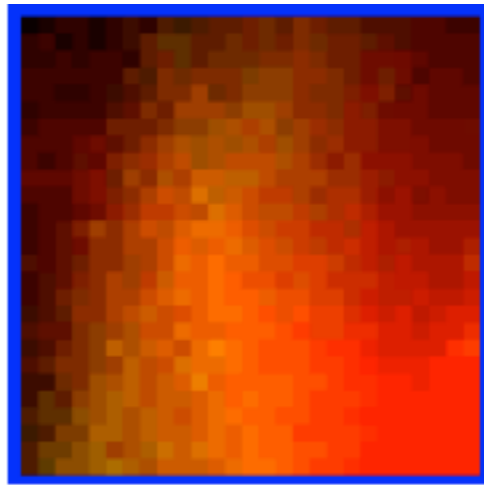
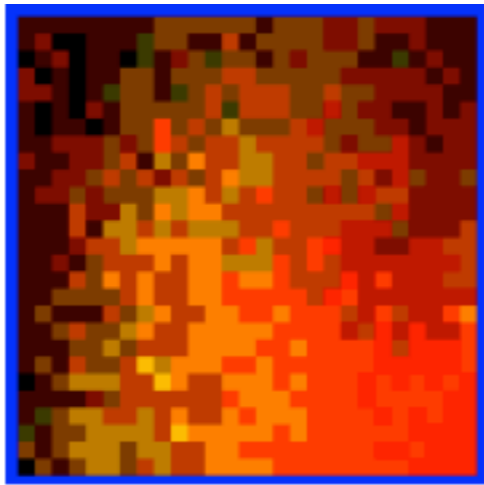
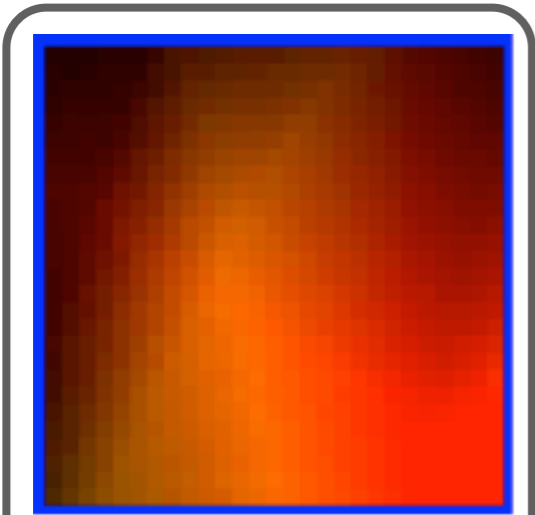
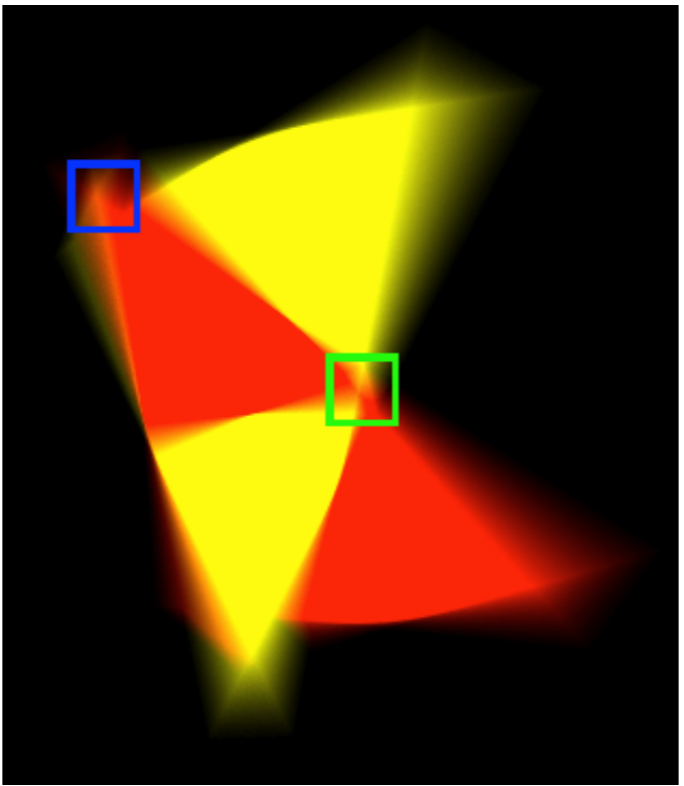


77 intervals



8 intervals

Results

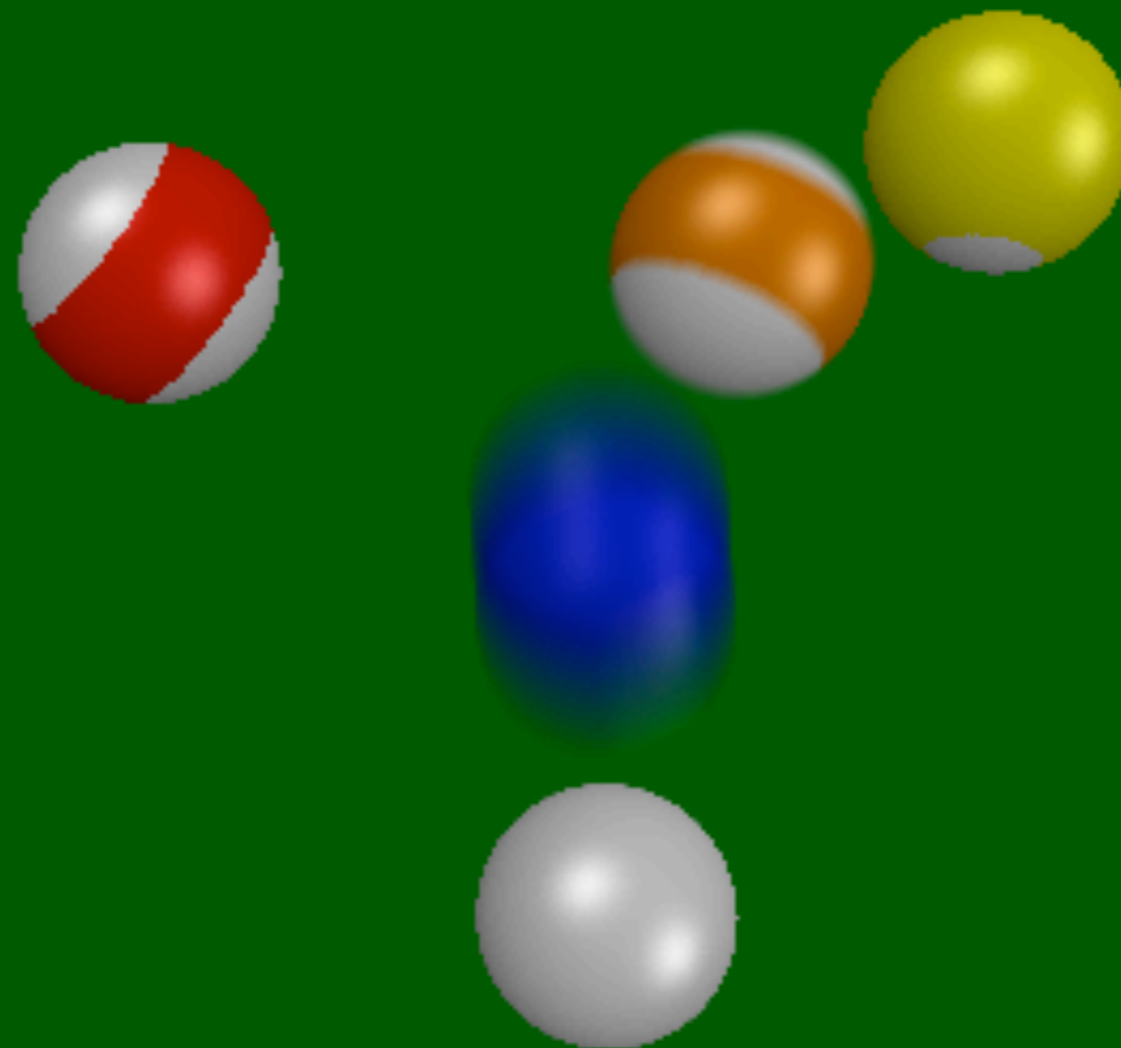


GT

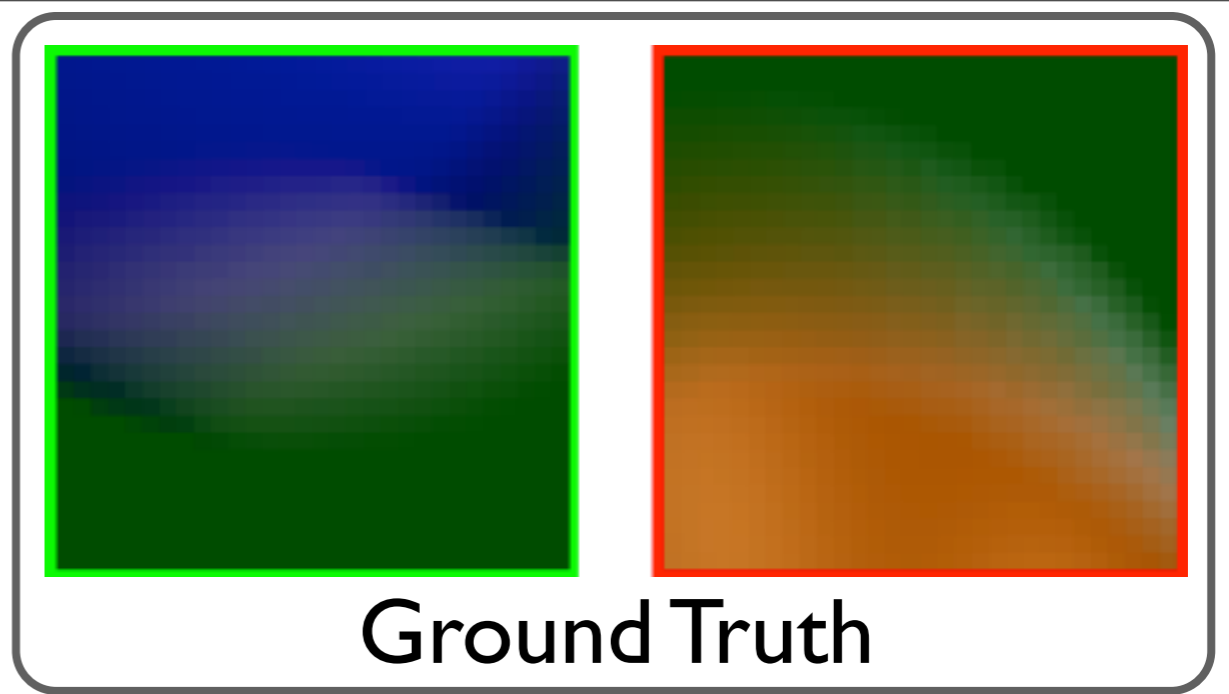
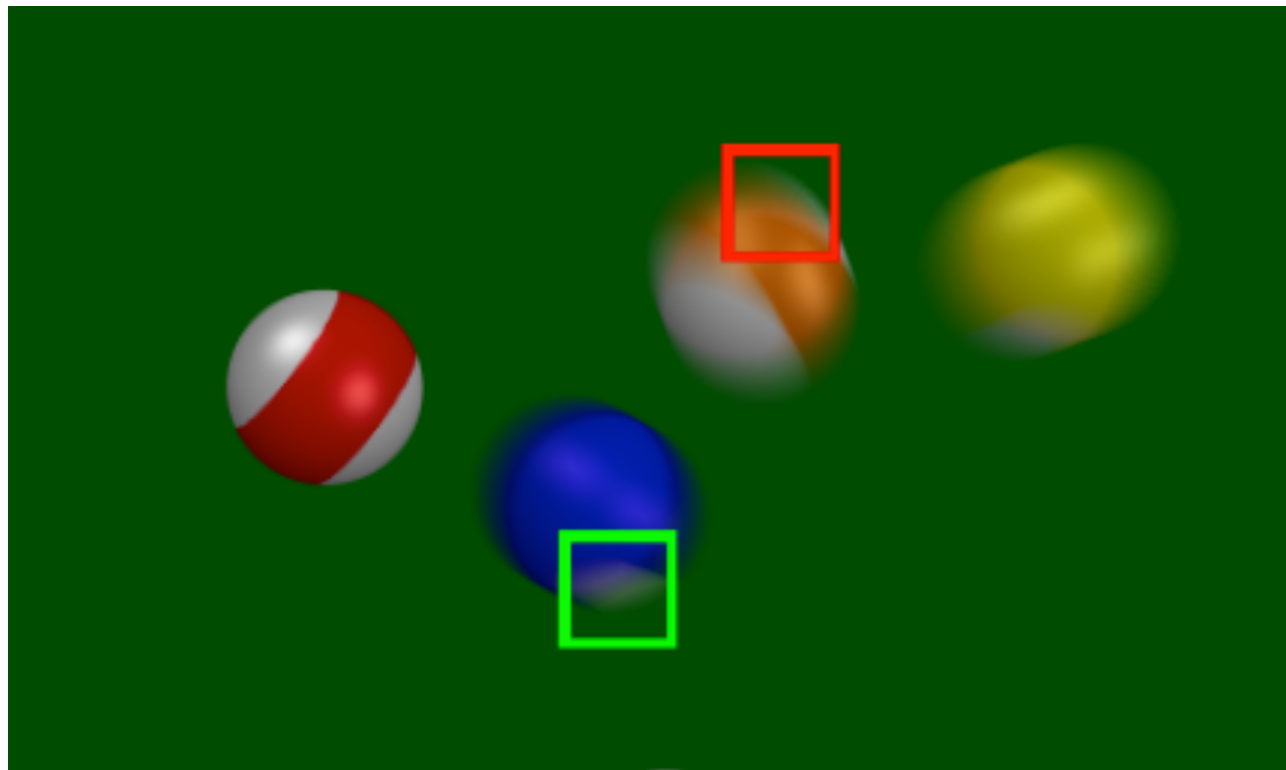
4 spp

16 spp

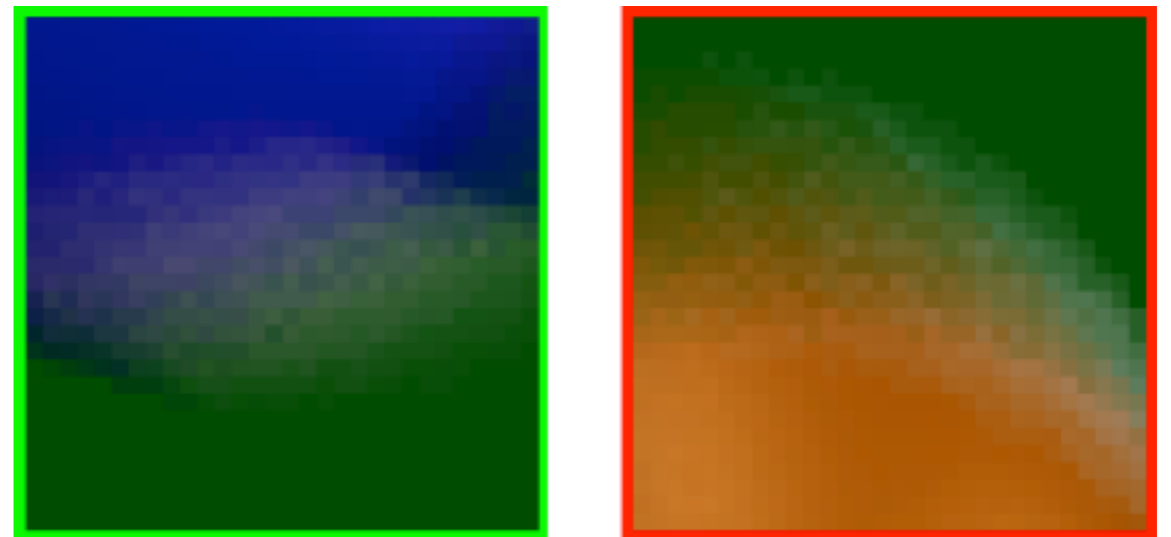
Analytical



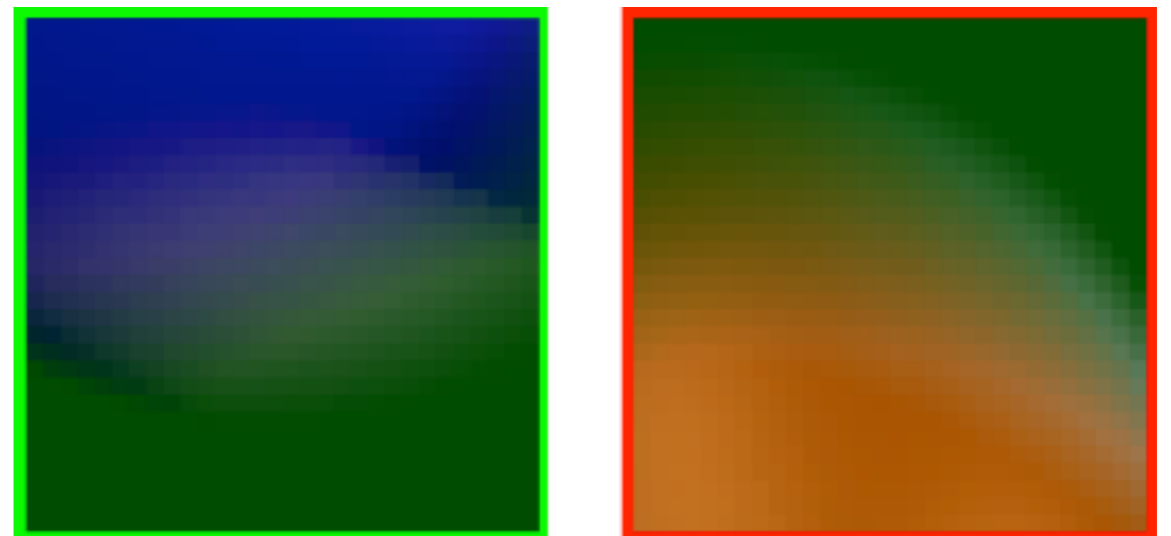
Compression: 8



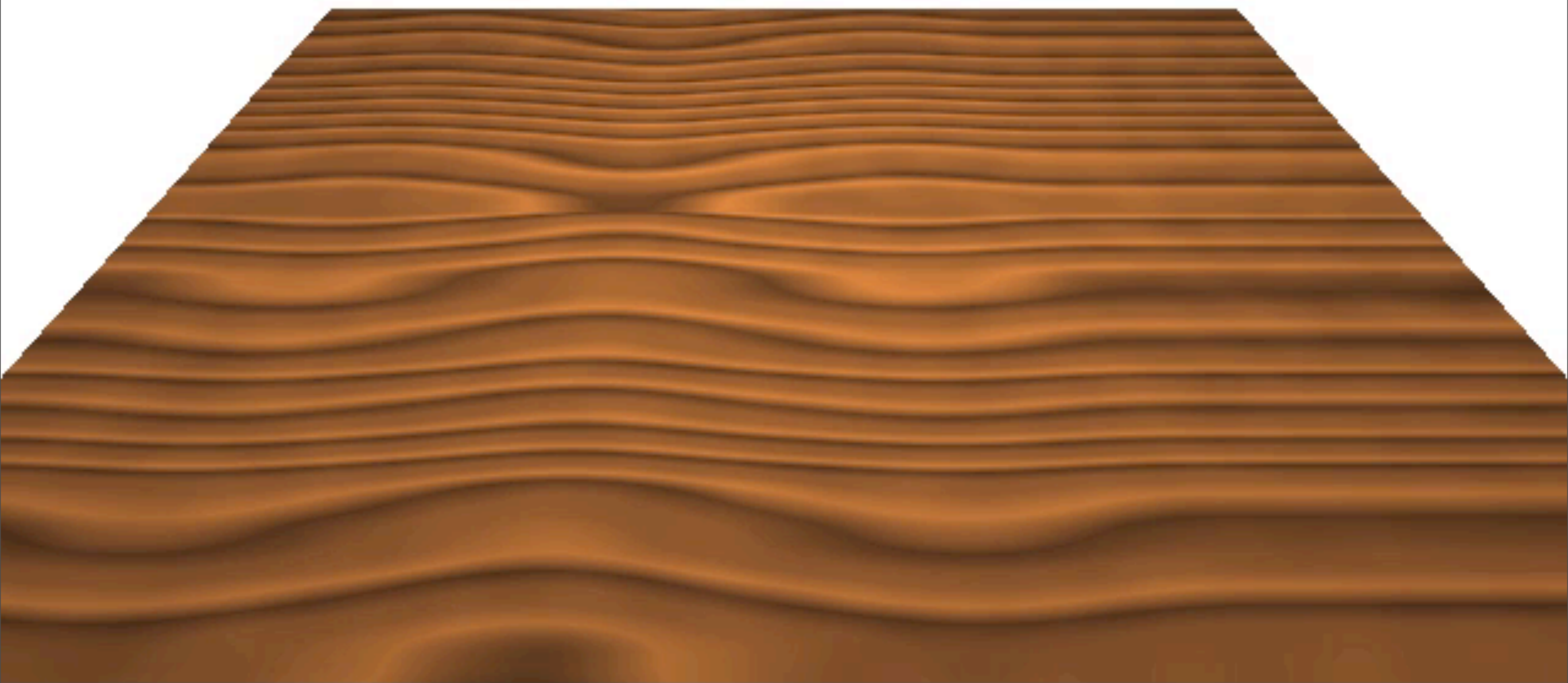
Ground Truth



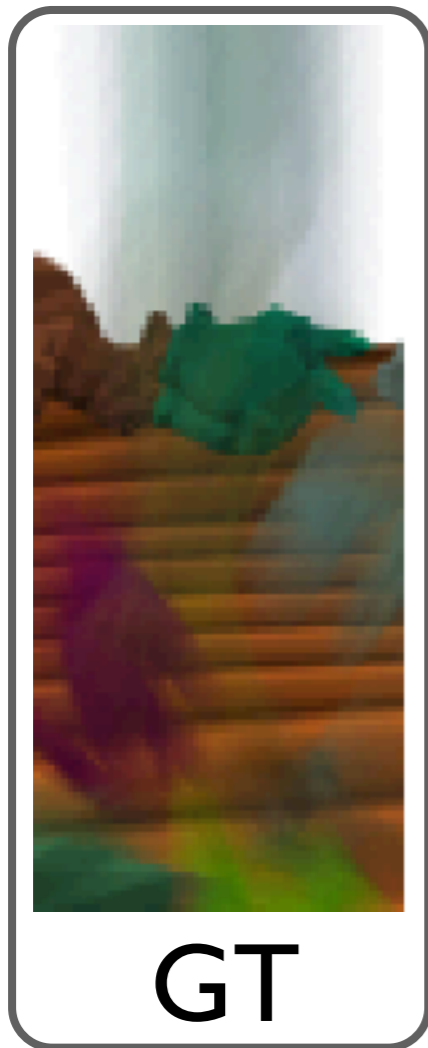
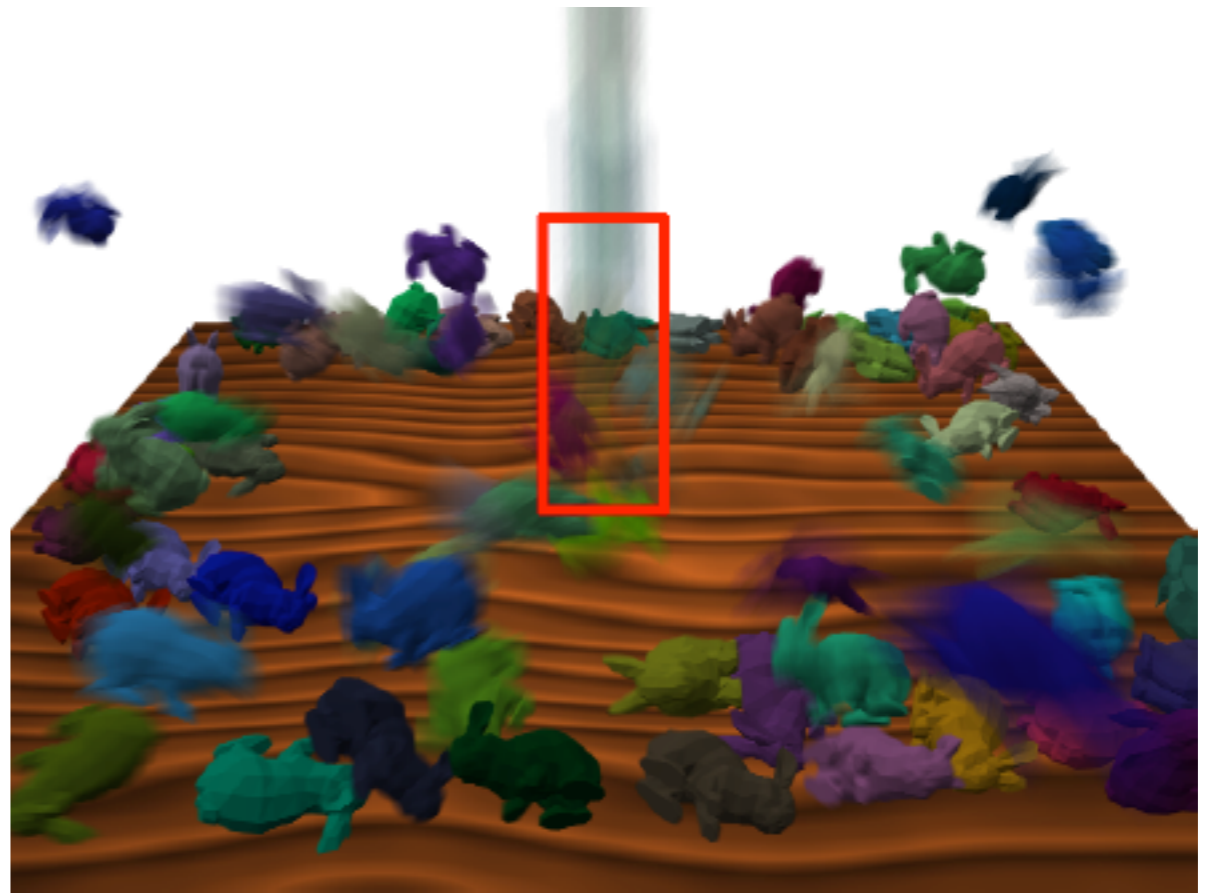
Stochastic 12 spp



Analytical, compression 8



Compression Rates



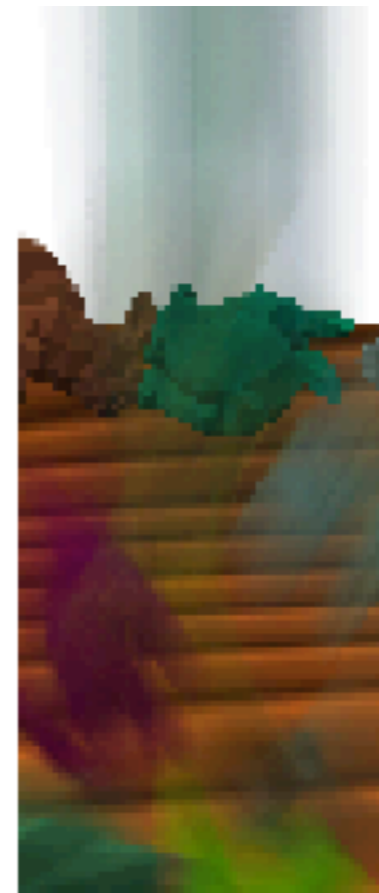
GT



Rate 4



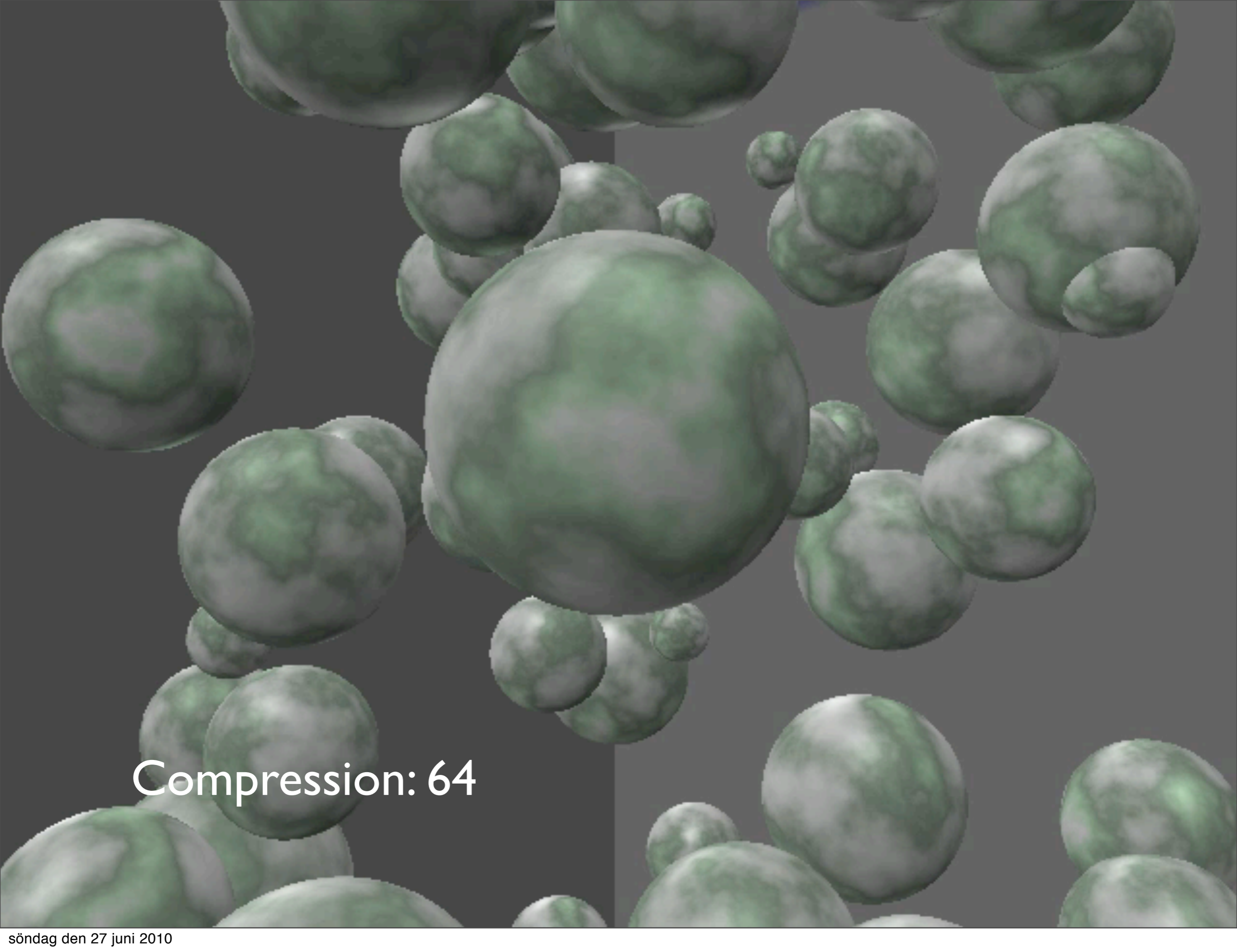
Rate 6



Rate 8



Rate 10



Compression: 64

Algorithm summary

- Time-dependent edge equations to compute exact exposure intervals
- Visibility management by using a linear approximation of a cubic rational depth function
- Blur translucent geometry by performing alpha-blending extended in time
- We propose a compression algorithm to address fixed memory requirements

Future work

- GPU implementation (in progress)
- Motion blurred shadows (in progress)
- Apply decoupled shading
- Analytical spatial anti-aliasing
- Higher order motion



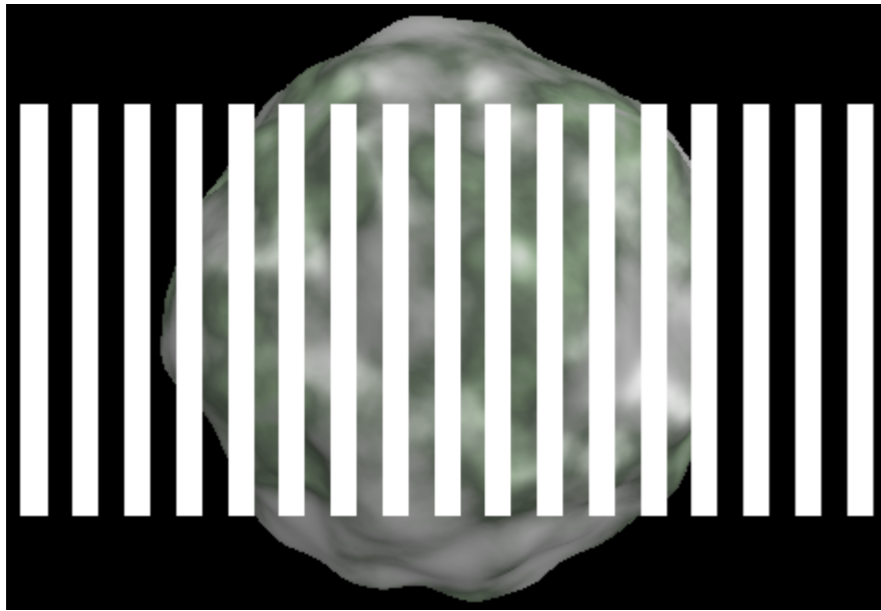
Danke !

Danke !

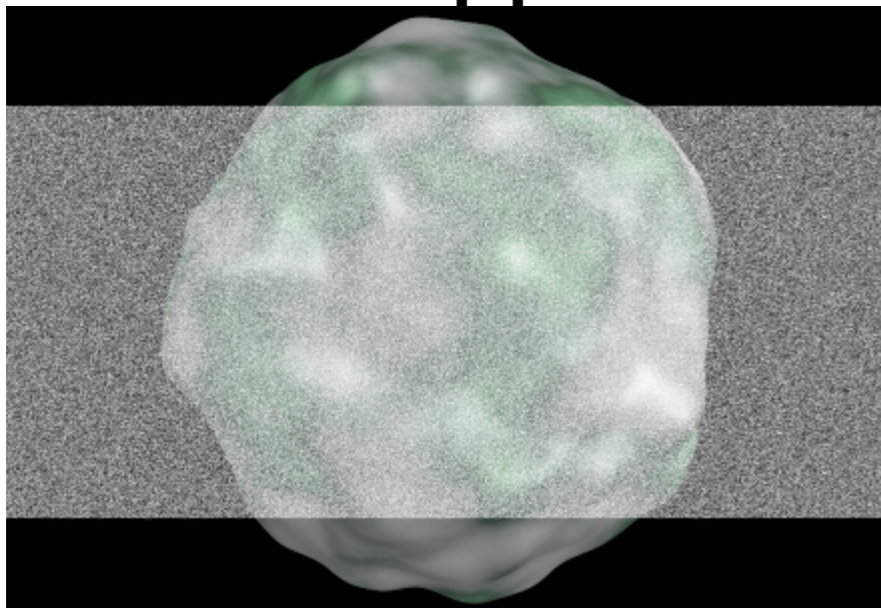
* No bunnies were harmed during the research

Compression failure case

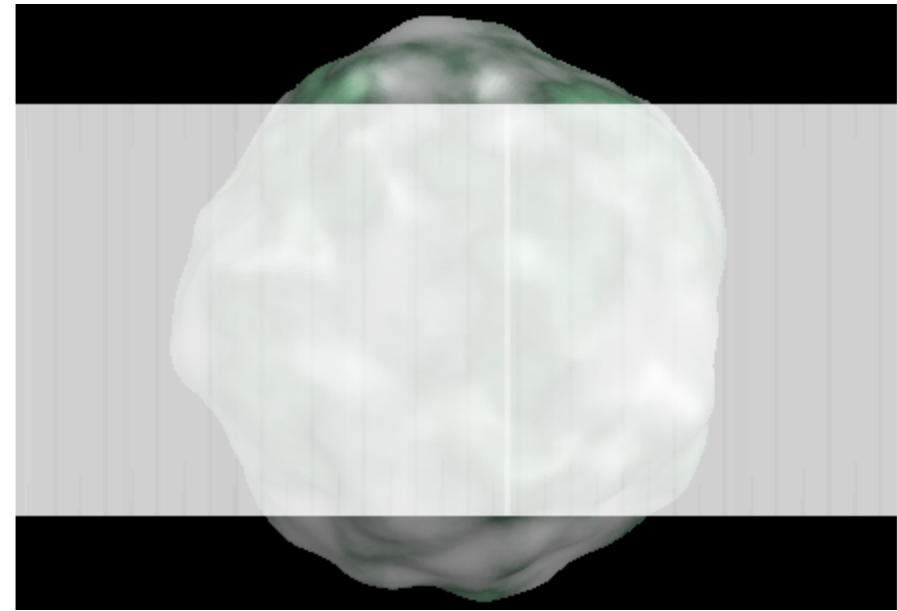
Fast moving fence



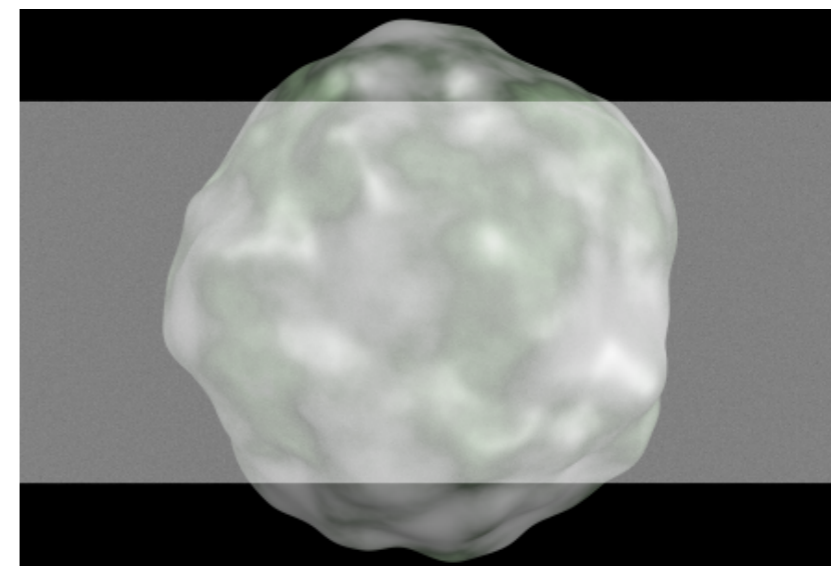
24 spp



Analytical 16 intervals



256 spp



Depth approximation error

