



# Decomposing Images into Layers with Advanced Color Blending

---

Yuki Koyama & Masataka Goto



# Quick Summary



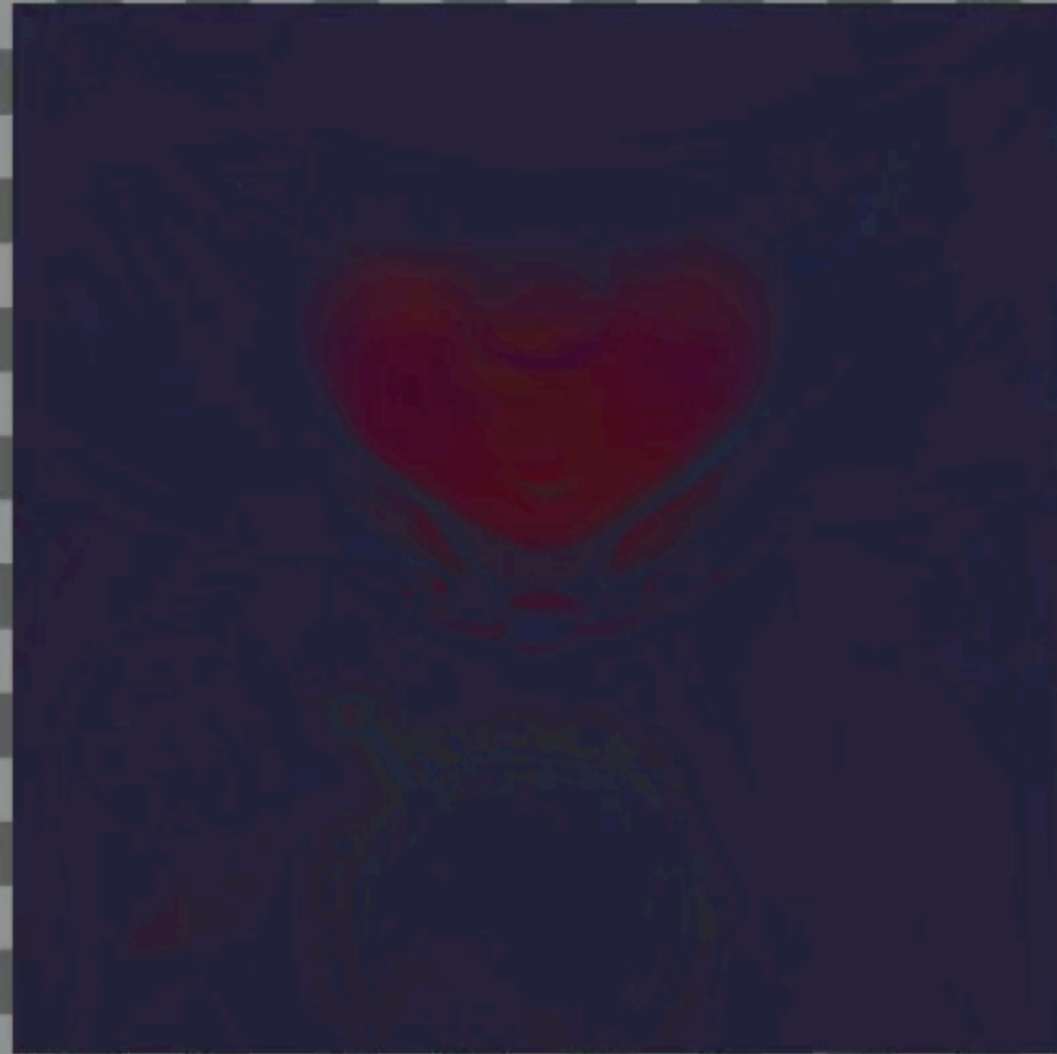
Our method decomposes a target image  
into semi-transparent layers ...



normal

hard-light

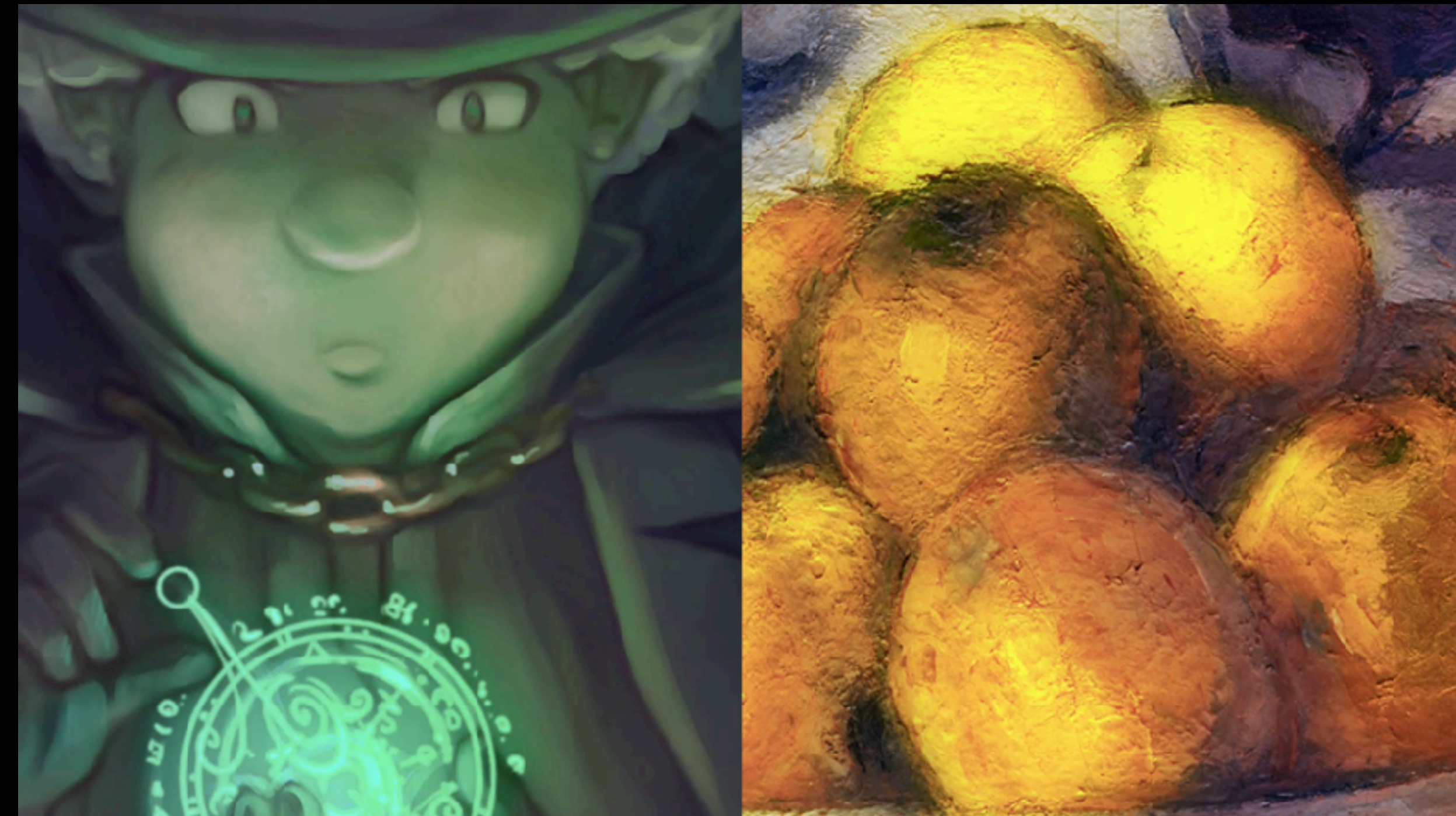
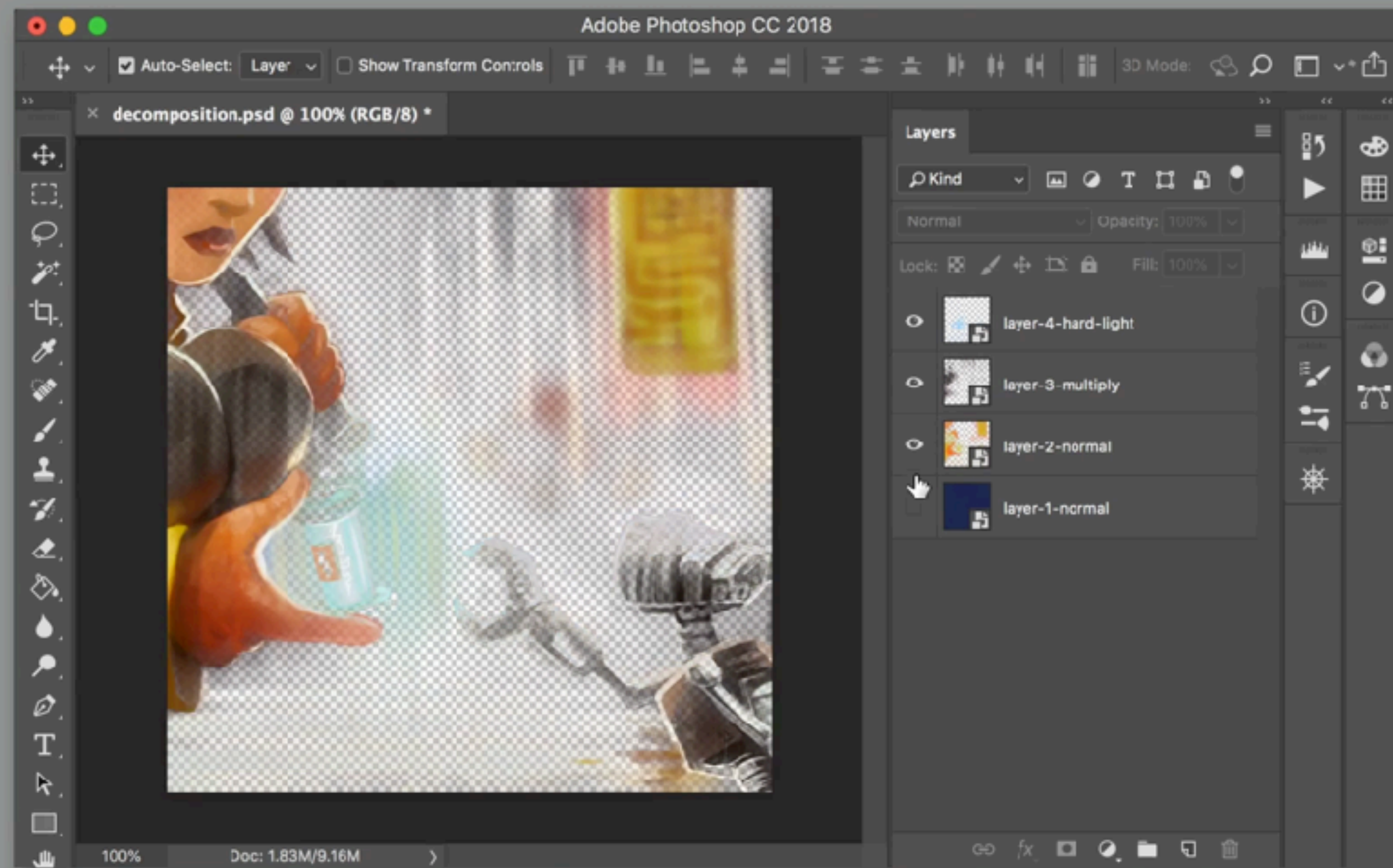
multiply



... that are associated with **advanced color-blend modes** such as **hard-light** and **multiply**.



# Usage Scenario



Import the resulting layers  
to Photoshop etc.

Perform non-trivial edits  
(e.g., lighting-aware hue change)

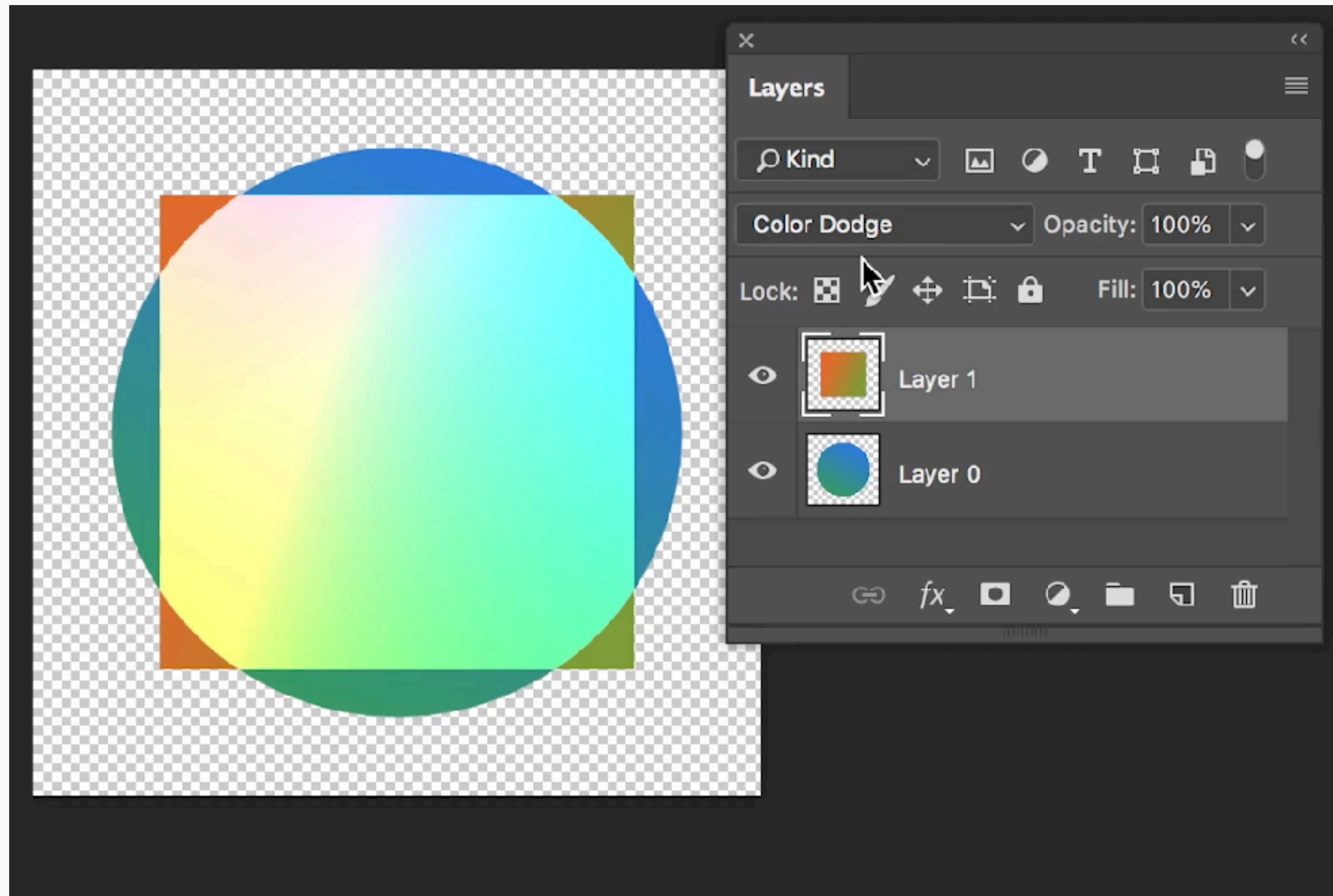


**Background**

**What is Advanced Color Blending?**



# Color Blending & Blend Mode



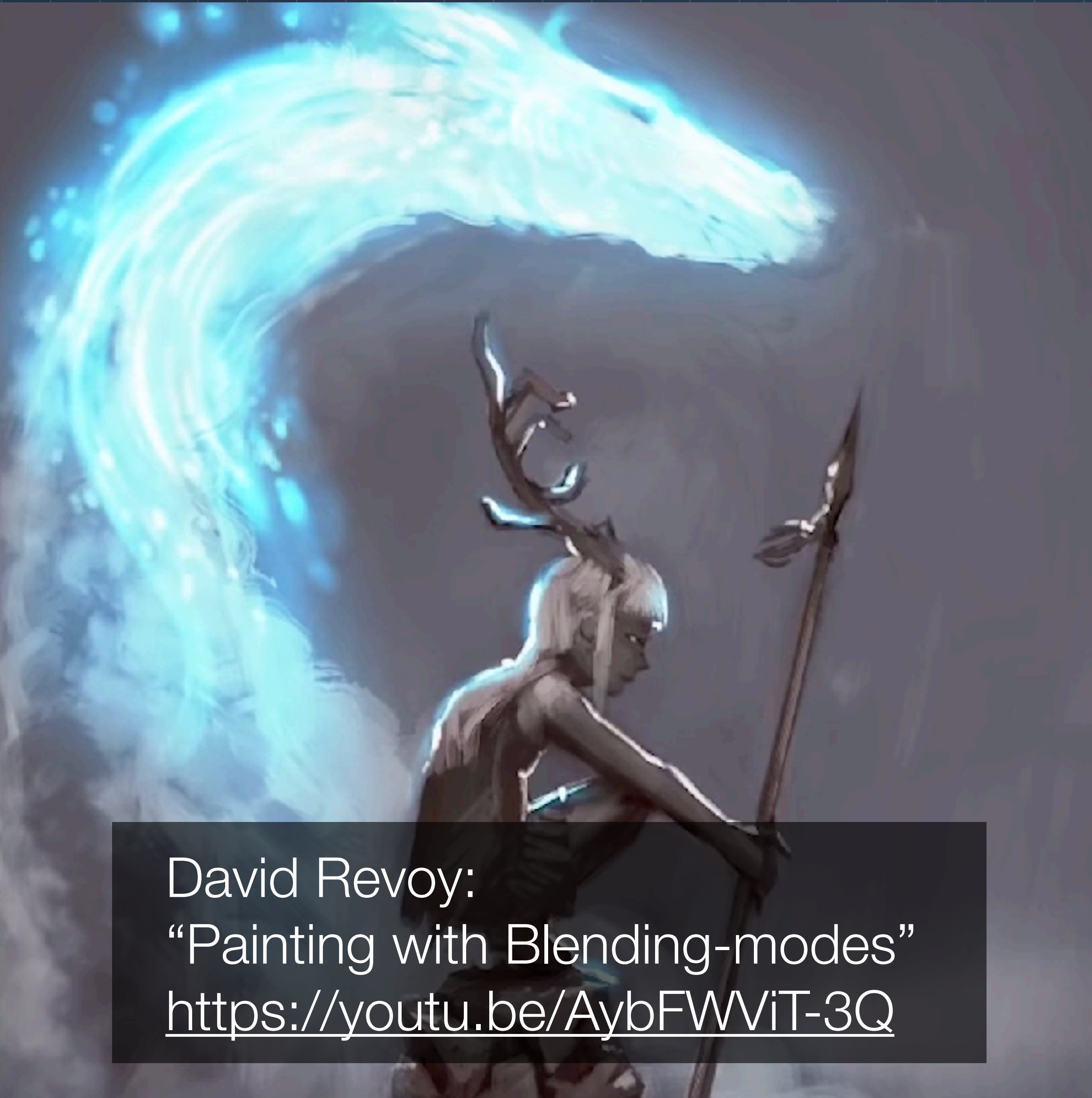
A ***color-blend mode*** defines the mapping from the colors of the layers to the color that will be rendered

Examples: **multiply, overlay, color-dodge, normal**, etc.

Available in     ...

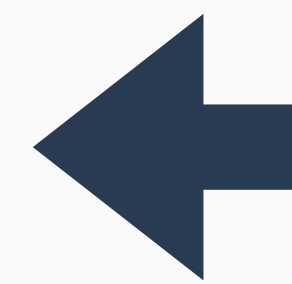


# Digital Drawing & Effects with Color Blending



David Revoy:  
“Painting with Blending-modes”  
<https://youtu.be/AybFWViT-3Q>

**Advanced (non-linear) color blending is used for creating interesting color effects**



**Example:**  
digital drawing using  
the **color-dodge** mode

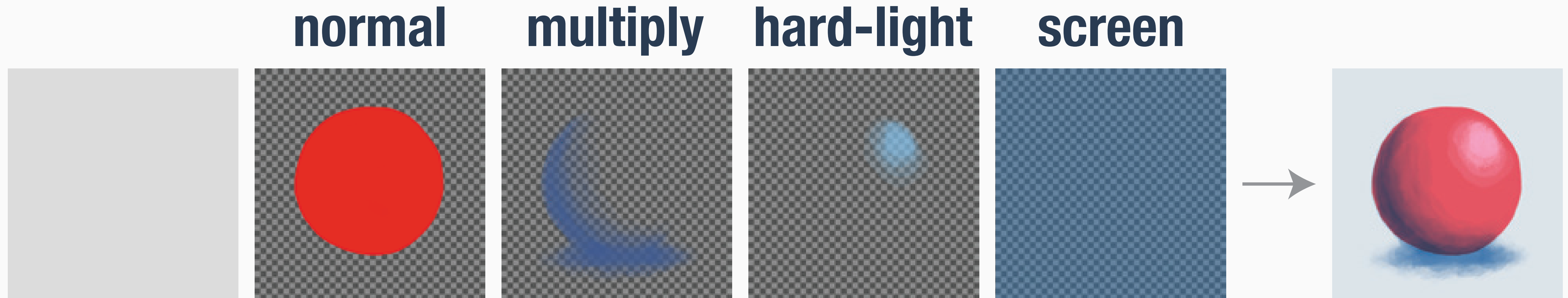


# Digital Drawing & Effects with Color Blending

**It is important to combine various blend modes**

# Digital Drawing & Effects with Color Blending

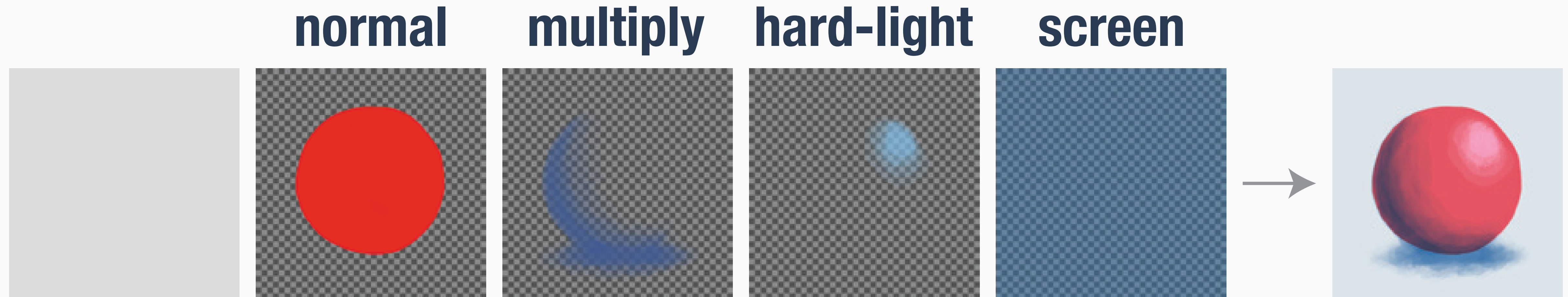
It is important to combine various blend modes





# Digital Drawing & Effects with Color Blending

It is important to combine various blend modes



Note: every digital artist has his/her own way to combine blend modes

# Our Motivation

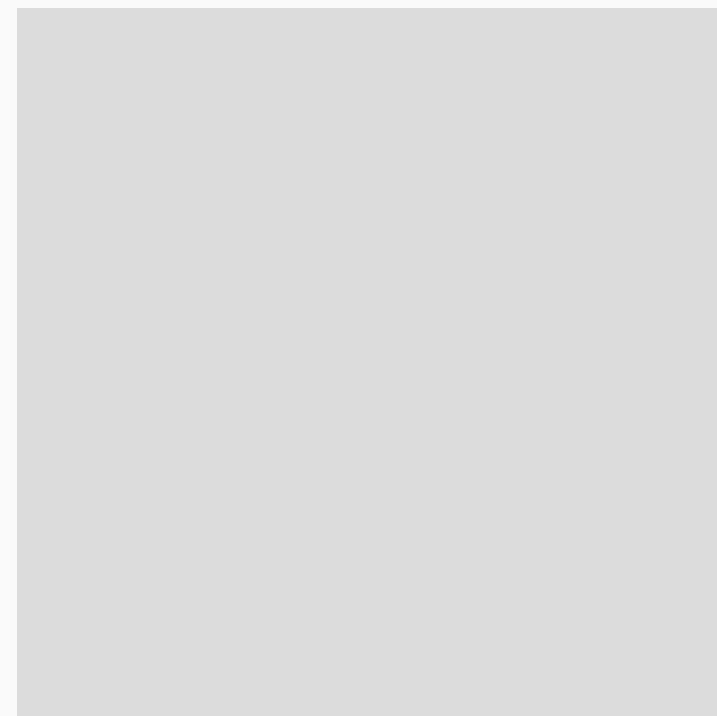
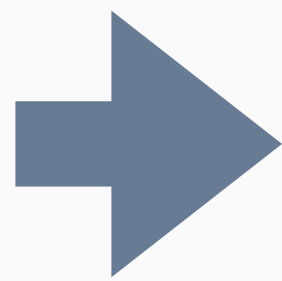
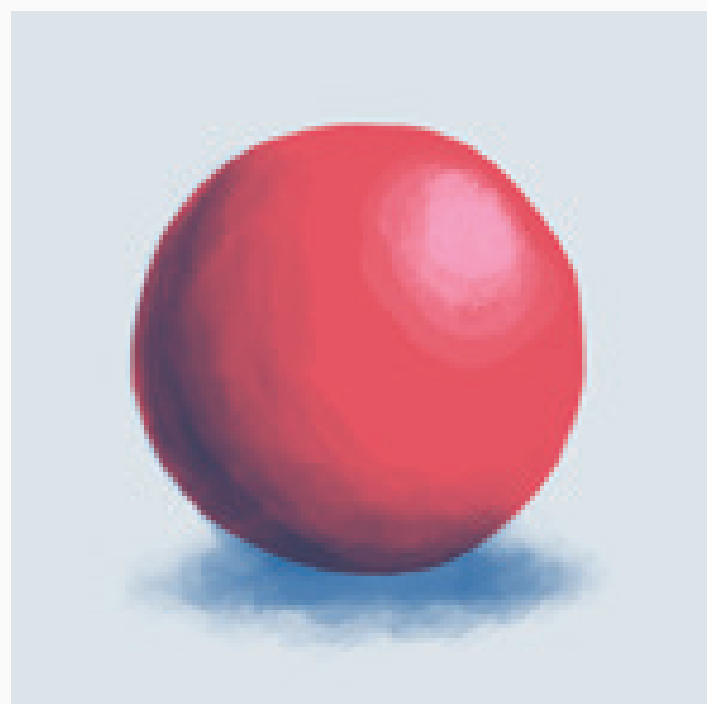


# Motivation: Color Unblending

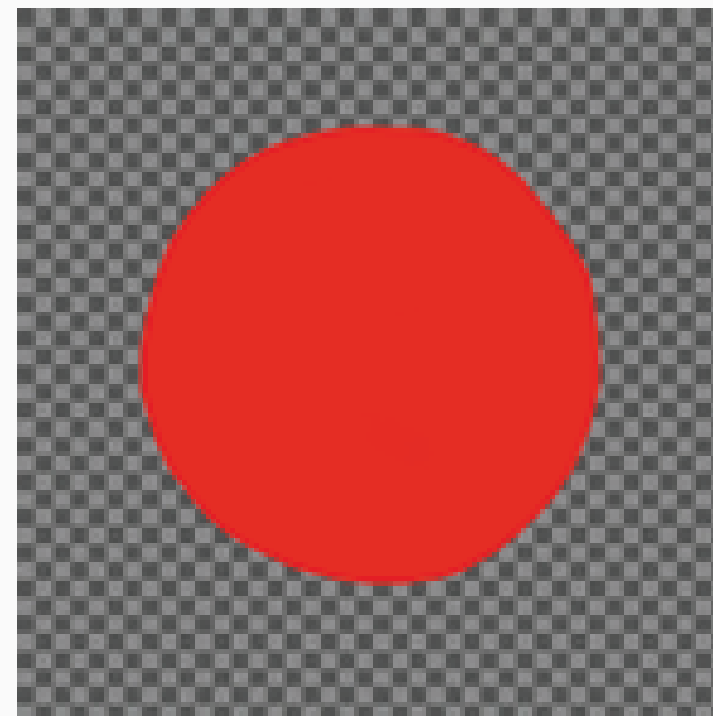
Arbitrary  
combination of  
blend modes

It should be useful if an existing image  
can be decomposed into layers with an  
arbitrary combination of advanced  
color-blend modes

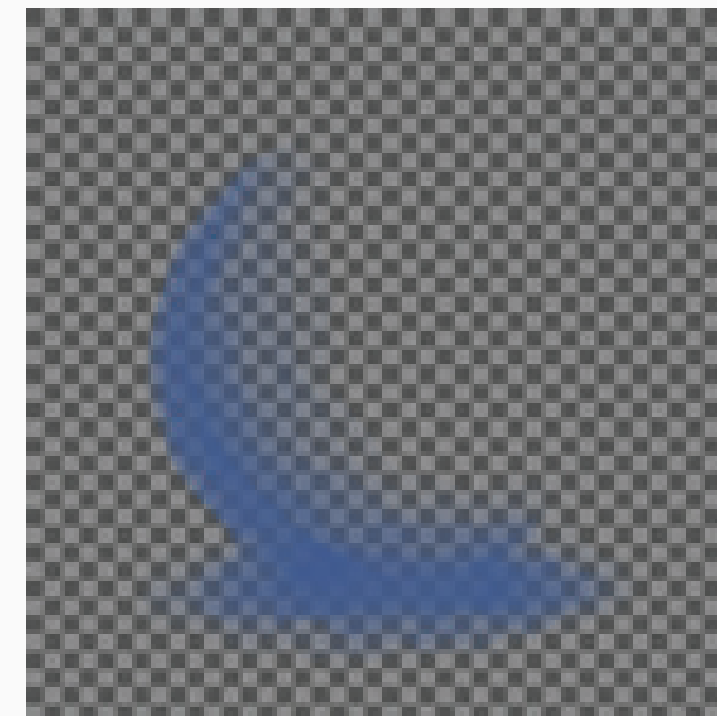
Unblending



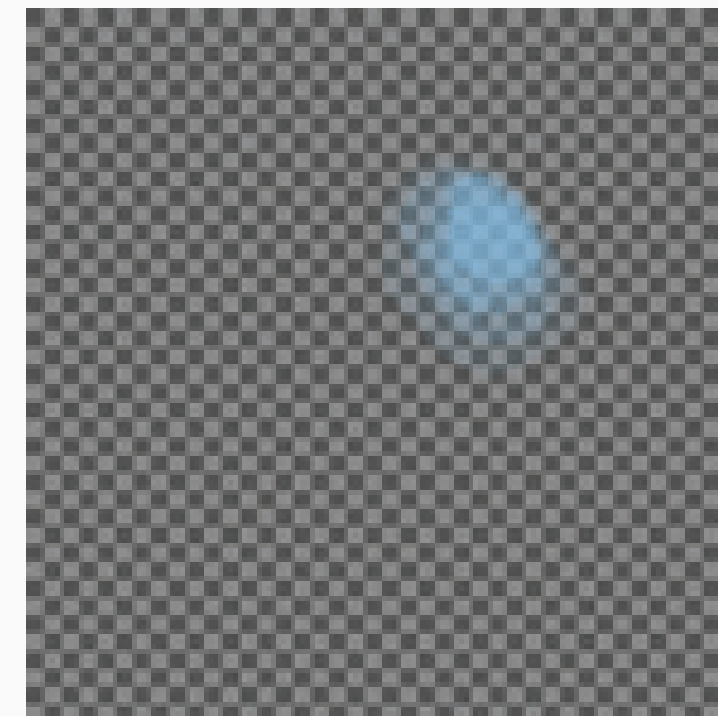
**normal**



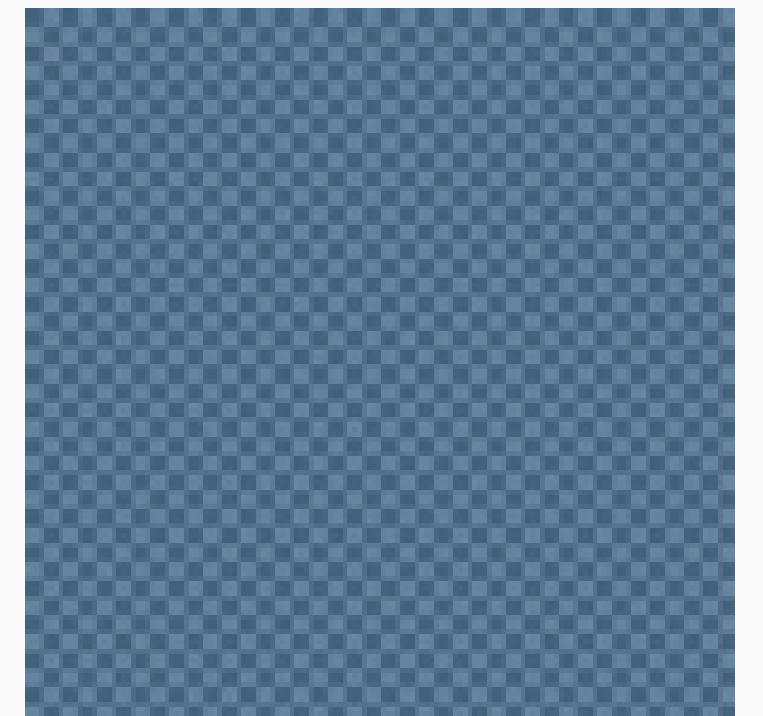
**multiply**



**hard-light**



**screen**

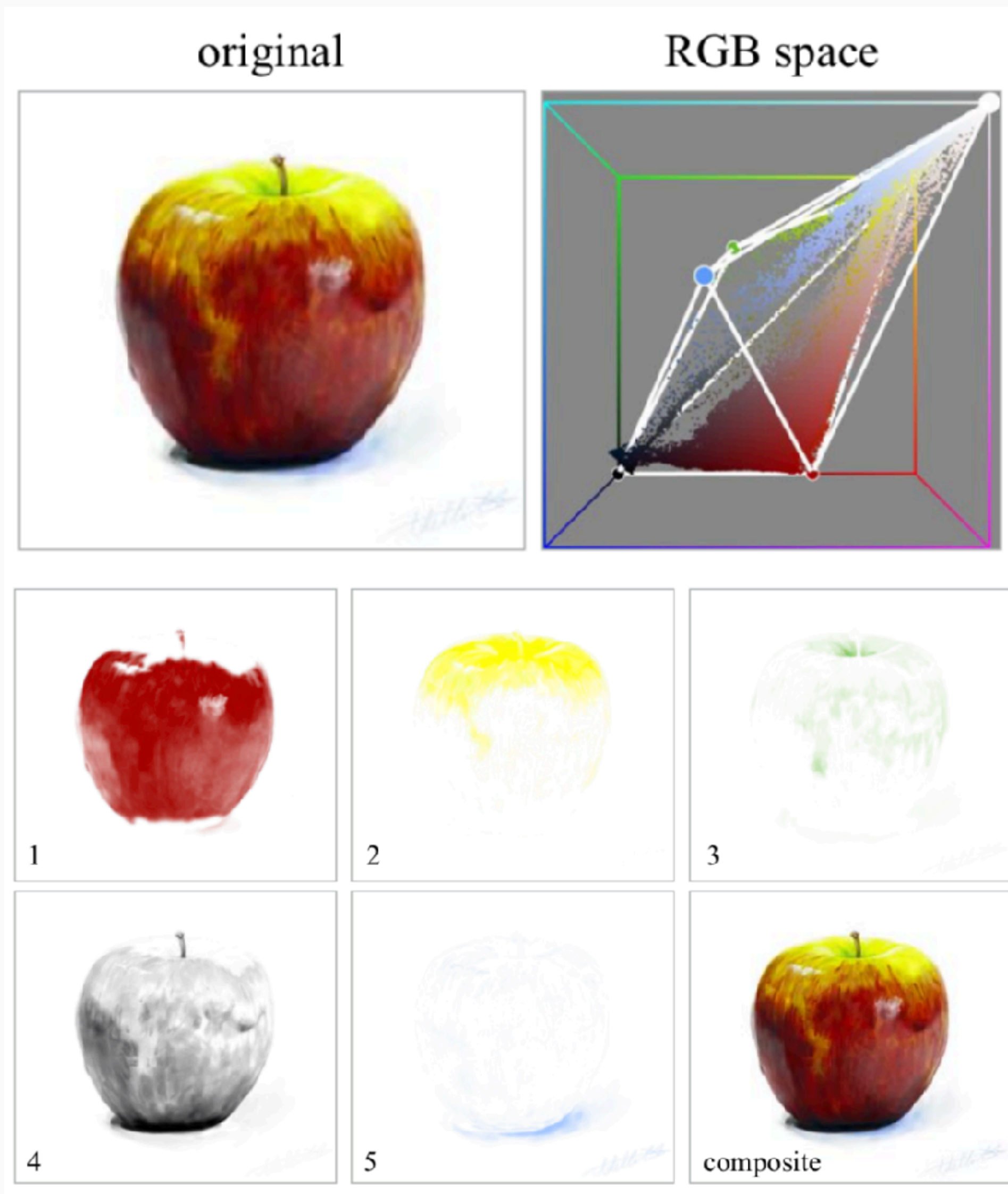


# **Previous Work**

## **Layer Decomposition by Color Unblending**



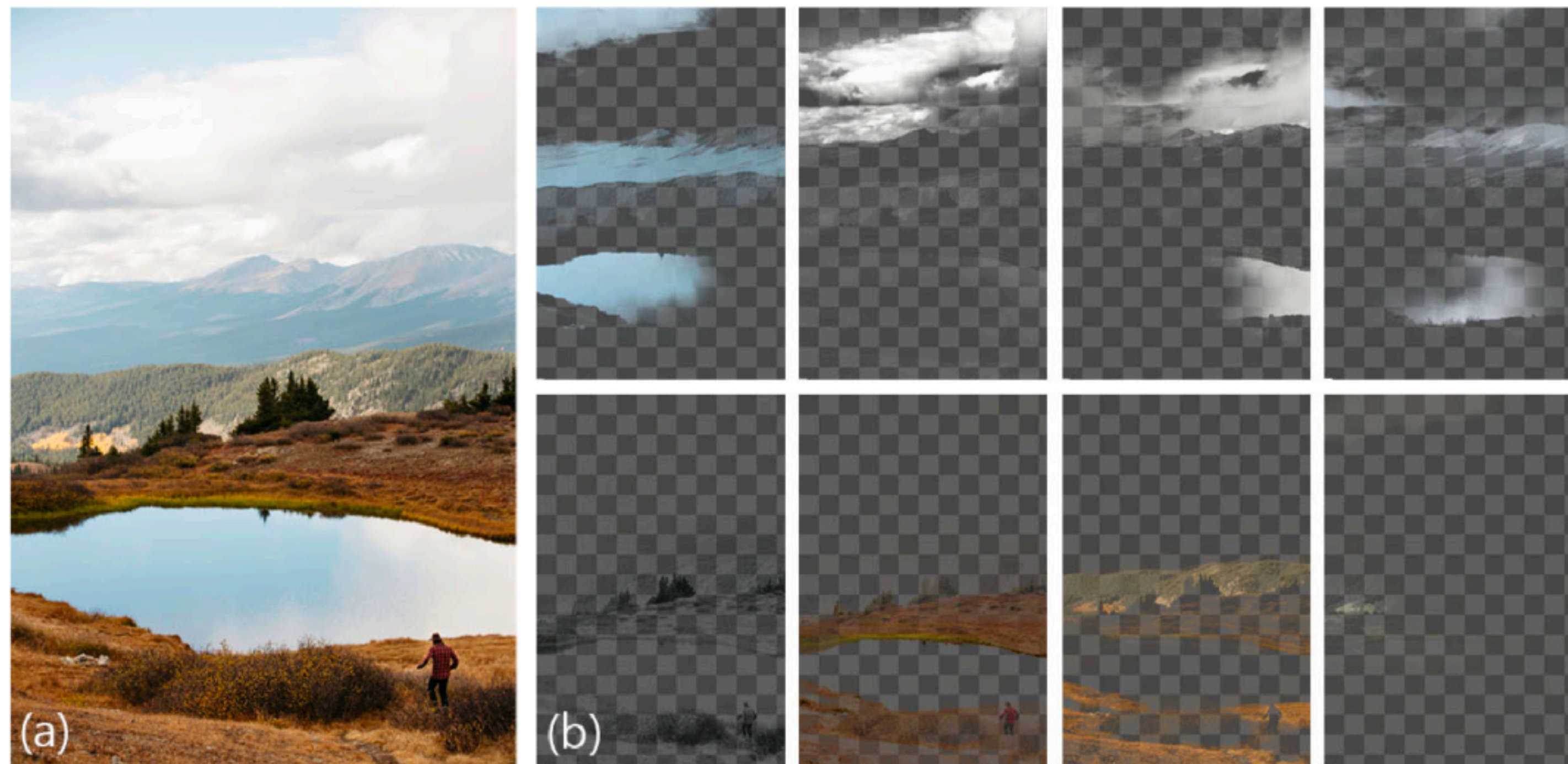
# RGB-Space Geometry [Tan+, TOG 16]



- Support **only a specific linear blend mode** (i.e., “normal”)
- Heavily **rely on the linearity** of the blend mode
- Cannot be (easily) extended for non-linear color-blend modes



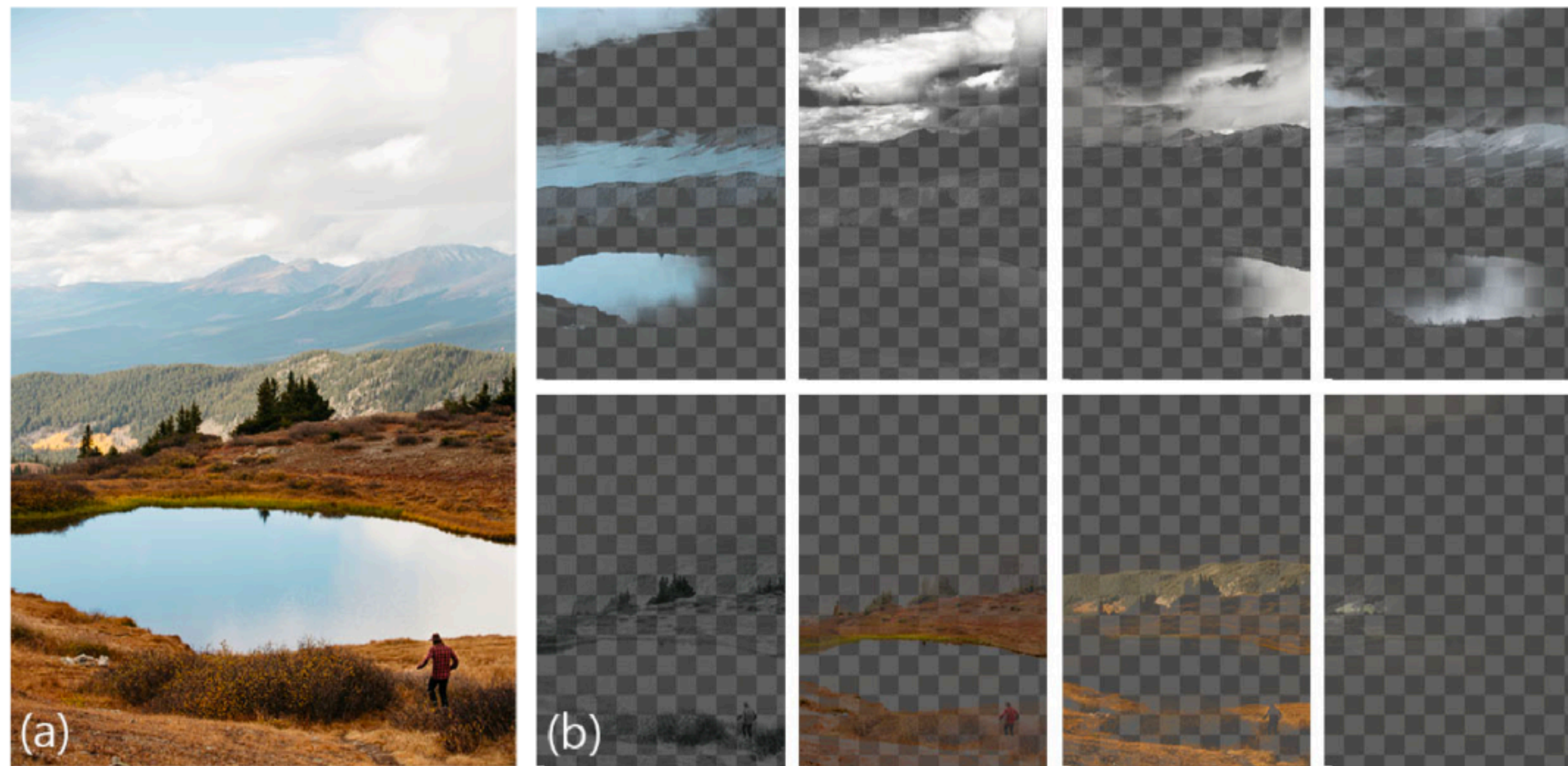
# Per-Pixel Optimization [Aksoy+, TOG 16; TOG 17]



- Support **only a specific linear blend mode** (and a linear alpha addition)



# Per-Pixel Optimization [Aksoy+, TOG 16; TOG 17]



- Support **only a specific linear blend mode** (and a linear alpha addition)

➔ We **generalize** this method to support advanced (non-linear) color-blend modes

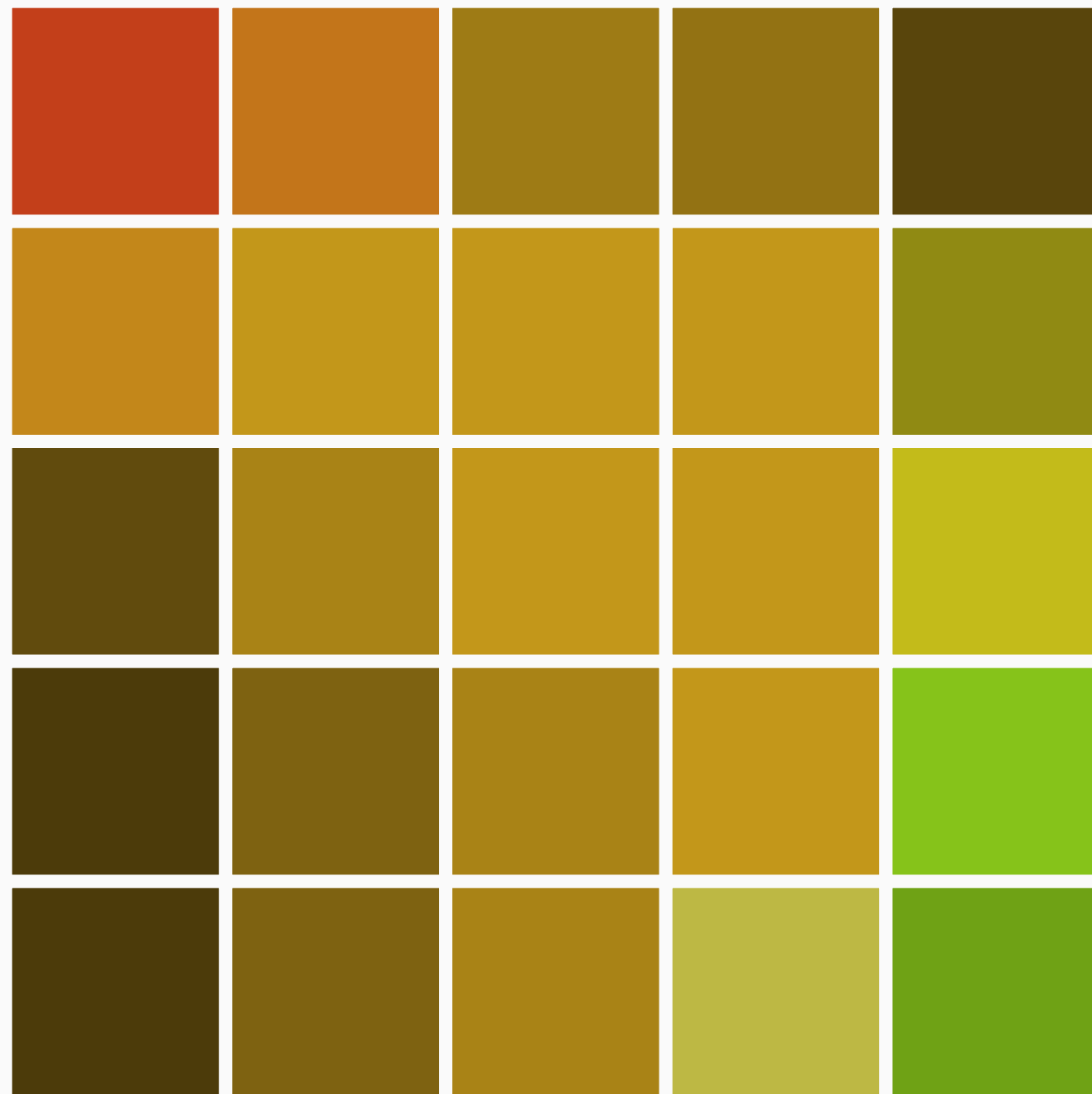
# **Details of Our Method [1/2]**

## **Per-Pixel Unblending Optimization (Concept-Level Explanation)**



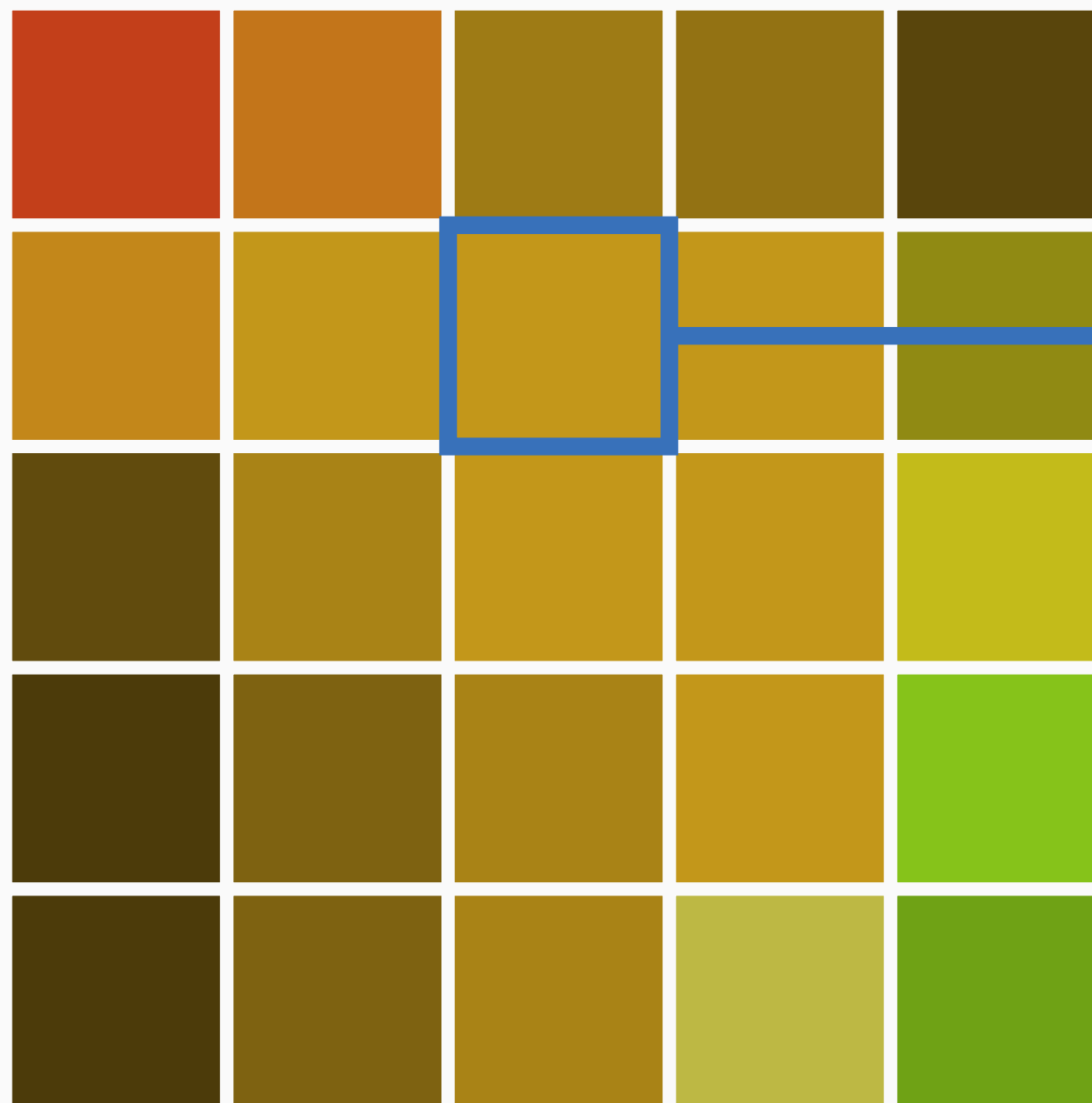
# Problem Setting: Per-Pixel Unblending

Target image

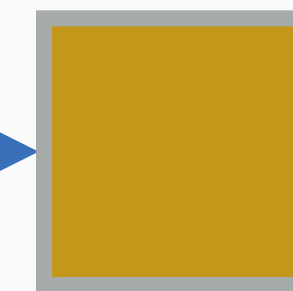


# Problem Setting: Per-Pixel Unblending

Target image



Target  
pixel color



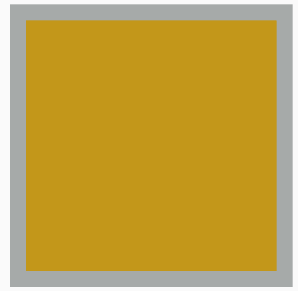
$$[r, g, b, \alpha]$$

The pixel that we are  
going to handle now



# Problem Setting: Per-Pixel Unblending

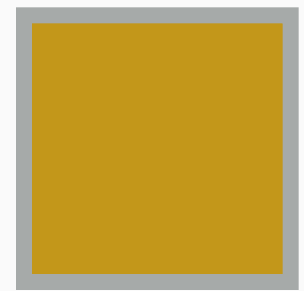
Target  
pixel color



$[r, g, b, \alpha]$

# Problem Setting: Per-Pixel Unblending

Target  
pixel color



$[r, g, b, \alpha]$

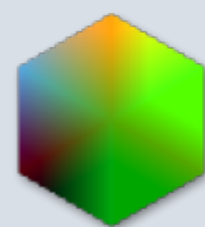
Un-  
blending

multiply



⋮

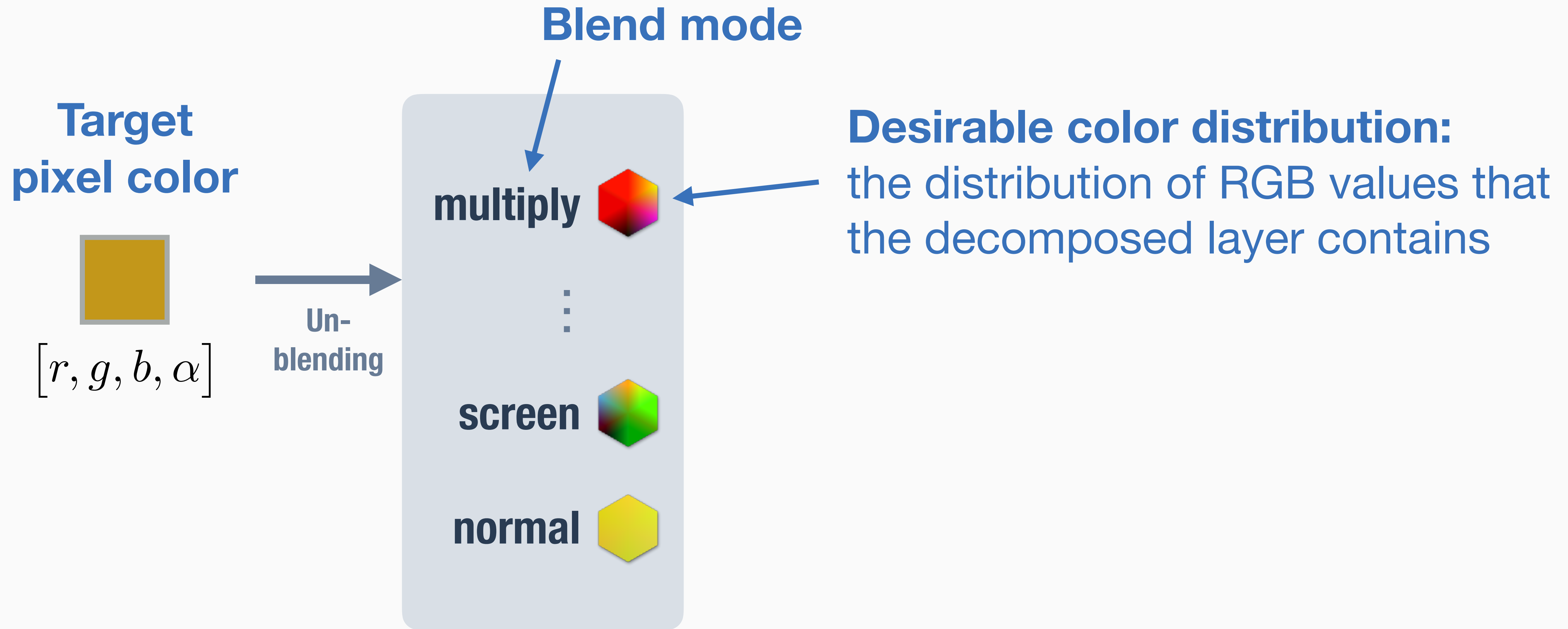
screen



normal



# Problem Setting: Per-Pixel Unblending

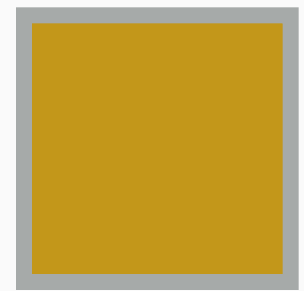




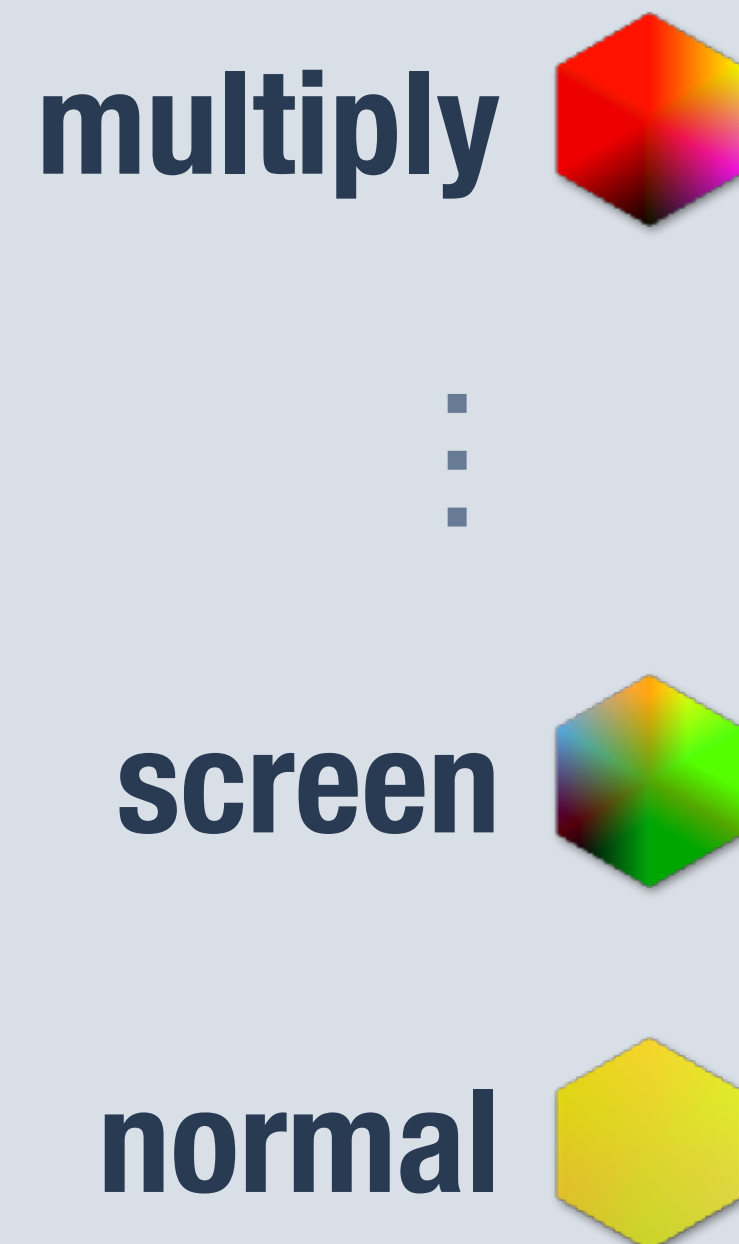
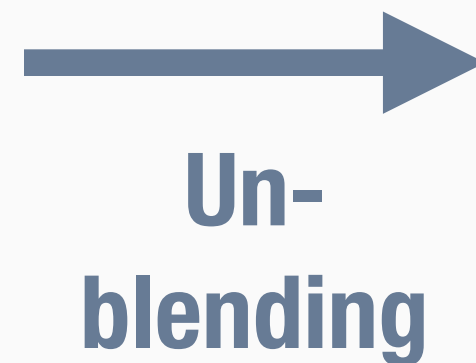
# Problem Setting: Per-Pixel Unblending

Input

Target  
pixel color



$[r, g, b, \alpha]$

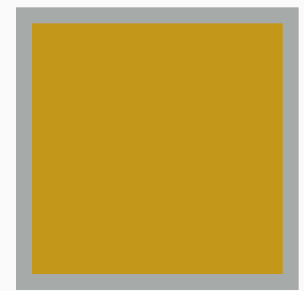


# Problem Setting: Per-Pixel Unblending

Input

Output

Target  
pixel color



$[r, g, b, \alpha]$

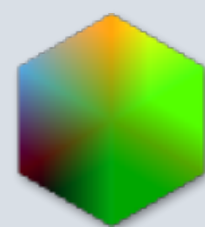
Un-  
blending

multiply



⋮

screen



normal

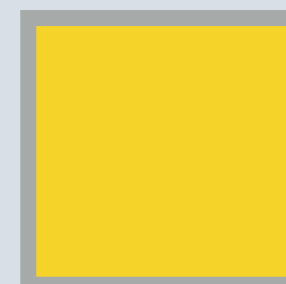


$[r_n, g_n, b_n, \alpha_n]$

⋮



$[r_2, g_2, b_2, \alpha_2]$



$[r_1, g_1, b_1, \alpha_1]$

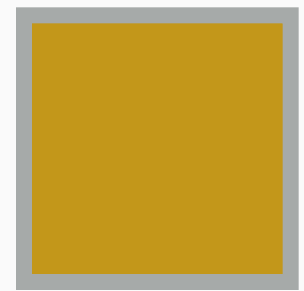


# Problem Setting: Per-Pixel Unblending

Input

Output

Target  
pixel color



$$[r, g, b, \alpha]$$

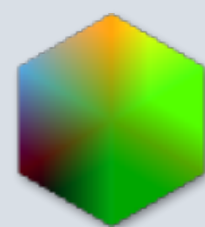
Un-  
blending

multiply



⋮

screen



normal

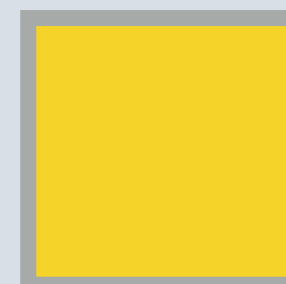


$$[r_n, g_n, b_n, \alpha_n]$$

⋮



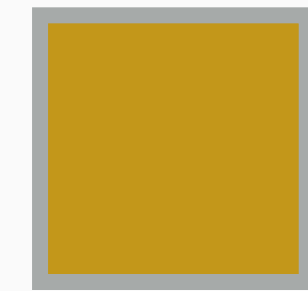
$$[r_2, g_2, b_2, \alpha_2]$$



$$[r_1, g_1, b_1, \alpha_1]$$

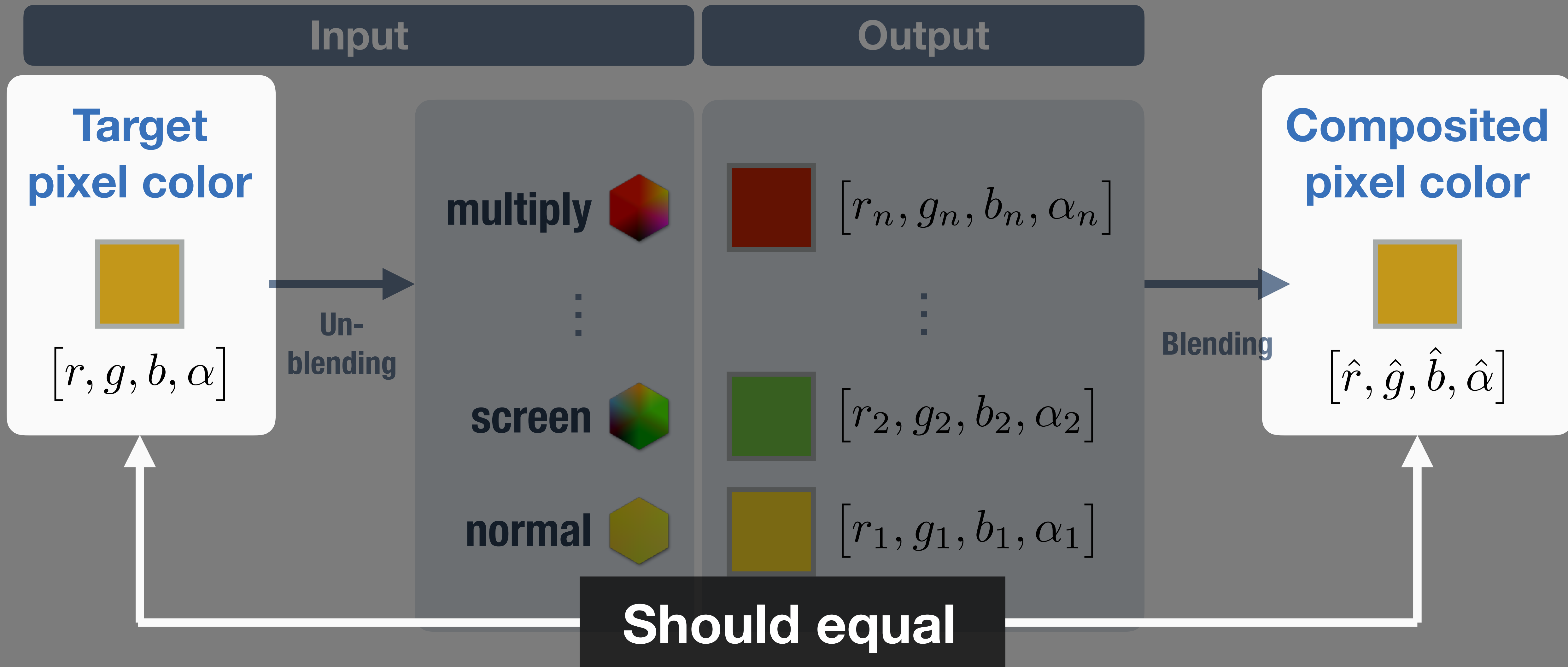
Blending

Composited  
pixel color



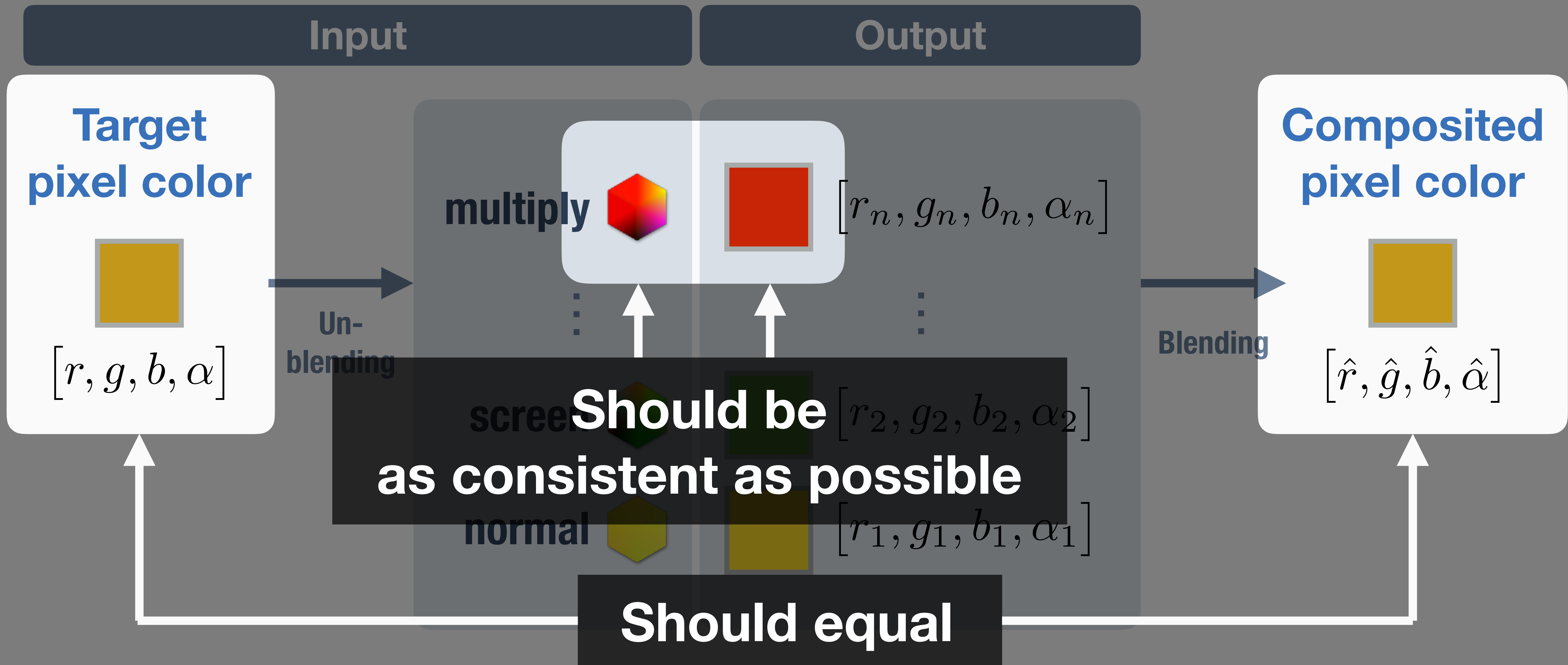
$$[\hat{r}, \hat{g}, \hat{b}, \hat{\alpha}]$$

# Problem Setting: Per-Pixel Unblending

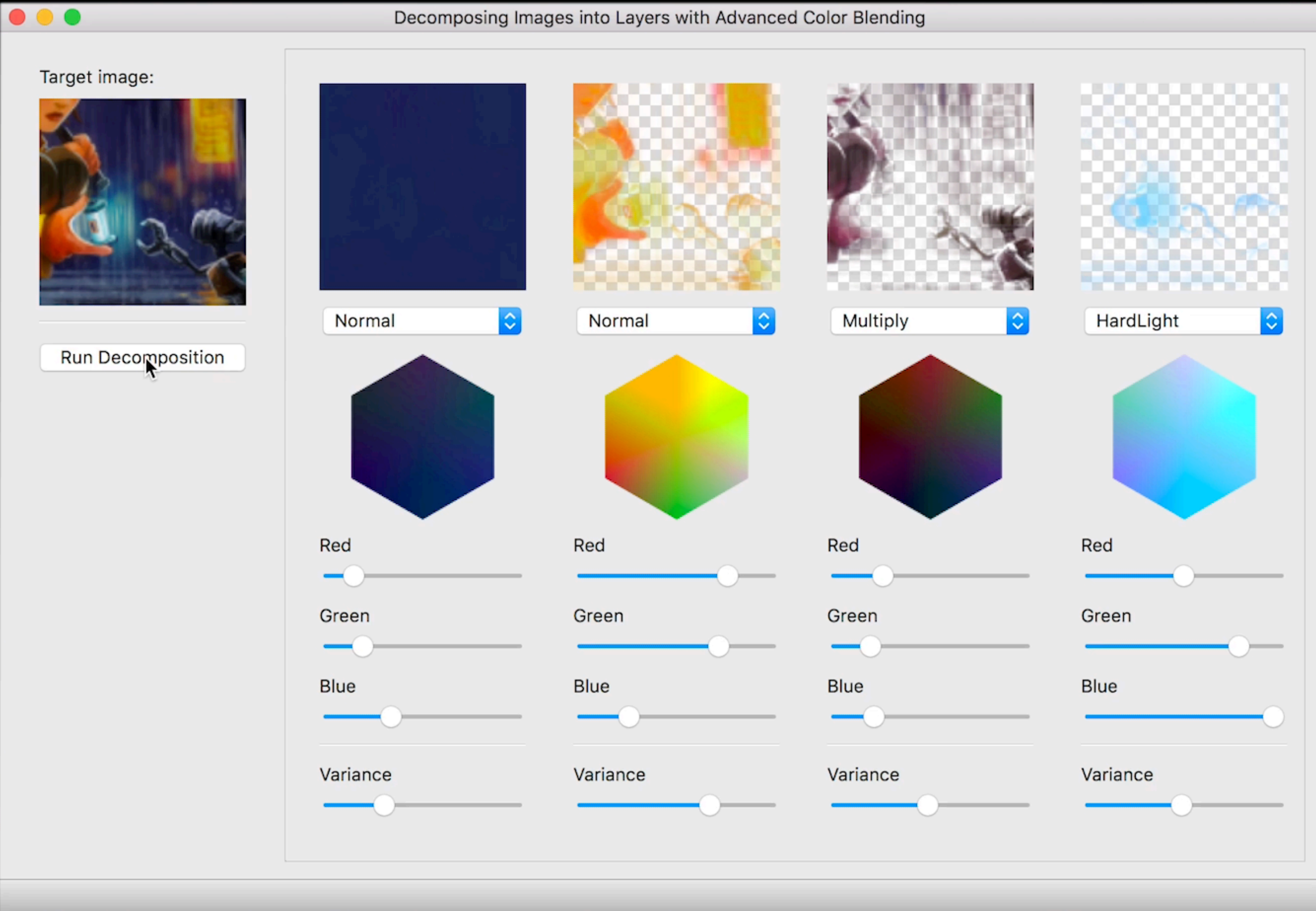




# Problem Setting: Per-Pixel Unblending



# Example User Interaction



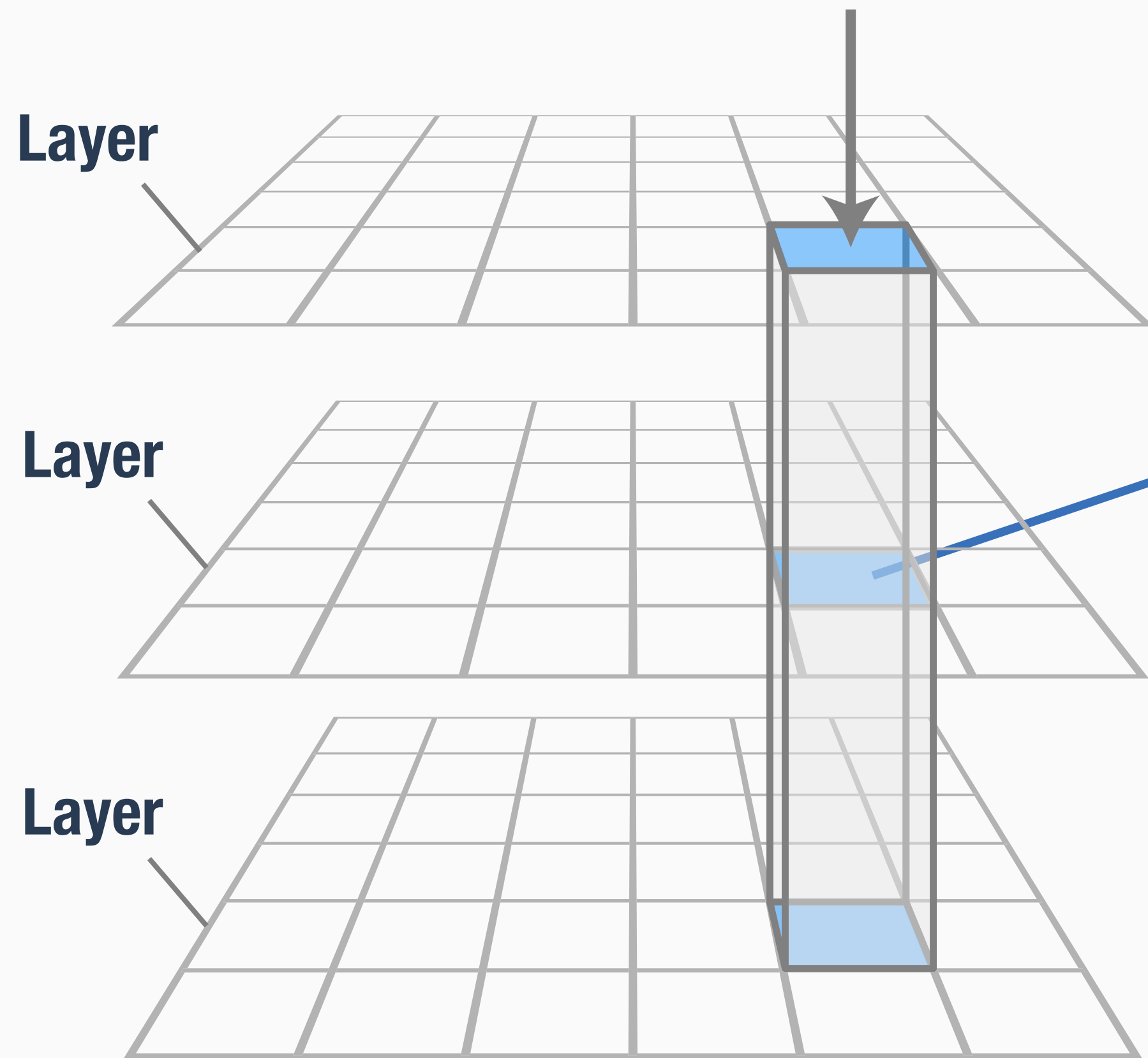


**Basics (Not Our Method)**

**Math of Advanced Color Blending**

# Descriptor of a Pixel Color

Target pixel

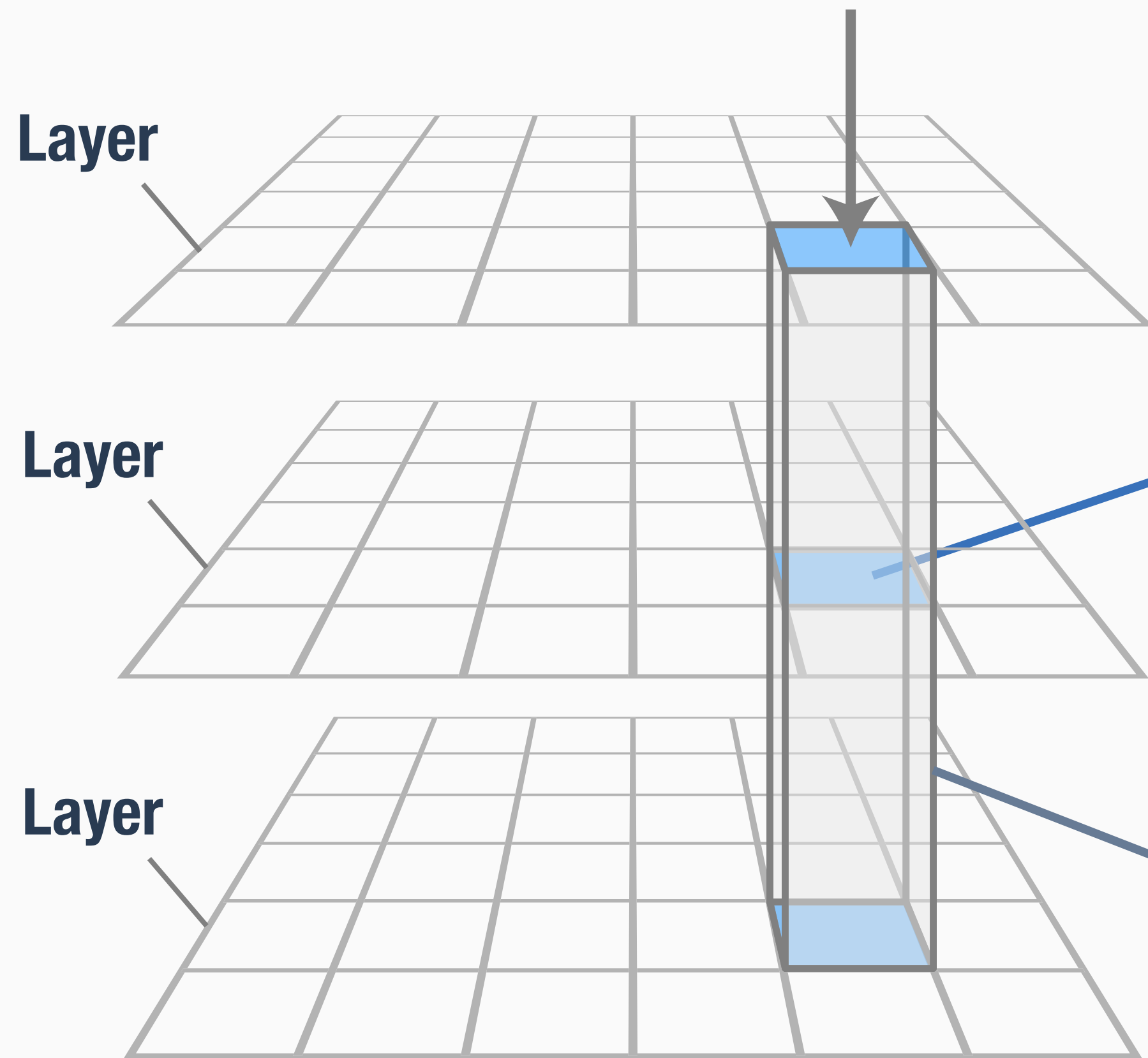


Descriptor of the  $i$ -th layer's pixel

$$\begin{aligned}\mathbf{x}_i &= [r_i \quad g_i \quad b_i \quad \alpha_i]^T \\ &= [\mathbf{c}_i^T \quad \alpha_i]^T \in \mathbb{R}^4\end{aligned}$$

# Descriptor of a Pixel Color

Target pixel



Descriptor of the  $i$ -th layer's pixel

$$\begin{aligned}\mathbf{x}_i &= [r_i \quad g_i \quad b_i \quad \alpha_i]^T \\ &= [\mathbf{c}_i^T \quad \alpha_i]^T \in \mathbb{R}^4\end{aligned}$$

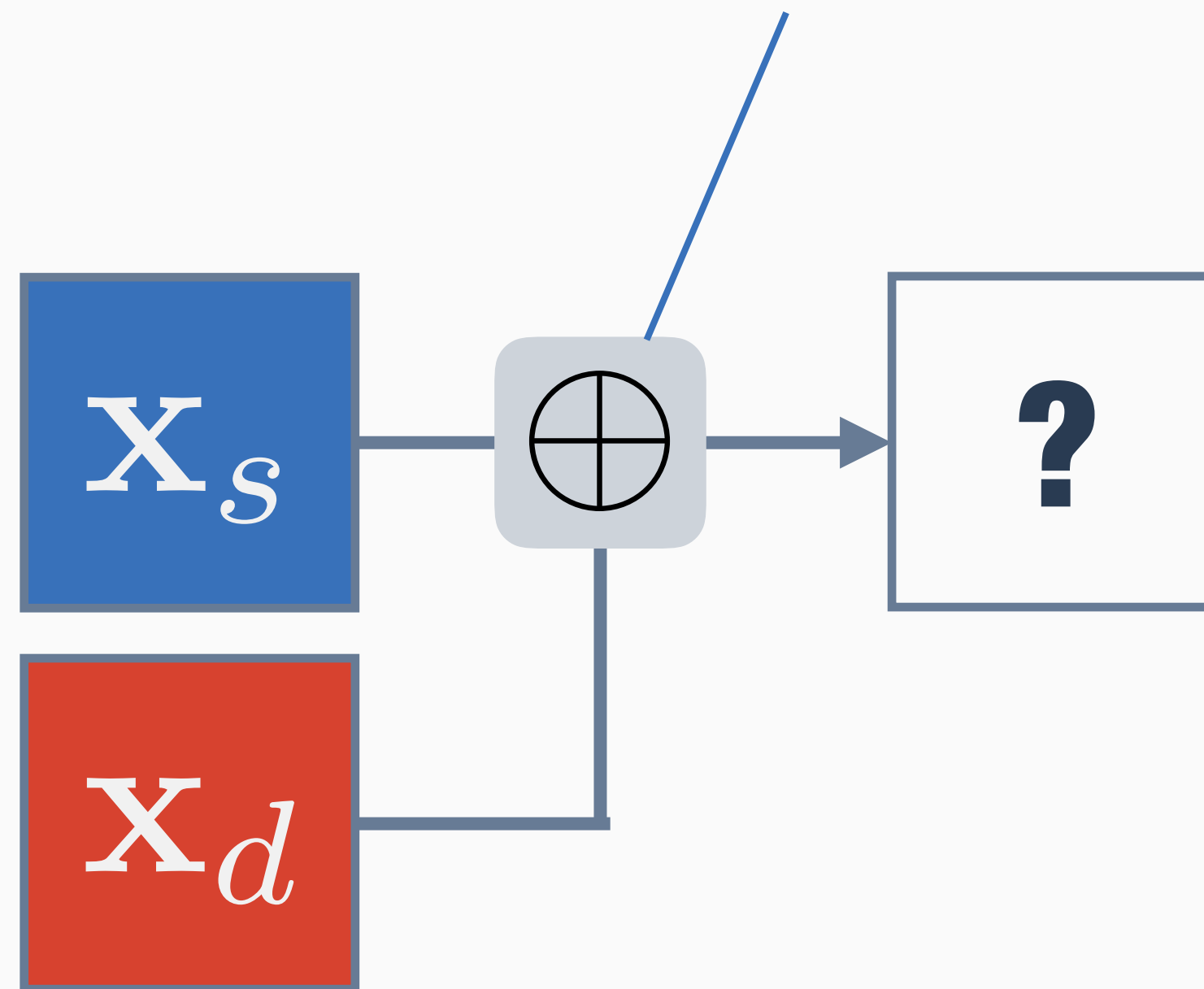
Descriptor of all the layers' pixels

$$\mathbf{x} = [\mathbf{x}_1^T \quad \dots \quad \mathbf{x}_n^T]^T \in \mathbb{R}^{4n}$$



# General Composition Operator ( $\oplus$ ) for Compositing Two Layers

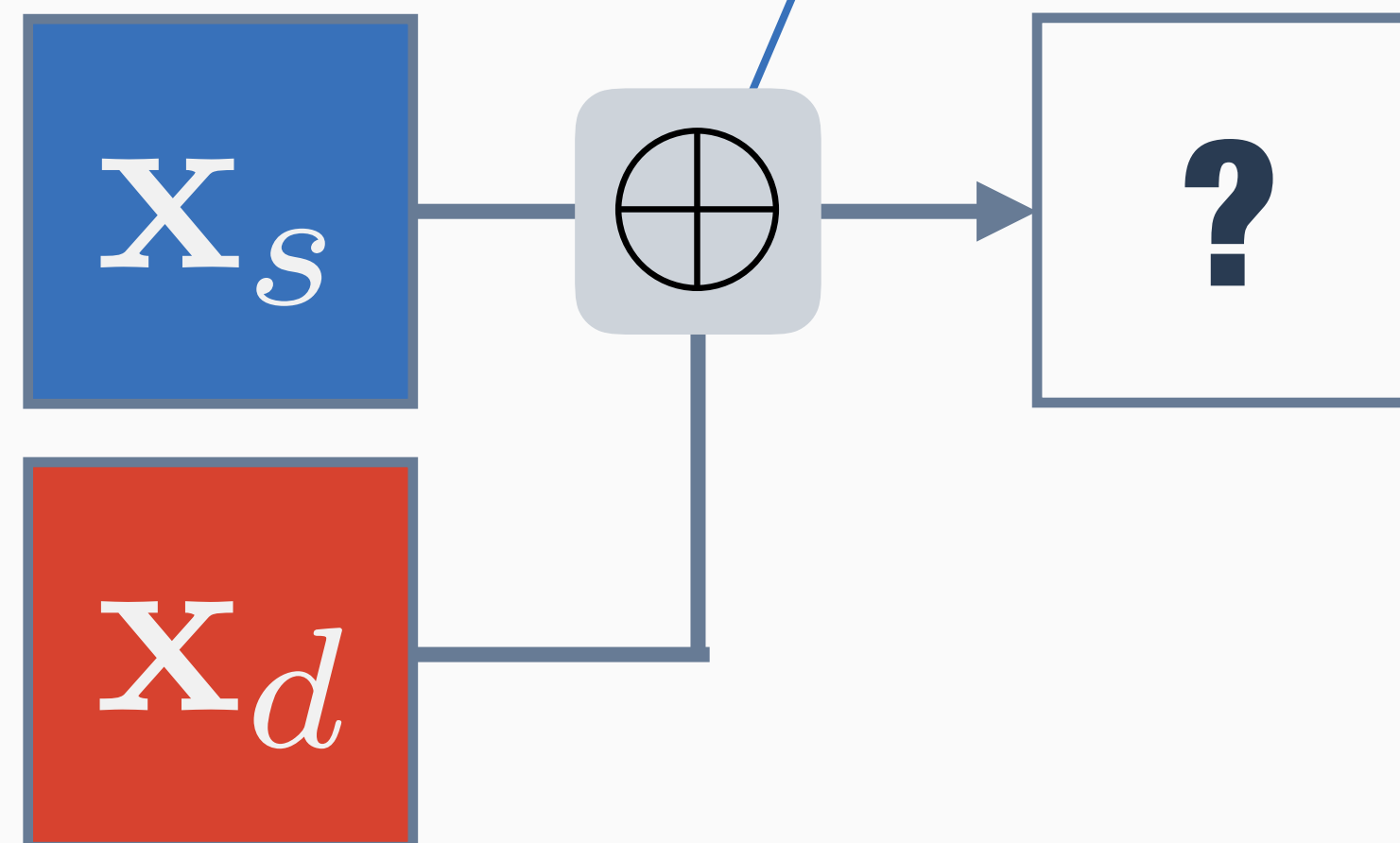
General composition operator



# General Composition Operator ( $\oplus$ ) for Compositing Two Layers

$$\mathbf{x}_s \oplus^{\text{rgb}} \mathbf{x}_d = \frac{f(\mathbf{c}_s, \mathbf{c}_d)\alpha_s\alpha_d + Y\alpha_s(1 - \alpha_d)\mathbf{c}_s + Z\alpha_d(1 - \alpha_s)\mathbf{c}_d}{\mathbf{x}_s \oplus^{\alpha} \mathbf{x}_d}$$

$$\mathbf{x}_s \oplus^{\alpha} \mathbf{x}_d = X\alpha_s\alpha_d + Y\alpha_s(1 - \alpha_d) + Z\alpha_d(1 - \alpha_s)$$



“SVG Compositing Specification” by W3C: <https://www.w3.org/TR/2011/WD-SVGCompositing-20110315/>

# General Composition Operator ( $\oplus$ ) for Compositing Two Layers

$$\mathbf{x}_s \oplus^{\text{rgb}} \mathbf{x}_d = \frac{f(\mathbf{c}_s, \mathbf{c}_d)\alpha_s\alpha_d + Y\alpha_s(1 - \alpha_d)\mathbf{c}_s + Z\alpha_d(1 - \alpha_s)\mathbf{c}_d}{\mathbf{x}_s \oplus^\alpha \mathbf{x}_d}$$
$$\mathbf{x}_s \oplus^\alpha \mathbf{x}_d = X\alpha_s\alpha_d + Y\alpha_s(1 - \alpha_d) + Z\alpha_d(1 - \alpha_s)$$

**Blend function:** Mode-specific function (can be non-linear)

$$f : [0, 1]^3 \times [0, 1]^3 \rightarrow [0, 1]^3$$

For example:  $f_{\text{normal}}$ ,  $f_{\text{multiply}}$ ,  $f_{\text{color-dodge}}$ ,  $\dots$

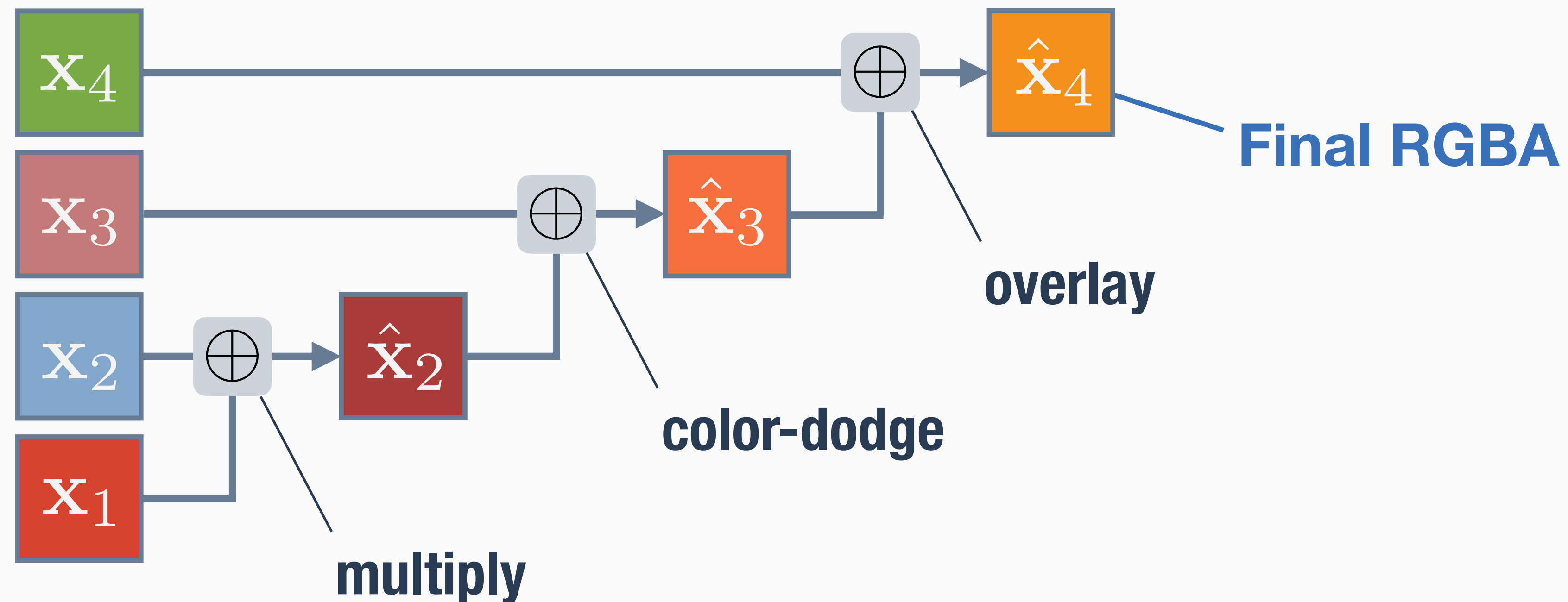
“SVG Compositing Specification” by W3C: <https://www.w3.org/TR/2011/WD-SVGCompositing-20110315/>



# More-Than-Two-Layers Case: **Recursive** Layering

**Recursive rule:**

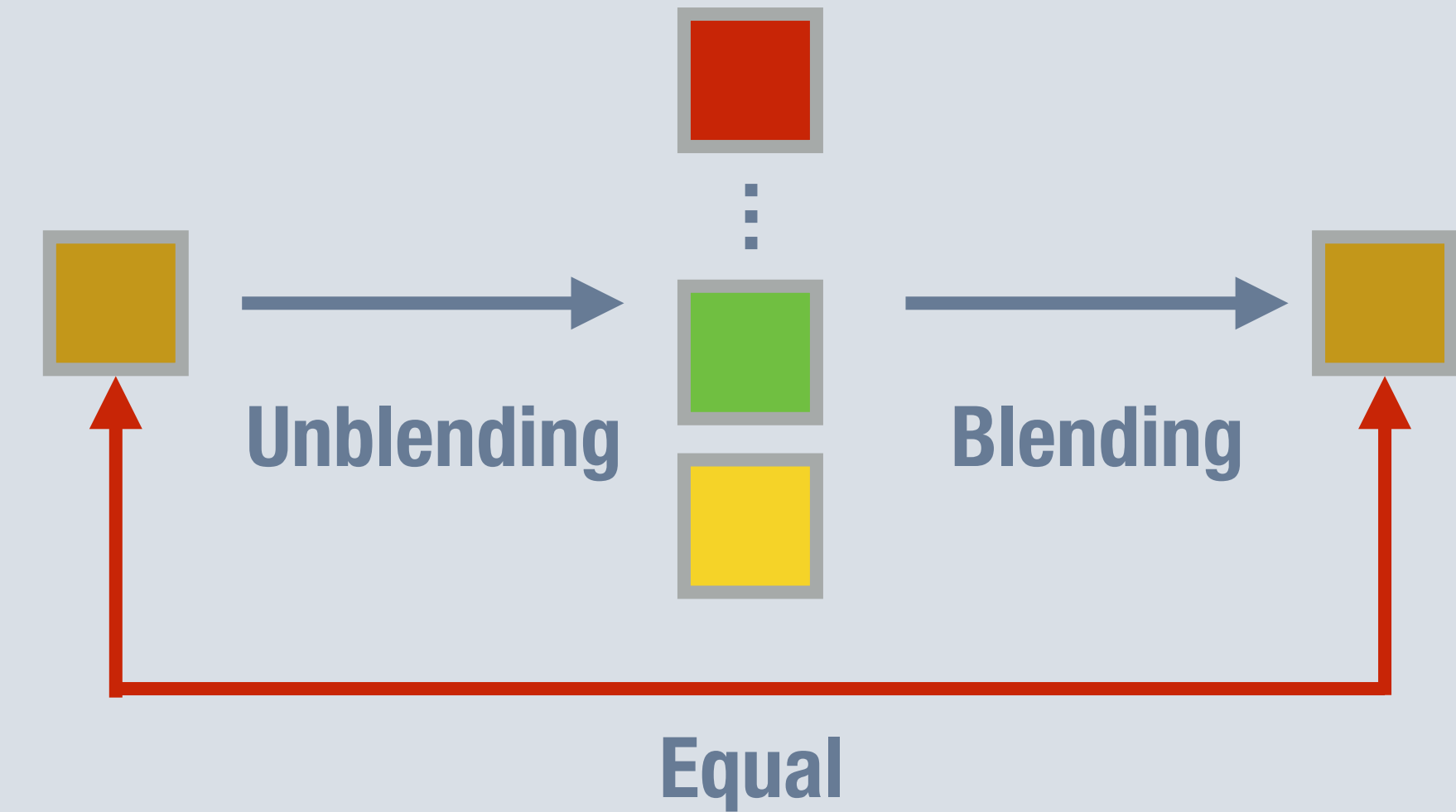
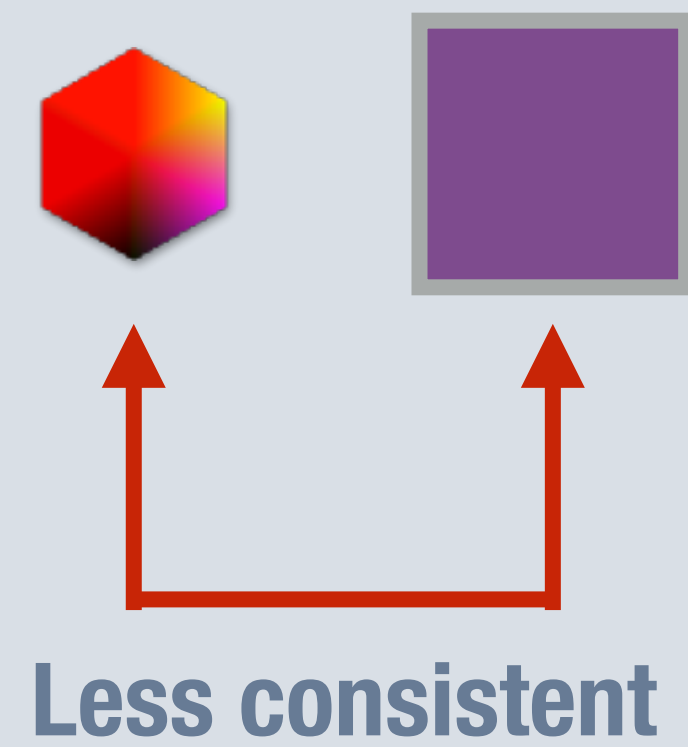
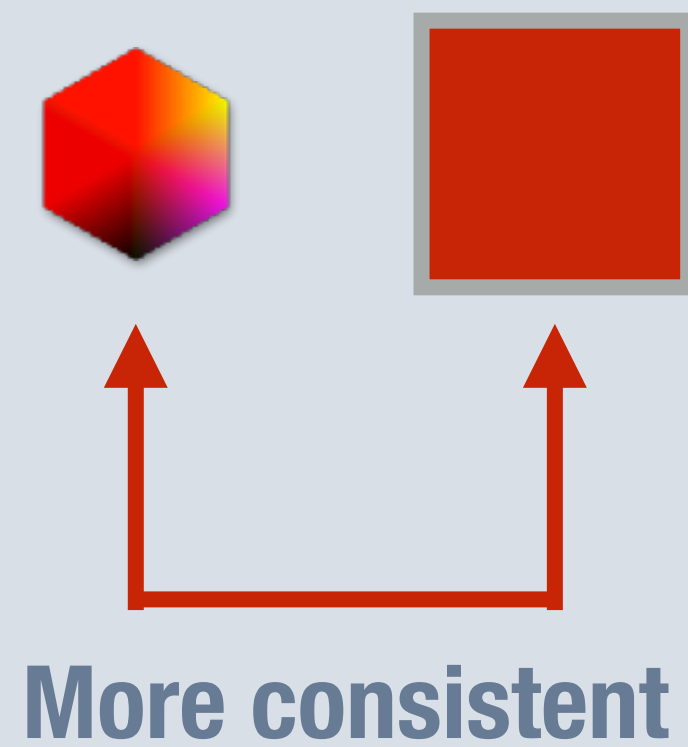
$$\hat{\mathbf{x}}_k = \begin{cases} \mathbf{x}_1 & (k = 1) \\ \mathbf{x}_k \oplus_k \hat{\mathbf{x}}_{k-1} & (\text{otherwise}) \end{cases}$$



# **Details of Our Method [2/2]**

## **Per-Pixel Unblending Optimization (Equation-Level Explanation)**

# Unblending Optimization



## Color consistency:

Each pixel color should be as consistent to the specified color distribution as possible

## Color reproducibility:

The target pixel color should be reproduced by compositing the resulting layer colors



# Unblending Optimization

## Equality-constrained optimization problem

$$\min_{\mathbf{x} \in \mathbb{R}^{4n}} E(\mathbf{x}) \quad \text{s.t.} \quad \mathbf{C}(\mathbf{x}) = \mathbf{0}$$

### Objective:

Each pixel color should be as consistent to the specified color distribution as possible

### Equality constraint:

The target pixel color should be reproduced by compositing the resulting layer colors

# Unblending Optimization

## Equality-constrained optimization problem

$$\min_{\mathbf{x} \in \mathbb{R}^{4n}} E(\mathbf{x}) \quad \text{s.t.} \quad \mathbf{C}(\mathbf{x}) = \mathbf{0}$$

### Objective:

Each pixel color should be as consistent to the specified color distribution as possible

### Equality constraint:

The target pixel color should be reproduced by compositing the resulting layer colors

# Constraint Function Definition

$$\mathbf{C}(\mathbf{x}) = \left( \mathbf{x}_n \oplus_n \left( \mathbf{x}_{n-1} \oplus_{n-1} \left( \mathbf{x}_{n-2} \oplus_{n-2} (\dots) \right) \right) \right) - \mathbf{c}^{\text{target}} = \mathbf{0}$$

**E.g., multiply      color-dodge      overlay ...**



# Constraint Function Definition

$$\mathbf{C}(\mathbf{x}) = \left( \mathbf{x}_n \oplus_n \left( \mathbf{x}_{n-1} \oplus_{n-1} \left( \mathbf{x}_{n-2} \oplus_{n-2} (\dots) \right) \right) \right) - \mathbf{c}^{\text{target}} = \mathbf{0}$$

E.g., multiply      color-dodge      overlay ...

**C.f., [Aksoy+16; 17]:** Assumed a **special case** of a linear model

$$\mathbf{C}(\mathbf{x}) = \sum_i^n \alpha_i \mathbf{c}_i - \mathbf{c}^{\text{target}} = \mathbf{0}$$

# Constraint Function Definition

$$\mathbf{C}(\mathbf{x}) = \left( \mathbf{x}_n \oplus_n \left( \mathbf{x}_{n-1} \oplus_{n-1} \left( \mathbf{x}_{n-2} \oplus_{n-2} (\dots) \right) \right) \right) - \mathbf{c}^{\text{target}} = \mathbf{0}$$

E.g., multiply      color-dodge      overlay ...

**C.f., [Aksoy+16; 17]:** Assumed a **special case** of a linear model

$$\mathbf{C}(\mathbf{x}) = \sum_i^n \alpha_i \mathbf{c}_i - \mathbf{c}^{\text{target}} = \mathbf{0}$$

Our method is a **generalization** of [Aksoy+16; 17]

# Solving Equality-Constrained Optimization [1/2]

## Algorithm: The augmented Lagrangian method

- Iteratively solve (unconstrained) optimizations

```
while (not converged) {  
    solve  $\min_{\mathbf{x}} \left\{ E(\mathbf{x}) - \boldsymbol{\lambda}^T \mathbf{C}(\mathbf{x}) + \frac{\rho}{2} \|\mathbf{C}(\mathbf{x})\|^2 \right\}$   
    update  $\boldsymbol{\lambda}$  and  $\rho$   
}
```



# Solving Equality-Constrained Optimization [1/2]

## Algorithm: The augmented Lagrangian method

- Iteratively solve (unconstrained) optimizations

```
while (not converged) {  
    solve  $\min_{\mathbf{x}} \left\{ E(\mathbf{x}) - \boldsymbol{\lambda}^T \mathbf{C}(\mathbf{x}) + \frac{\rho}{2} \|\mathbf{C}(\mathbf{x})\|^2 \right\}$   
    update  $\boldsymbol{\lambda}$  and  $\rho$   
}
```

Need the  
**derivative of  $\mathbf{C}(\mathbf{x})$**   
to efficiently solve  
this optimization

# Solving Equality-Constrained Optimization [2/2]

It is possible to algorithmically calculate the derivative of  $\mathbf{C}(\mathbf{x})$  by **recursively** applying the **chain rule**

$$\frac{\partial \mathbf{C}(\mathbf{x})}{\partial \mathbf{x}_i} = \begin{cases} \mathbf{I} & (i = k = 1) \\ \frac{\partial}{\partial \mathbf{x}_k} (\mathbf{x}_k \oplus_k \hat{\mathbf{x}}_{k-1}) & (i = k \neq 1) \\ \frac{\partial \hat{\mathbf{x}}_{k-1}}{\partial \mathbf{x}_i} \cdot \frac{\partial}{\partial \hat{\mathbf{x}}_{k-1}} (\mathbf{x}_k \oplus_k \hat{\mathbf{x}}_{k-1}), & (\text{otherwise}) \end{cases}$$

# Solving Equality-Constrained Optimization [2/2]

It is possible to algorithmically calculate the derivative of  $\mathbf{C}(\mathbf{x})$  by **recursively** applying the **chain rule**

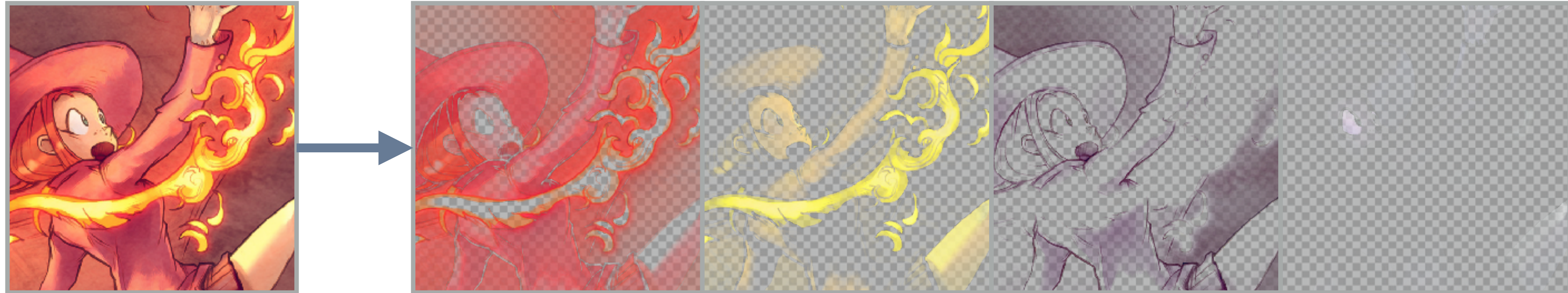
$$\frac{\partial \mathbf{C}(\mathbf{x})}{\partial \mathbf{x}_i} = \begin{cases} \mathbf{I} & (i = k = 1) \\ \frac{\partial}{\partial \mathbf{x}_k} (\mathbf{x}_k \oplus_k \hat{\mathbf{x}}_{k-1}) & (i = k \neq 1) \\ \frac{\partial \hat{\mathbf{x}}_{k-1}}{\partial \mathbf{x}_i} \cdot \frac{\partial}{\partial \hat{\mathbf{x}}_{k-1}} (\mathbf{x}_k \oplus_k \hat{\mathbf{x}}_{k-1}), & (\text{otherwise}) \end{cases}$$

➡ Our unblending optimization problem can be efficiently solved by **gradient-based algorithms** (e.g., L-BFGS)



# Unblending using linear blend mode

➔ Identical to the results by [Aksoy+17]



# Unblending using various blend modes



**color-dodge**

**multiply**

**overlay**



# **Post-Processing**

## **Refinement for Smoothness**

# Refinement Step for Smoothness

## ■ Problem:

Resulting layers may not be smooth because the unblending optimization is “per-pixel” (no inter-pixel consideration)





# Refinement Step for Smoothness

## ■ Problem:

Resulting layers may not be smooth because the unblending optimization is “per-pixel” (no inter-pixel consideration)



## ■ Solution:

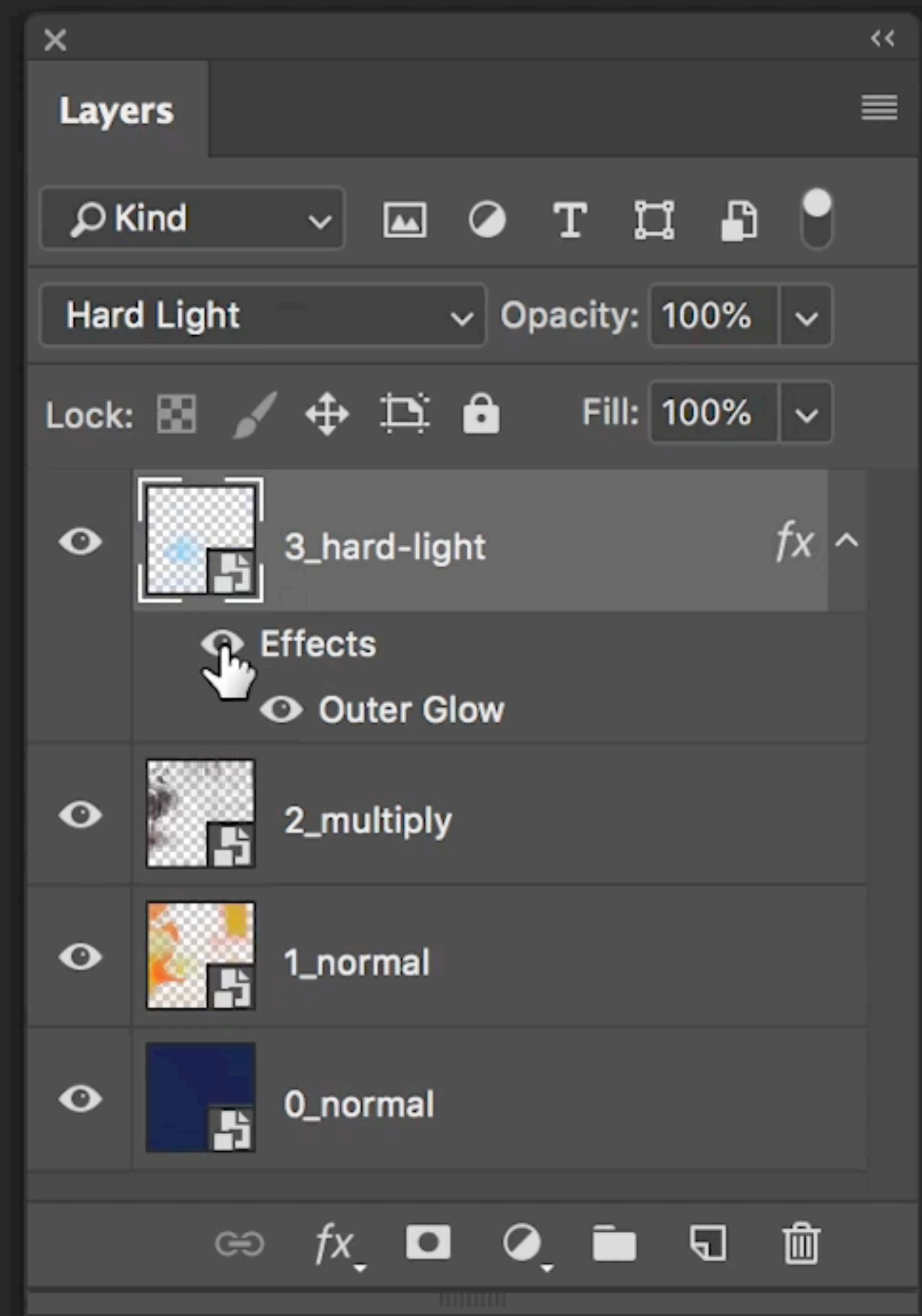
Performing a procedure called “**matte refinement**” [Aksoy+17] as a post processing to enforce smoothness



# Usage Scenario

## Example: Photoshop





Input image courtesy of David Revoy

# **Applications #1**

**Lighting-Aware Editing of  
an Existing Illustration**





**Input Image**



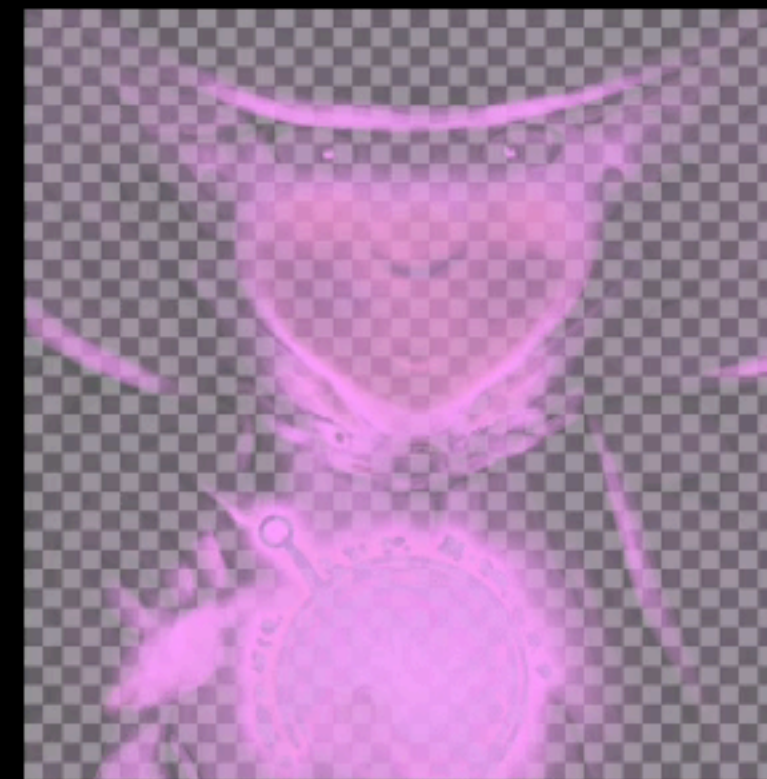
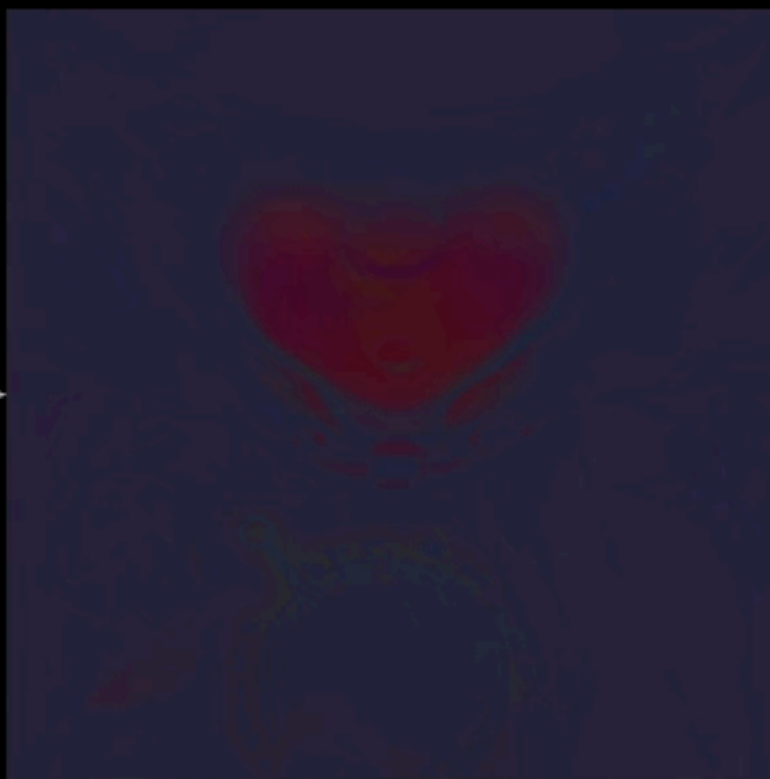
**Layer-Based Edit Example**

(N/A)

normal

hard-light

multiply



**Decomposed Layers**



# **Applications #2**

**Bringing a Physical Painting  
into Digital Workflow**





**Input Image**



**Layer-Based Edit Example**

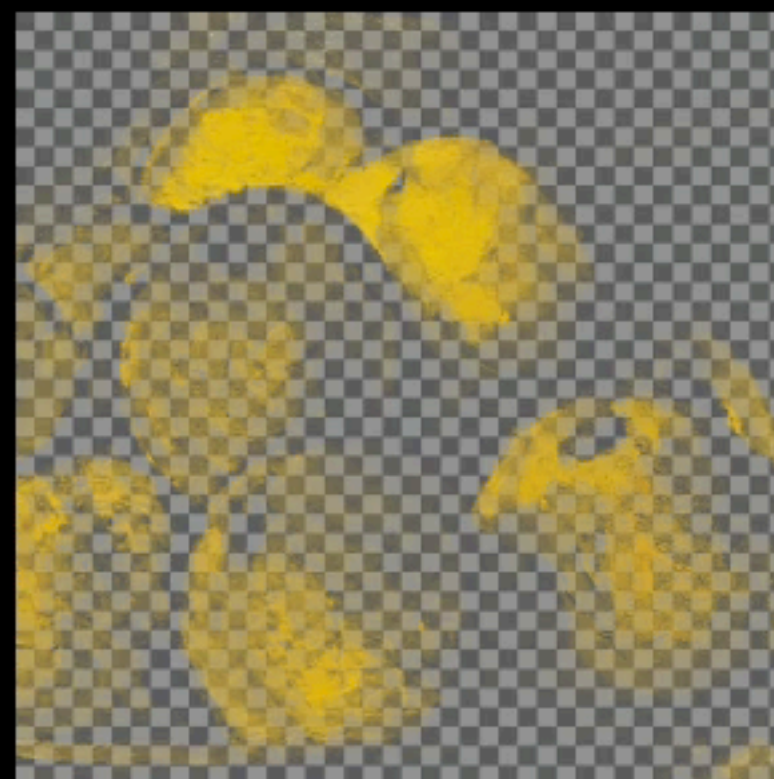
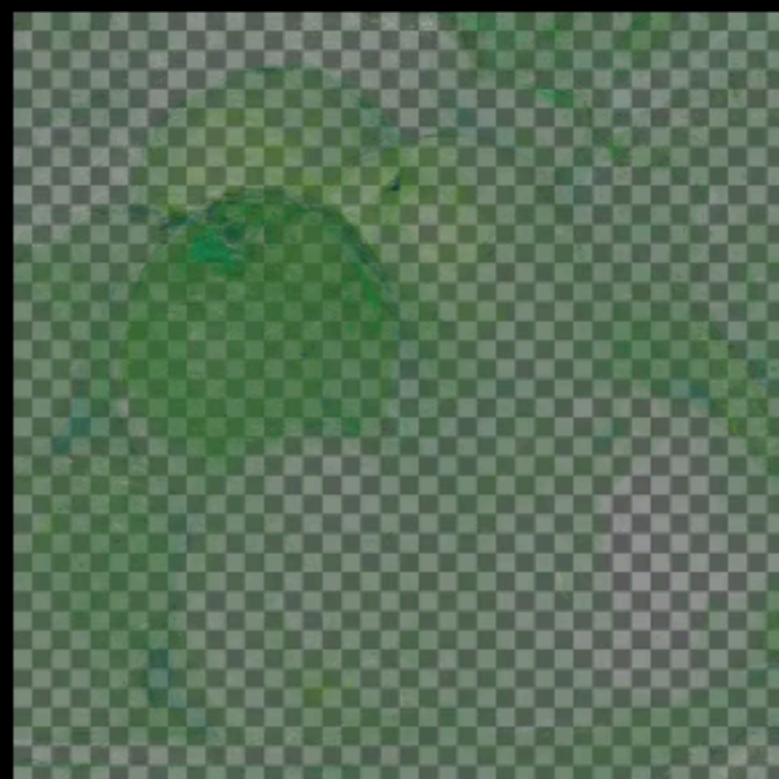
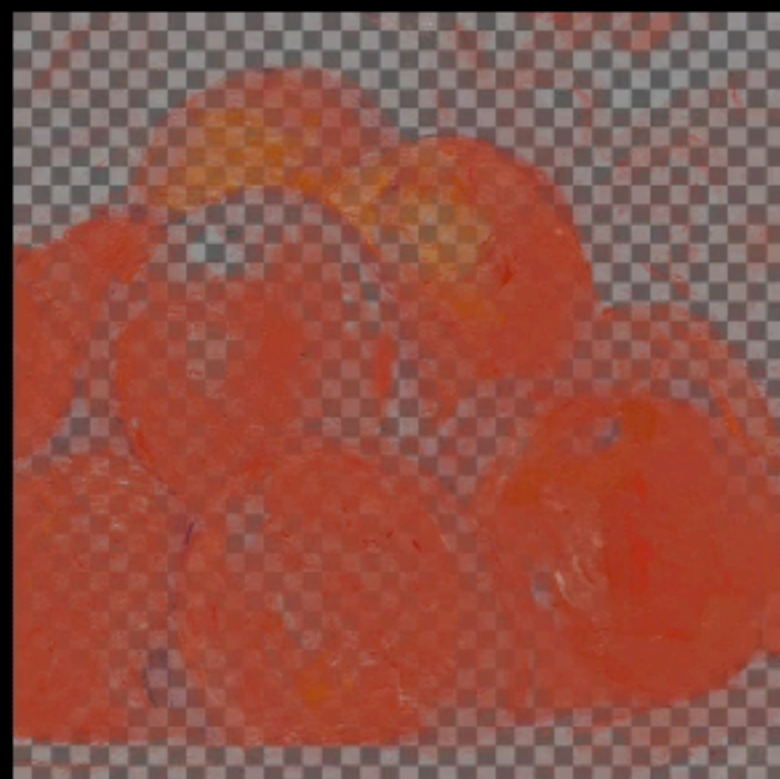
(N/A)

normal

normal

soft-light

multiply

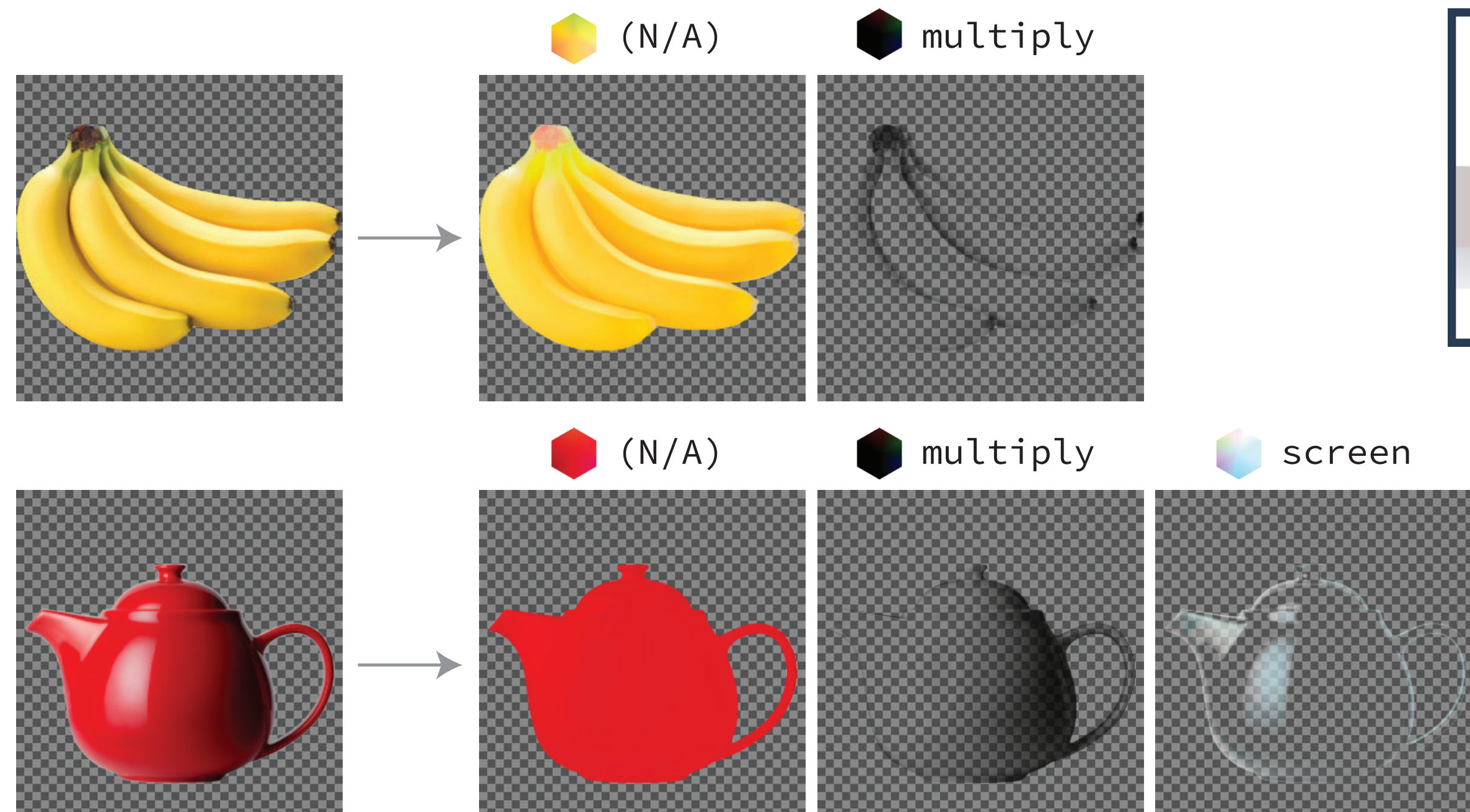


**Decomposed Layers**

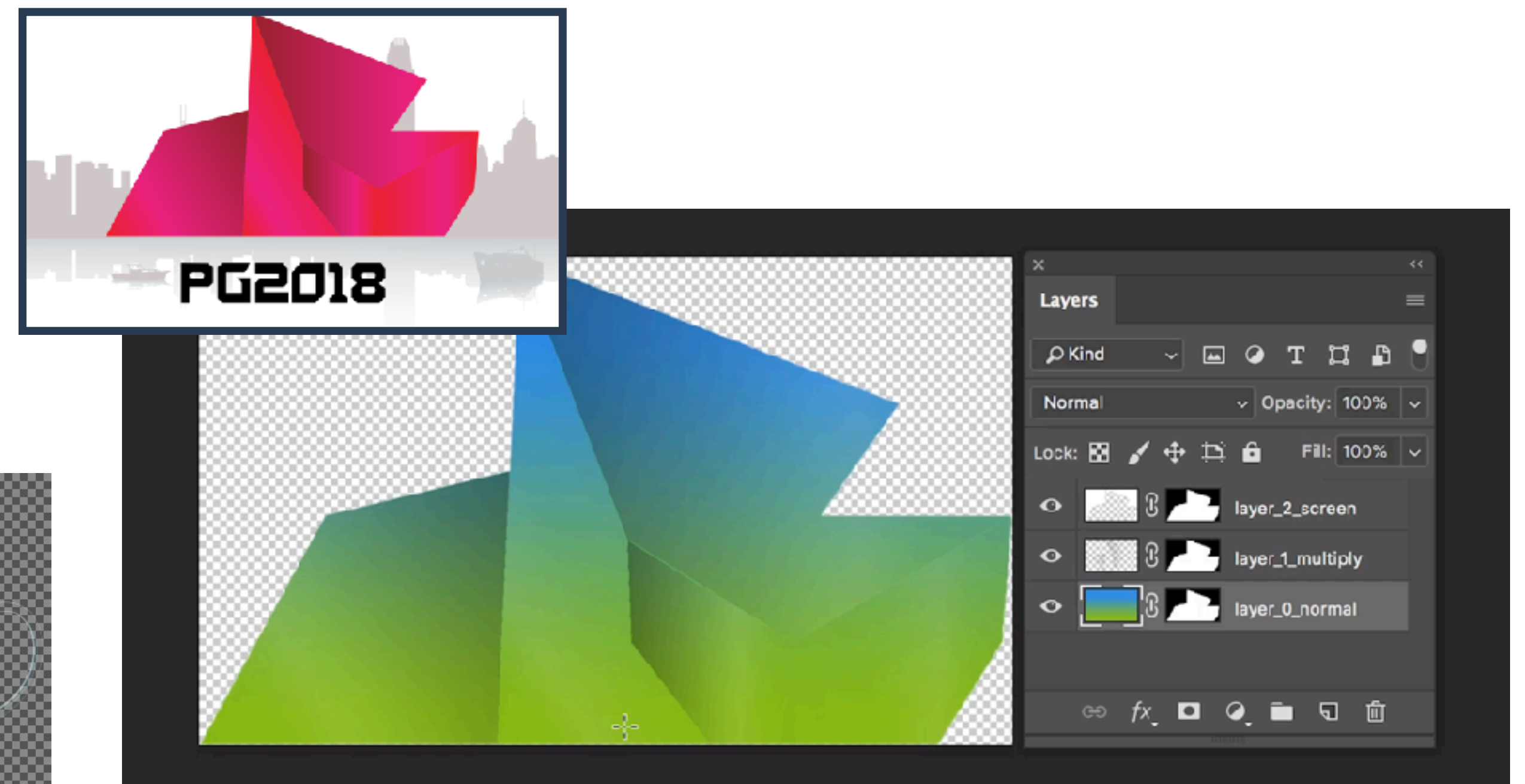


# Other Applications

**Intrinsic image decomposition:**  
decomposing an image into  
reflectance and shading layers



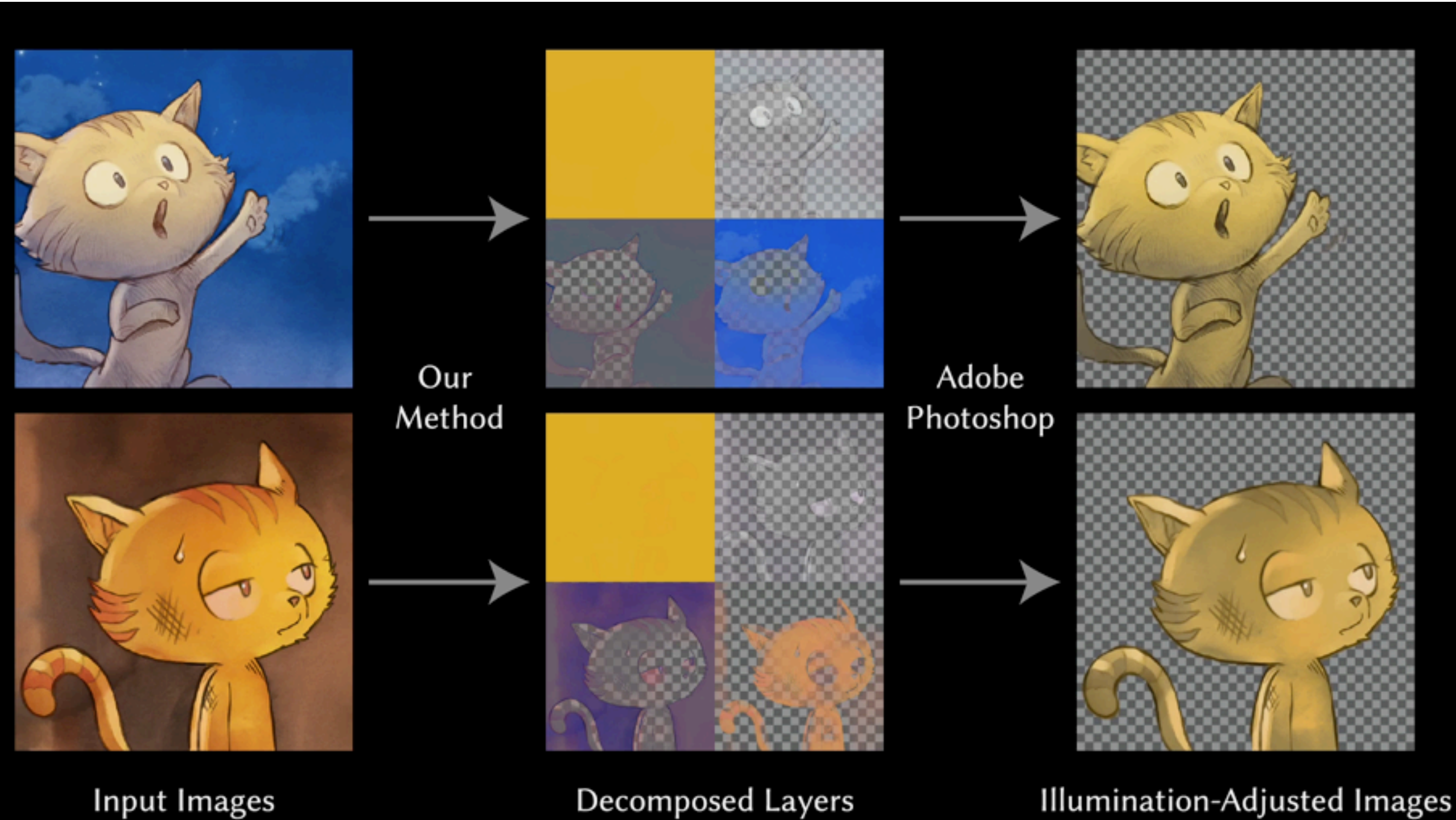
**Adding gradation to an  
existing conference logo:**  
extracting a “fill” layer from the  
logo and re-fill with gradation





# Remixing existing illustrations:

Harmonize colors of images from different scenes to put them together in a derivative work





# Summary

# Summary

## ■ Solution for a new problem

- Decomposing an image into layers with **advanced (non-linear) color blending**

## ■ Techniques

- **Generalizing** Aksoy+'s method [16; 17] for handling advanced (non-linear) color blend modes

## ■ Future work

- Automatic determination of optimal color distributions etc.





**In preparation...**

We are preparing for releasing our source codes at **GitHub**  
so please check our project page later

## **Decomposing Images into Layers with Advanced Color Blending**

---

Yuki Koyama & Masataka Goto

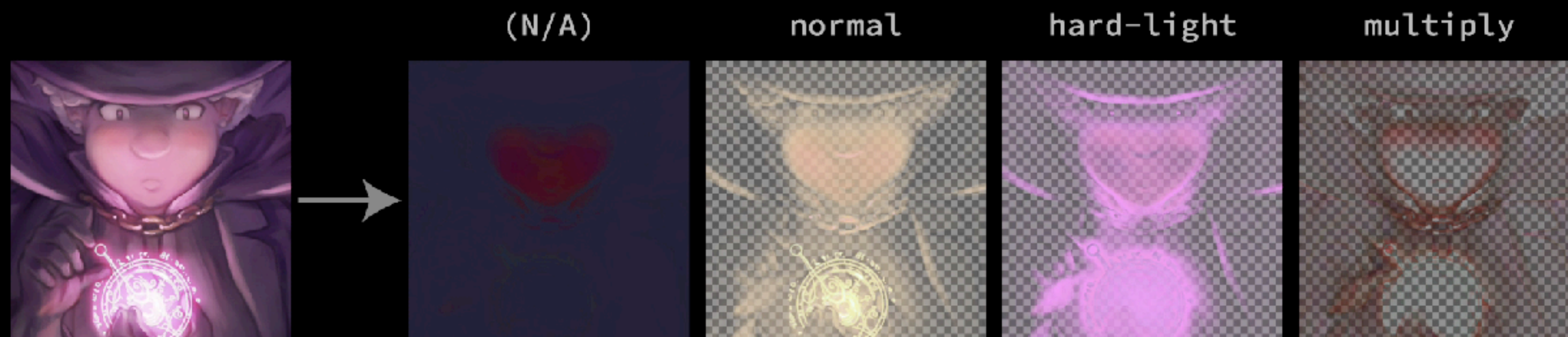




Computer Graphics Forum (Pacific Graphics 2018)

# Decomposing Images into Layers with Advanced Color Blending

Yuki Koyama & Masataka Goto 

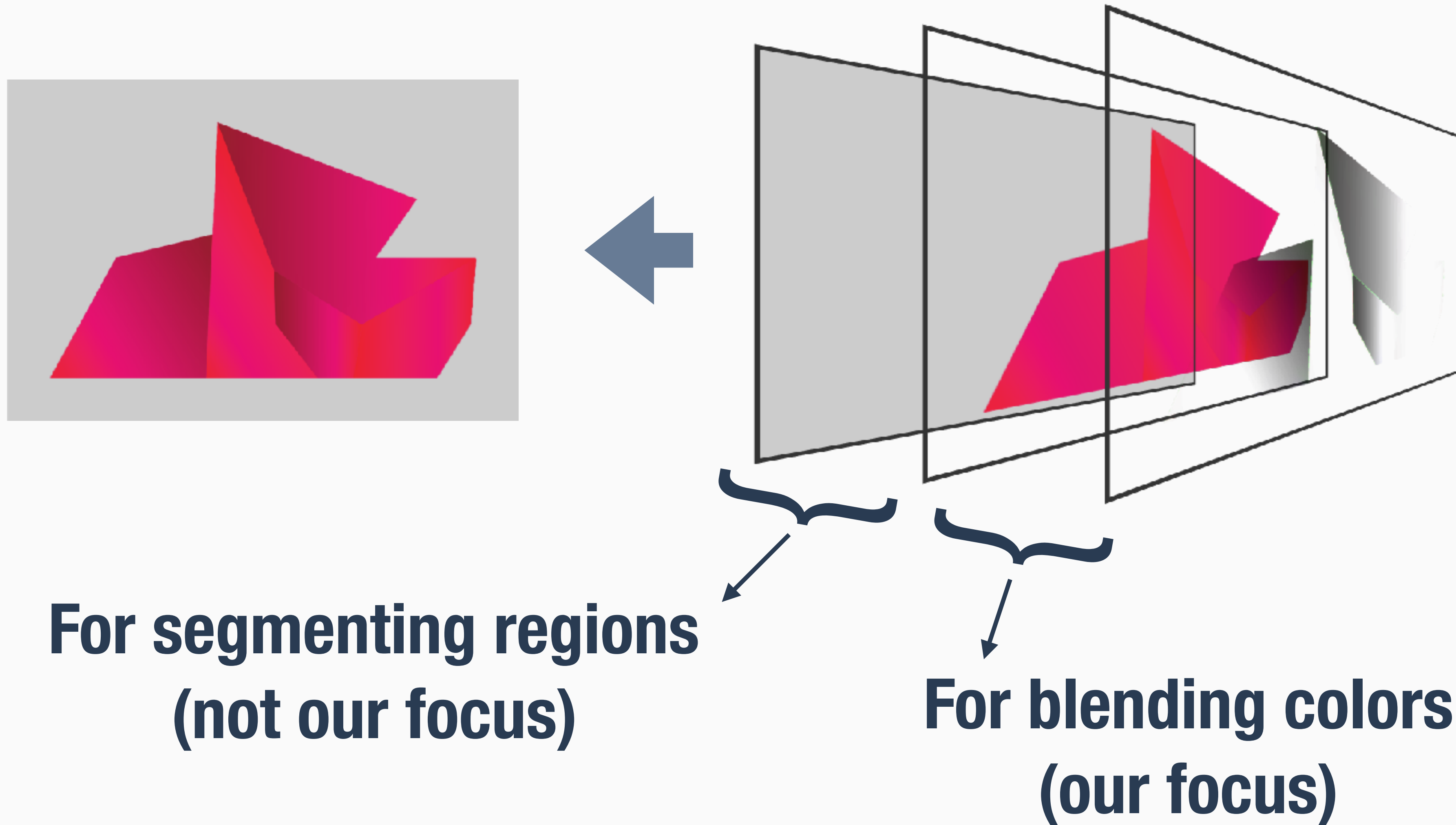


Input image courtesy of David Revoy | [www.davidrevoy.com](http://www.davidrevoy.com)

## Decomposing Images into Layers with Advanced Color Blending

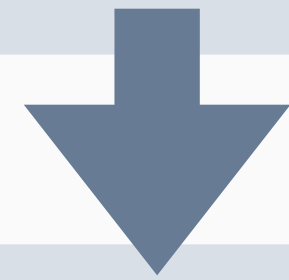
Yuki Koyama & Masataka Goto 

# Two Main Usages of Layers



# Specialization of the Composite Operator

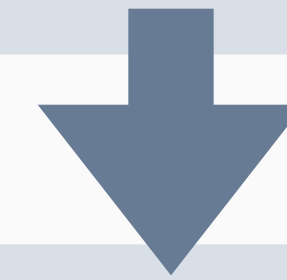
$$f(\mathbf{c}_s, \mathbf{c}_d) = \mathbf{c}_s,$$
$$X = Y = Z = 1$$



$$\mathbf{x}_s \oplus^{\text{rgb}} \mathbf{x}_d = \frac{\alpha_s \mathbf{c}_s + (1 - \alpha_s) \alpha_d \mathbf{c}_d}{\alpha_s \oplus^{\alpha} \alpha_d},$$
$$\alpha_s \oplus^{\alpha} \alpha_d = \alpha_s + (1 - \alpha_s) \alpha_d$$

**Alpha blending model**  
C.f., [Tan+16]

$$f(\mathbf{c}_s, \mathbf{c}_d) = \mathbf{c}_s + \mathbf{c}_d,$$
$$X = 2, Y = Z = 1$$

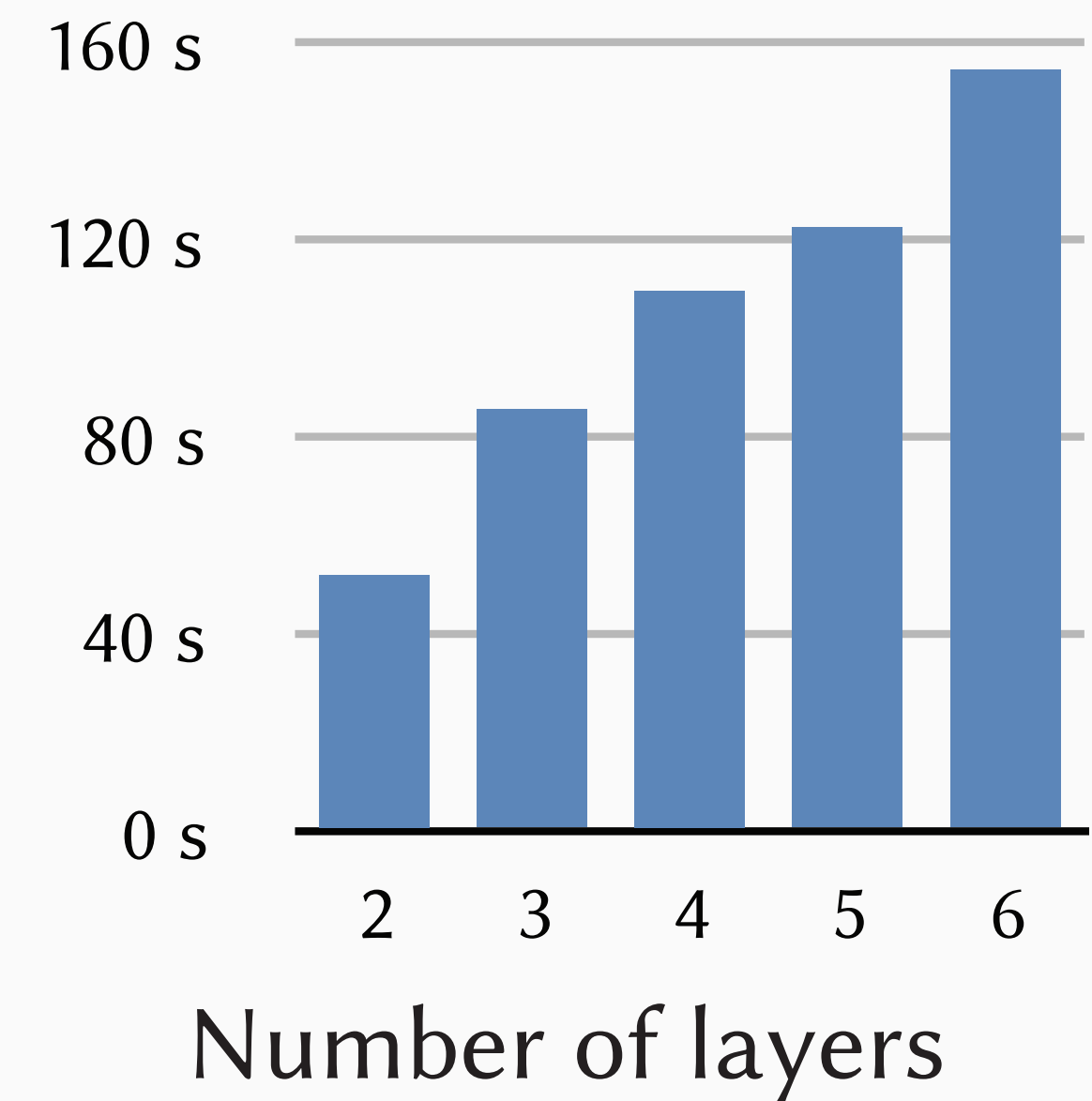
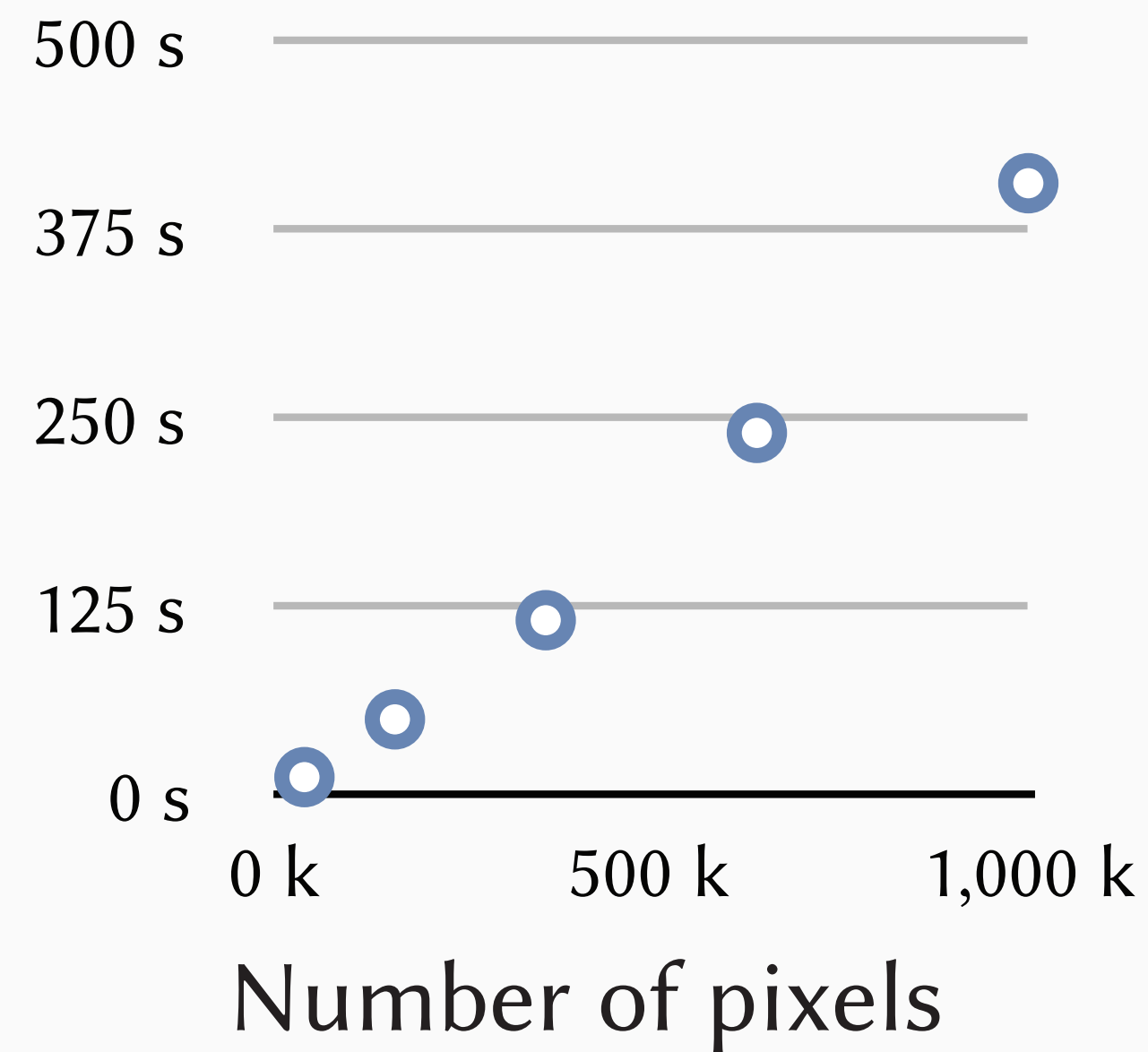
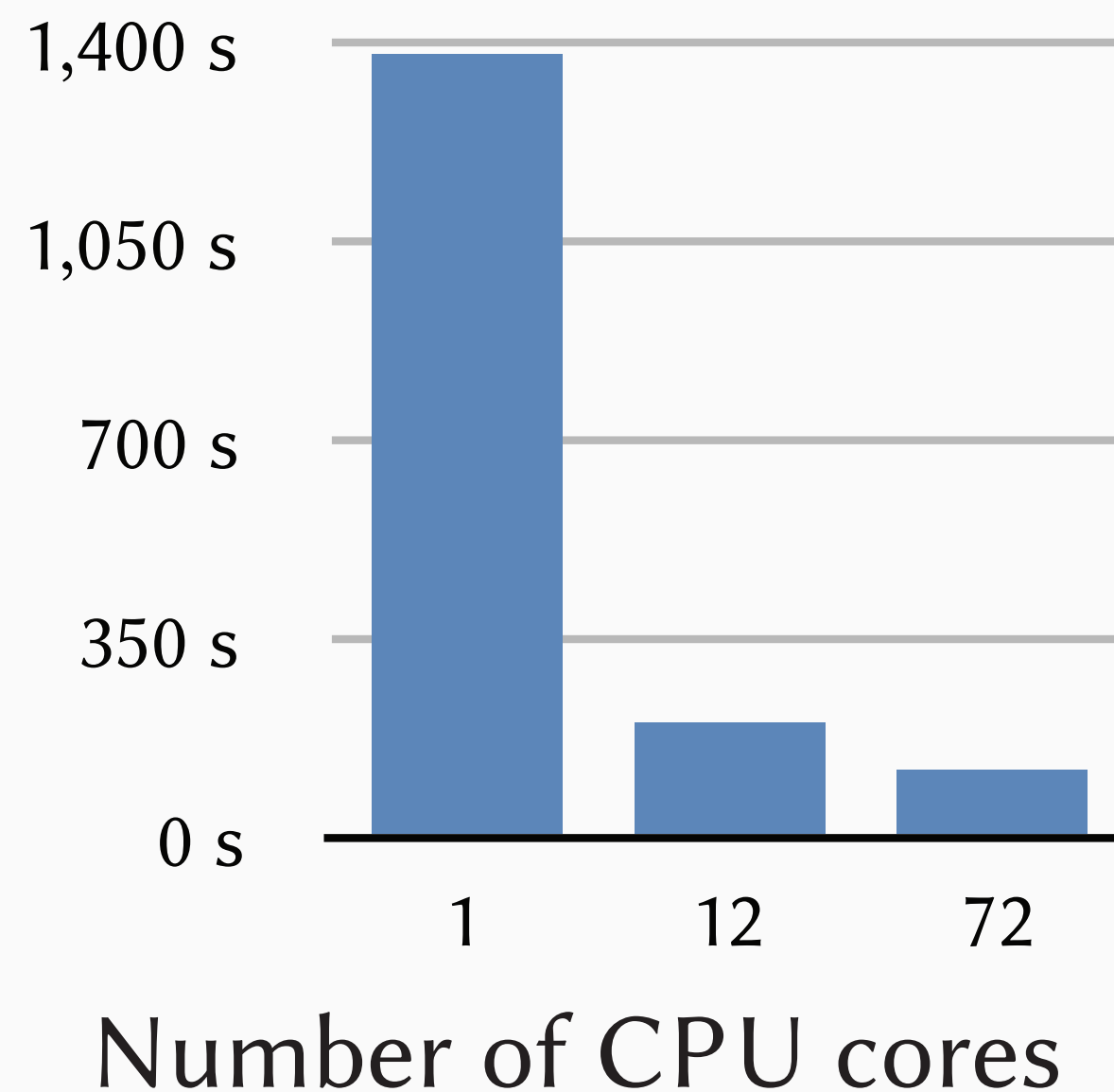


$$\mathbf{x}_s \oplus^{\text{rgb}} \mathbf{x}_d = \frac{\alpha_s \mathbf{c}_s + \alpha_d \mathbf{c}_d}{\alpha_s \oplus^{\alpha} \alpha_d},$$
$$\alpha_s \oplus^{\alpha} \alpha_d = \alpha_s + \alpha_d$$

**Linear additive model**  
C.f., [Aksoy+16; 18]

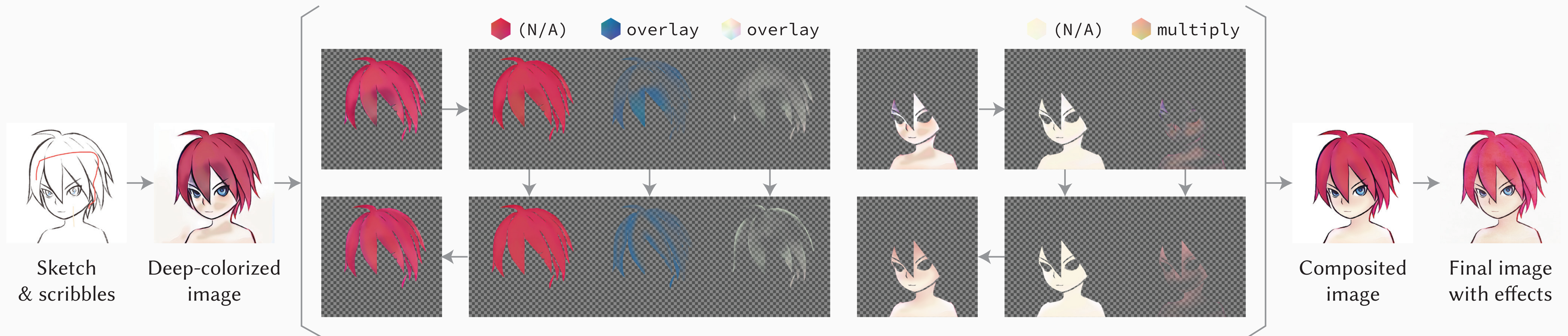


# Computational Cost



- Parallelization is effective
- Increase cost along with the number of pixels
- Increase cost along with the number of layers (due to the recursive calculation)

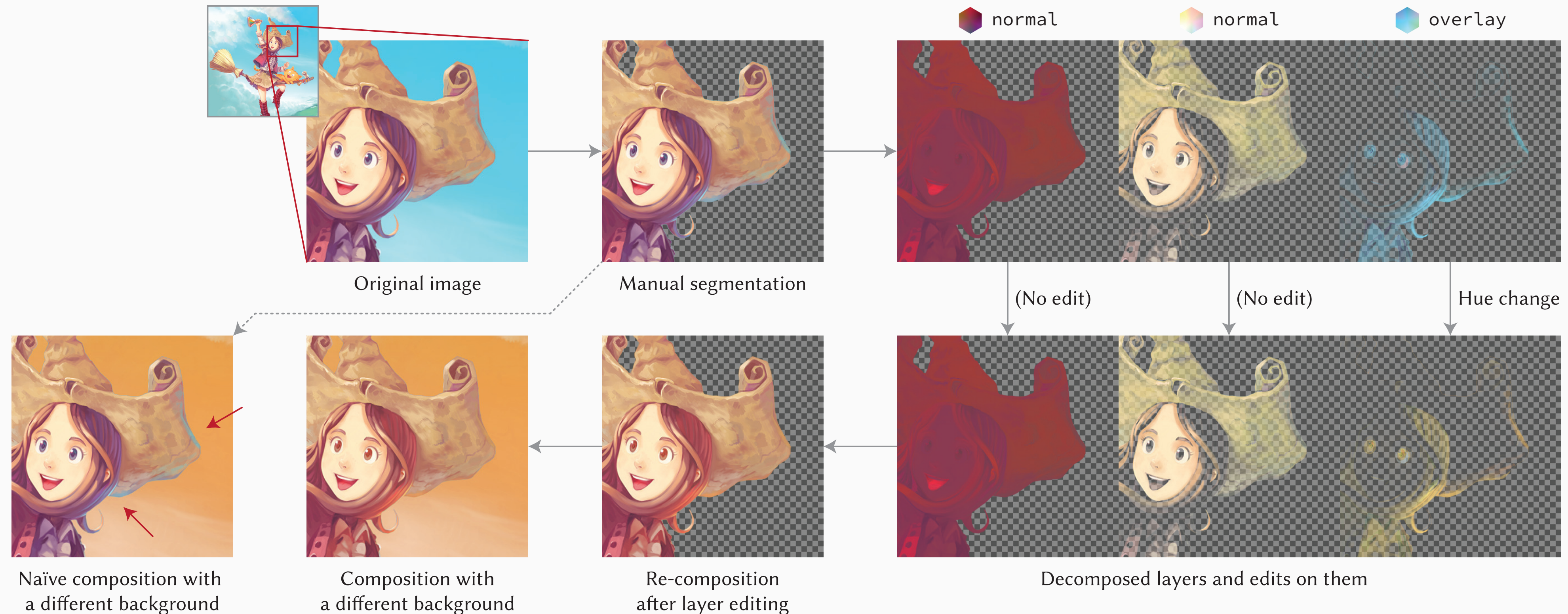
# Fixing Deep Colorization via Layer Decomposition



Our method enables a new digital painting workflow by effectively using such a deep-colored image as a good starting point. Instead of directly editing the generated single-layered image, the user can decompose it into layers with familiar advanced color blending then individually edit each unsatisfactory layer by adding strokes, changing hue, etc.



# Lighting-Aware Replacement of Background



Example use of our method, in which the blue sky background is replaced with a sunset one. With our method, the character region is decomposed into three layers: the bottom two layers roughly correspond to albedo, and the top layer roughly corresponds to lighting by the blue sky represented as an overlay layer. To match the lighting in sunset, the top layer's hue is modified. Finally, the layers are re-composited with the sunset background. Note that naïvely compositing the character region with the sunset background (the leftmost image) does not look good because lighting is mismatched. Input image courtesy of David Revoy.