

Vectorising Bitmaps into Semi-Transparent Gradient Layers

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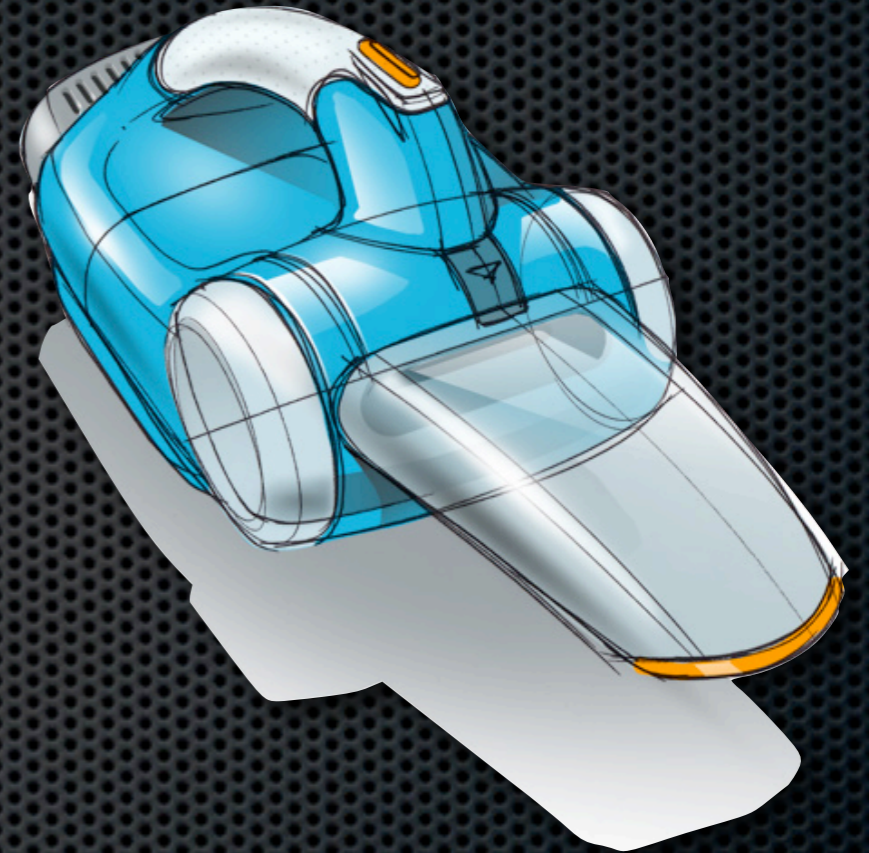
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Rey Juan Carlos

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UNIVERSITY OF CALIFORNIA



photos



drawings

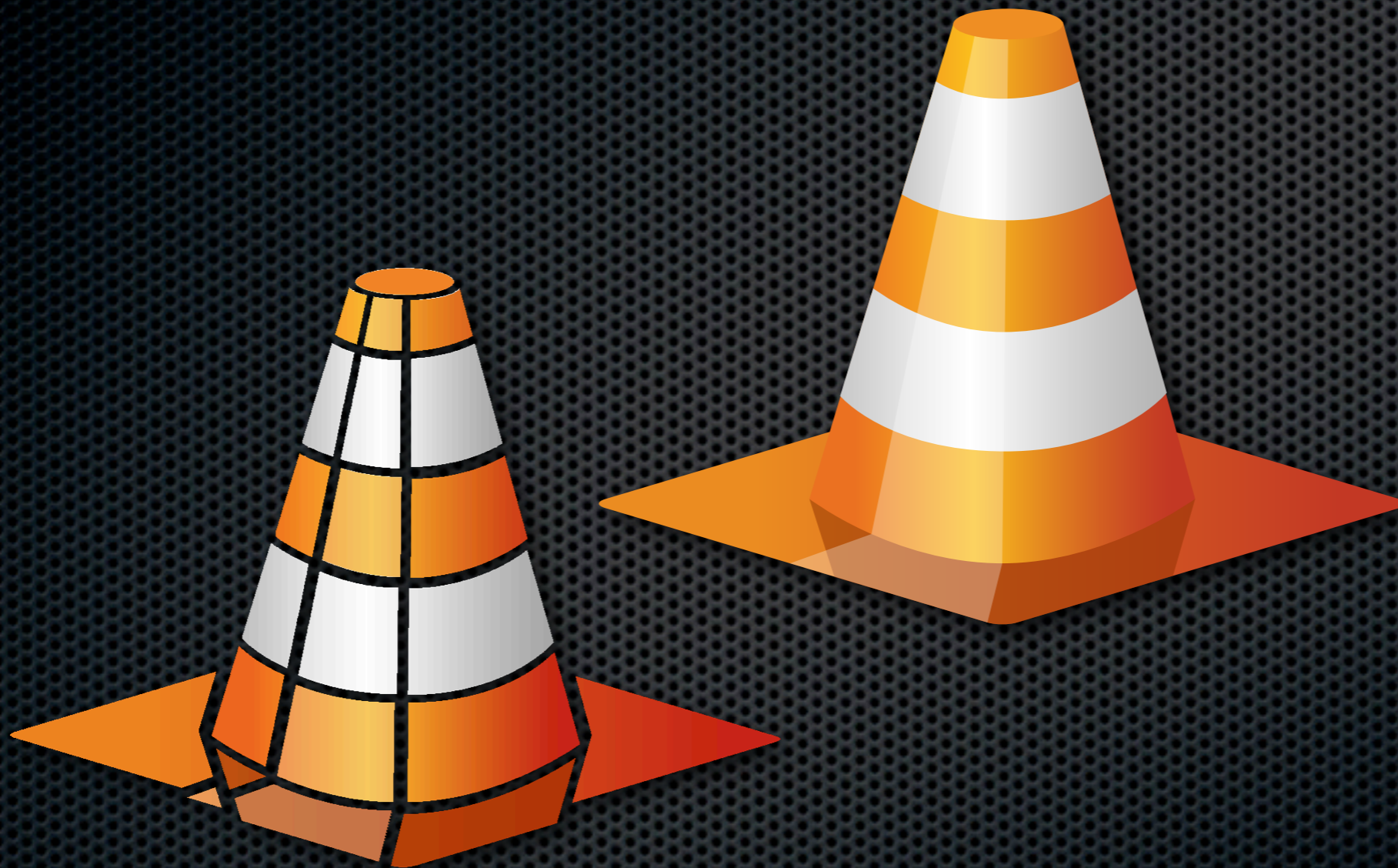


vector art

Vector art representations



Vector art representations



single layer

Vector art representations



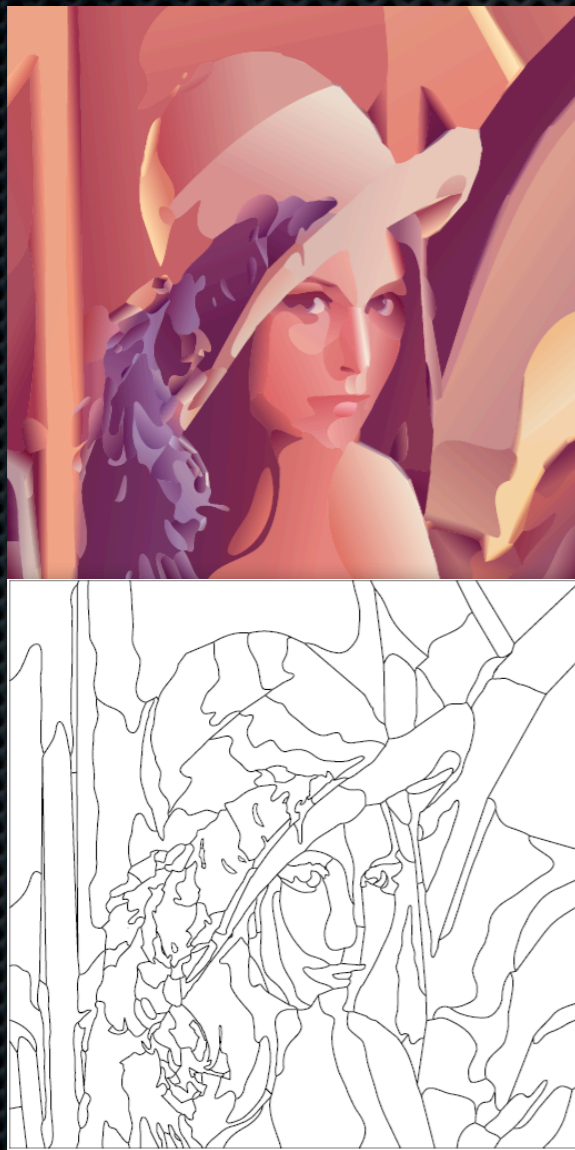
single layer

multiple layers

Image vectorisation

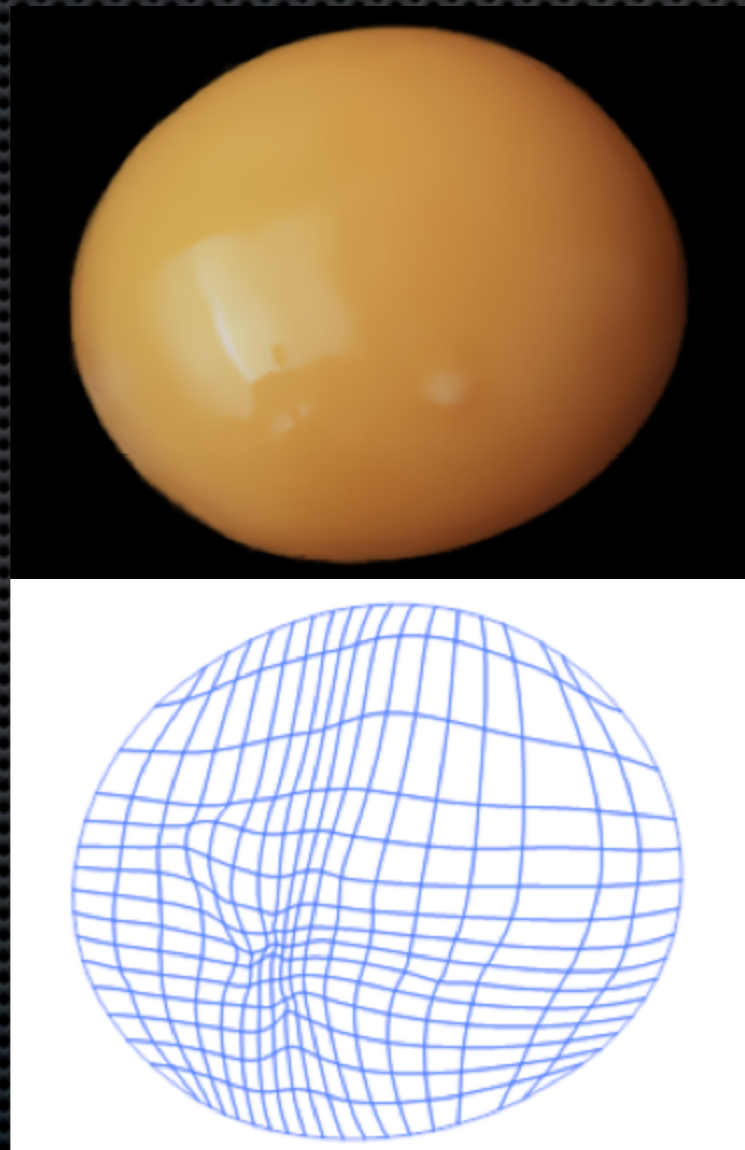
Ardeco

[Lecot & Lévy 2006]



Gradient meshes

[Sun+ 2007]



Diffusion curves

[Orzan+ 2008]



Our interactive workflow



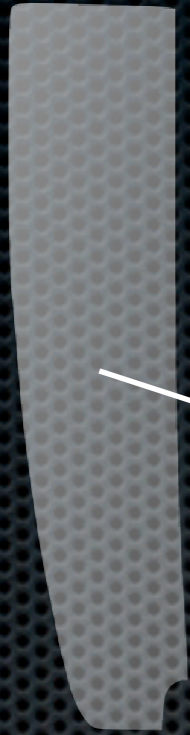
Shutterstock/George Dolgikh

Our interactive workflow



Shutterstock/George Dolgikh

Our interactive workflow



Shutterstock/George Dolgikh

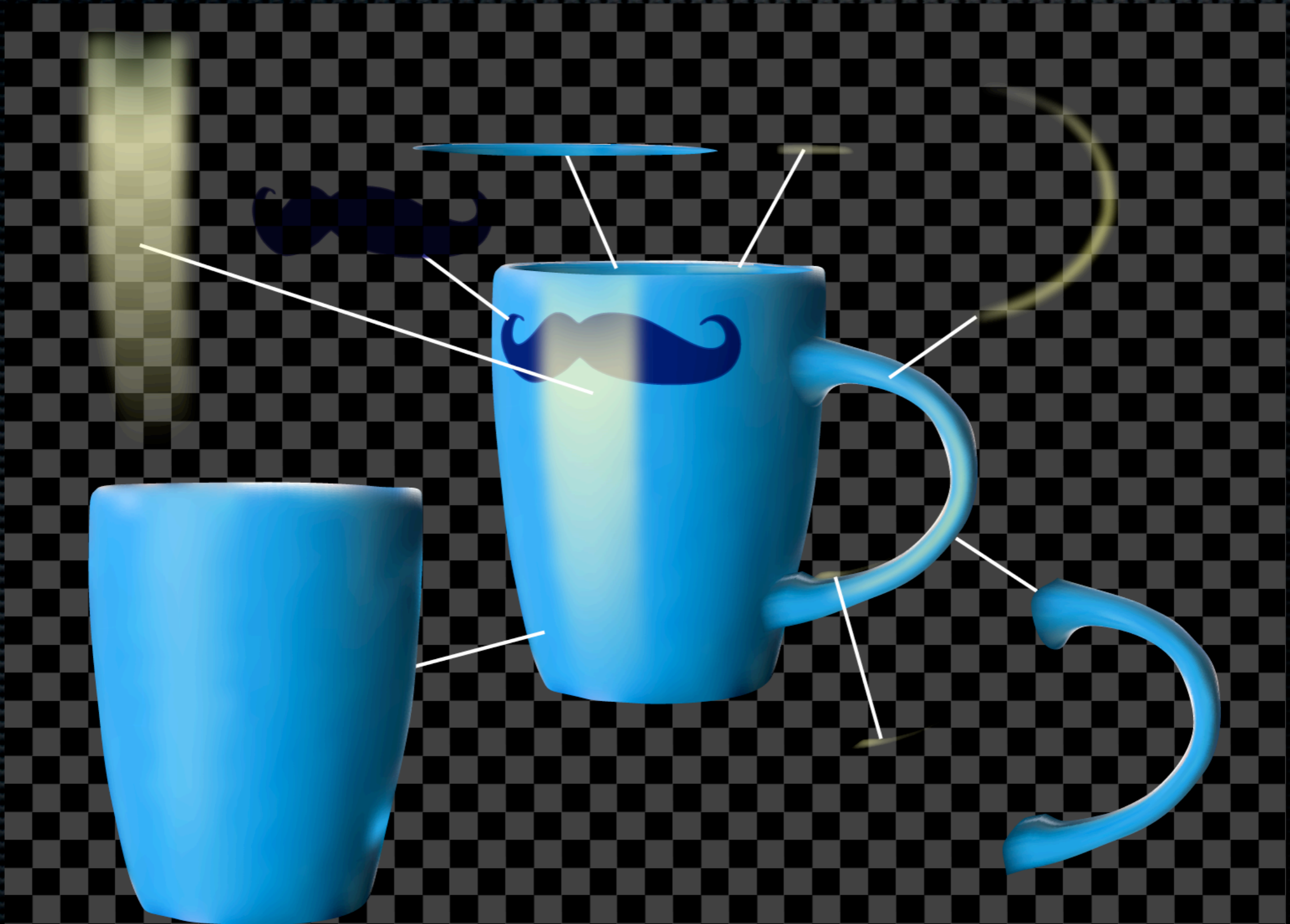
Our interactive workflow



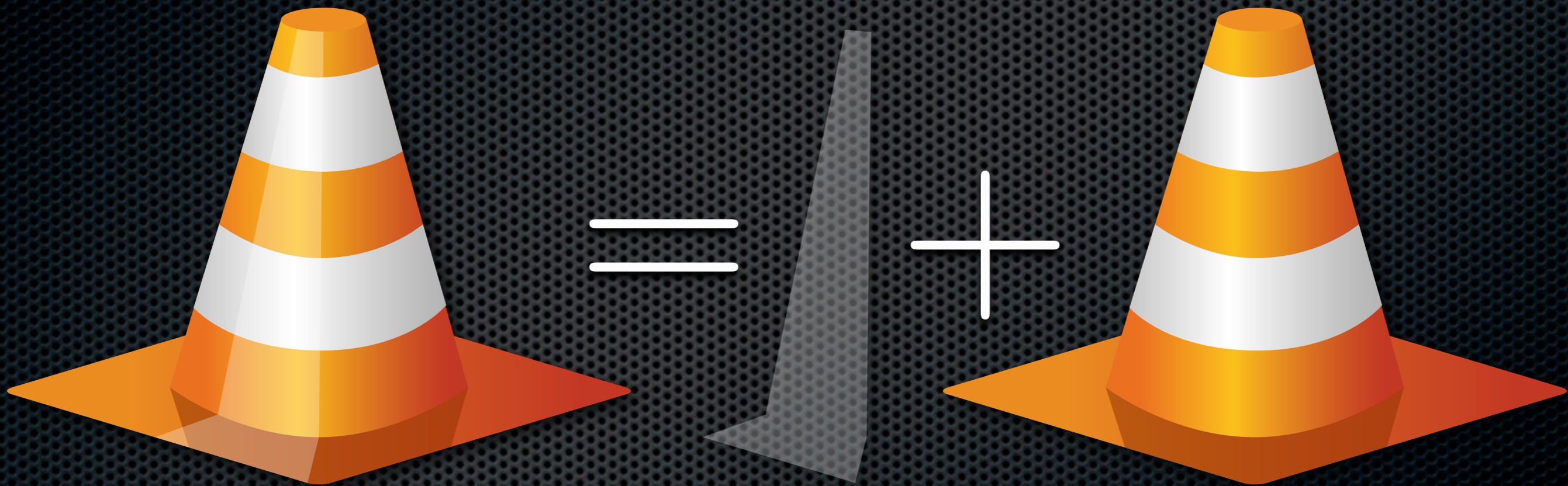
Vectorised result



Editing result



Similarity to matting



$$\mathbf{I} = \alpha \cdot \mathbf{F} + (1 - \alpha) \cdot \mathbf{B}$$

compositing equation

[Porter & Duff 1984]

The matting problem

[Smith & Blinn 1996]

$$\mathbf{I} = \alpha \cdot \mathbf{F} + (1 - \alpha) \cdot \mathbf{B}$$

we have 3 equations: one each for R, G, B

The matting problem

[Smith & Blinn 1996]

$$\mathbf{I} = \alpha \cdot \mathbf{F} + (1 - \alpha) \cdot \mathbf{B}$$

we have 3 equations: one each for R, G, B

know

solve for

unknowns

\mathbf{I} α \mathbf{F} \mathbf{B}

7

underconstrained

\mathbf{I} \mathbf{B} α \mathbf{F}

4

underconstrained

Solving the matting problem

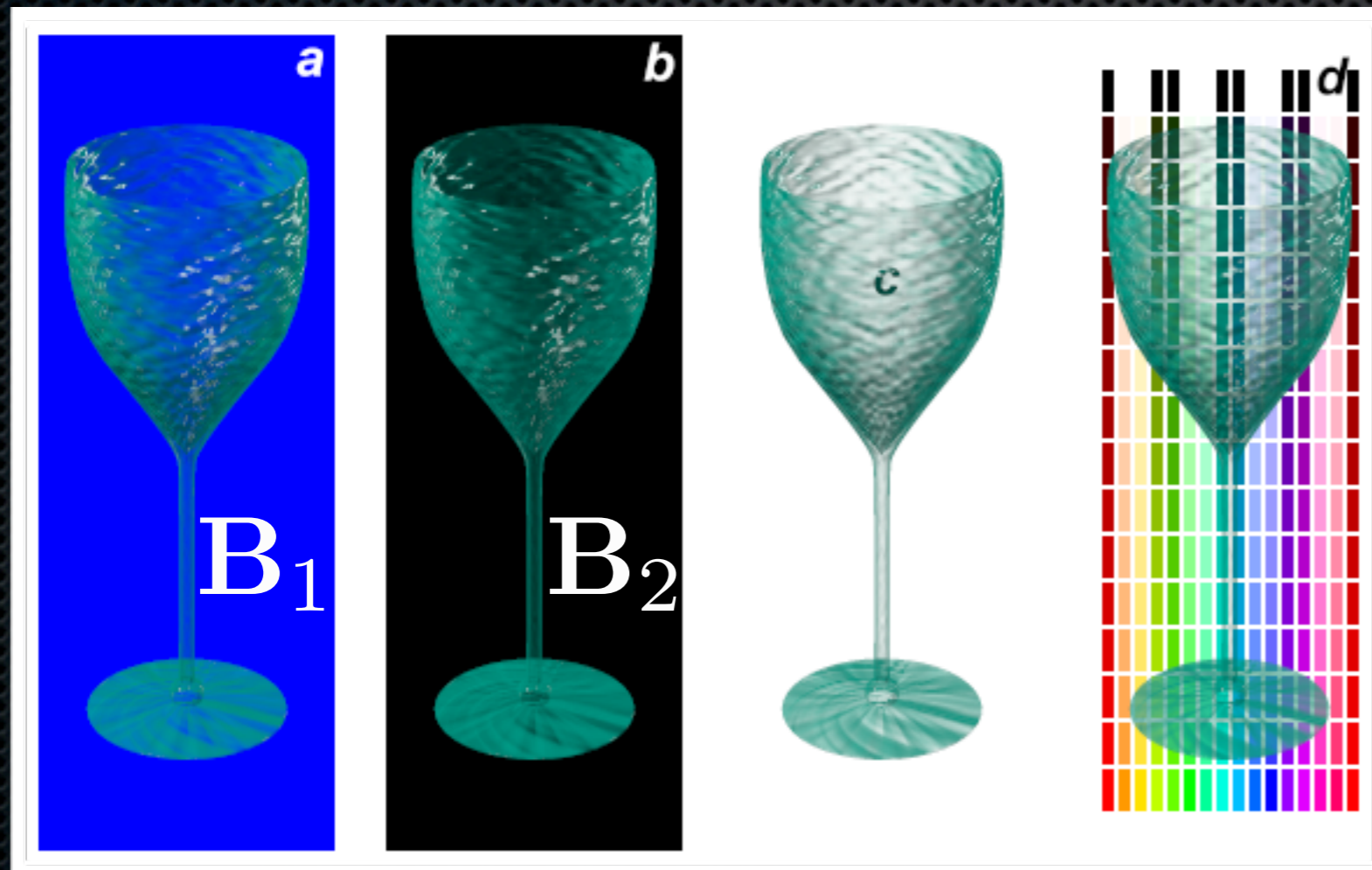
[Smith & Blinn 1996]

$$\mathbf{I}_1 = \alpha \cdot \mathbf{F} + (1 - \alpha) \cdot \mathbf{B}_1$$

$$\mathbf{I}_2 = \alpha \cdot \mathbf{F} + (1 - \alpha) \cdot \mathbf{B}_2$$

we know:

\mathbf{I}_1 \mathbf{I}_2 \mathbf{B}_1 \mathbf{B}_2



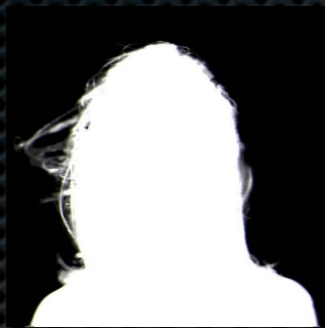
solve for:

α \mathbf{F}

6 equations,
4 unknowns

great!

Image decompositions



Alpha matting

[e.g. Smith+ 1996, Chuang+ 2001, Levin+ 2008]



Reflection separation

[e.g. Levin+ 2004/2007, Kim+ 2013, Li & Brown 2014]



Intrinsic images

[e.g. Bousseau+ 2009, Carroll+ 2011]

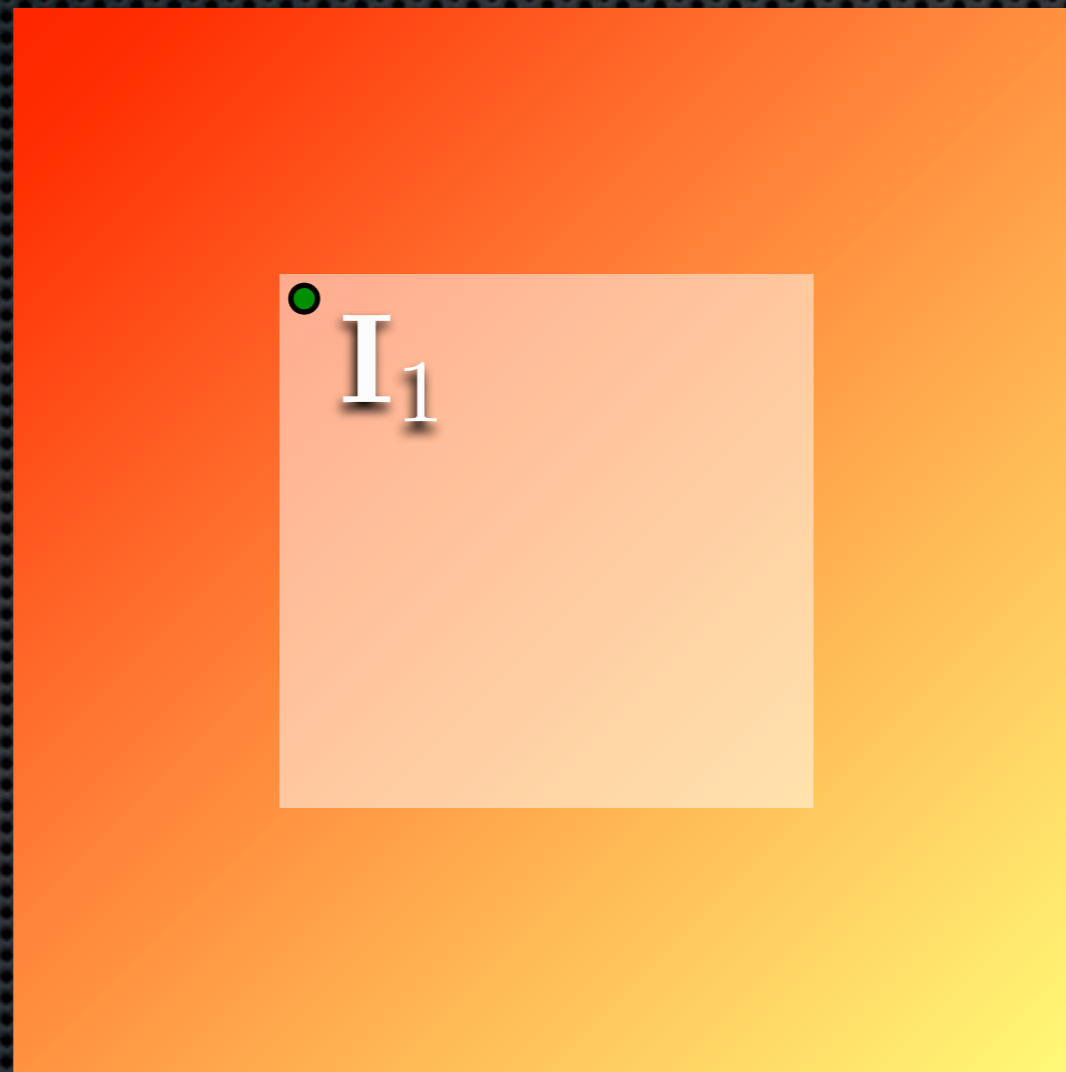
Decomposing



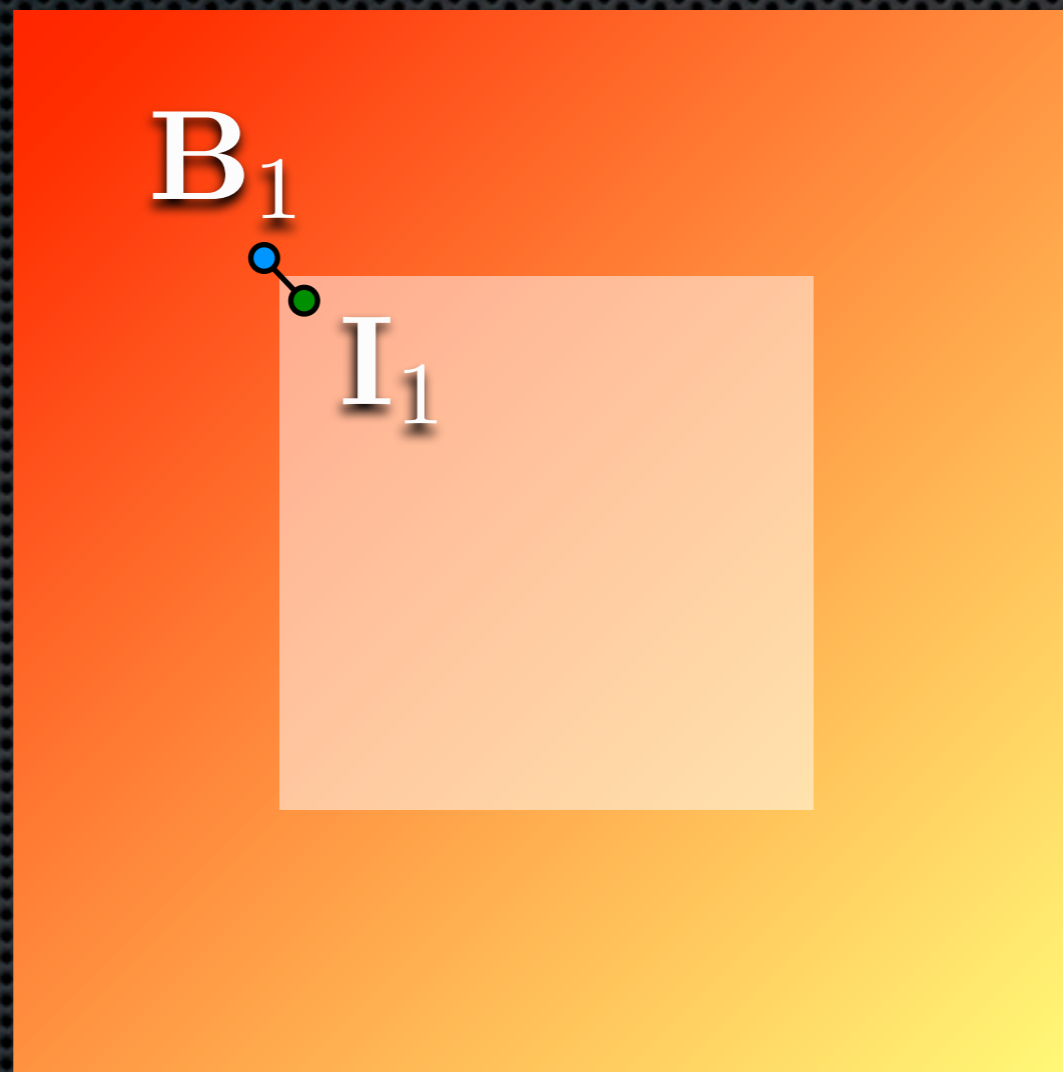
Decomposing



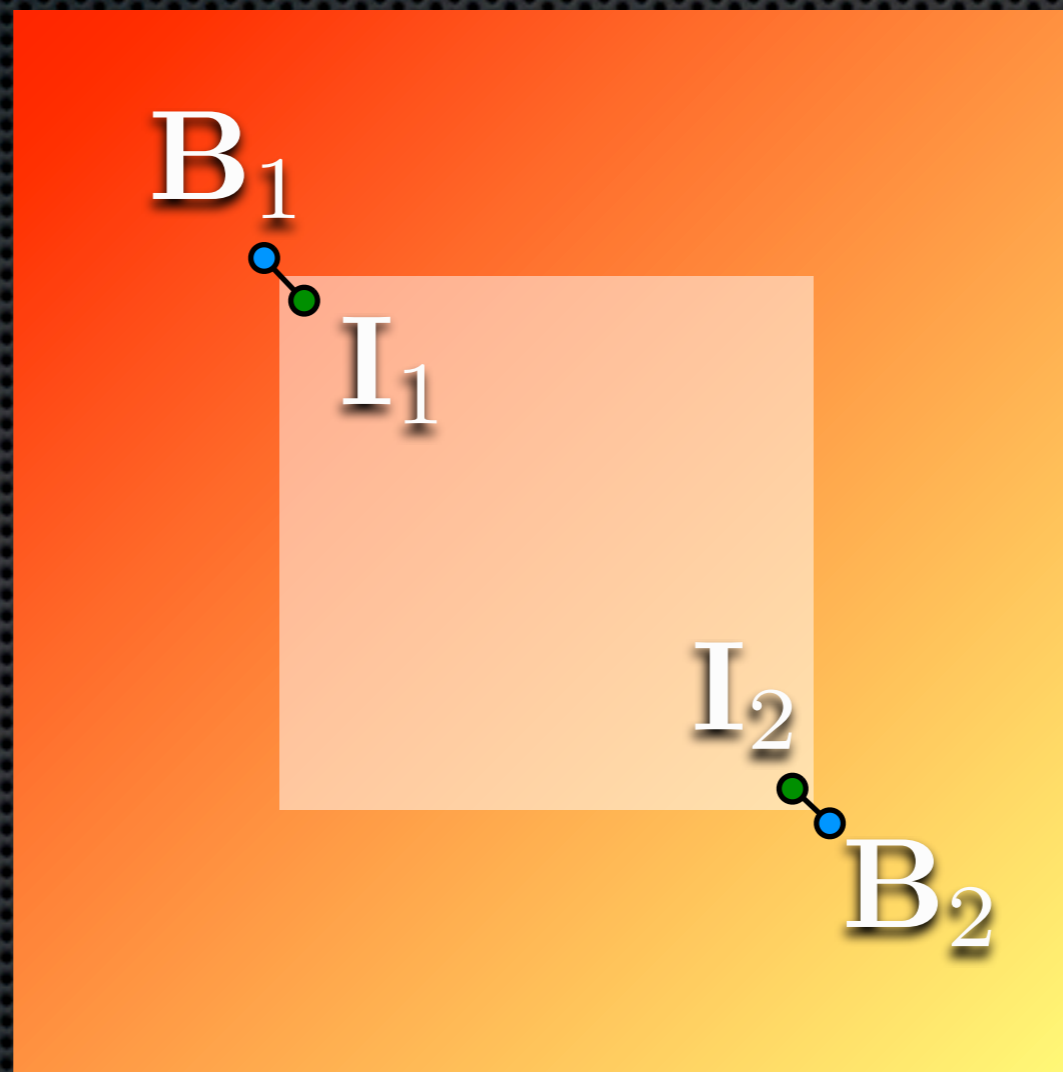
Decomposing



Decomposing



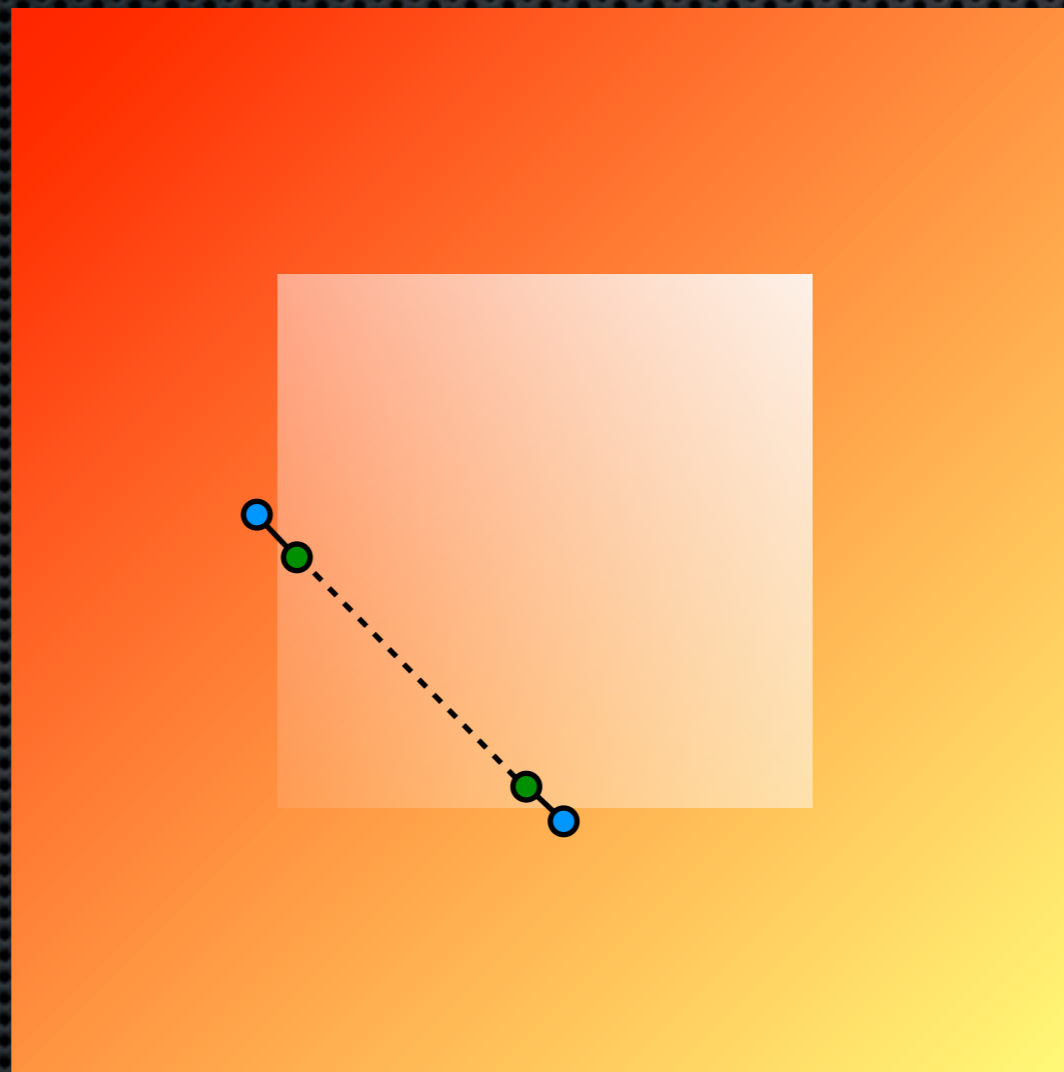
Decomposing



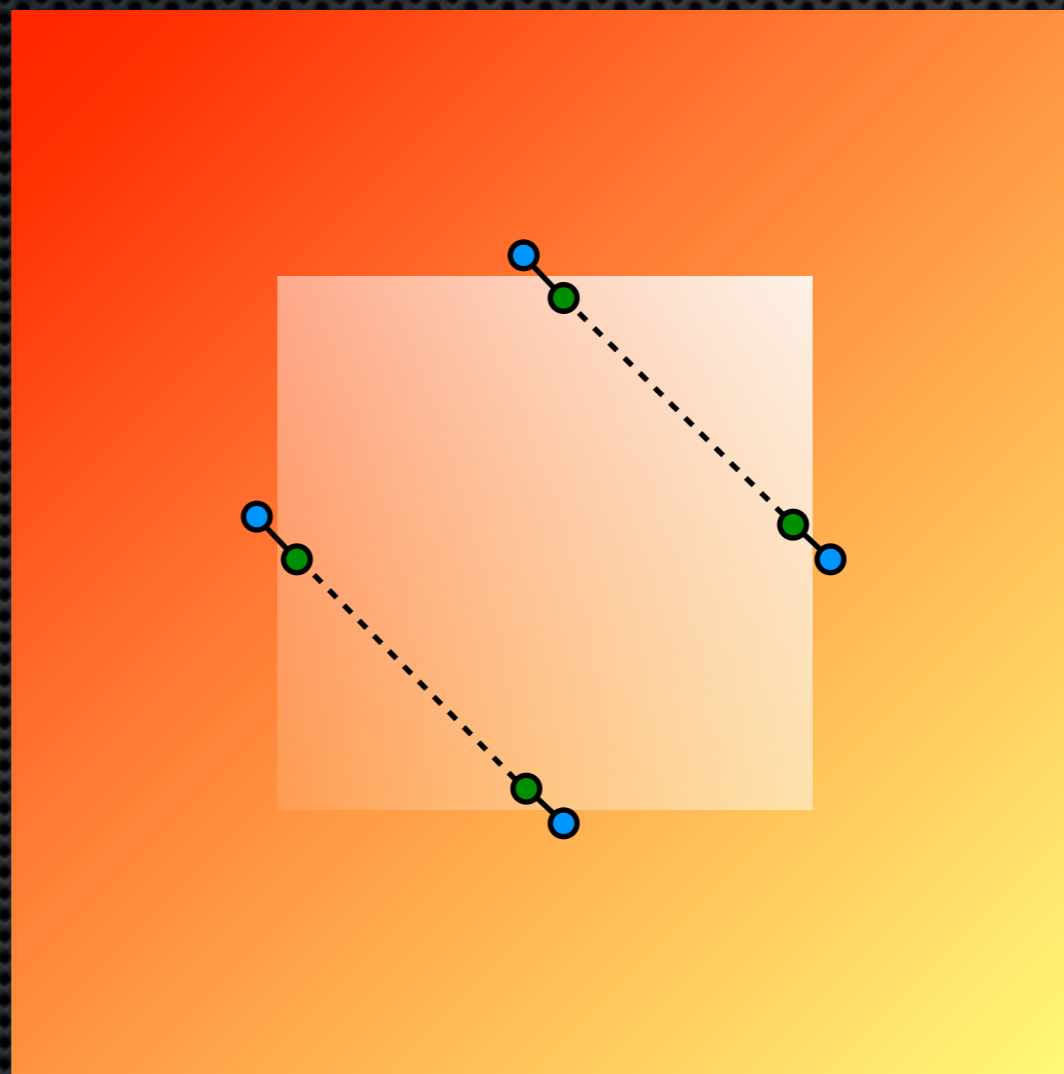
Decomposing



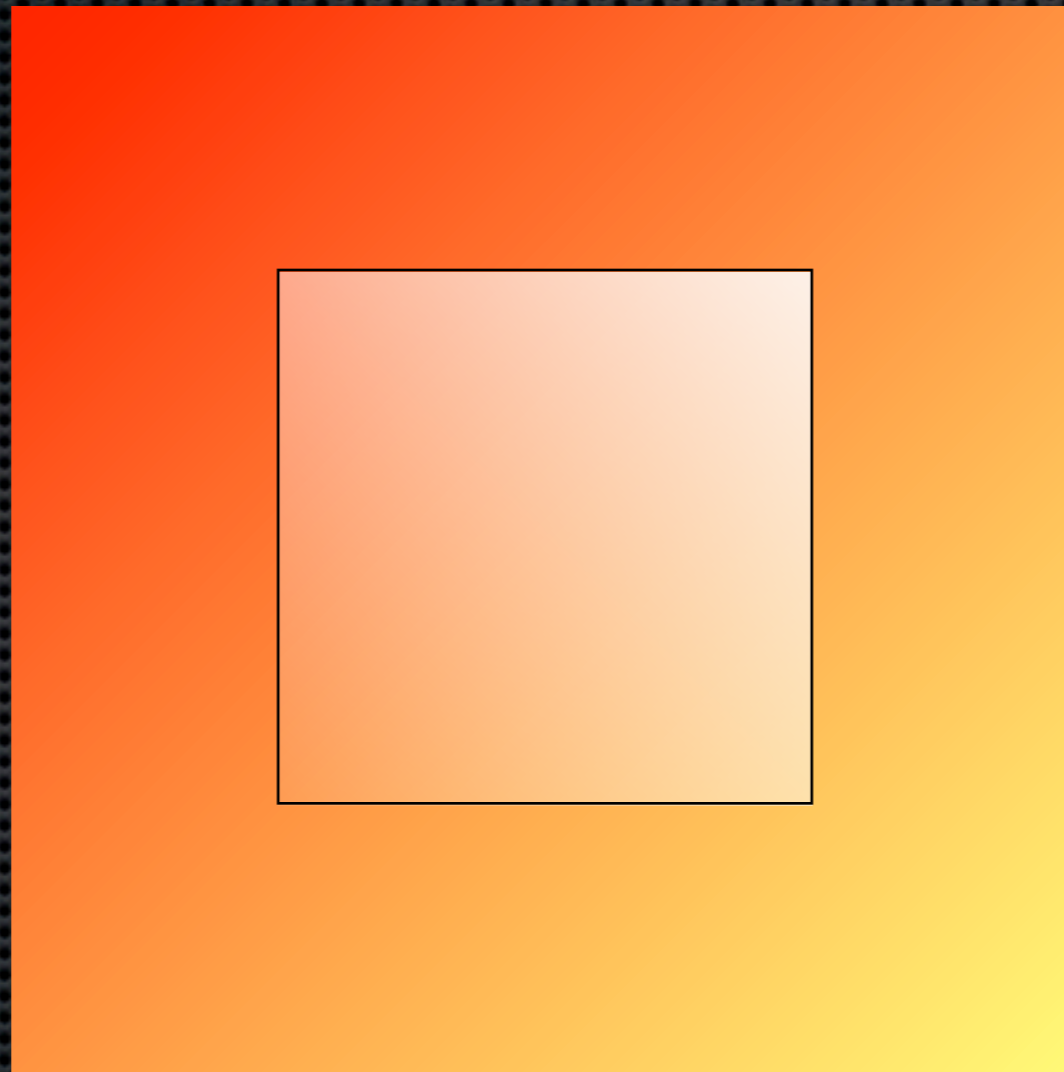
Decomposing



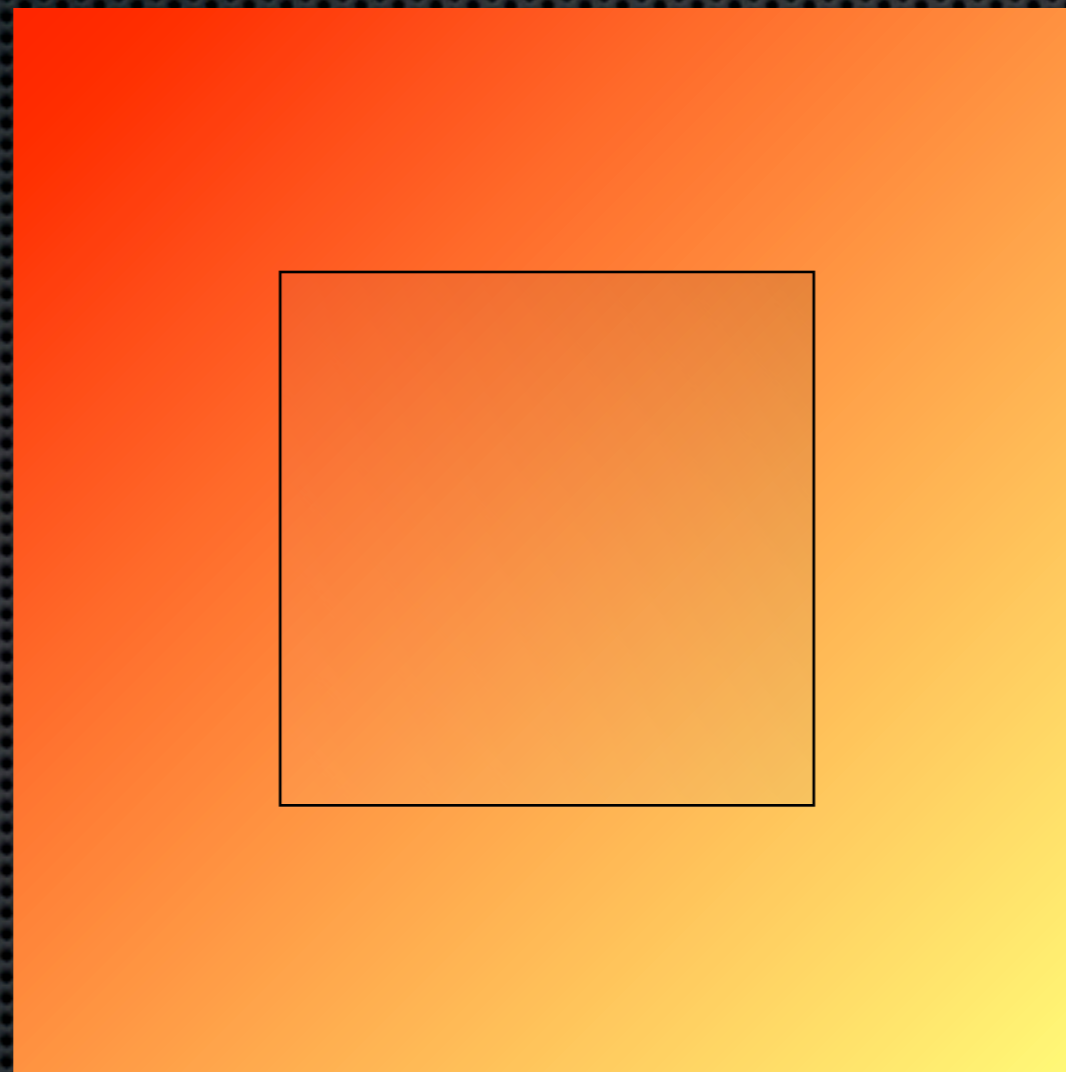
Decomposing



Decomposing



Decomposing



Decomposing



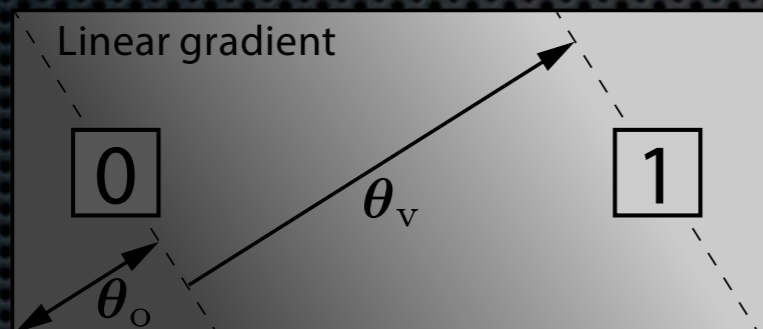
Parametric gradient functions

$$f = c \circ g = c(g(\mathbf{x}, \boldsymbol{\theta}), \boldsymbol{\theta})$$

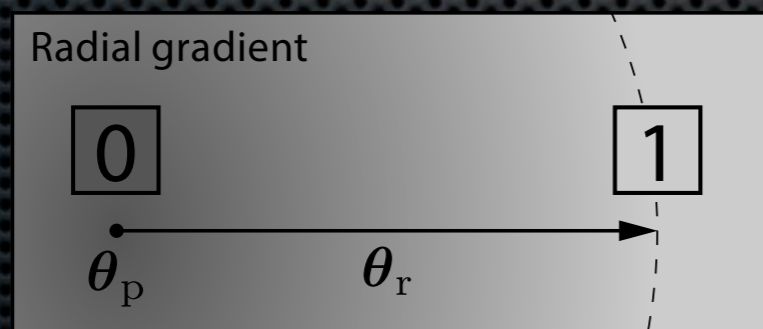
Parametric gradient functions

$$f = c \circ g = c(g(\mathbf{x}, \theta), \theta)$$

gradient function $g(\mathbf{x}, \theta)$



$$g_{\text{linear}}(\mathbf{x}, \theta) = \frac{\mathbf{x} \cdot \theta_v}{\|\theta_v\|^2} + \theta_o$$

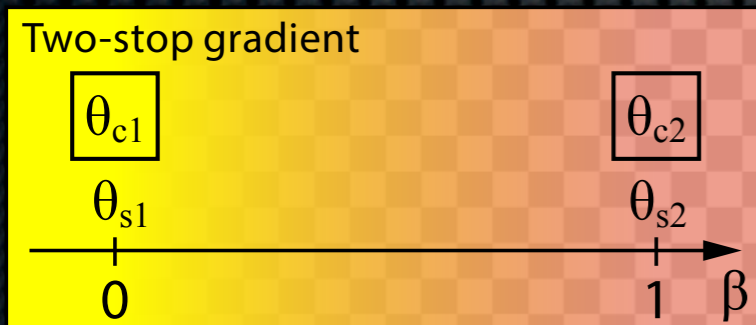


$$g_{\text{radial}}(\mathbf{x}, \theta) = \frac{\|\mathbf{x} - \theta_p\|}{\theta_r}$$

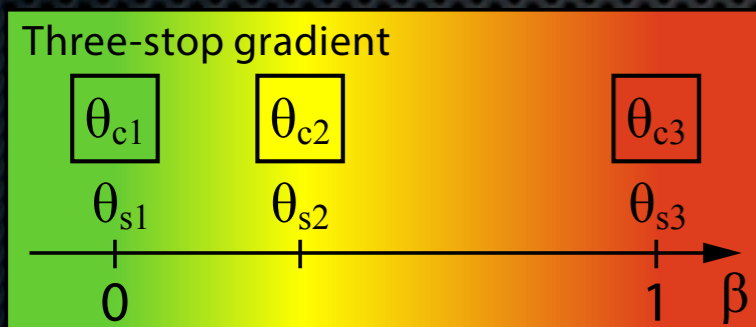
Parametric gradient functions

$$f = \mathbf{c} \circ \mathbf{g} = \mathbf{c}(\mathbf{g}(\mathbf{x}, \boldsymbol{\theta}), \boldsymbol{\theta})$$

colour function $\mathbf{c}(\beta, \boldsymbol{\theta})$



$$\mathbf{c}_2(\beta, \boldsymbol{\theta}) = \text{mix}(\boldsymbol{\theta}_{c1}, \boldsymbol{\theta}_{c2}, \beta)$$



$$\mathbf{c}_3(\beta, \boldsymbol{\theta}) = \begin{cases} \text{mix}\left(\boldsymbol{\theta}_{c1}, \boldsymbol{\theta}_{c2}, \frac{\beta}{\boldsymbol{\theta}_{s2}}\right) & \beta \leq \boldsymbol{\theta}_{s2} \\ \text{mix}\left(\boldsymbol{\theta}_{c2}, \boldsymbol{\theta}_{c3}, \frac{\beta - \boldsymbol{\theta}_{s2}}{1 - \boldsymbol{\theta}_{s2}}\right) & \beta > \boldsymbol{\theta}_{s2} \end{cases}$$

$$\text{mix}(\mathbf{a}, \mathbf{b}, t) = (1 - t) \cdot \mathbf{a} + t \cdot \mathbf{b}$$

Foreground estimation

$$\mathbf{I} = \alpha \cdot \mathbf{F} + (1 - \alpha) \cdot \mathbf{B} = \mathbf{F} \circ \mathbf{B}$$

$$\mathbf{I}(\mathbf{x}) = \mathbf{F}(\mathbf{x}) \circ \mathbf{B}(\mathbf{x})$$

$$\mathbf{I}(\mathbf{x}) = f(\mathbf{x}, \boldsymbol{\theta}) \circ \mathbf{B}(\mathbf{x})$$

$$\arg \min_{\boldsymbol{\theta}, \mathbf{B}} \sum_{\mathbf{x} \in R} \left(\mathbf{I}(\mathbf{x}) - f(\mathbf{x}, \boldsymbol{\theta}) \circ \mathbf{B}(\mathbf{x}) \right)^2$$

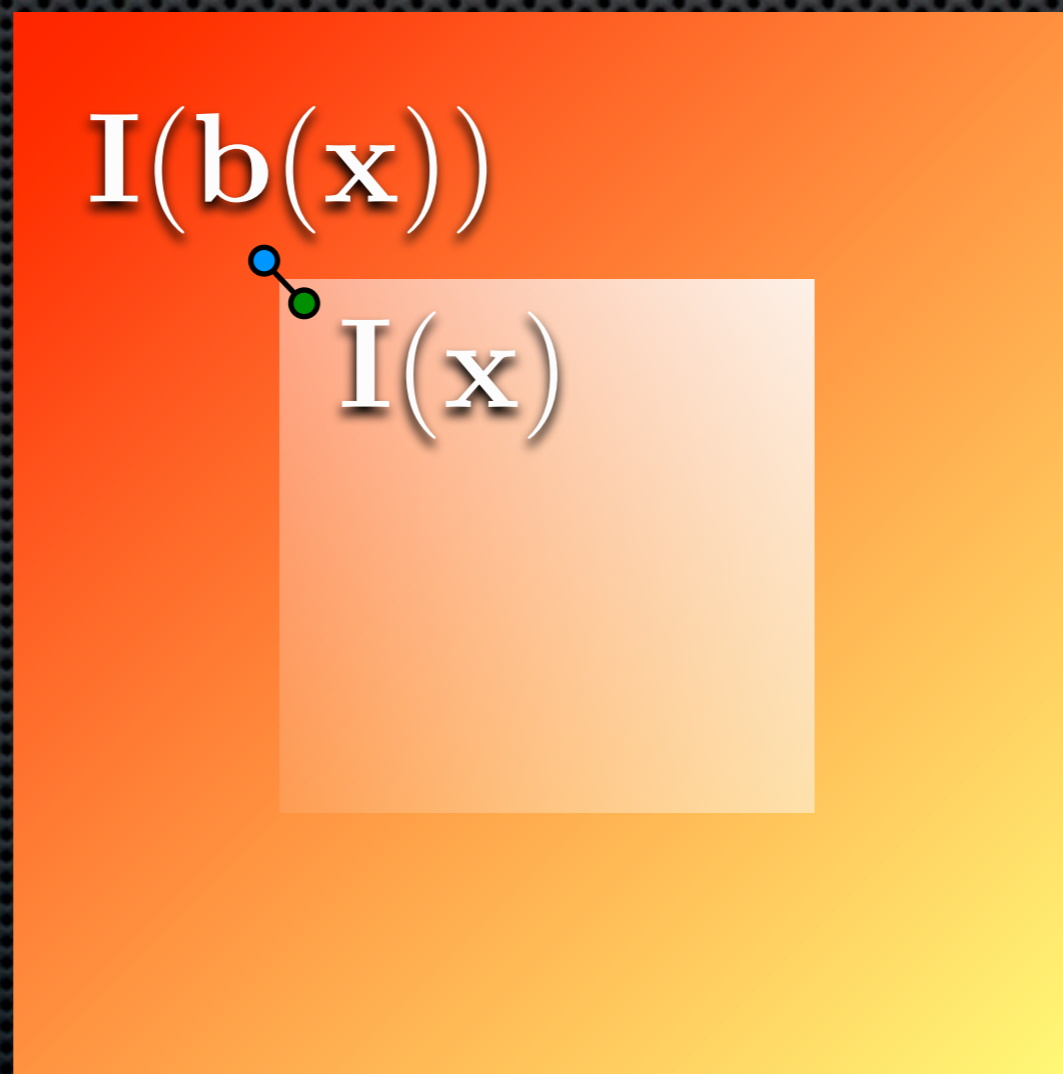
\mathbf{x} pixel position

$\boldsymbol{\theta}$ gradient parameters

R selected image region

Foreground estimation

$$\arg \min_{\theta, \mathbf{B}} \sum_{\mathbf{x} \in R} \left(\mathbf{I}(\mathbf{x}) - f(\mathbf{x}, \theta) \circ \mathbf{B}(\mathbf{x}) \right)^2$$

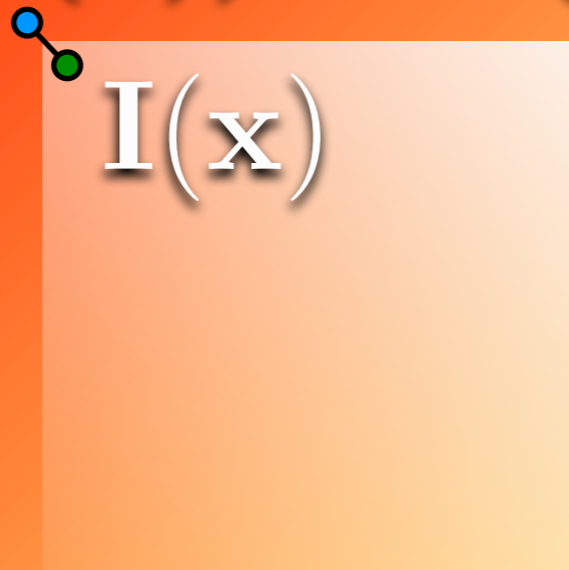


- \mathbf{x} pixel position
- θ gradient parameters
- R selected image region
- \mathbf{b} background sample

Foreground estimation

$$\arg \min_{\theta} \sum_{\mathbf{x} \in \partial R} \left(\mathbf{I}(\mathbf{x}) - f(\mathbf{x}, \theta) \circ \mathbf{I}(\mathbf{b}(\mathbf{x})) \right)^2$$

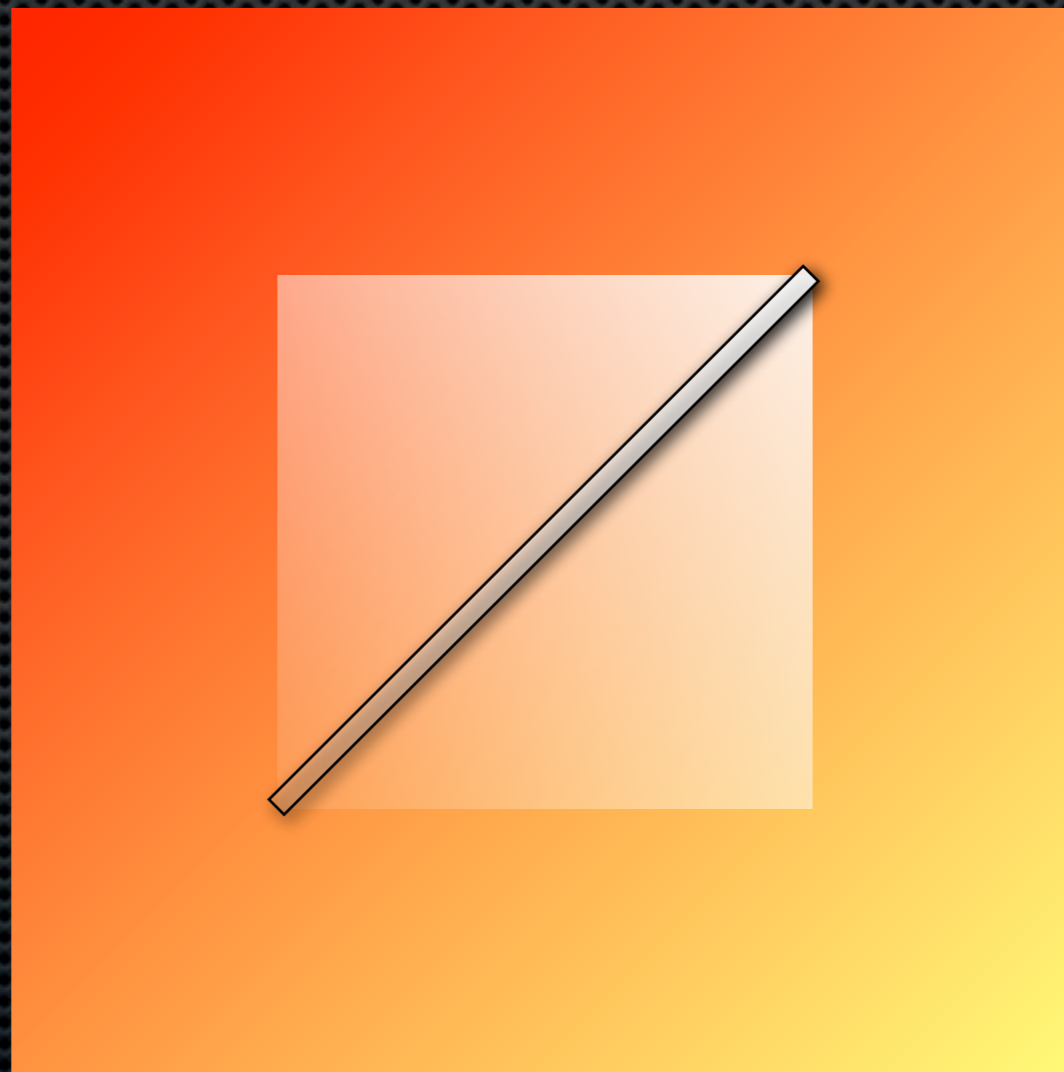
$$\mathbf{I}(\mathbf{b}(\mathbf{x})) \approx \mathbf{B}(\mathbf{x})$$



- \mathbf{x} pixel position
- θ gradient parameters
- R selected image region
- \mathbf{b} background sample
- ∂R region boundary

Foreground estimation

$$\arg \min_{\theta} \sum_{\mathbf{x} \in \partial R} \left(\mathbf{I}(\mathbf{x}) - f(\mathbf{x}, \theta) \circ \mathbf{I}(\mathbf{b}(\mathbf{x})) \right)^2$$



- \mathbf{x} pixel position
- θ gradient parameters
- R selected image region
- \mathbf{b} background sample
- ∂R region boundary

Background estimation



Input photo (slightly blurred)

Background estimation



Input photo (slightly blurred)

Background estimation



Input photo (slightly blurred)

Background estimation



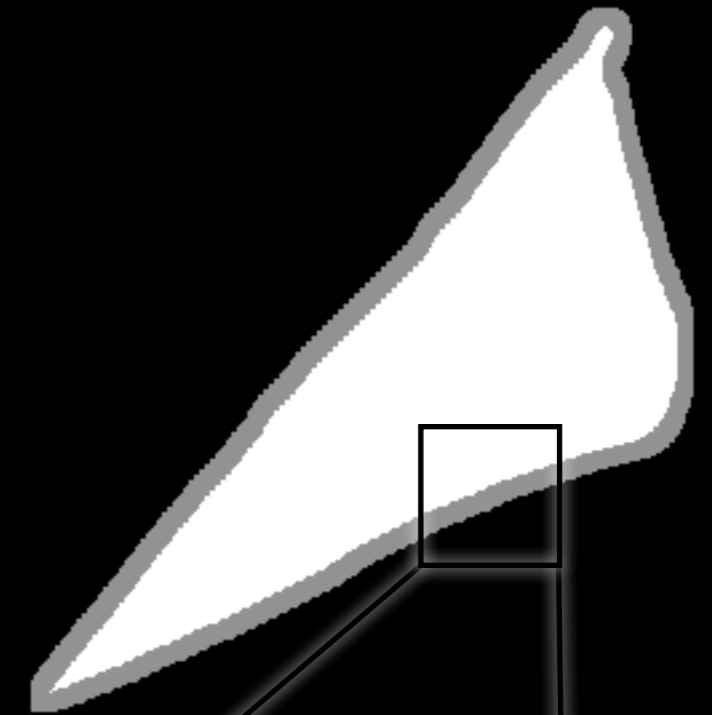
with hard region boundary

Background estimation



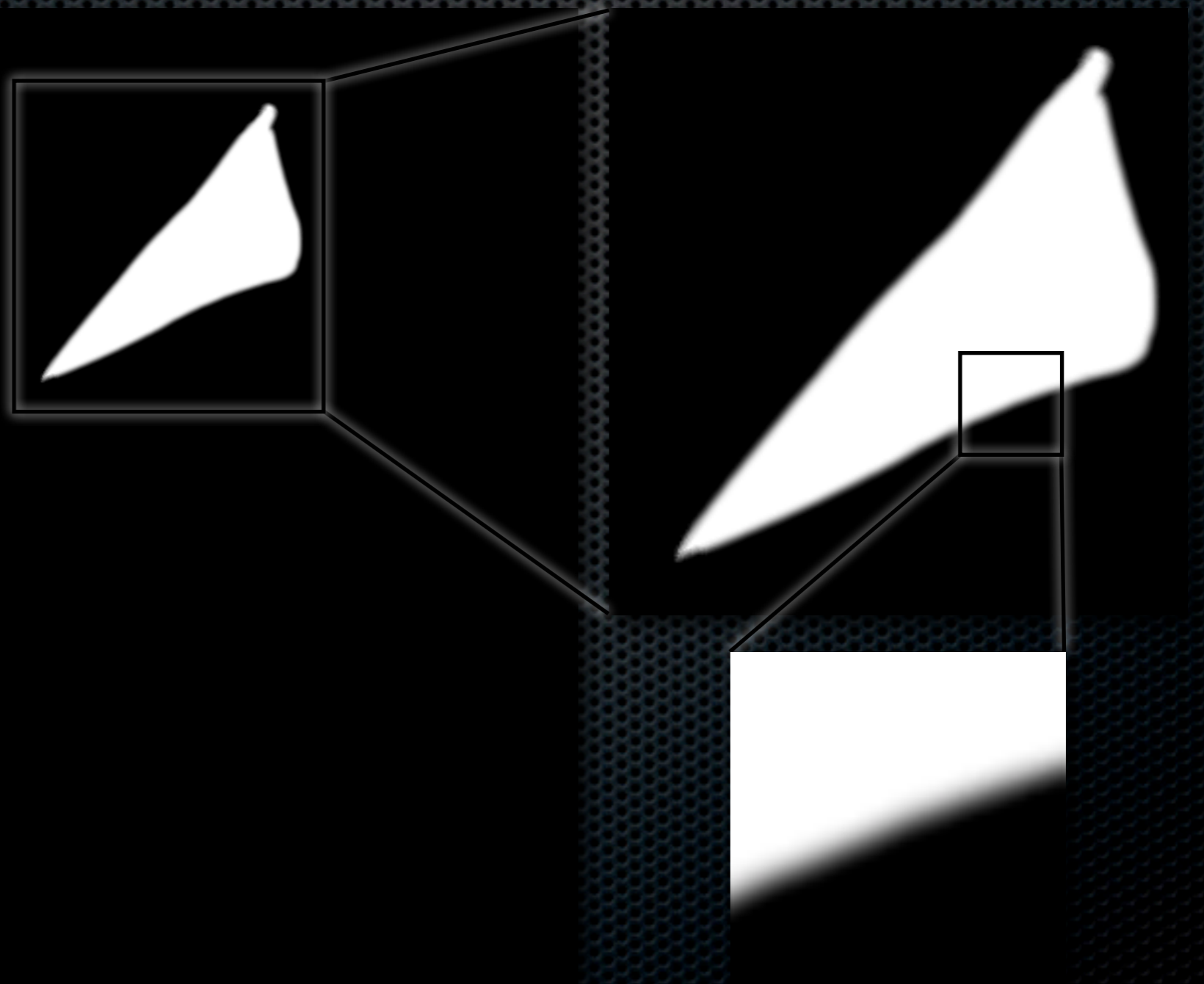
hard region boundary

Background estimation



trimap from hard region boundary

Background estimation



matted region boundary

Background estimation



with matted region boundary

Background estimation



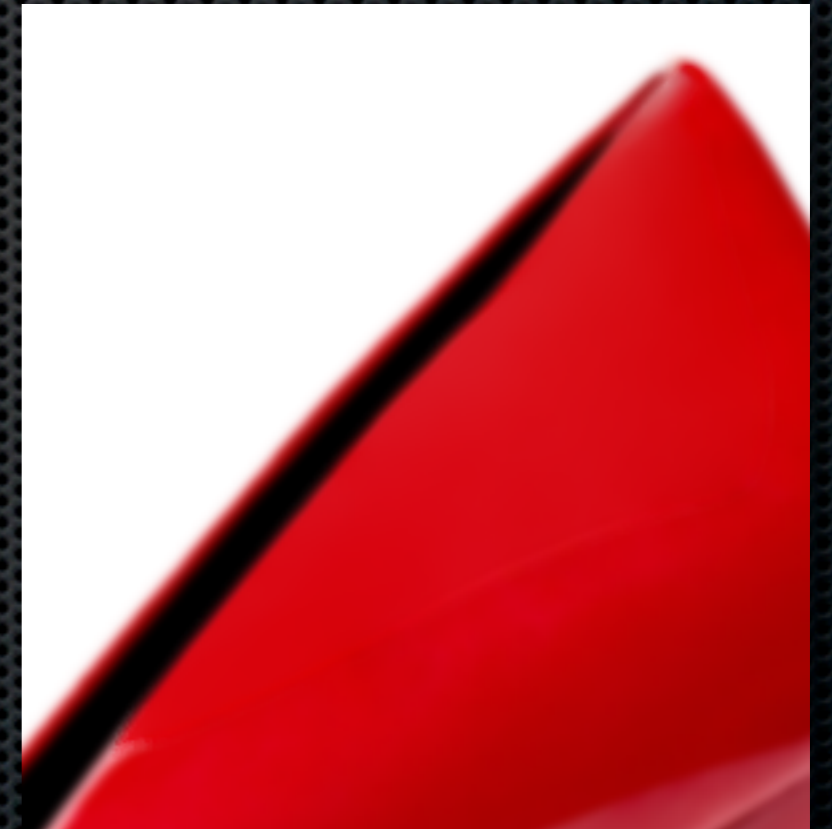
plus TV-smoothed region

Background estimation



plus Poisson blending

Background estimation



plus Poisson blending

Summary

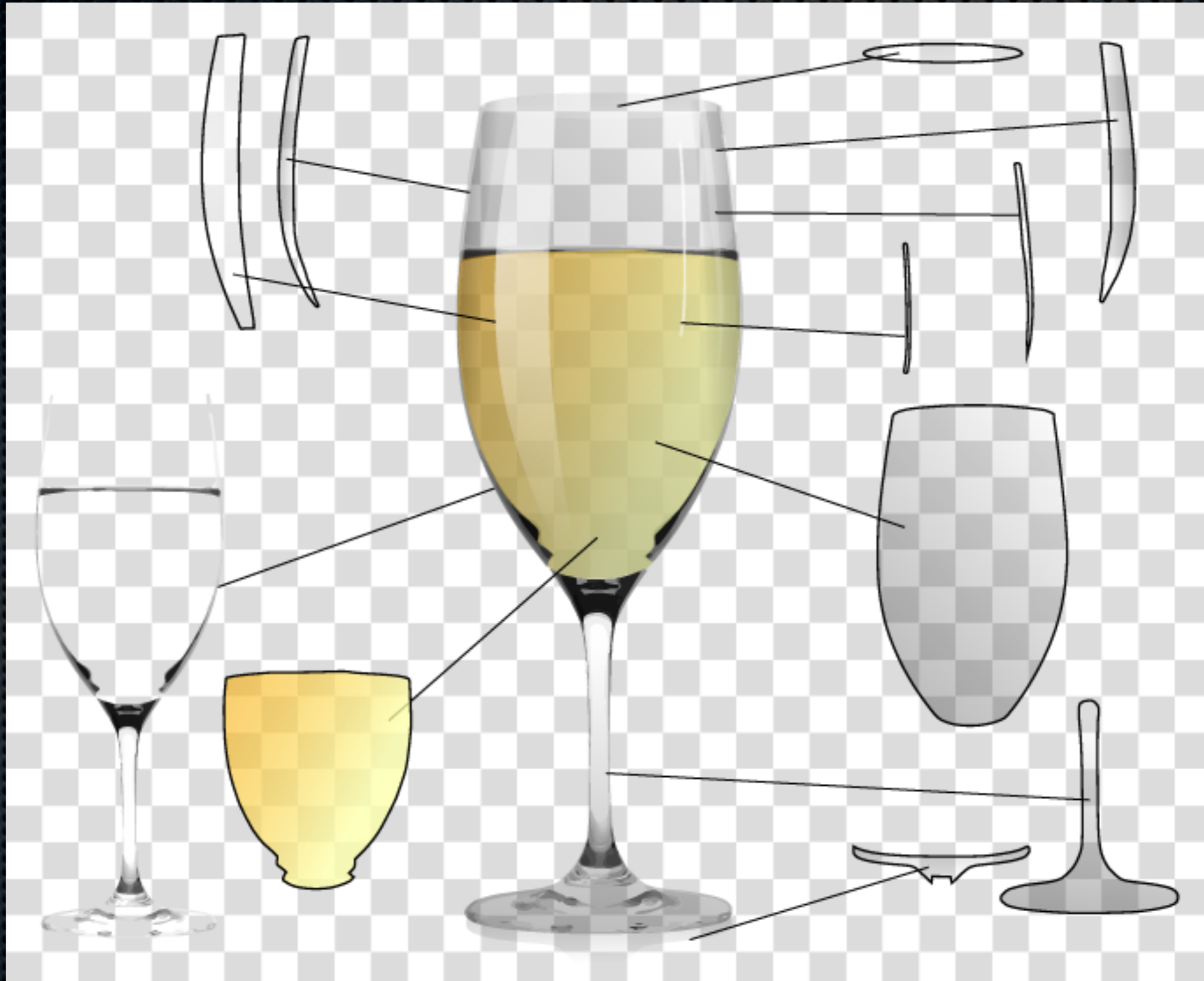
1. joint decompositing and vectorisation of foreground:
 - ✦ solving the matting problem around region boundary
 - ✦ strong prior on foreground + user input
2. background estimation by optimisation:
 - ✦ similar to inversion of compositing equation
 - ✦ additional terms to remove residuals:
TV smoothness + Poisson blending

Input photo

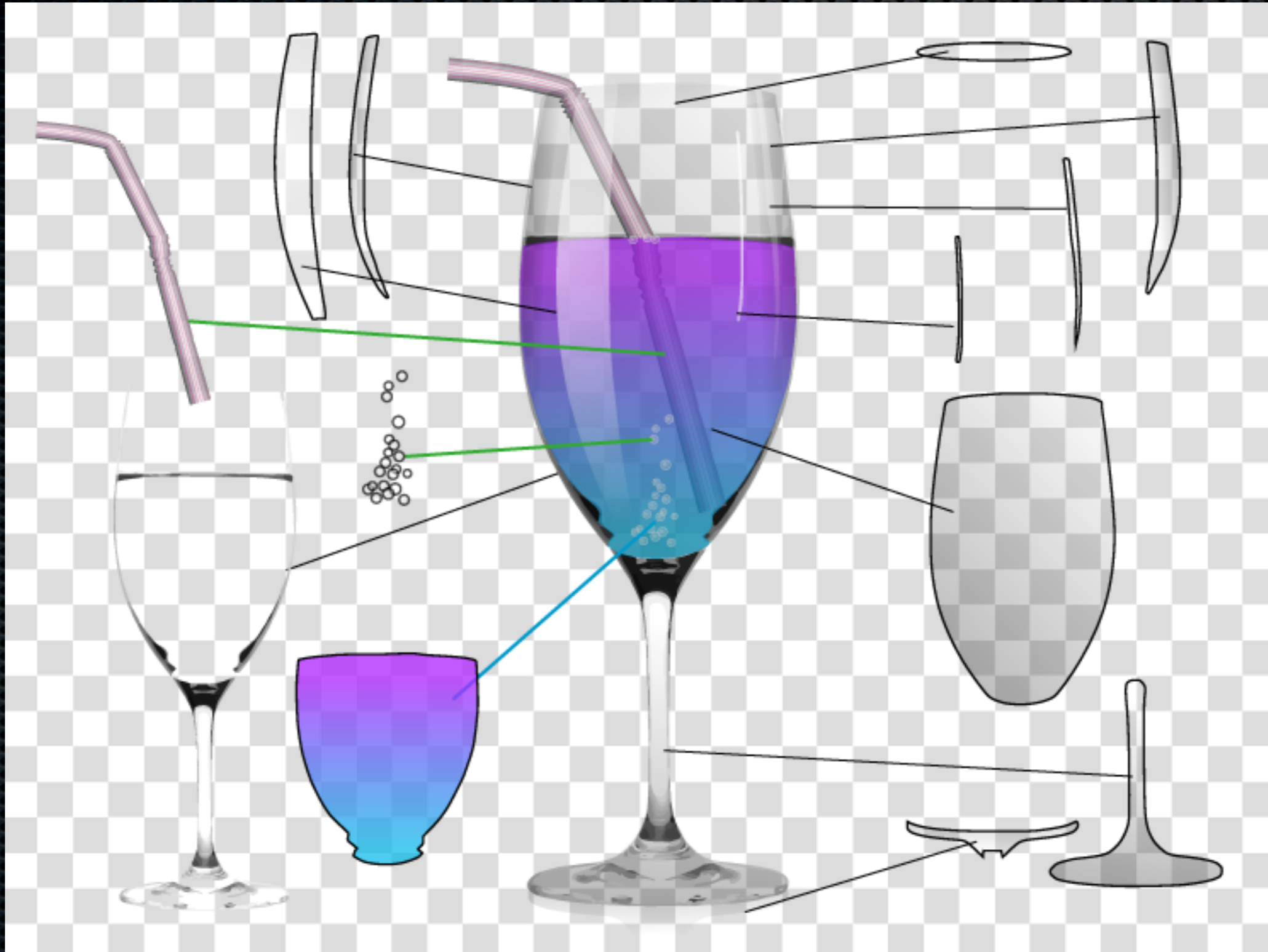


Shutterstock/Givaga

Vectorised result



Editing result

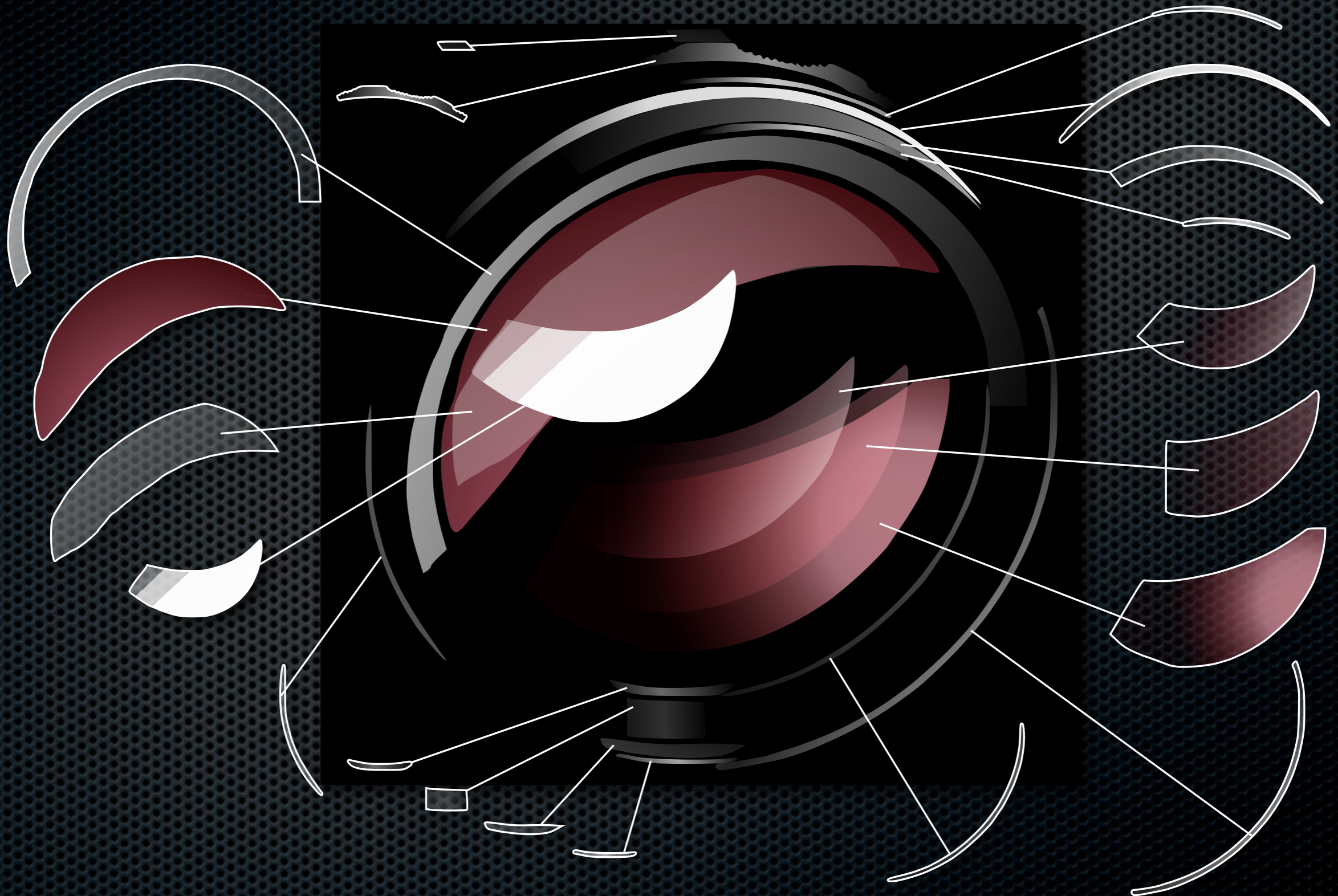


Input photo

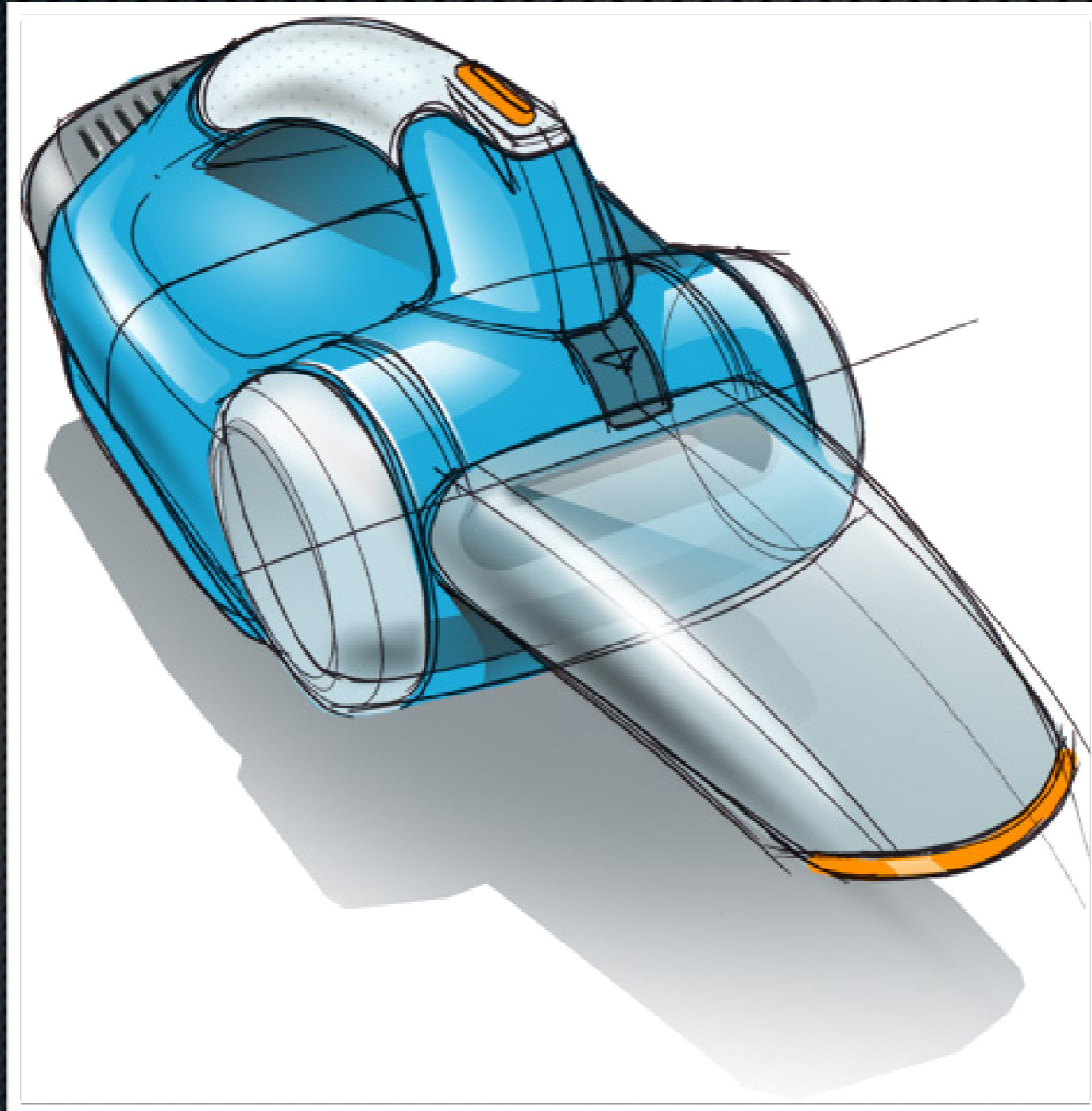


Flickr/squinza (CC BY-SA 2.0)

Vectorised result

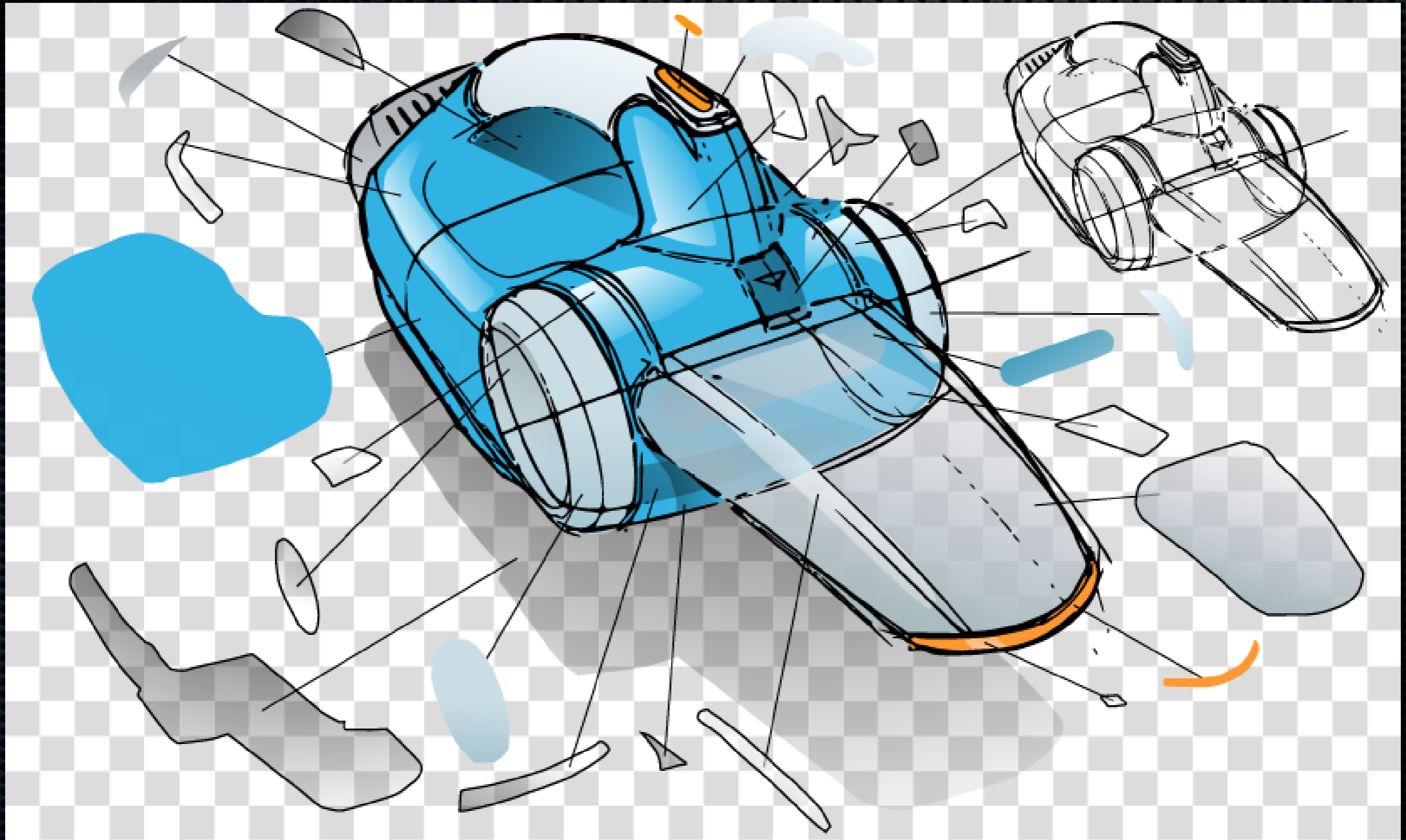


Input drawing

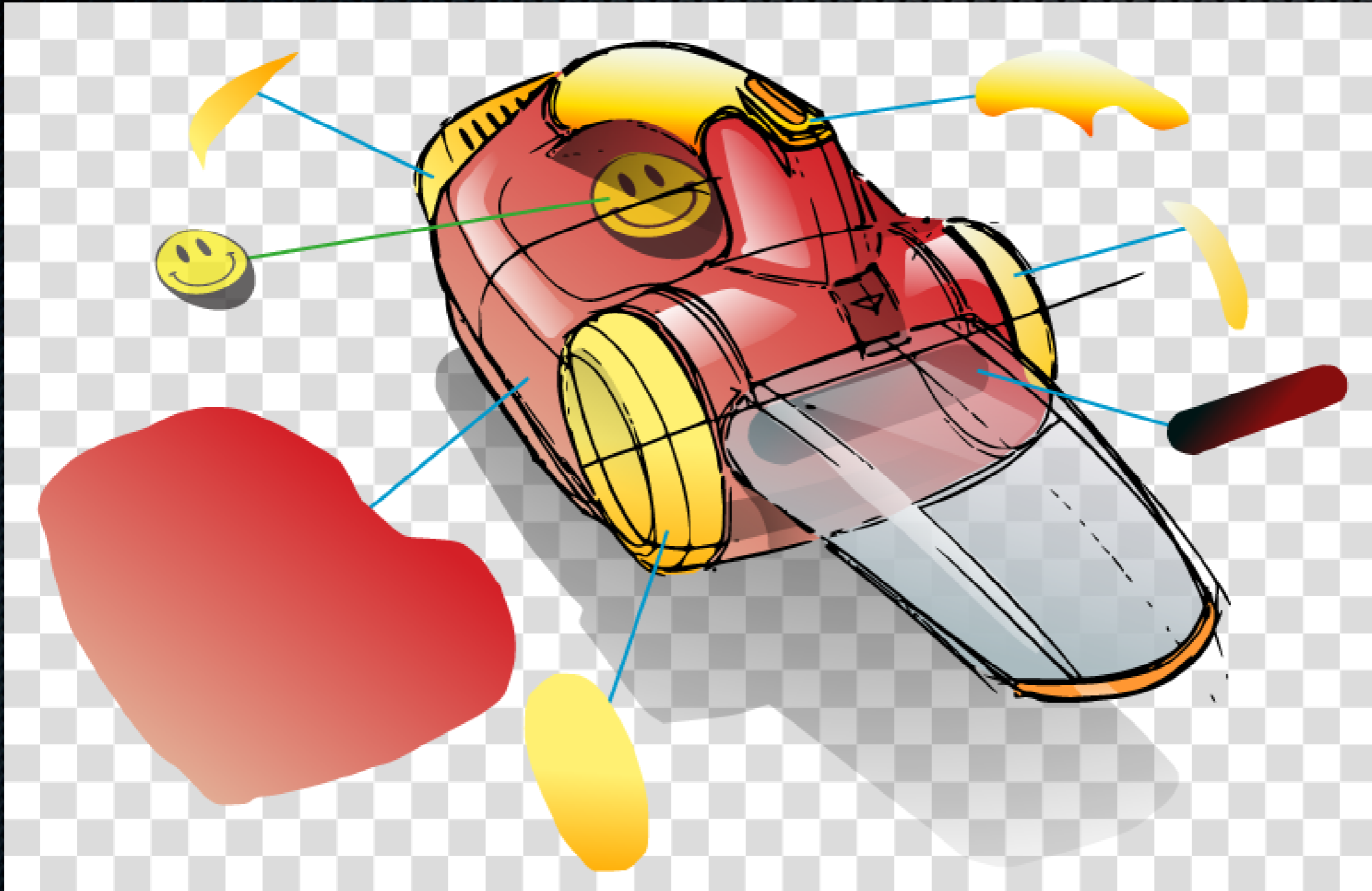


Spencer Nugent

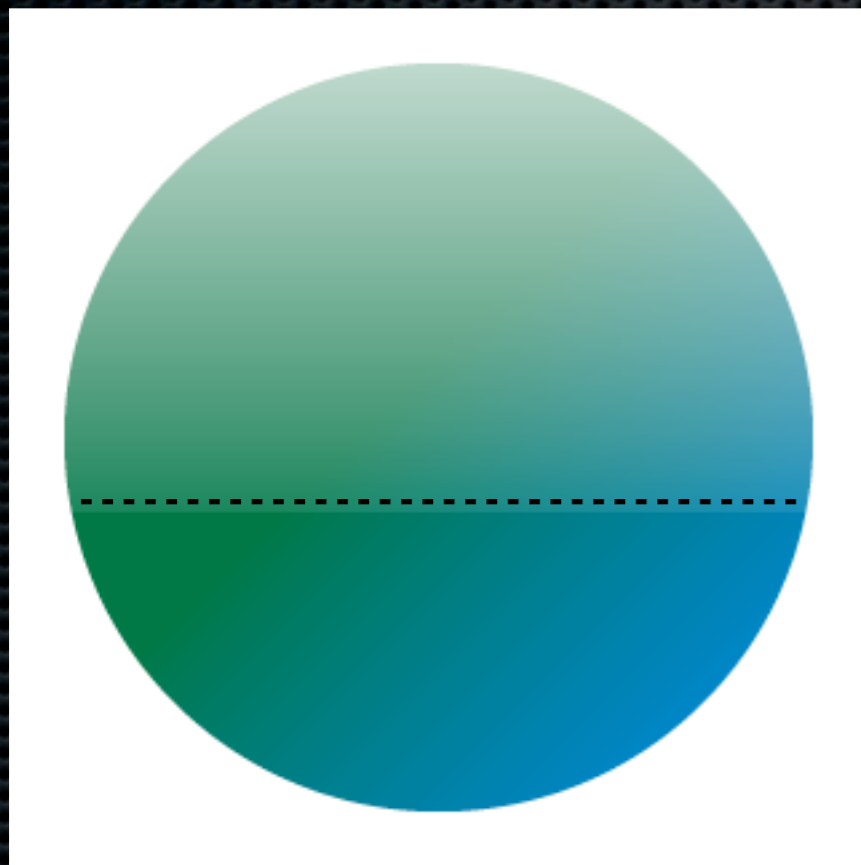
Vectorised result



Editing result

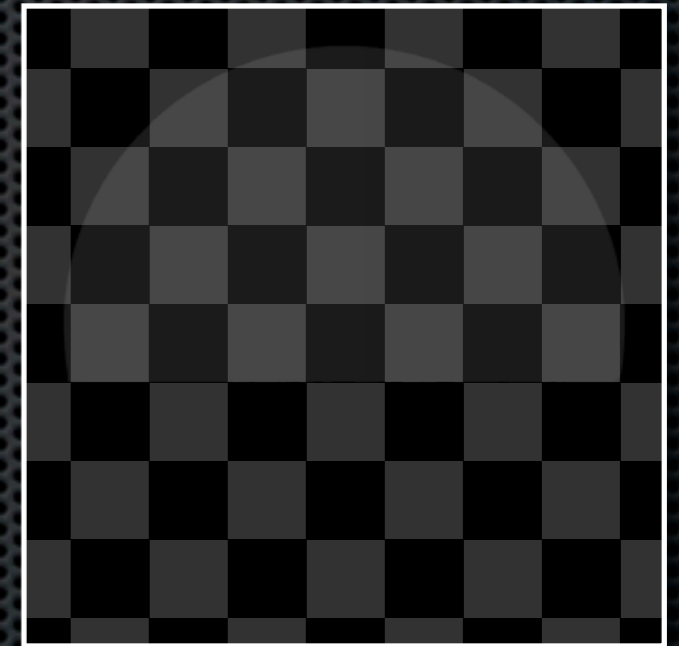
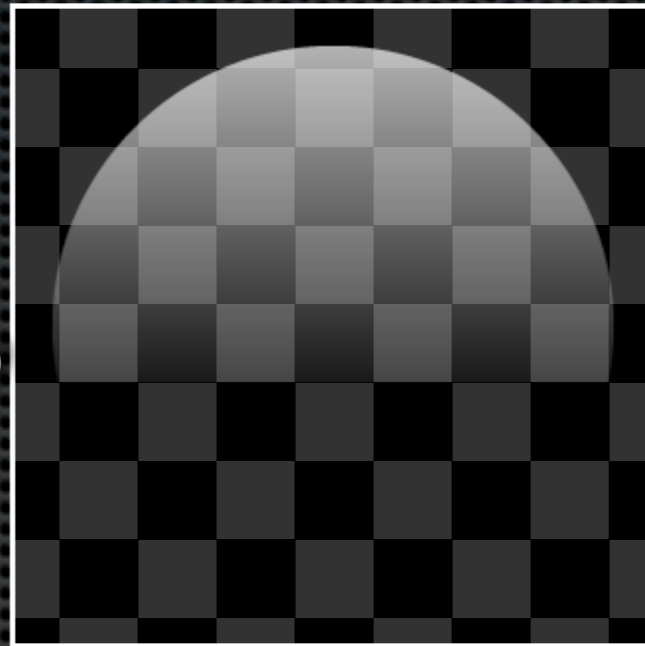


Limitation: few iso-contours

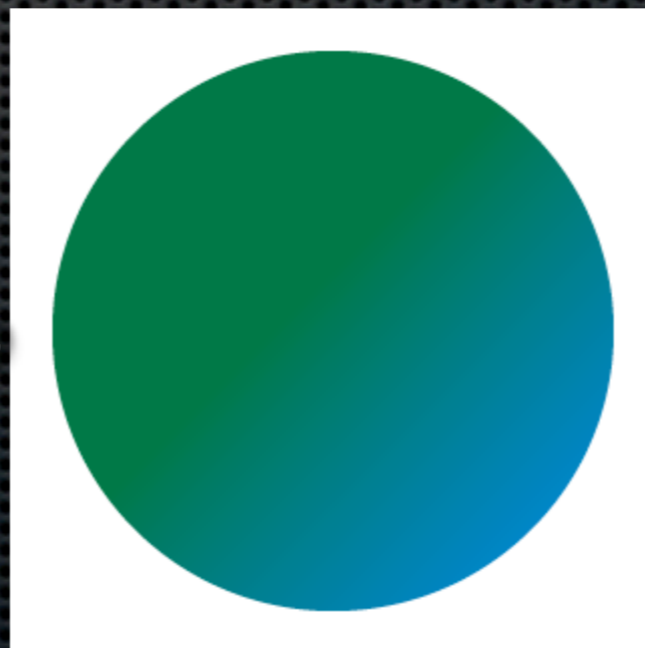


input image

foreground



background



ground truth

our decomposition

Limitation: background textures



input photo



our recomposited result

Limitation: background textures



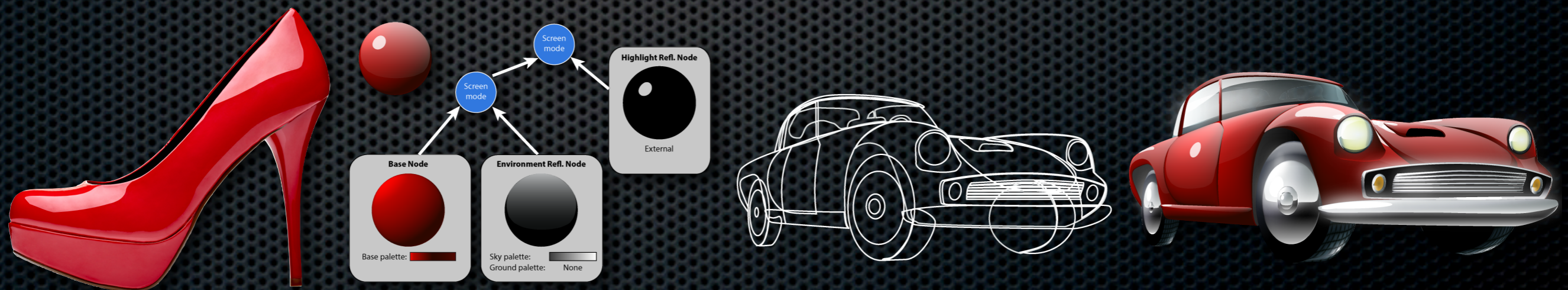
input photo



estimated background

Future work

- ✦ more complex semi-transparent vector primitives
- ✦ automatic segmentation and decompositing
- ✦ extract Vector Shade Trees [Lopez-Moreno+ 2013] from exemplar materials





Conclusion



- ✦ key insight: complex images can often be explained by stacking simple layers
- ✦ first approach creating layered vector art from bitmaps:
 - ✦ opaque and semi-transparent gradient layers
 - ✦ produces a simple, editable stack of vector layers
 - ✦ valuable for professionals and novices alike