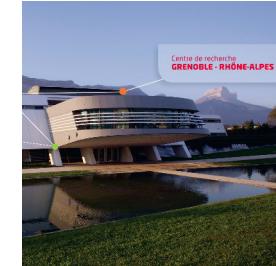




# Background



Team MAVERICK at LJK, University of Grenoble (LJK) & INRIA Grenoble

Realistic rendering



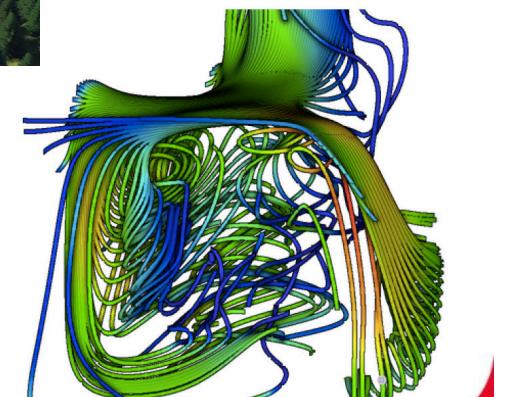
Complex scenes



Expressive rendering



Visualization



# Advancing fronts in SciVi

- Theoretical approaches:
  - Specific scientific domain
  - Multiresolution
  - Topology
    - EuroVis2017, Giri Nucha (IISc), G.-P. Bonneau & al.
    - Computing Contour Tree for Higher Order Polynomials
- Visual Perception and Cognition awareness in Visualization



# Visual Perception

motivation, basic facts

1

# Motivation

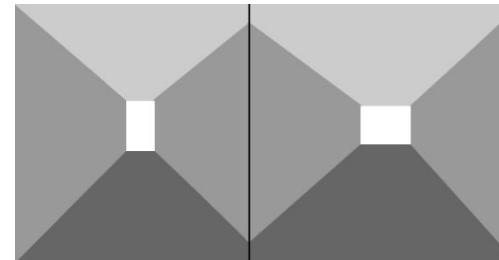


<http://nivea.psycho.univ-paris5.fr/>

# Visual Perception

Perception is the organization, identification and interpretation of sensory information in order to represent and understand the environment.

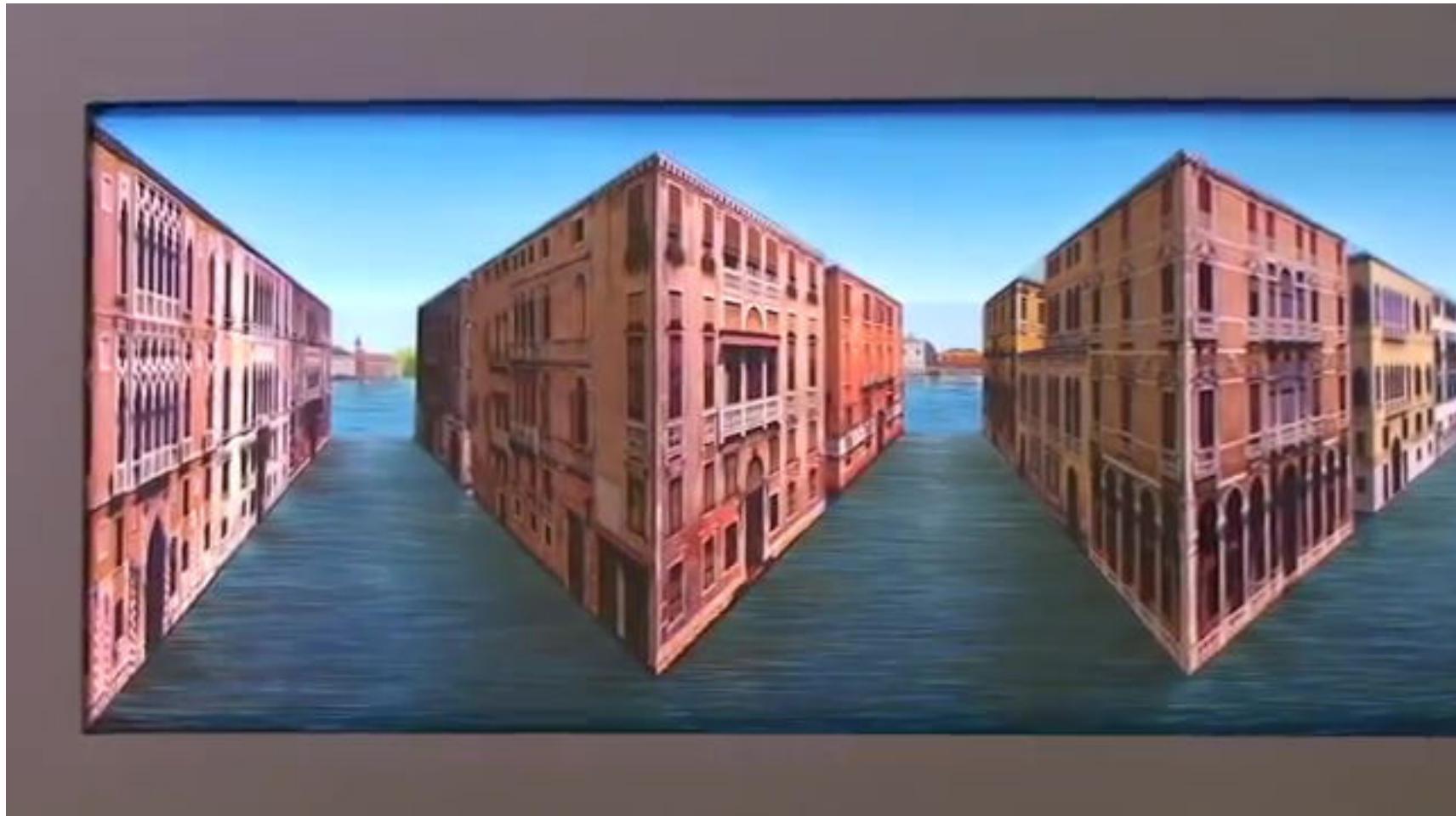
« Perceiving is identify something somewhere, and this need for localized identification persists even when we look at some abstract configurations, for which we cannot build on any previous experience » (Gombrich, in *Art and Illusion*)



Reverspective  
<http://www.patrickhughes.co.uk/>



# Reverspective



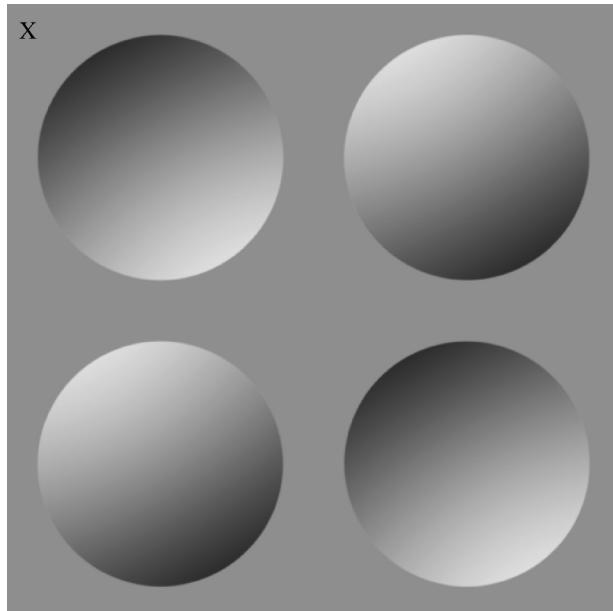
<http://www.patrickhughes.co.uk/>

# A priori in Visual perception

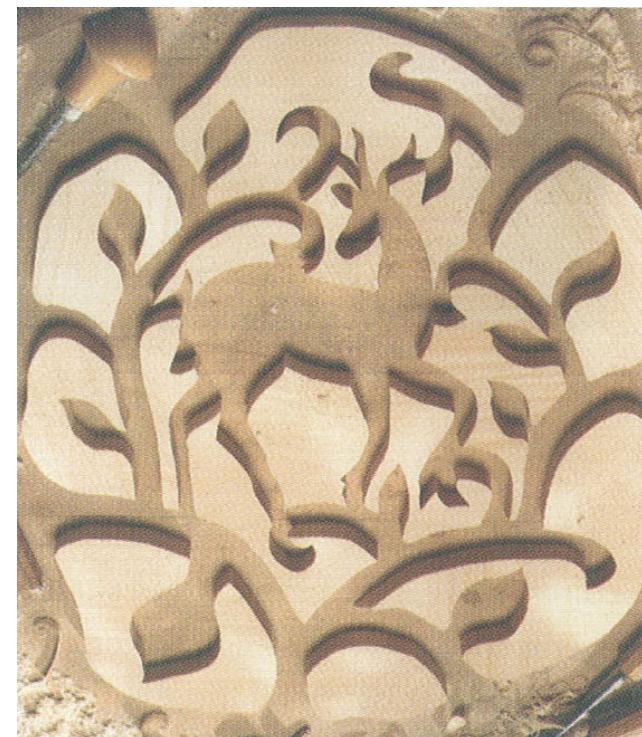


<http://michaelbach.de/>

# A priori in Visual perception



[Stone&al.,2009]

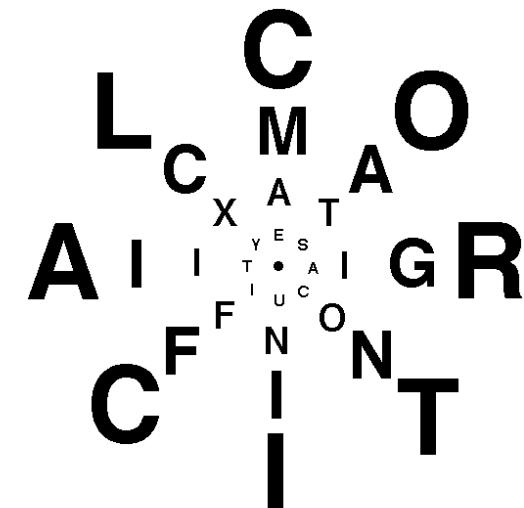
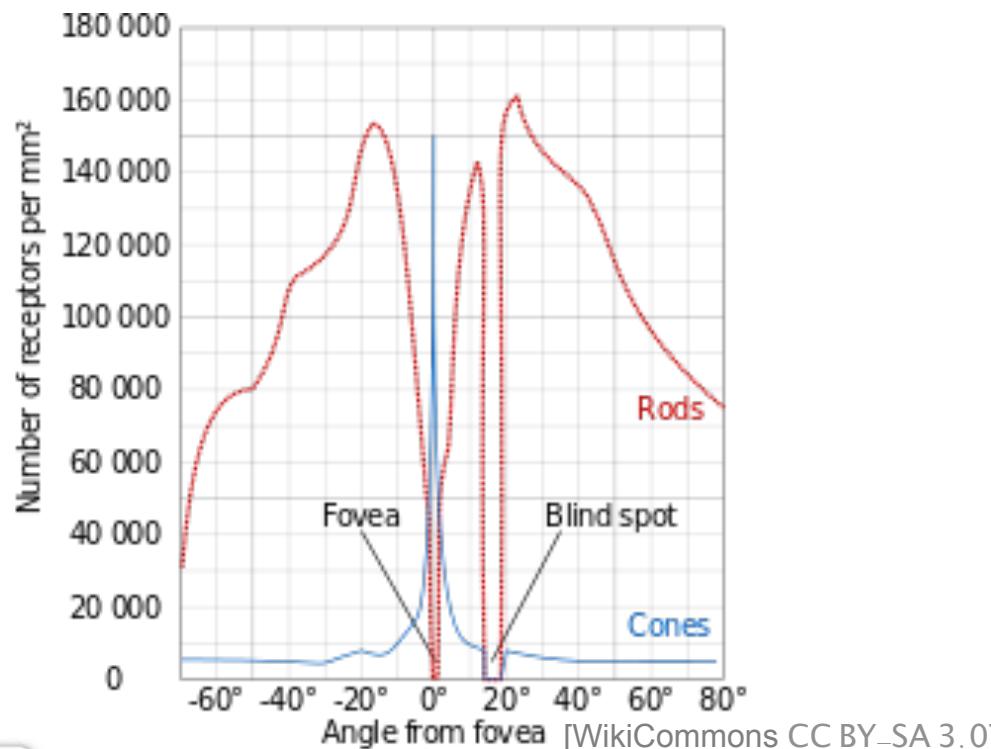


Basic Vision, pp. 226

# Visual Acuity

Contrast: 120 millions rods

Color: 5 millions cones

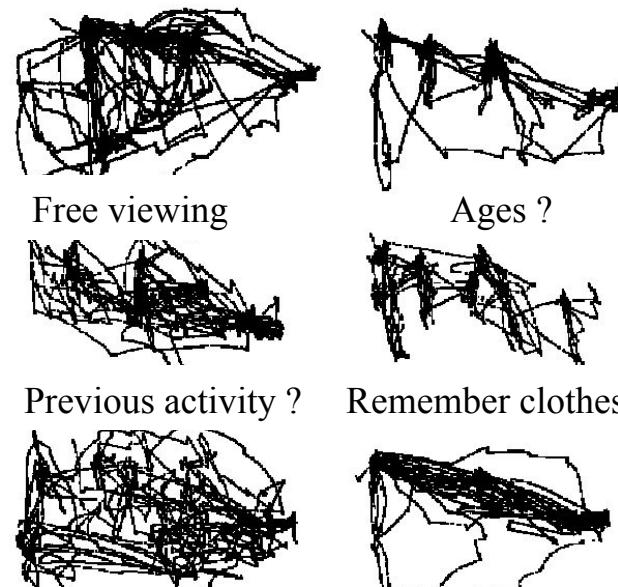


Anstis, S. (1974).

# Saccadic eye movements

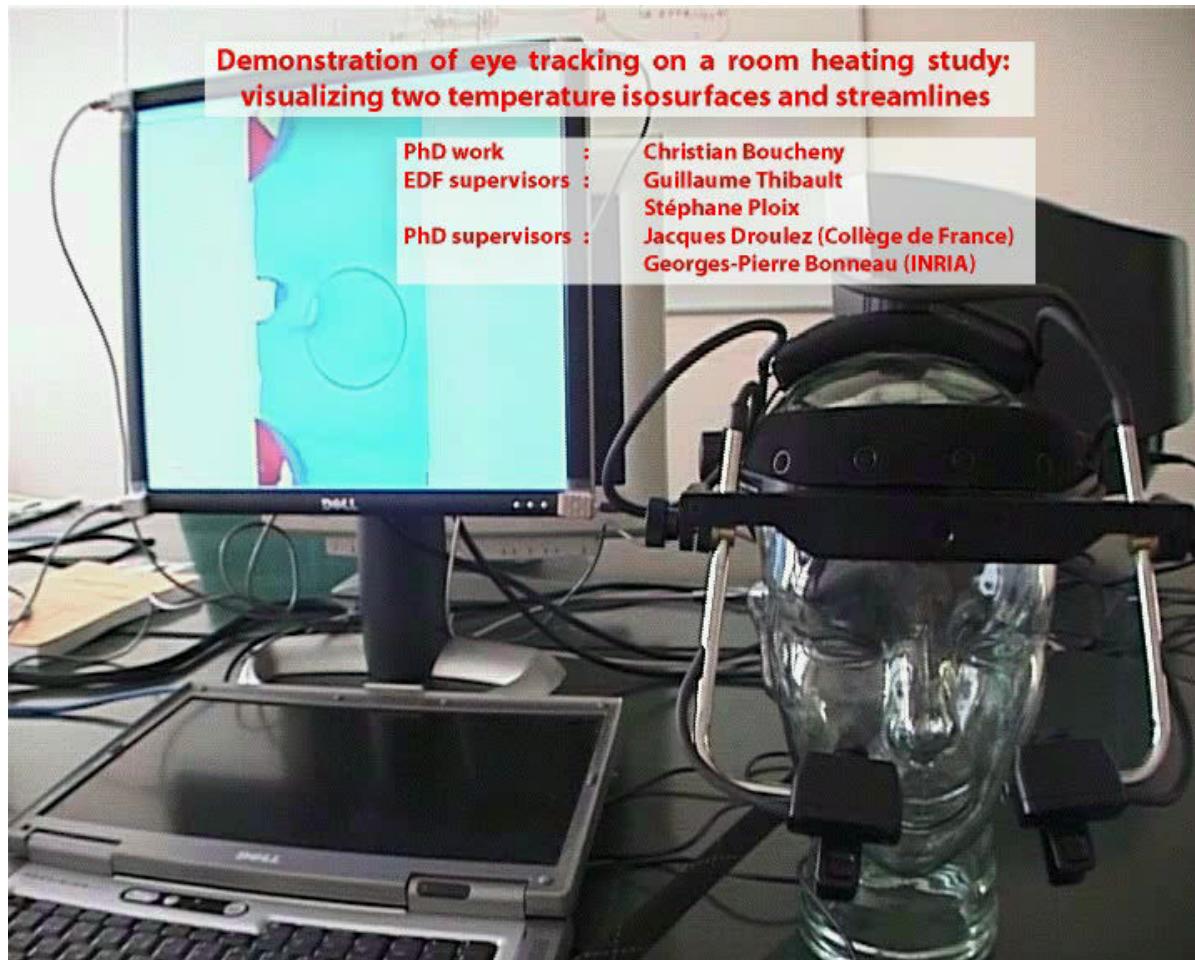


I. E. Repin, *Un visiteur inattendu* (1884)



Yarbus, Eye movement and Vision, 1967, fig. 109

# Gaze guided visualization





# Depth perception

In Direct Volume Rendering

# Depth perception



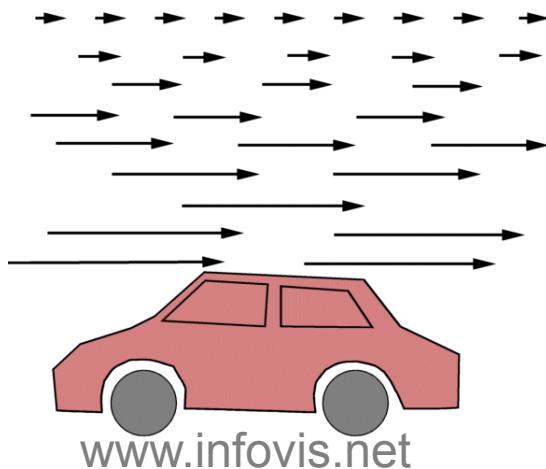
wikipedia

Ten monocular cues

Two binocular cues



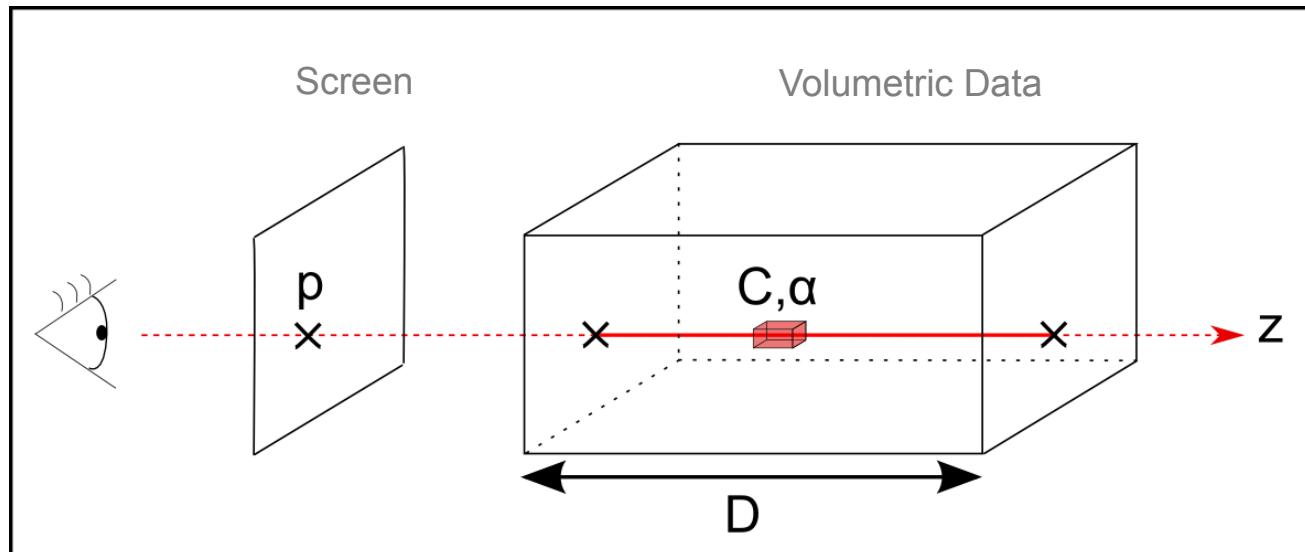
wikipedia



[www.infovis.net](http://www.infovis.net)

# Direct Volume Rendering

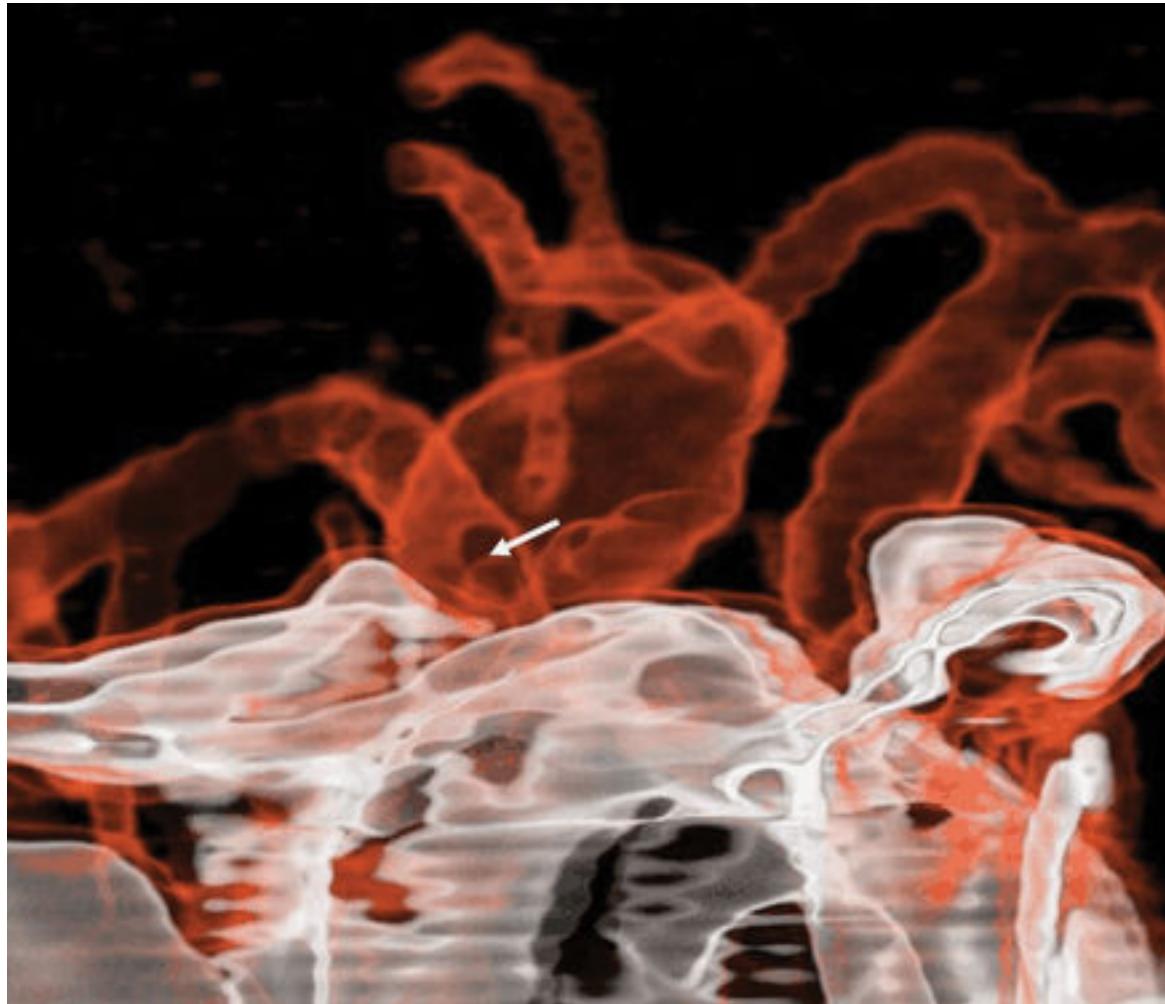
$$\text{TF} : \mathbf{s} \longrightarrow (\mathbf{C}, \alpha)$$



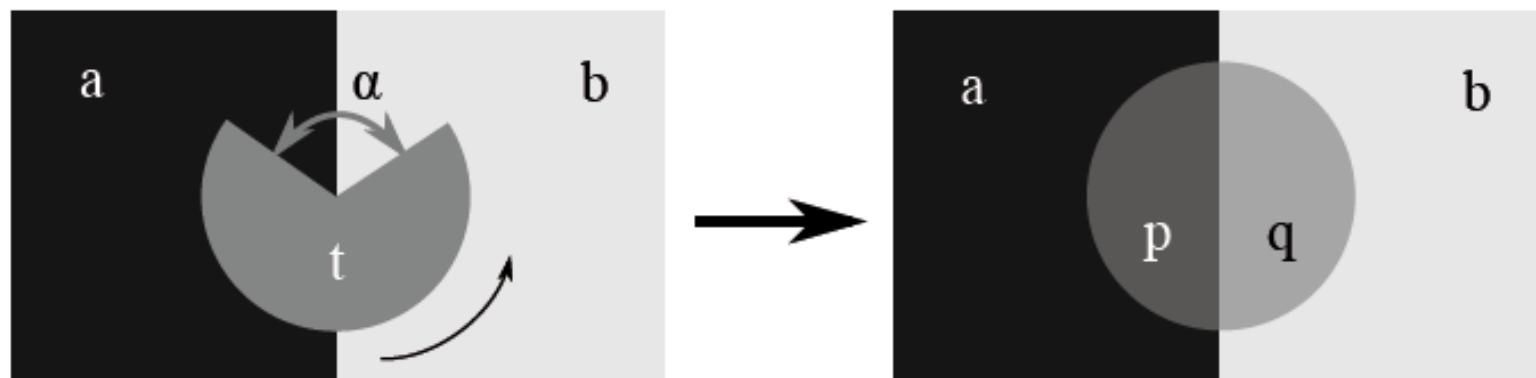
$$C(p) \approx \sum_{i=0}^n C_i \prod_{j=0}^{i-1} (1 - \alpha_j)$$

# Direct Volume Rendering

[radiographics.rsna.org](http://radiographics.rsna.org)



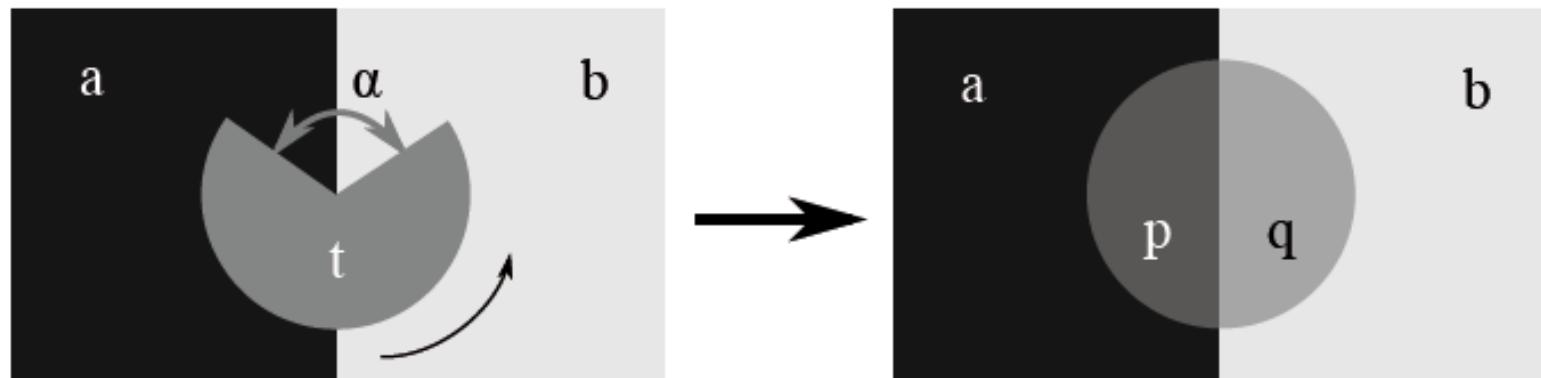
# Transparency perception



$p=?$   
 $q=?$

[Metelli, 1974]

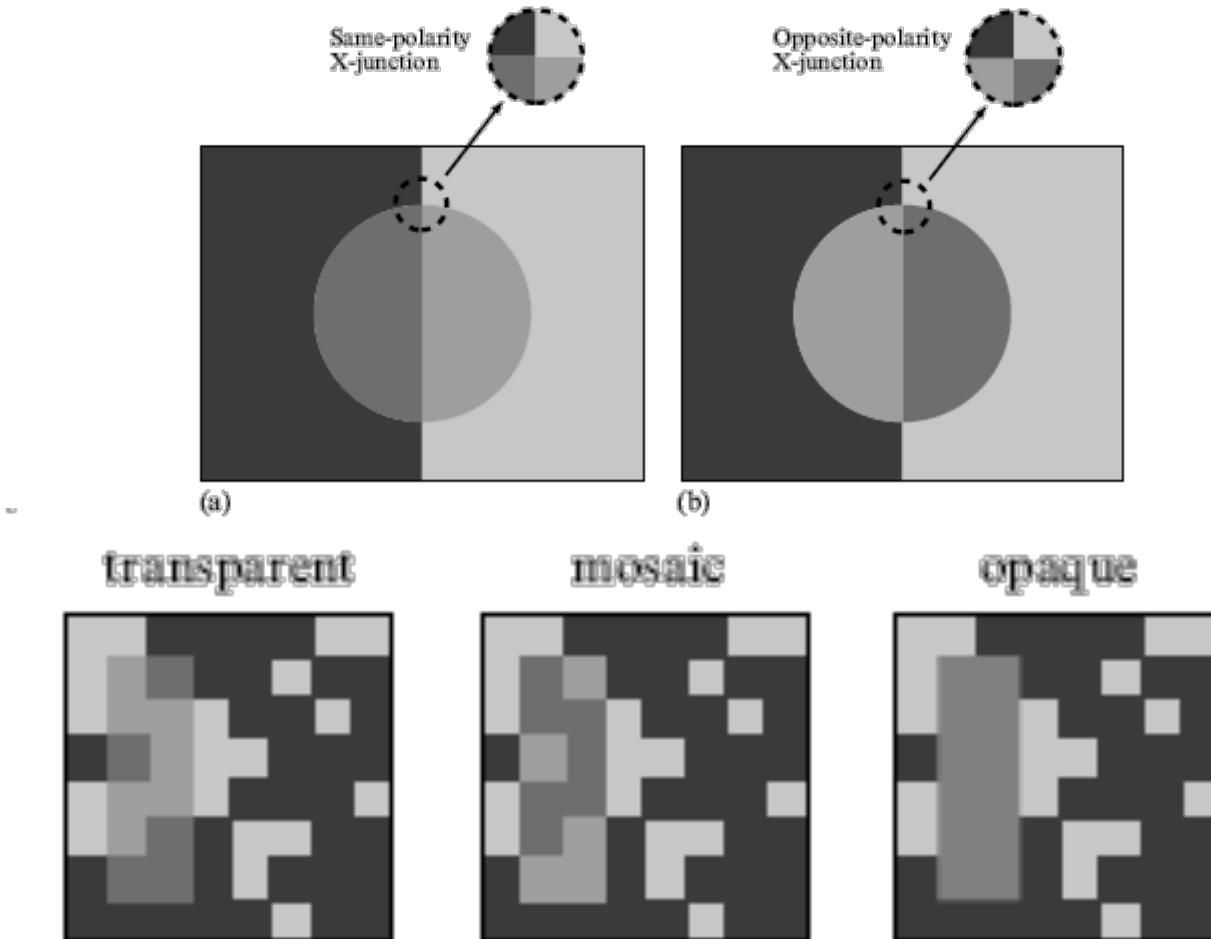
# Transparency perception



$$p = (1-\alpha) t + \alpha a$$
$$q = (1-\alpha) t + \alpha b$$

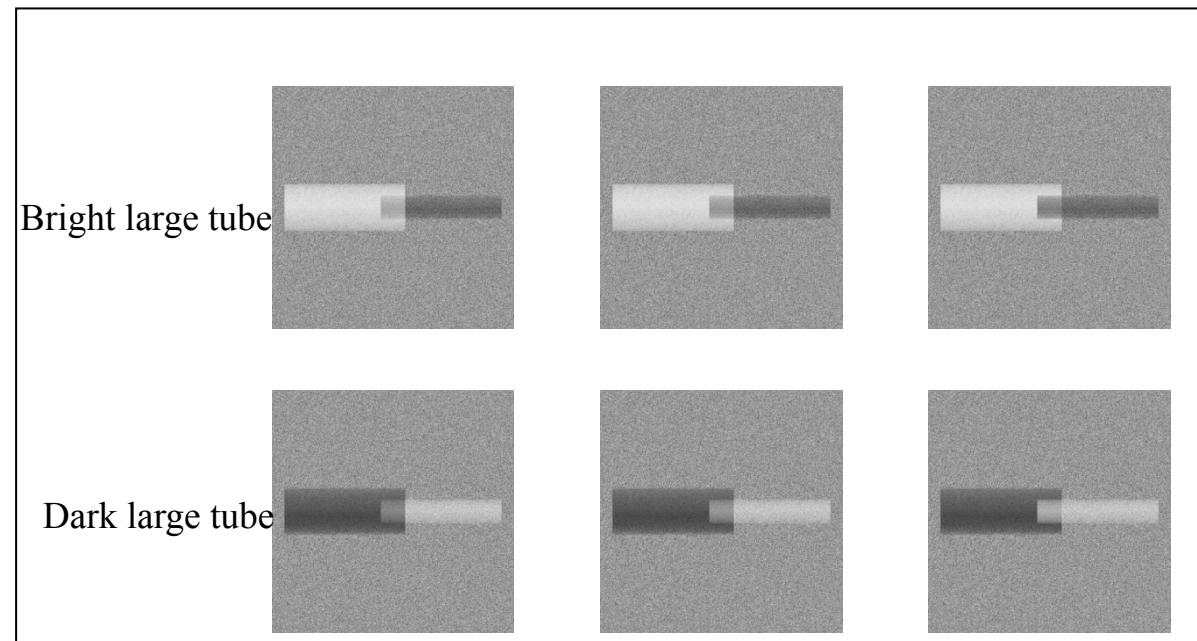
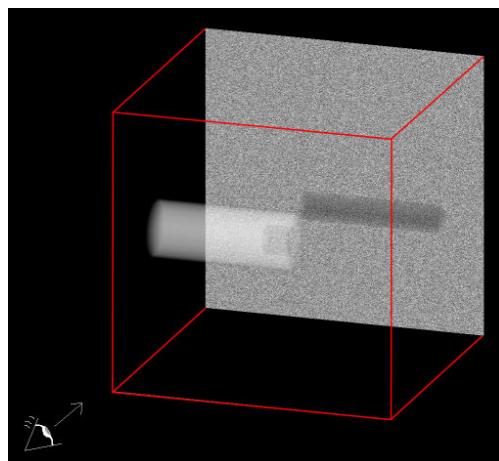
[Metelli, 1974]

# Transparency perception



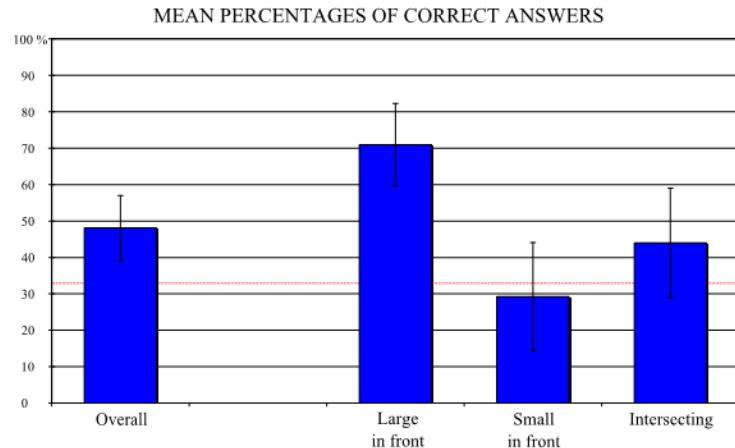
# Evaluation of Depth Perception in DVR

## Static experiment

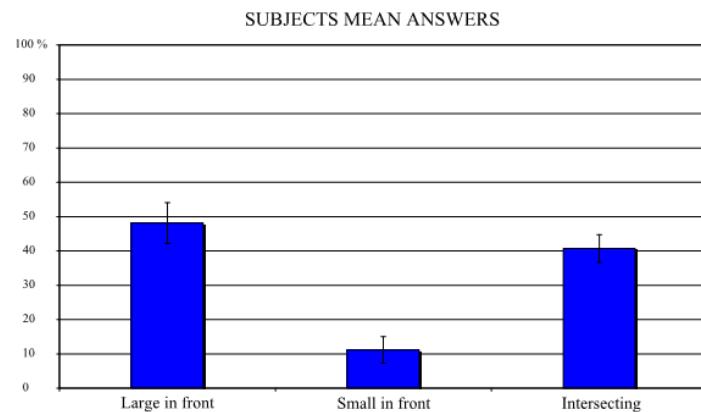


Boucheny, Bonneau & al, APGV 2007

# Static experiment results

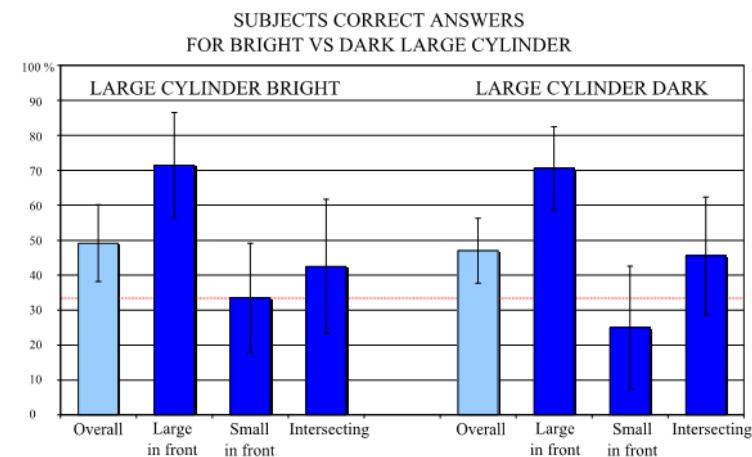


- Poor global performances



Bias for Large tube in front

Boucheny, Bonneau & al, APGV 2007



Luminance does not impact

# Dynamic experiment

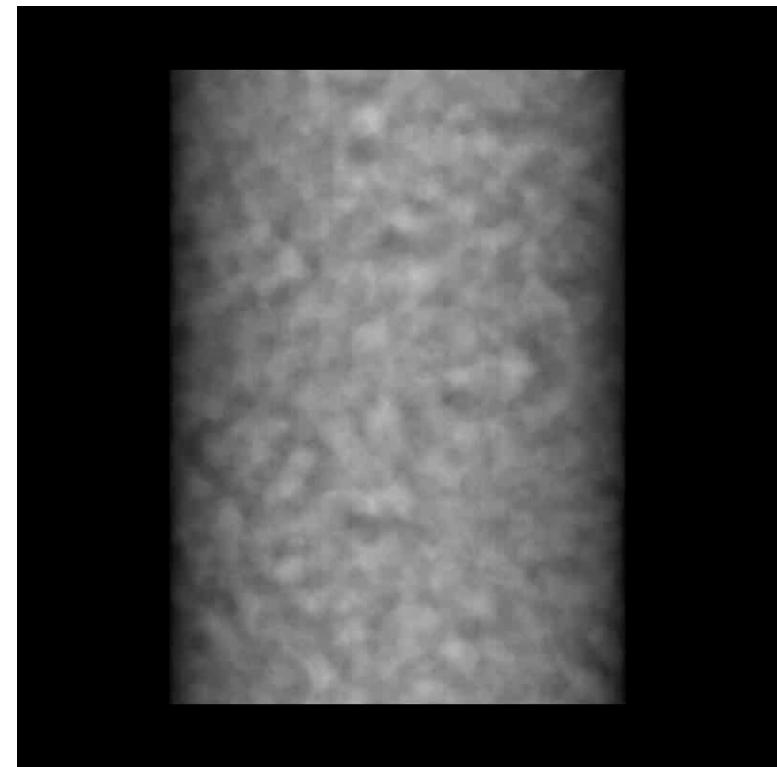
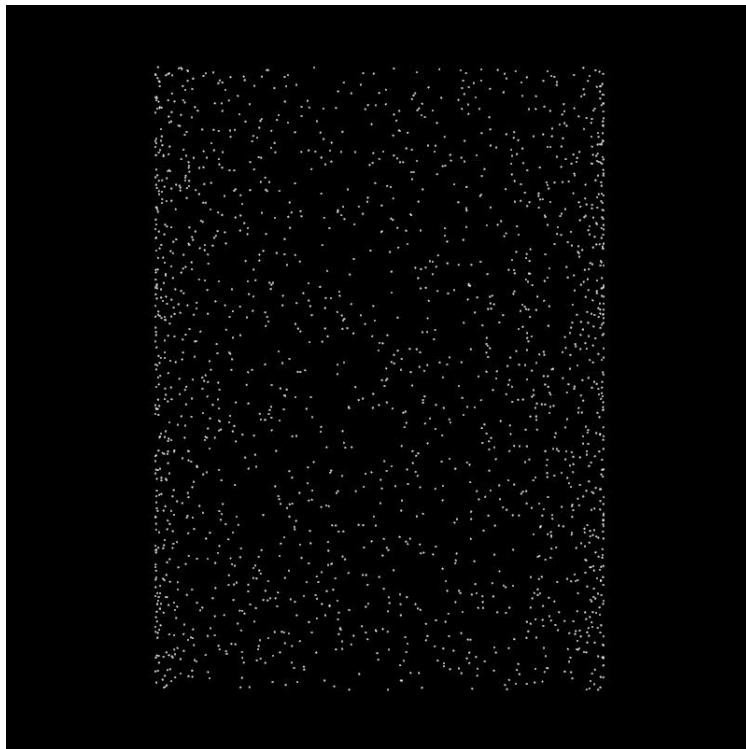
related to the Kinetic Depth Effect



Conical Pendulum Motion

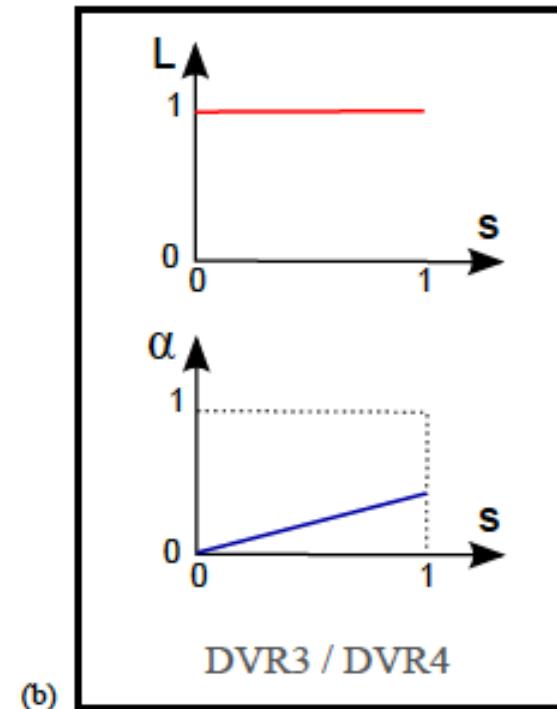
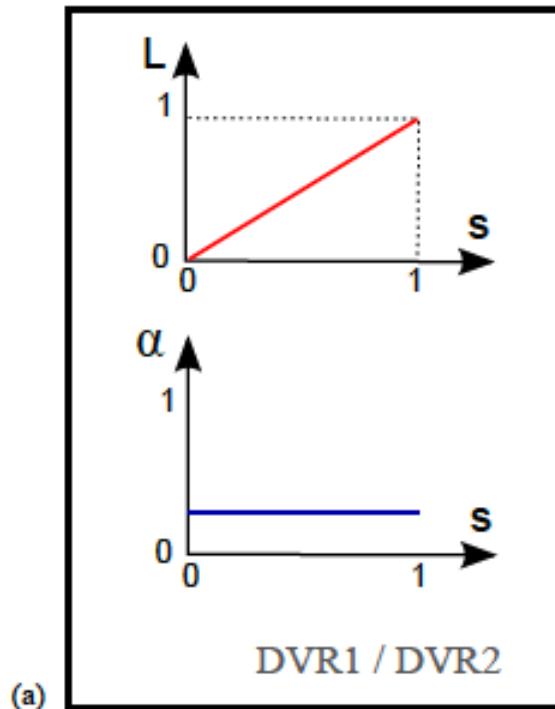
[Bista&al 2016]

# Dynamic experiment

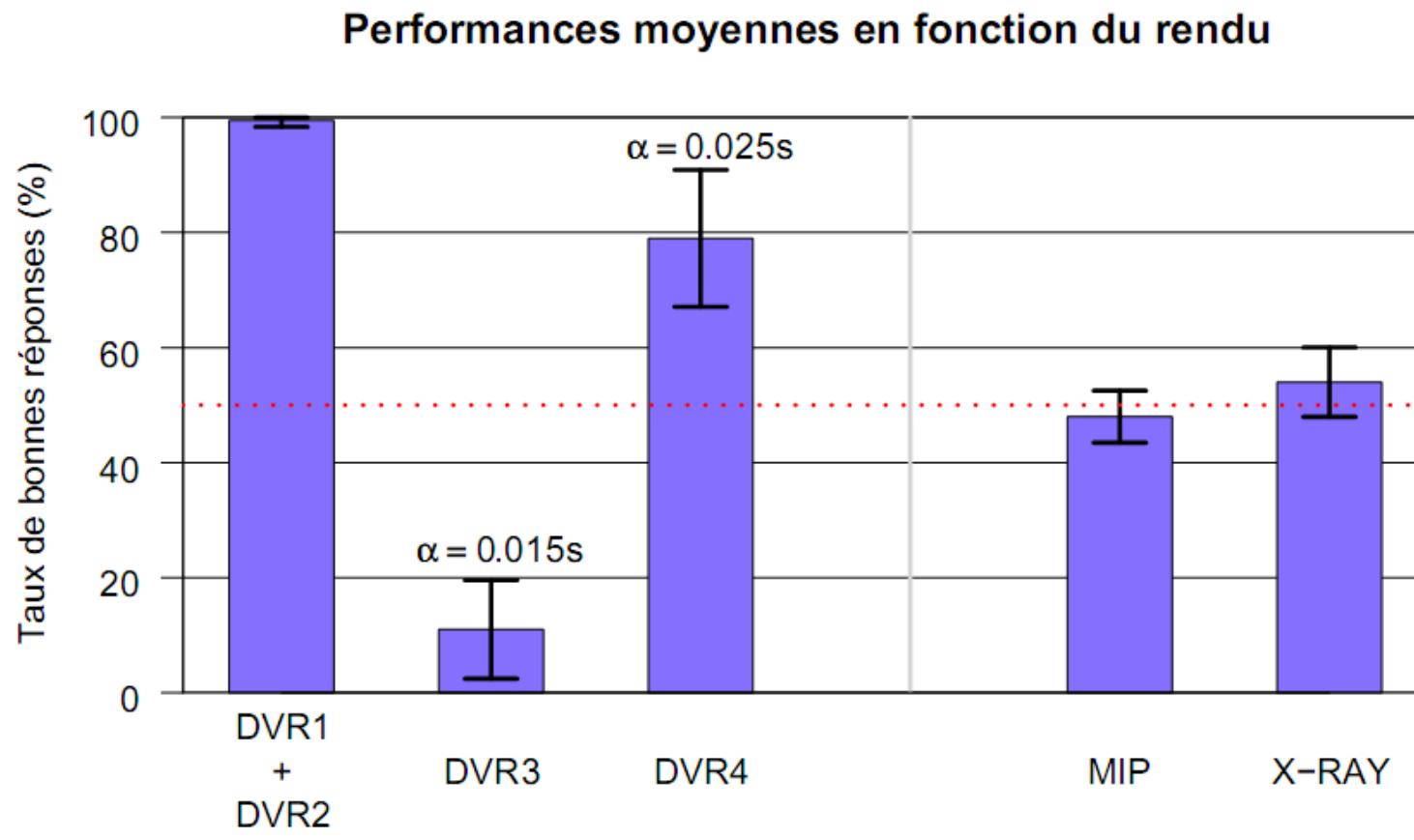


Boucheny, Bonneau & al ACM TAP 2009

# Choice of transfer functions

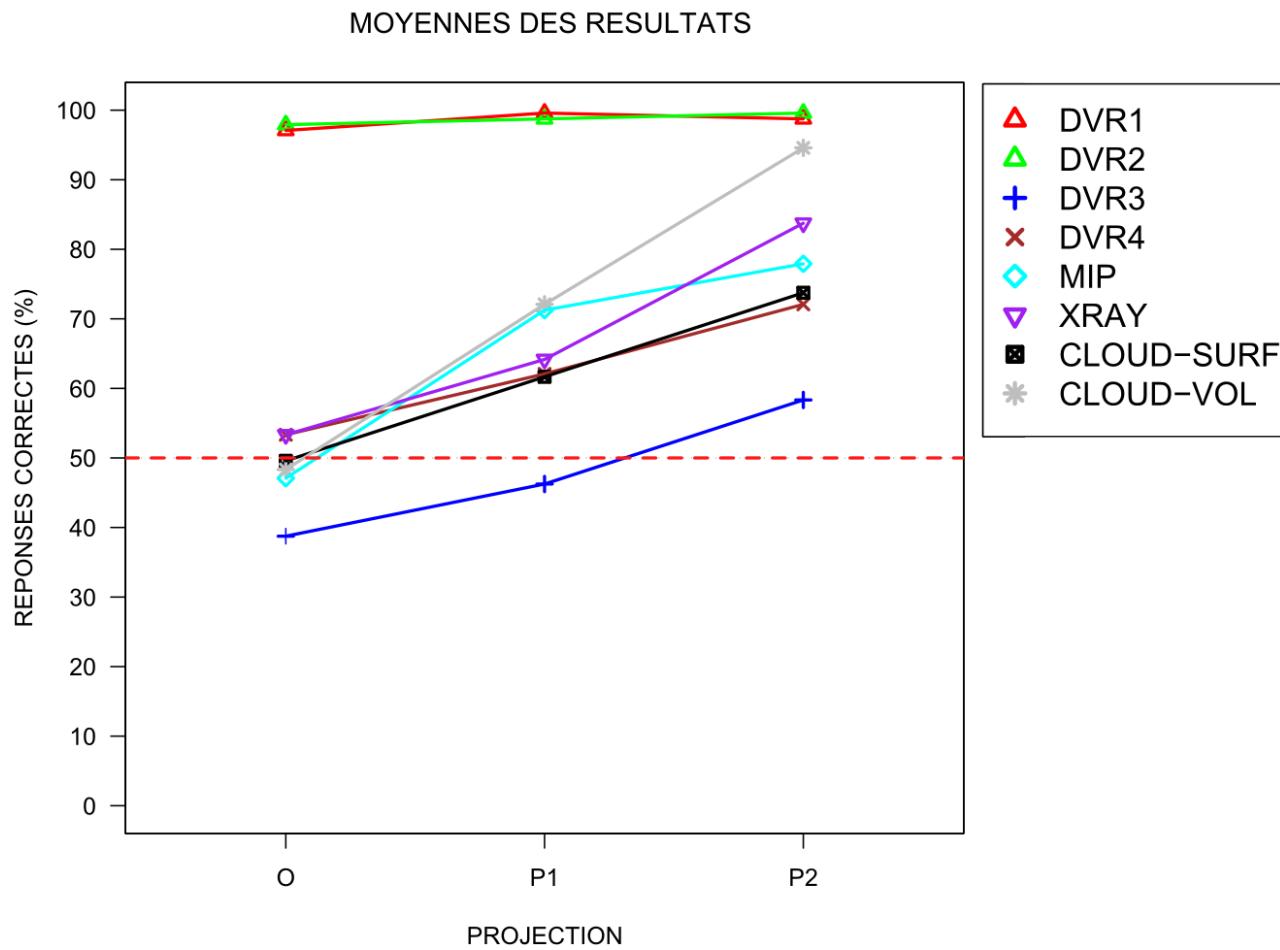


# Dynamic experiment results



Boucheny, Bonneau & al ACM TAP 2009

# Dynamic experiment results

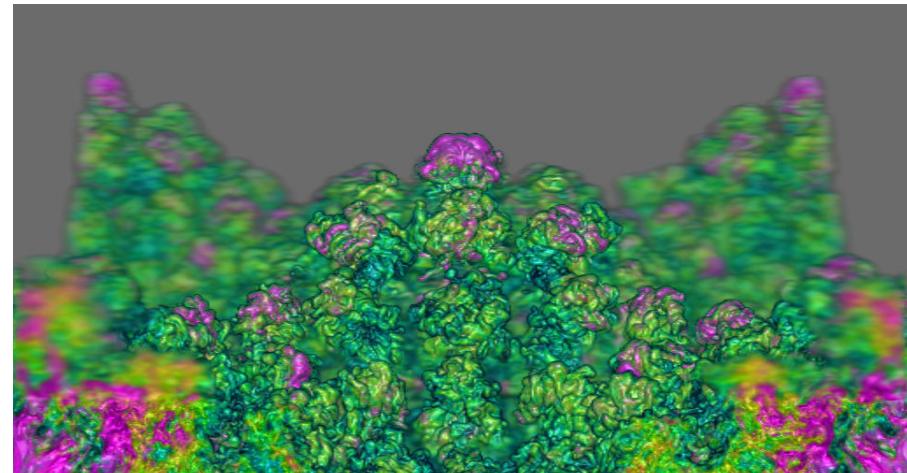
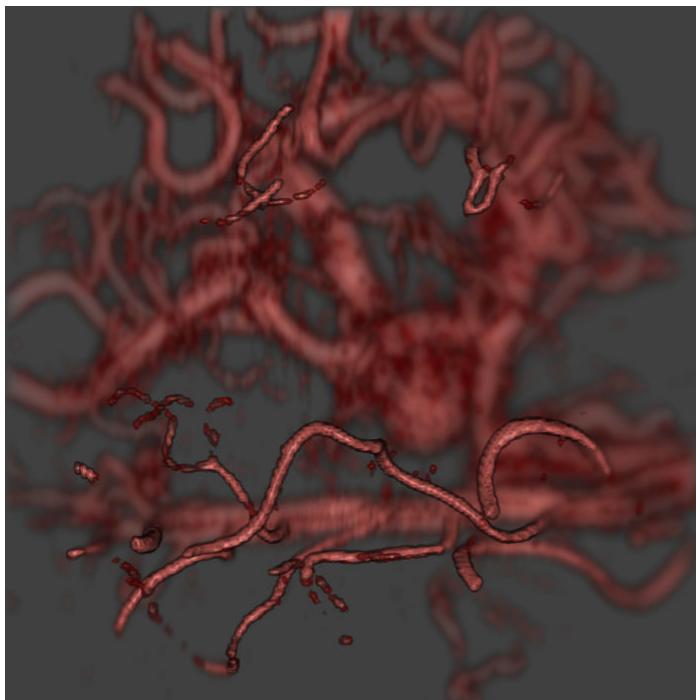


Boucheny, Bonneau & al ACM TAP 2009

# Depth is difficult to perceive in DVR

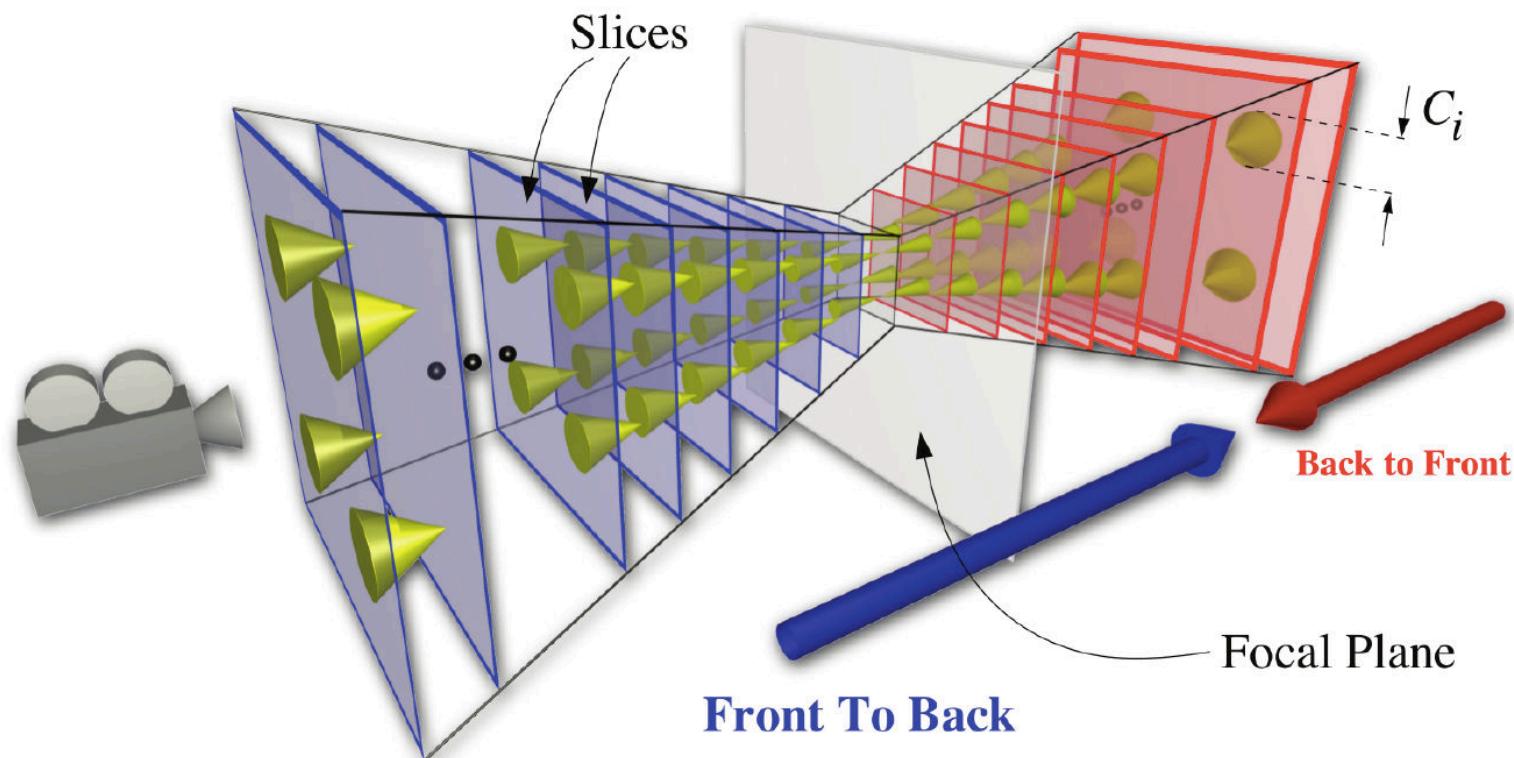
## Can we improve this?

# Depth Blur in Direct Volume Rendering



Grosset, Schott, Bonneau, Hansen, Pacific Visualization 2013

# Depth Blur in Direct Volume Rendering



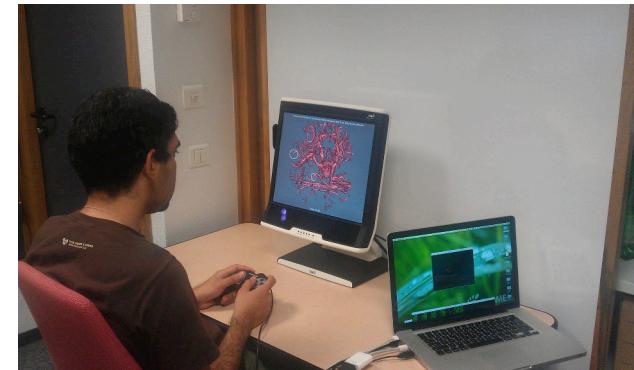
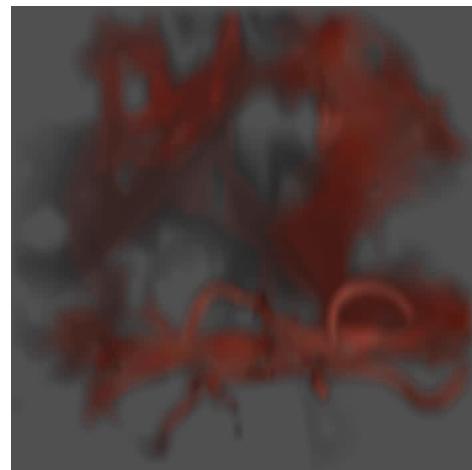
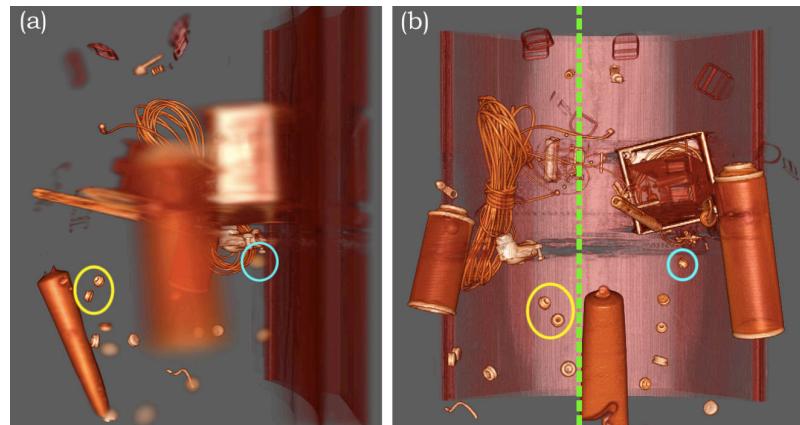
Schott, Grosset, martin, Pegoraro, Smith, Hansen, CGF 2009

# Hypotheses



- **HYP 1:** DoF will improve the accuracy of ordinal depth perception in a volume rendered image where there are multiple features.
- **HYP 2:** DoF will improve the speed of ordinal depth perception in a volume rendered image where there are multiple features.
- **HYP 3:** If users view a moving focal plane, correct perception of ordinal depth will improve.

# Depth Blur Experiment in DVR

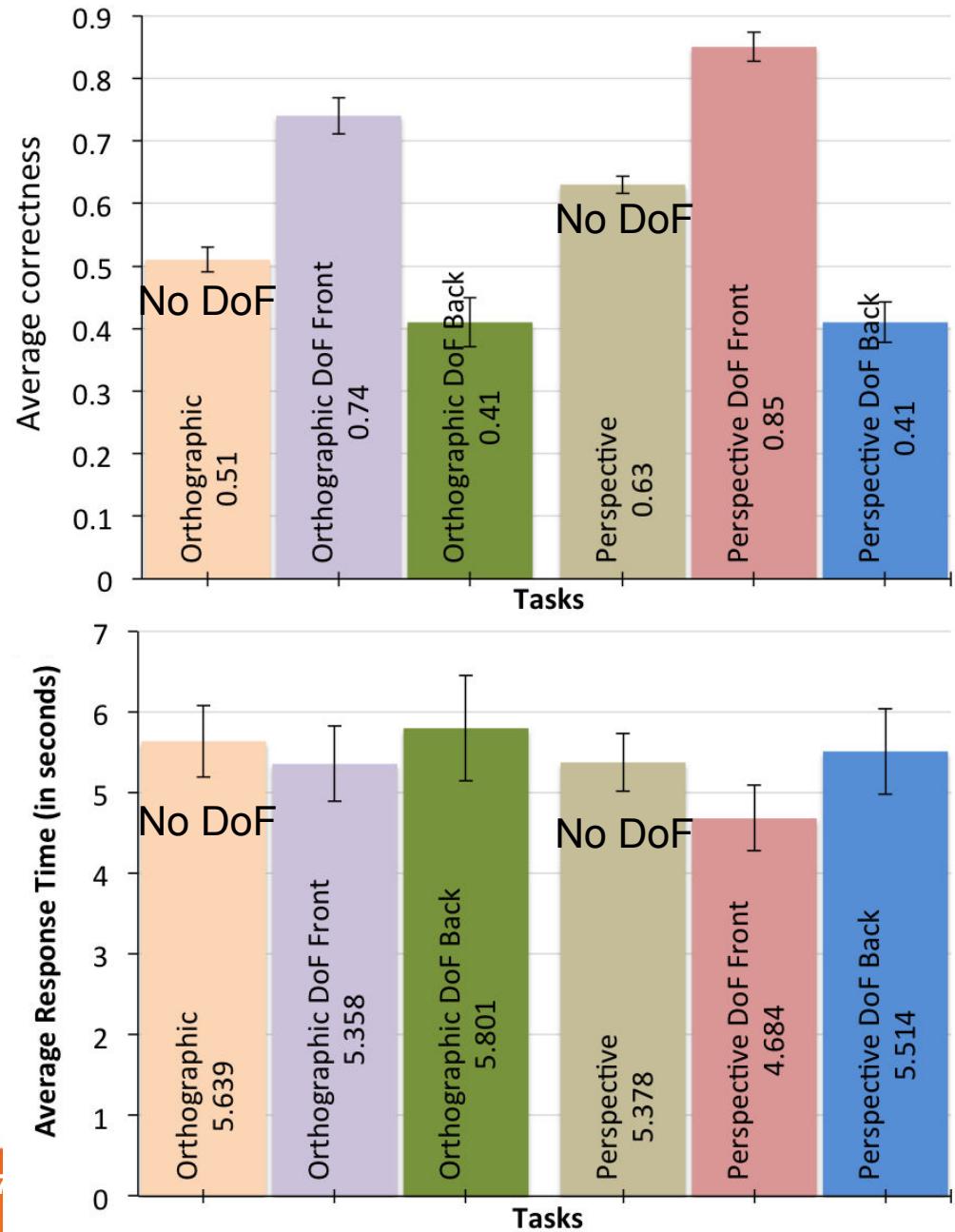


Grosset, Schott, Bonneau, Hansen, Pacific Visualization 2013

GDR MASCOT-NUM, 22nd May 2017, Visual perception in Visualization, G.-P. Bonneau

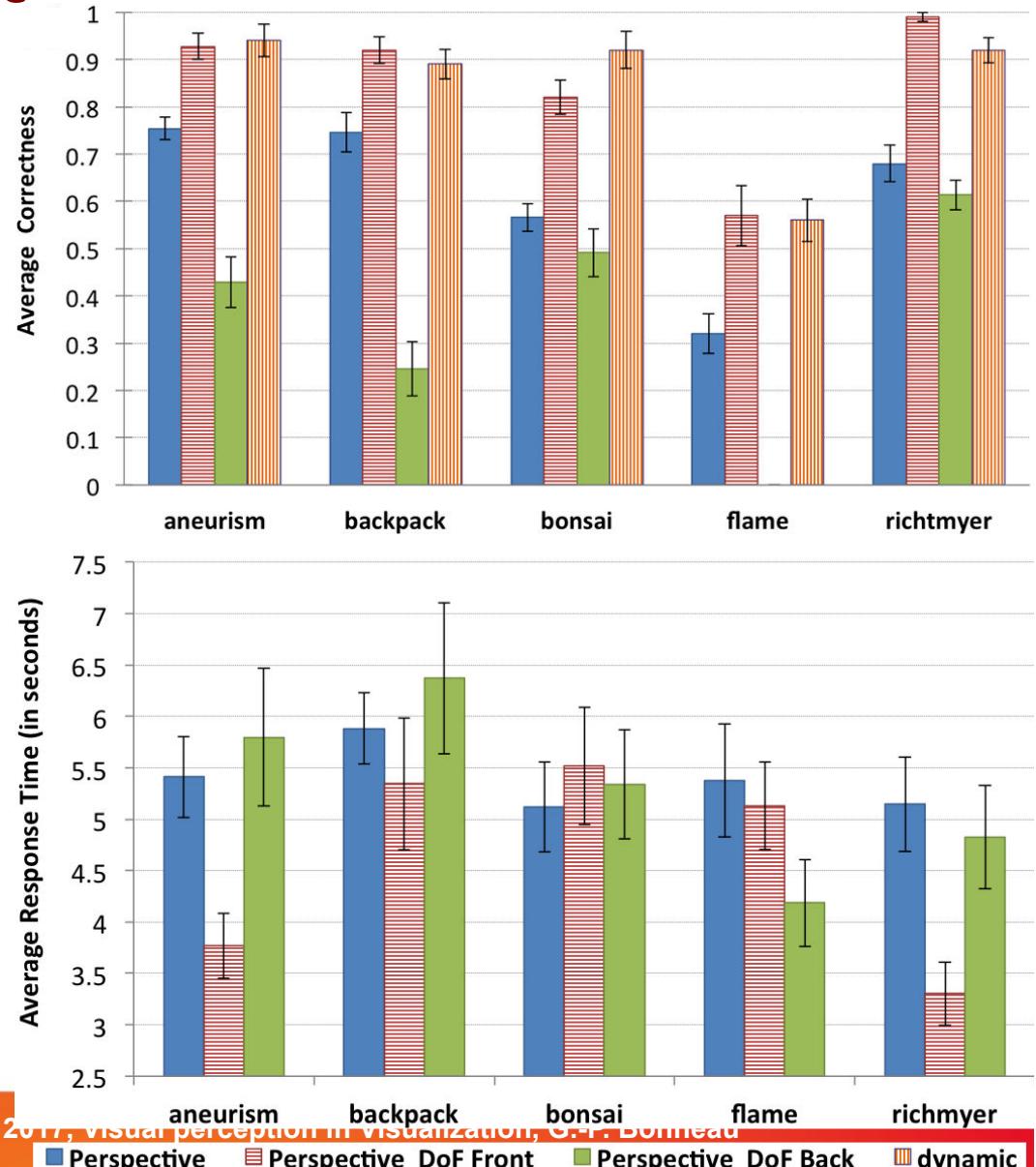
# Results: Overall

- HYP1 & HYP2
- Depth of Field is NOT always beneficial
  - DoF on Front feature helps
  - DoF on Back feature has reverse effects
  - Why?
    - Depth Cue conflict?



# Results: Dataset

- DoF on Front feature is helpful
- Not equally beneficial
  - Translucent datasets are hard to understand



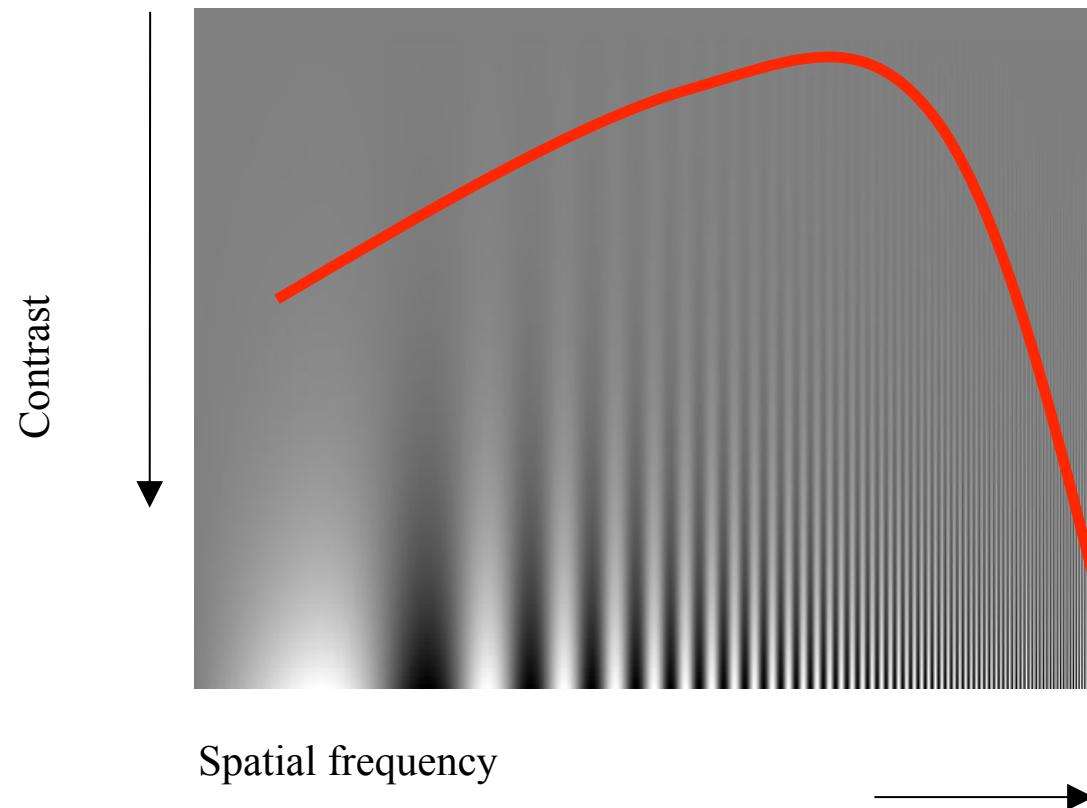


# Contrast sensitivity perception

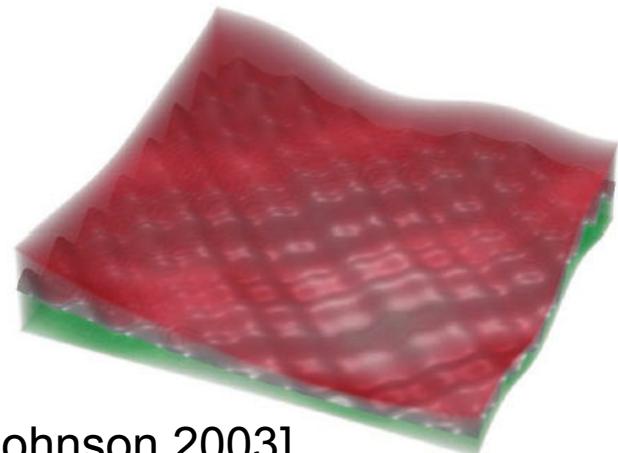
3

Visualization of Data with Uncertainty

# Contrast sensitivity function

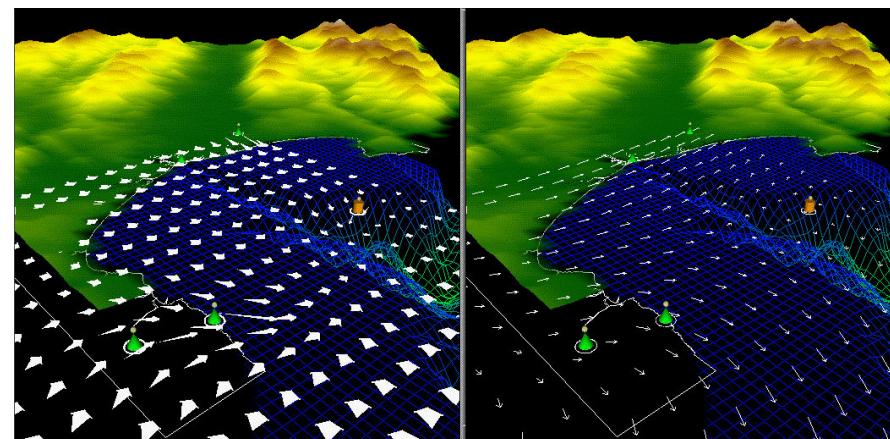
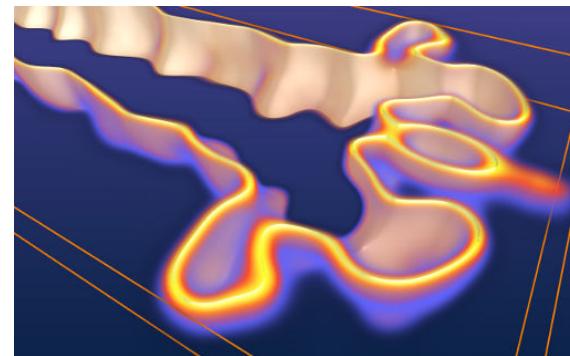


# Visualization of Data with Uncertainty



[Johnson 2003]

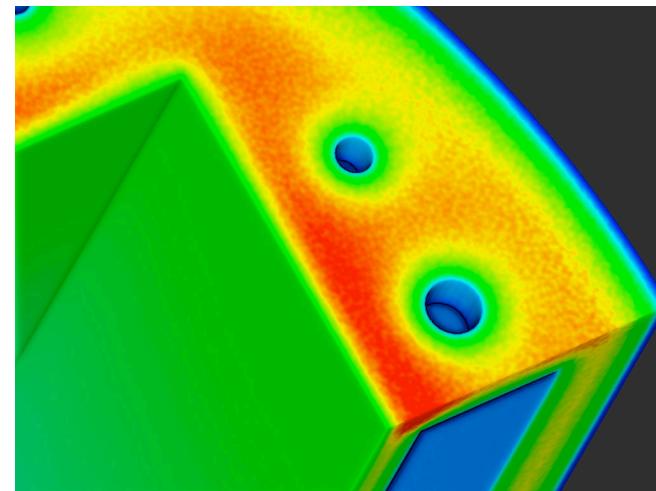
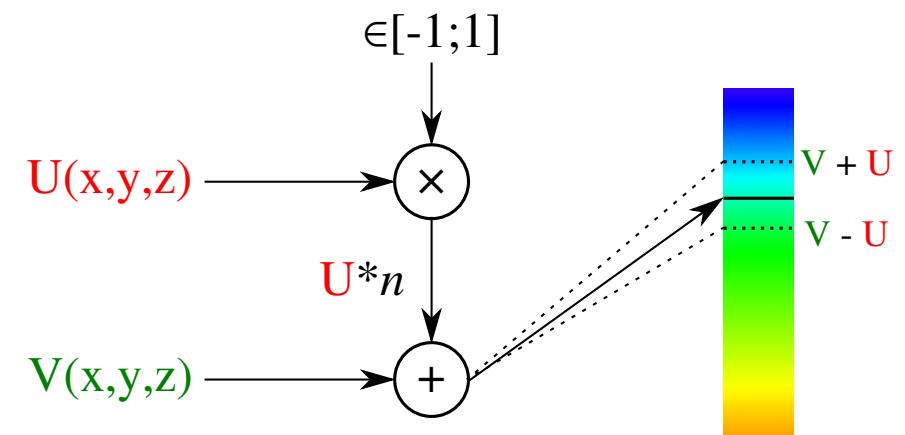
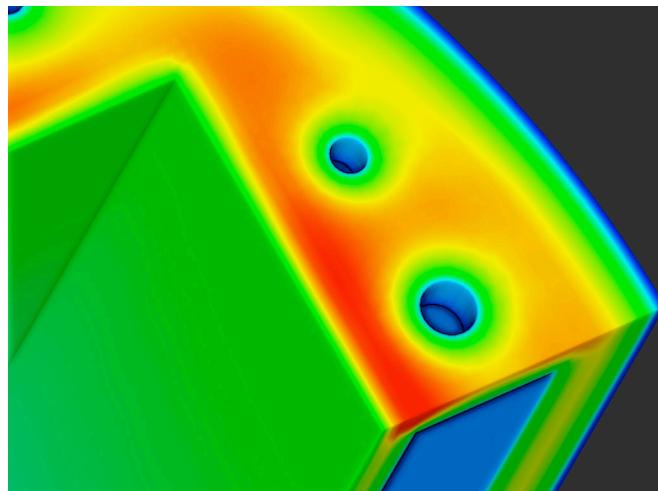
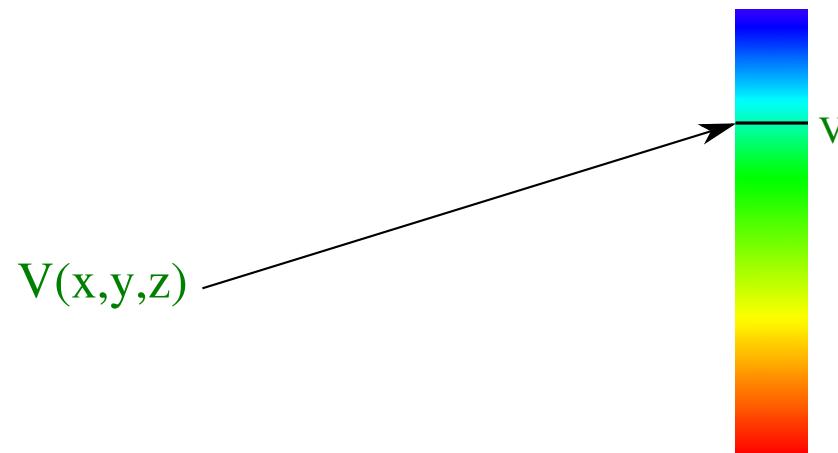
ZIB 2011



Wittenbrink & al. 1995

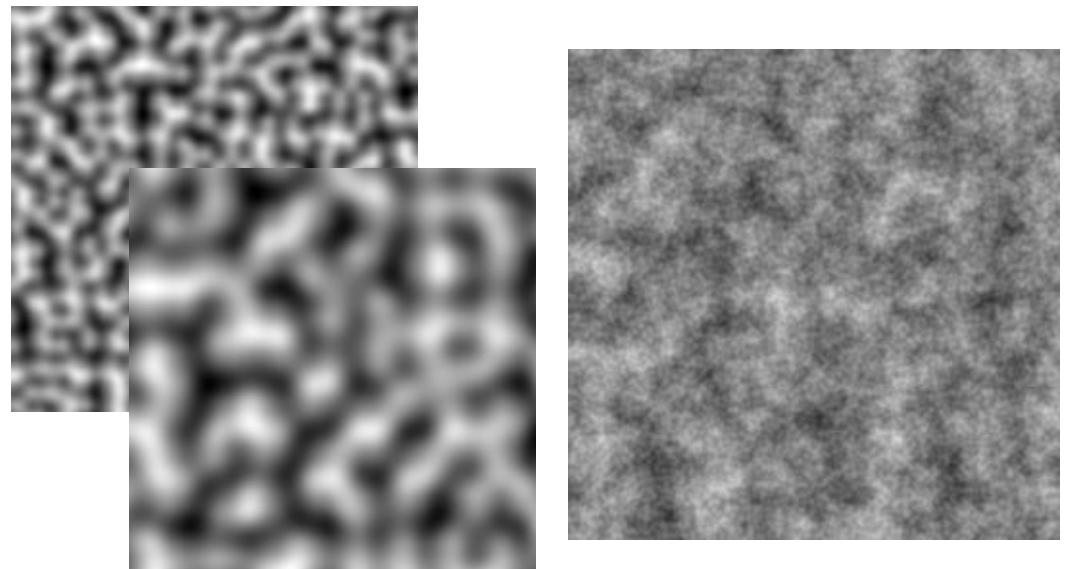
# Uncertainty Visualization using Noise

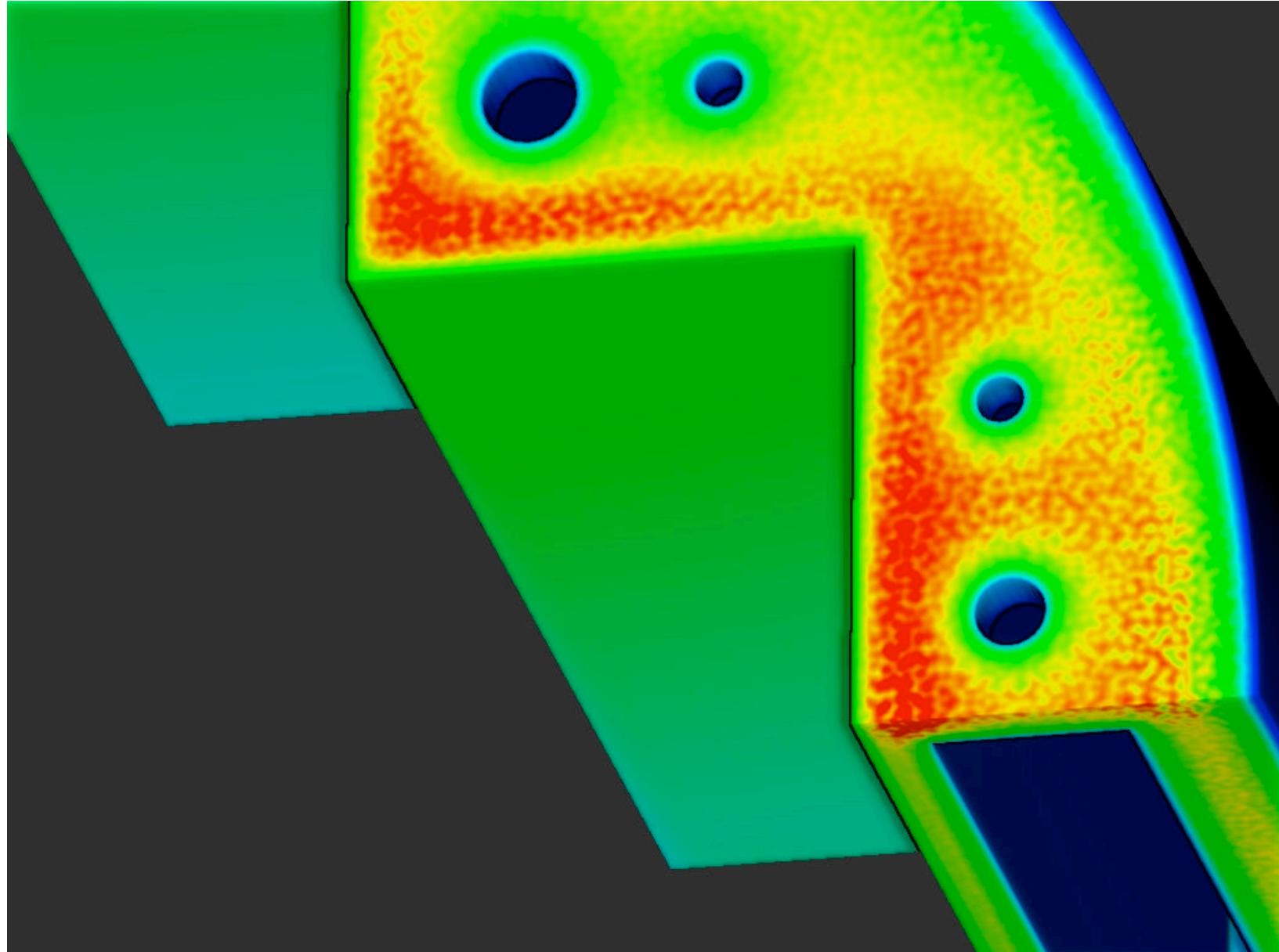
Coninx, Bonneau & al., APGV 2011



# Perlin Noise

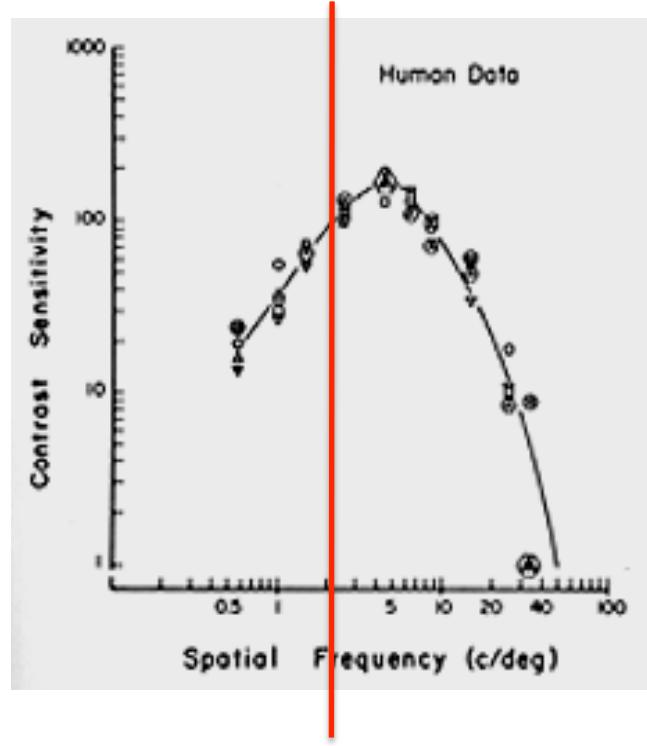
$$n_{f_0,p}(\vec{x}, t) = \sum_i p^i N(f_0 2^i \vec{x}, t)$$



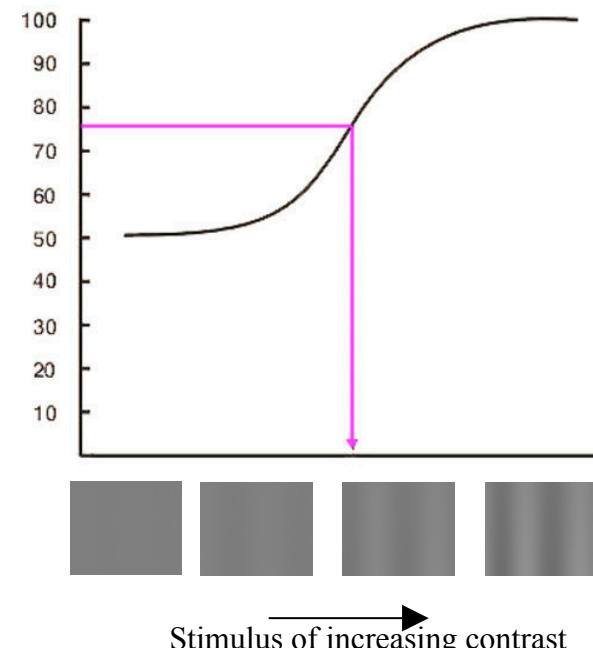


Coninx, Bonneau & al., APGV 2011

GDR MASCOT-NUM, 22nd May 2017, Visual perception in Visualization, G.-P. Bonneau



DeValois & DeValois (1988)  
*Spatial Vision*. Oxford: Oxford University Press



# Stimulus of high contrast



Exemple of stimulus with high contrast.

# Stimulus near threshold contrast



Exemple of stimulus with near-threshold contrast.

# Psychophysik experiment

- Choice of 4 values of frequency  $f_0$  and persistence  $p$   
=> 16 stimuli
- $f_0 = \{ 2, 4, 8, 16 \}$   $p = \{ 0., 0.25, 0.5, 0.75 \}$
- Stimulus of size 256x256
- LCD with pixel size 0.295mm, screen-subject distance 2m  
=> 120 pixels per degree
- 30 questions to measure a threshold =>  $30 * 16 = 480$  quest., ~5s for each quest. => ~40mn duration for each subject

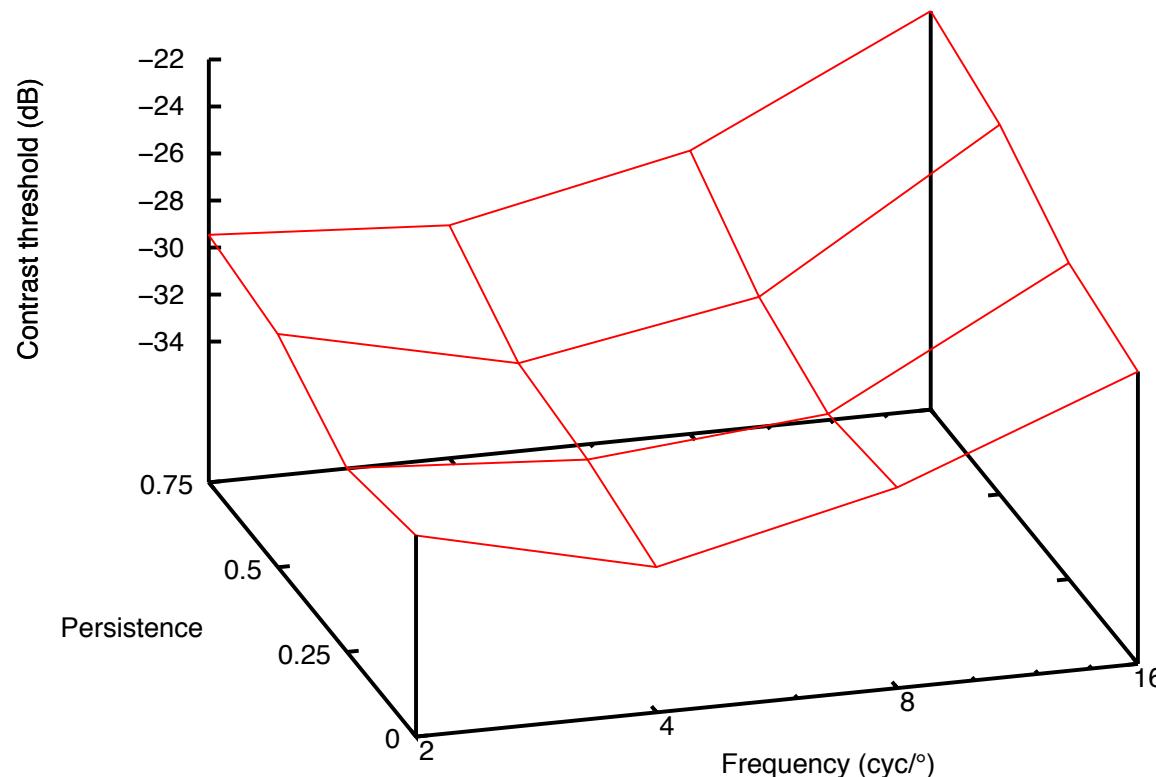
# Psychophysik experiment pitfalls

- Eliminate all biases in the stimuli
- Calibrate screen precisely
- resolution in gray value must be < 0.25% of measured threshold

...

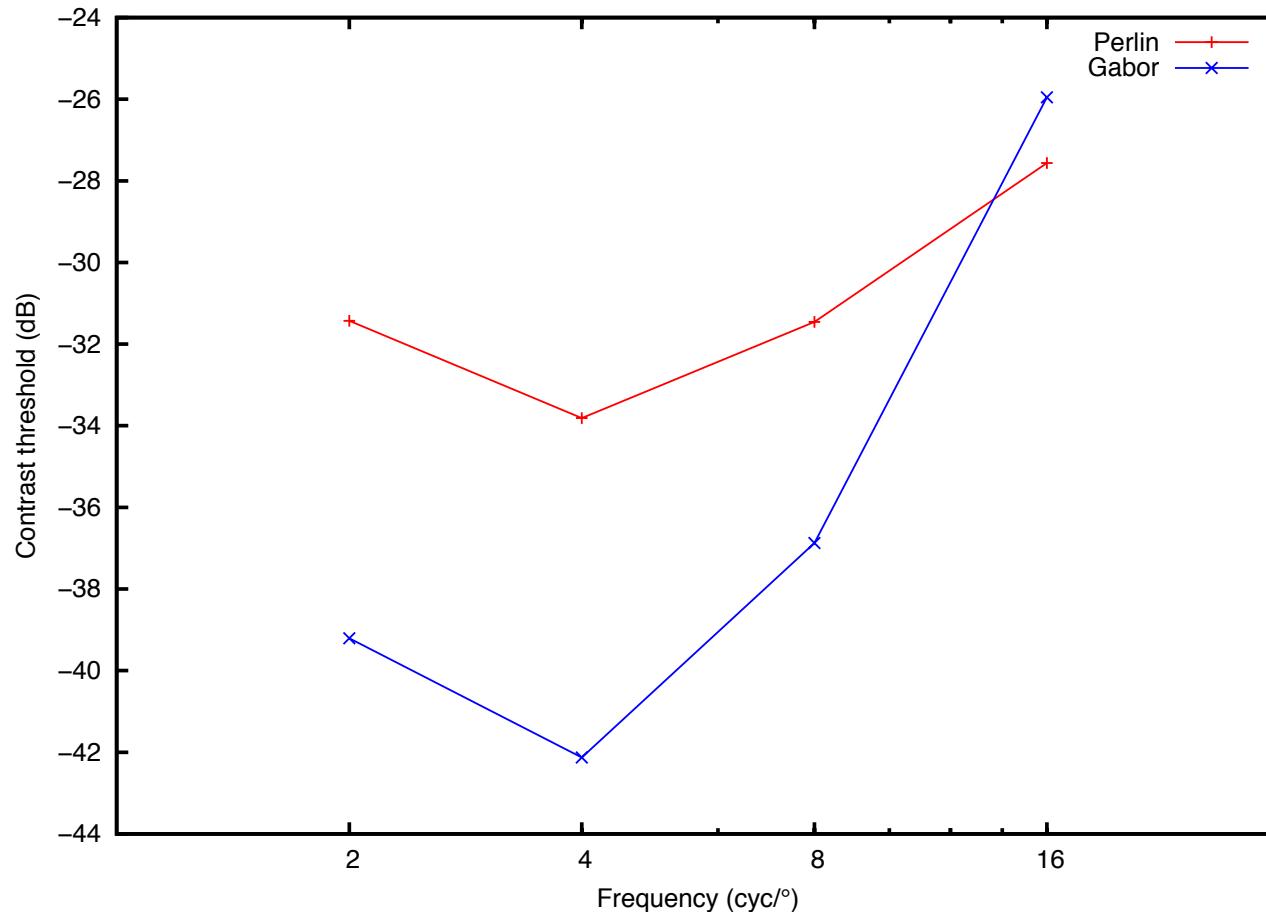
=> necessary to collaborate with perception experts!

# Sensitivity Thresholds for Perlin Noise

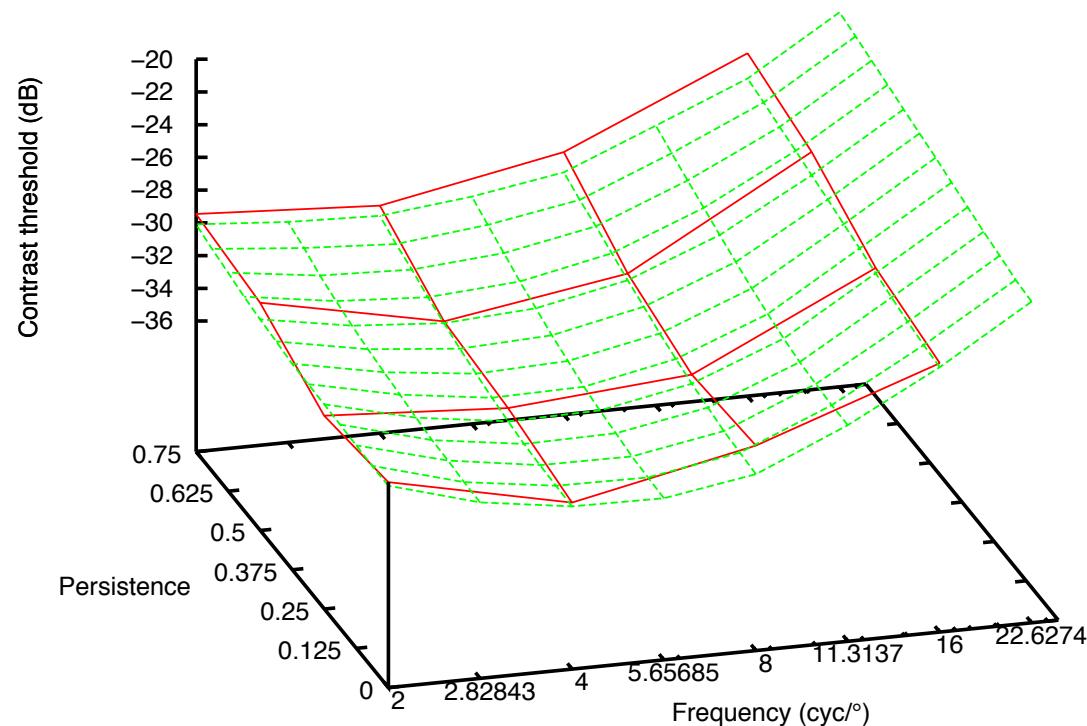


Coninx, Bonneau & al., APGV 2011

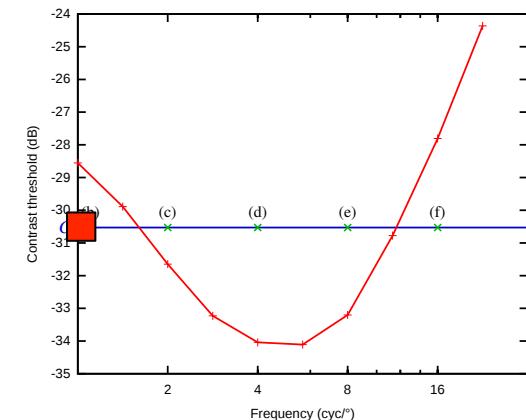
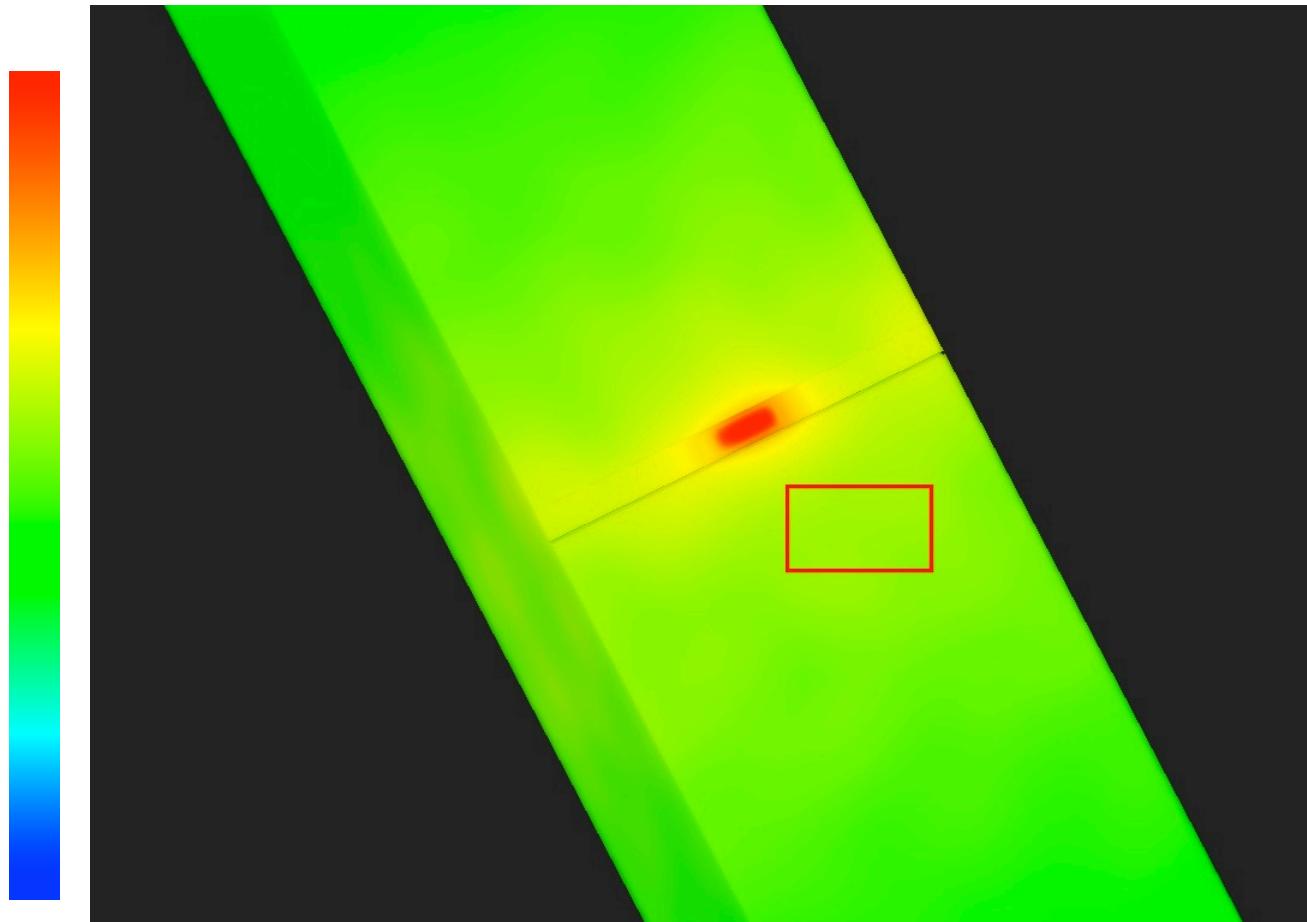
# Threshold for Regular vs. Stochastic Stimuli



# Model for Automatic Threshold Computation

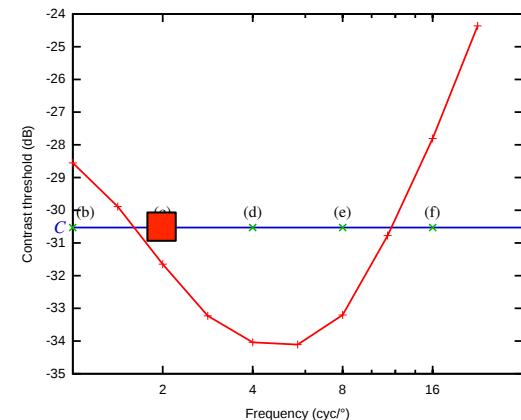
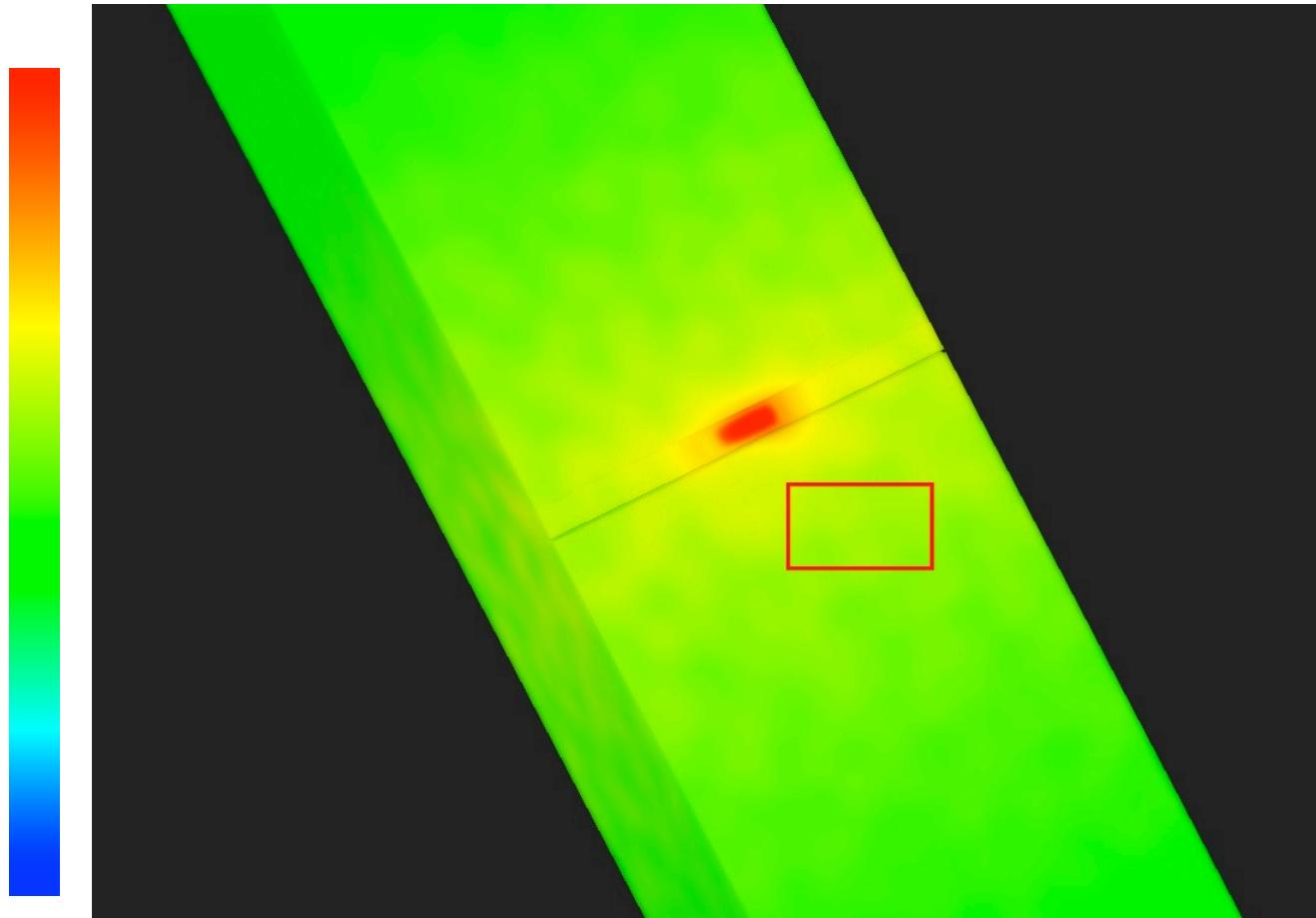


# Using Sensitivity Thresholds



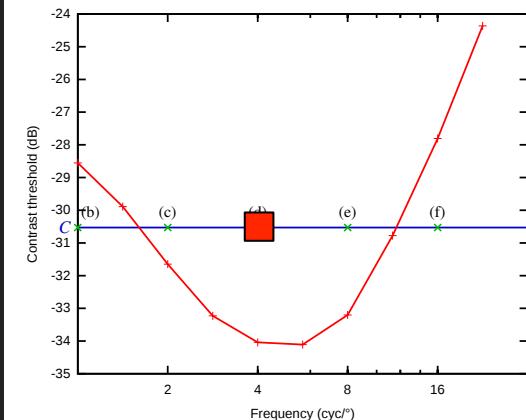
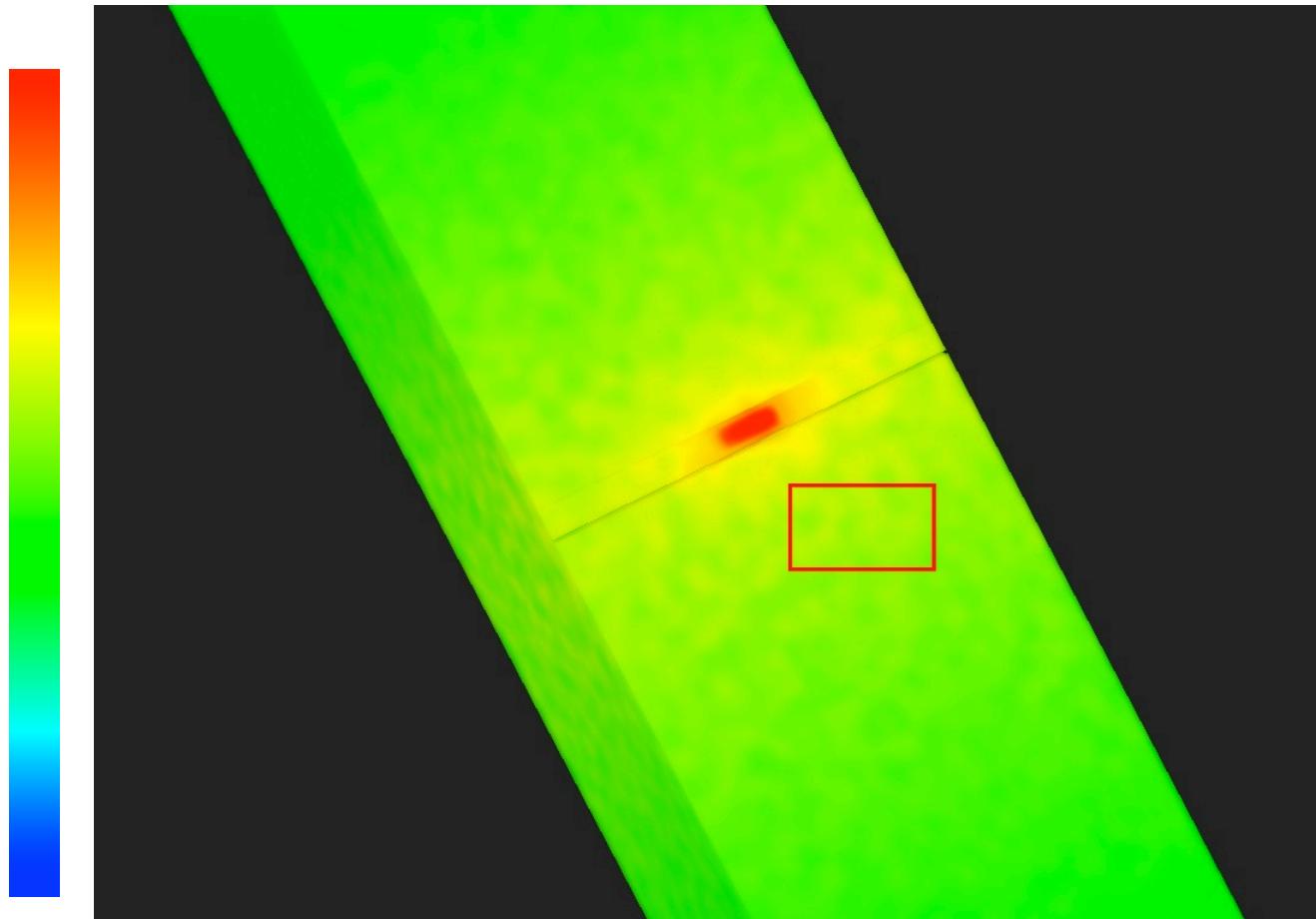
Coninx, Bonneau & al., APGV 2011

# Using Sensitivity Thresholds



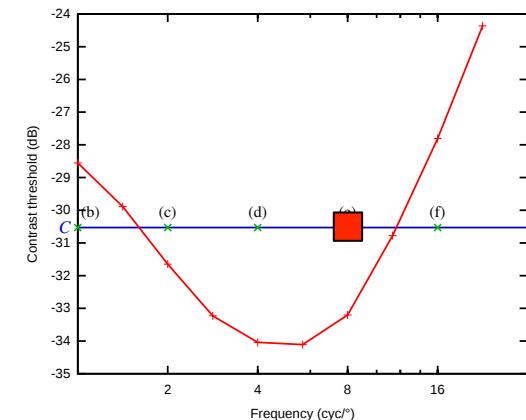
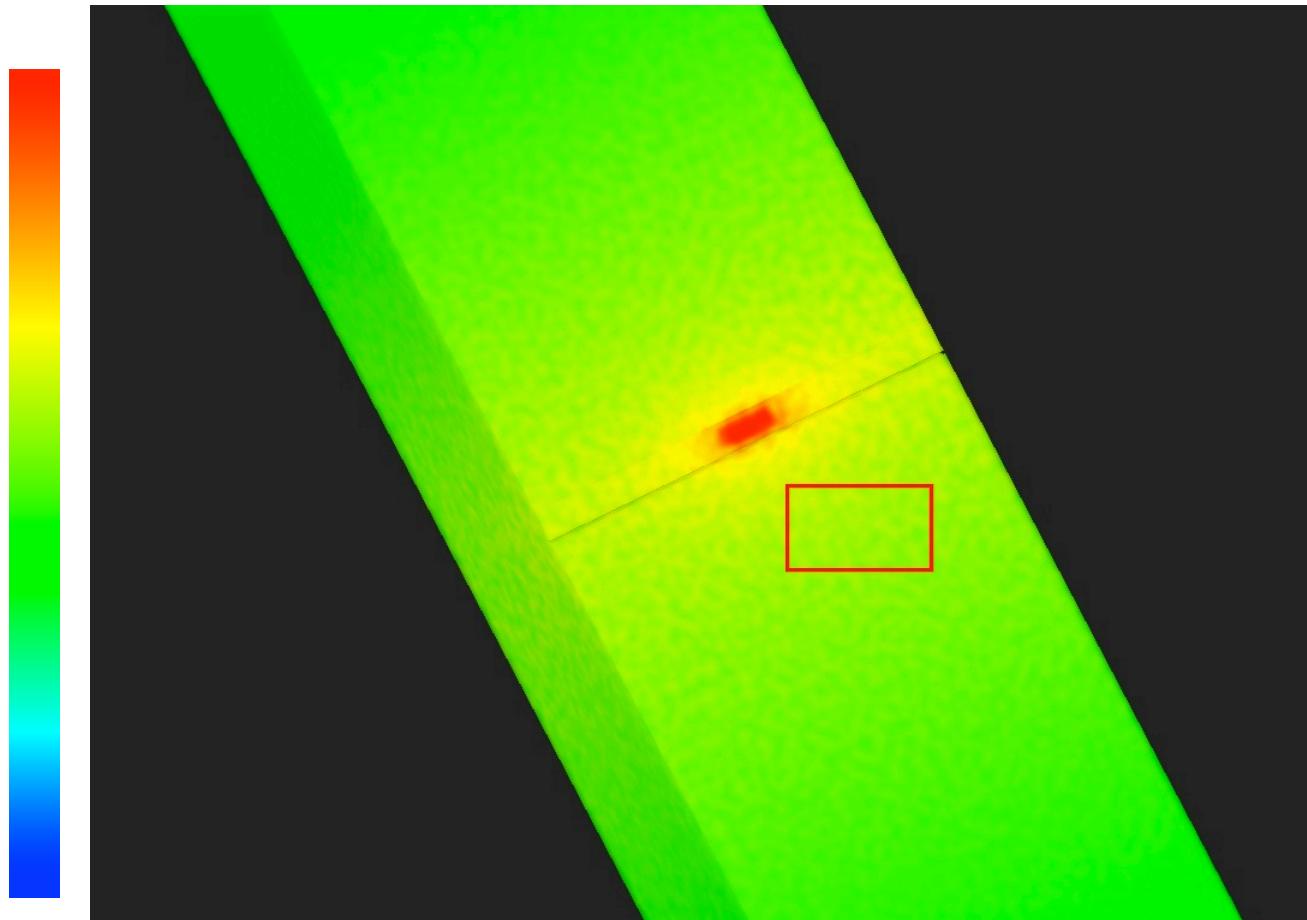
Coninx, Bonneau & al., APGV 2011

# Using Sensitivity Thresholds



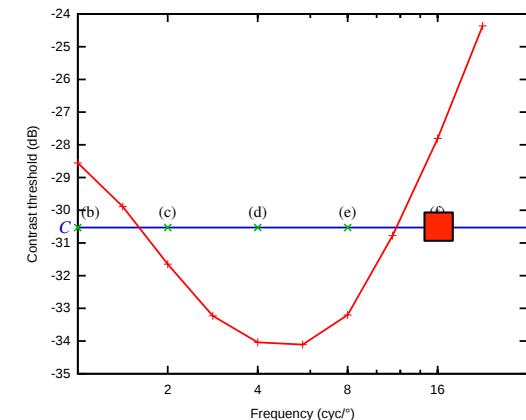
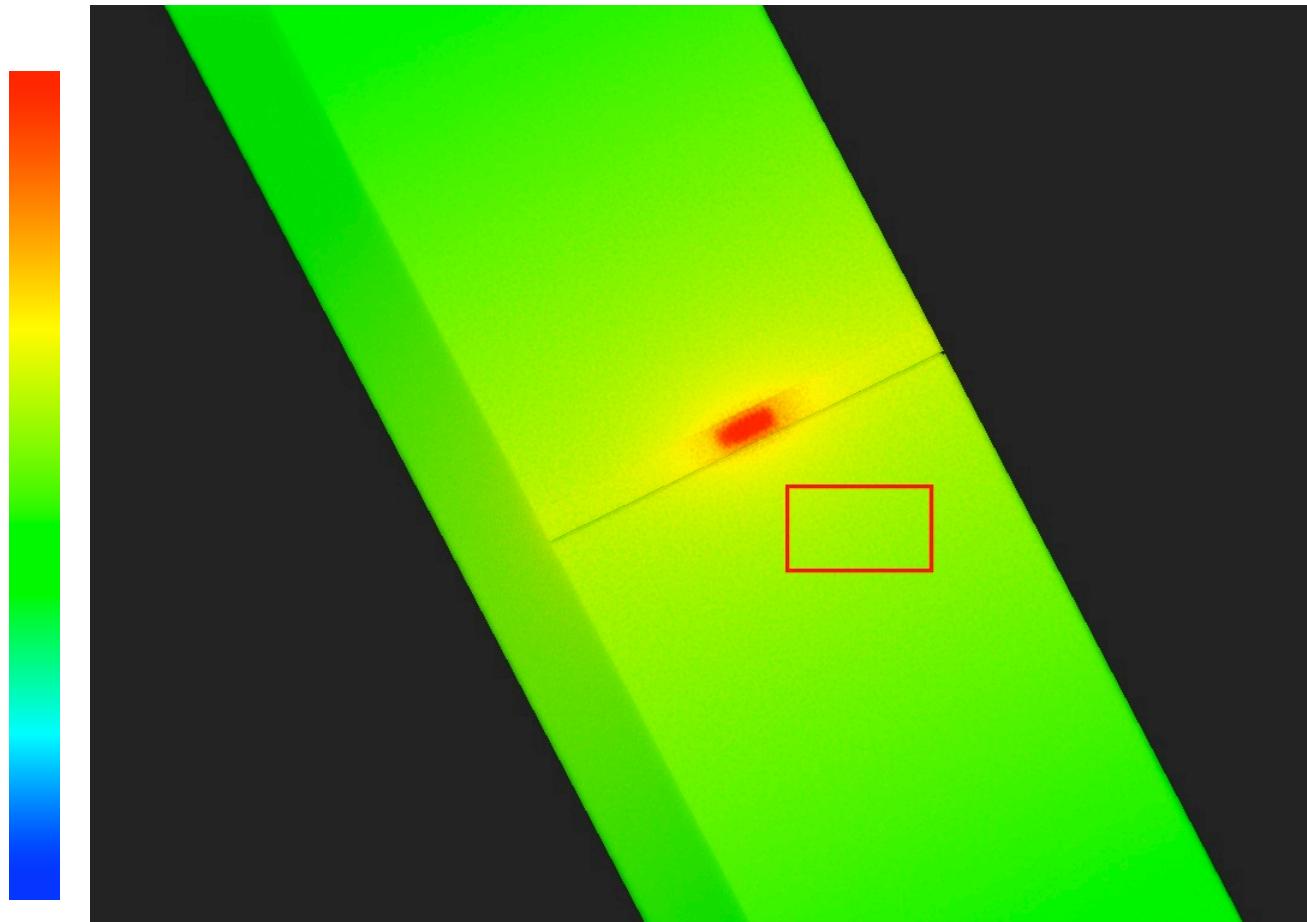
Coninx, Bonneau & al., APGV 2011

# Using Sensitivity Thresholds



Coninx, Bonneau & al., APGV 2011

# Using Sensitivity Thresholds



Coninx, Bonneau & al., APGV 2011

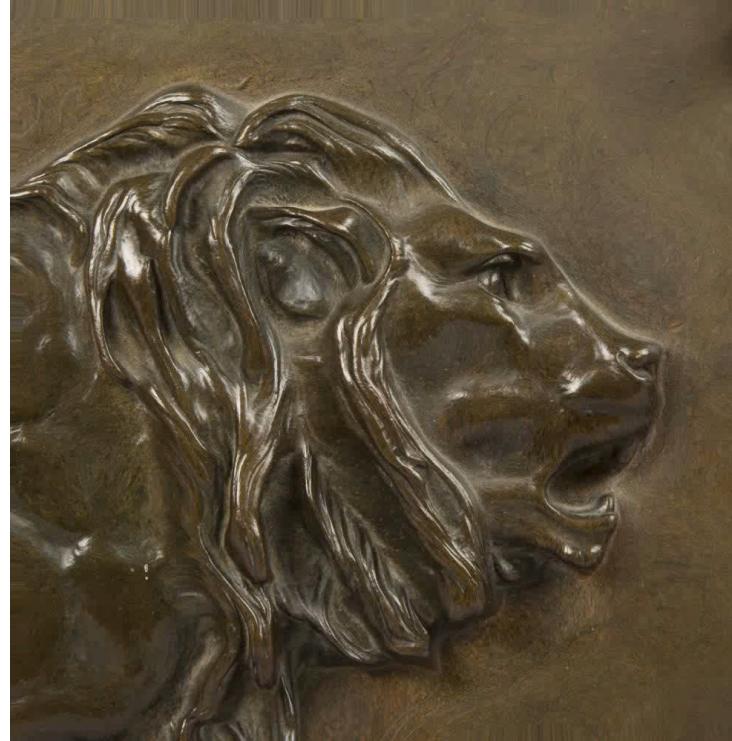


# Perception of shape

4

Image based shape manipulation

# Modifying Perceived Shape by Warping

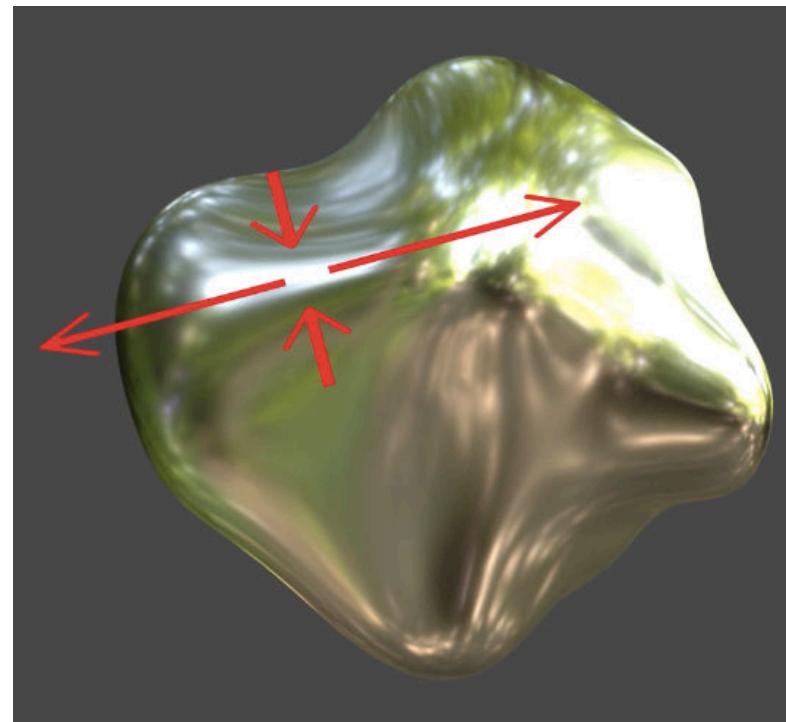


R. Vergne, P. Barla, G.-P. Bonneau, R.W. Fleming  
*Flow-Guided Warping for Image-Based Shape Manipulation*  
Siggraph 2016

# Related Works

- Image manipulation
  - [Boyadzhiev et al. 2015] [Wu et al. 2012] [Wadhwa et al. 2013] [Wadhwa et al. 2015]
  - [Dekel et al. 2015] [Bousseau et al. 2009] [Yeung et al. 2011]
- 3D Reconstruction
  - [Khan et al. 2006] [Gutierrez et al. 2008] [Luft et al. 2006] [Kholgade et al. 2014] [Vergne et al. 2012]
- Shape perception
  - [Horn and Brooks 1989] [Oren and Nayar 1997] [Koenderink and van Doorn 1980] [Fleming et al. 2004] [Fleming et al. 2011] [Caniard and Fleming 2007] [Mooney and Anderson 2014].

# How do we perceive shape?

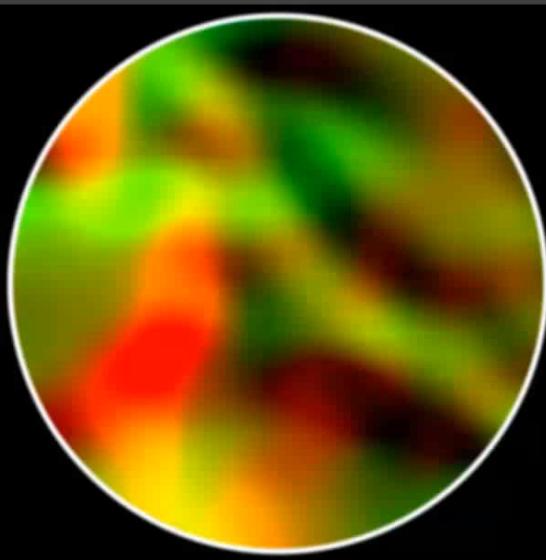


# Measuring Orientation Patterns

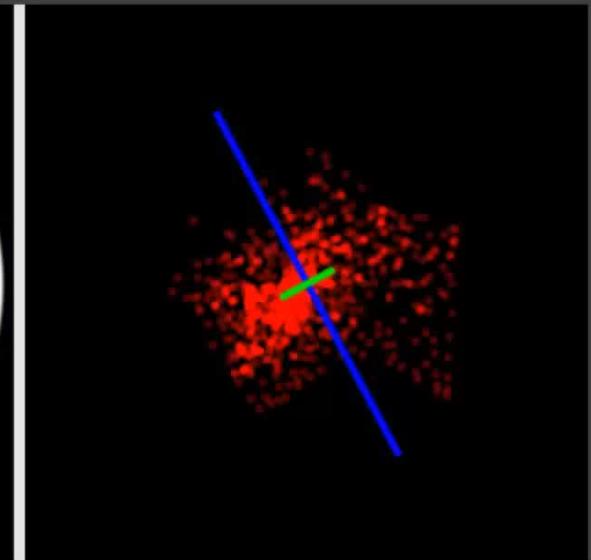
Orientation patterns are measured by the structure tensor



circle centered  
at pixel P

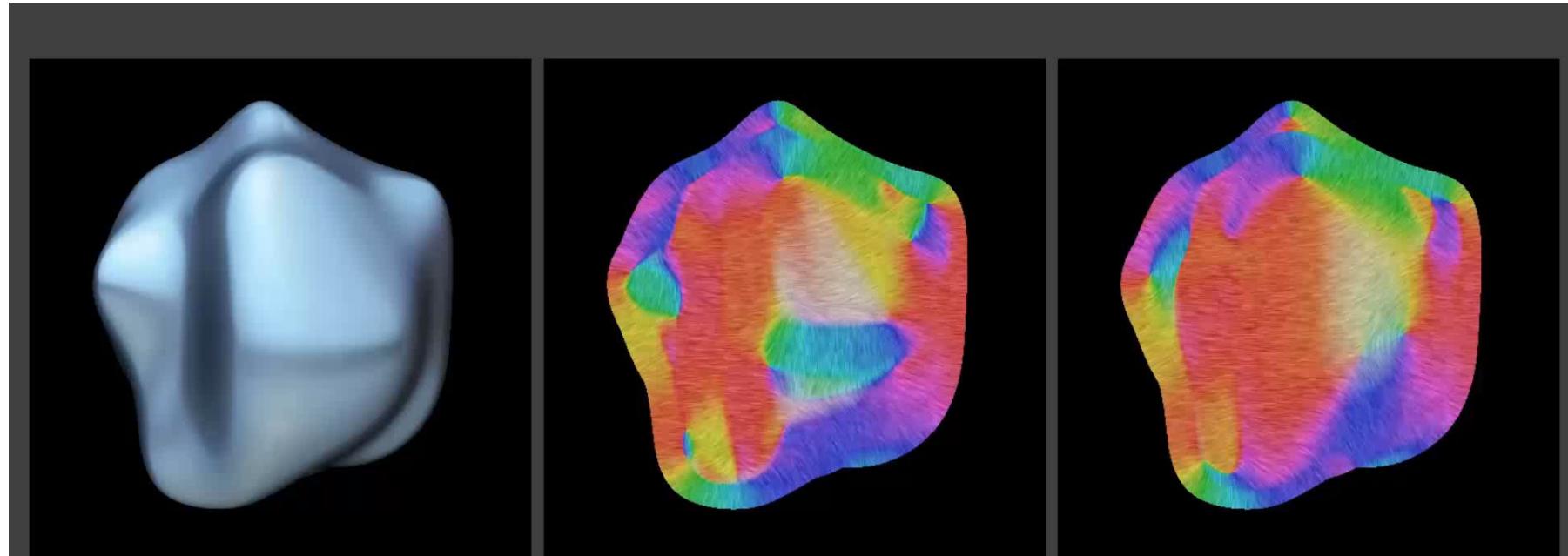


intensity  
gradients in  
circle



structure tensor  
at P

# Surface ST vs Image ST

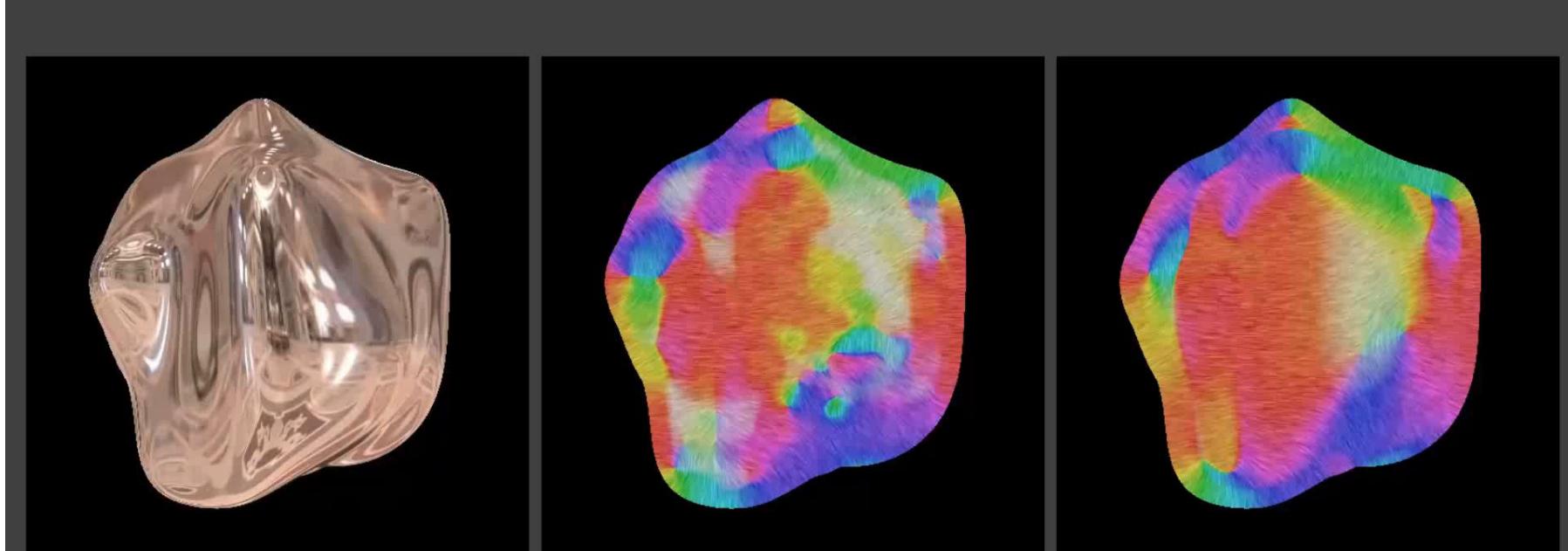


CG rendering

Image based ST

Surface based ST

# Shape ST vs Image ST

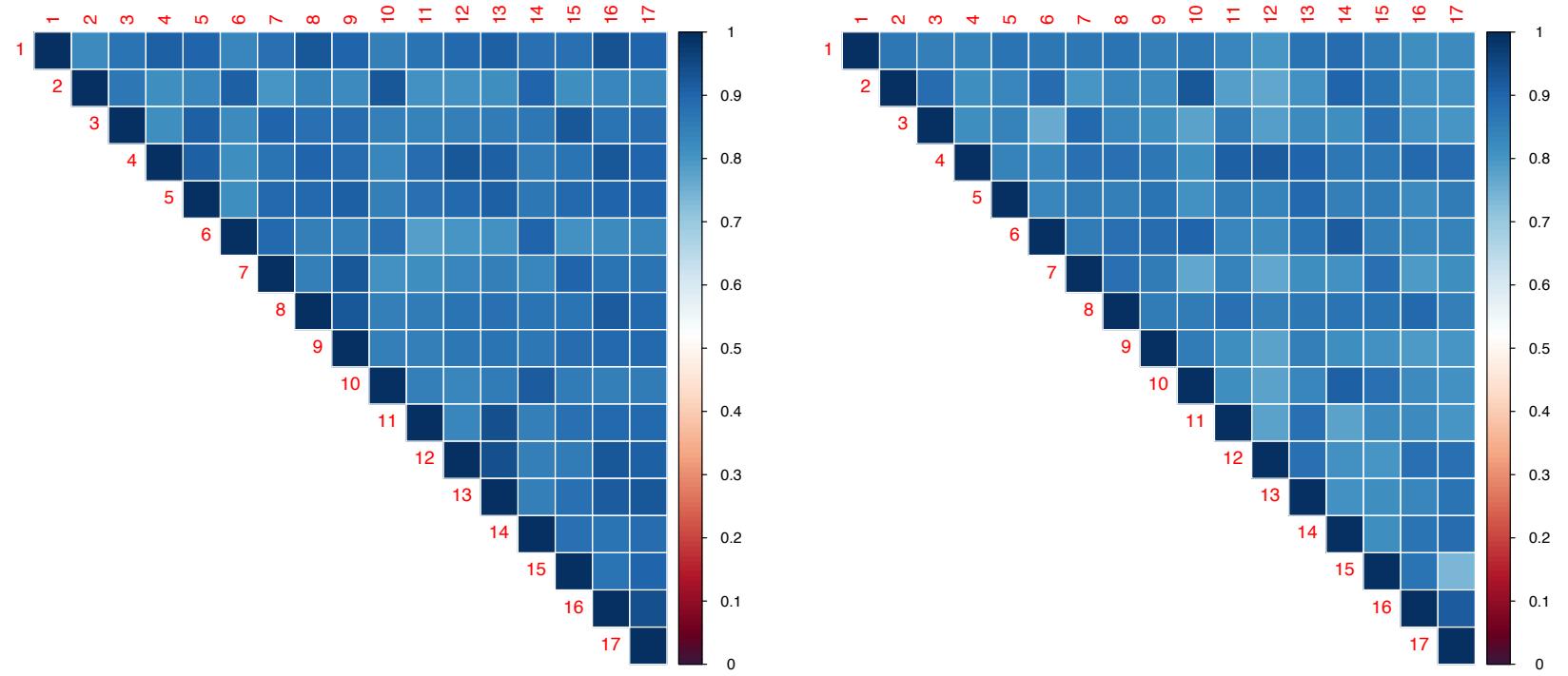


CG rendering

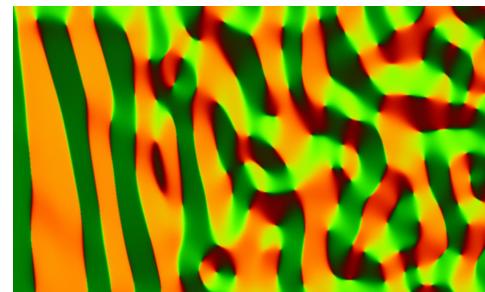
Image based ST

Surface based ST

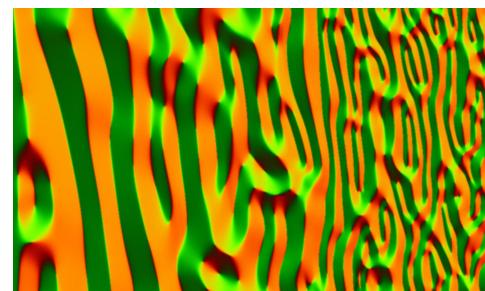
# Shape ST vs Image ST correlation



# Scale selection



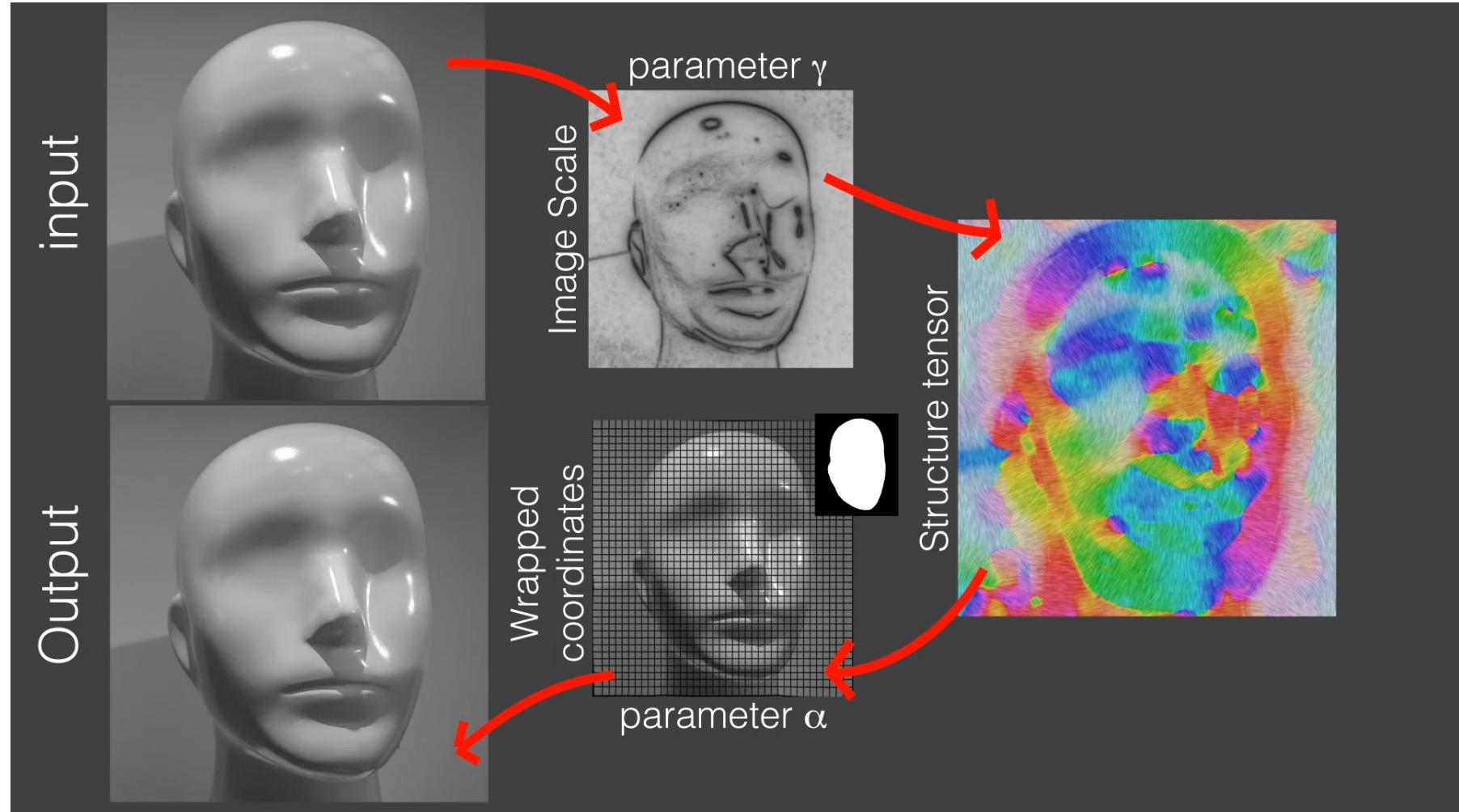
Uniform



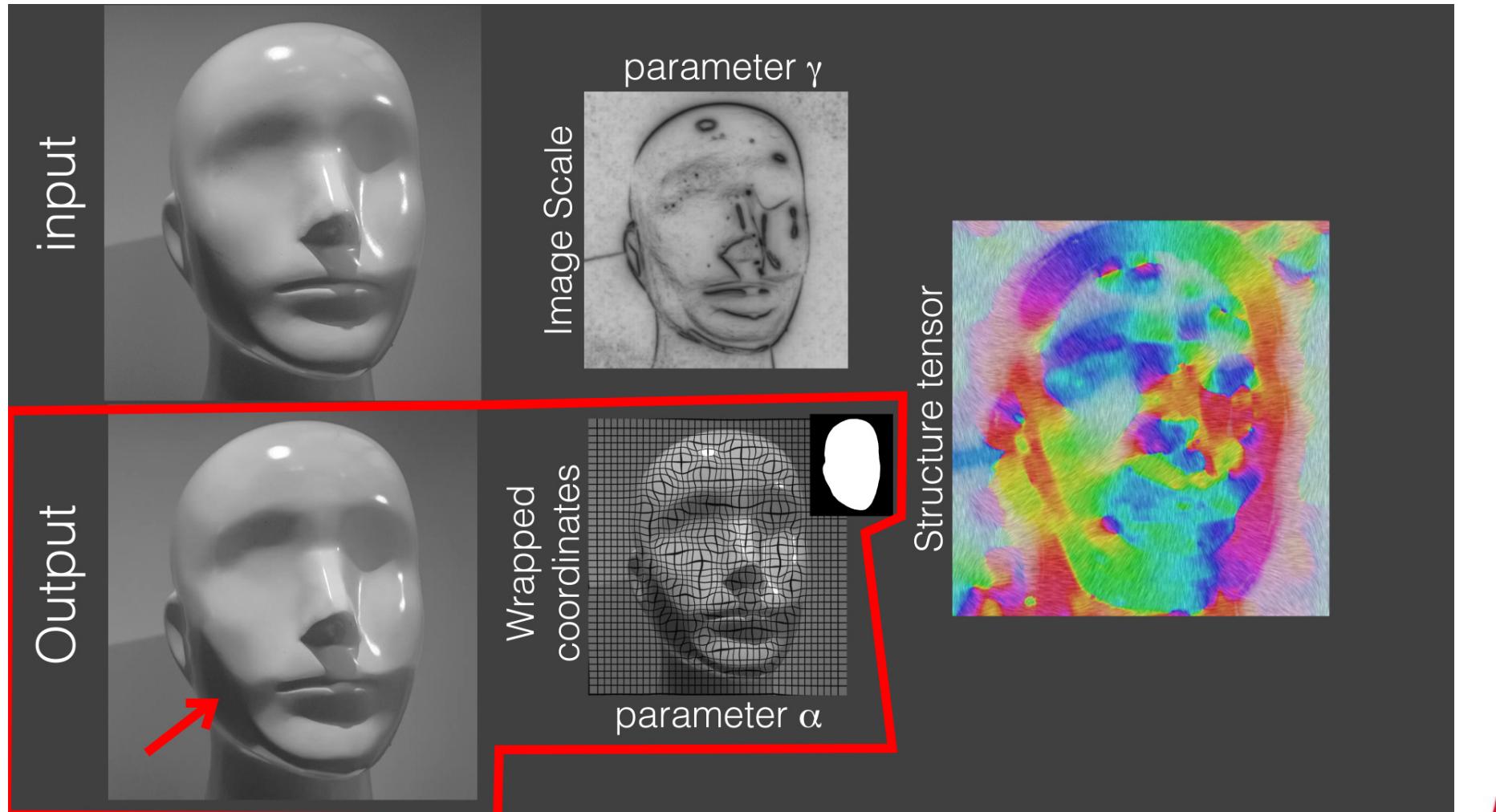
Local

$$\sigma_I(\mathbf{p}, \gamma) = \frac{\int \sigma^{2\gamma} \beta_I(\mathbf{p}, \sigma) d\sigma}{\int \beta_I(\mathbf{p}, \sigma) d\sigma}$$

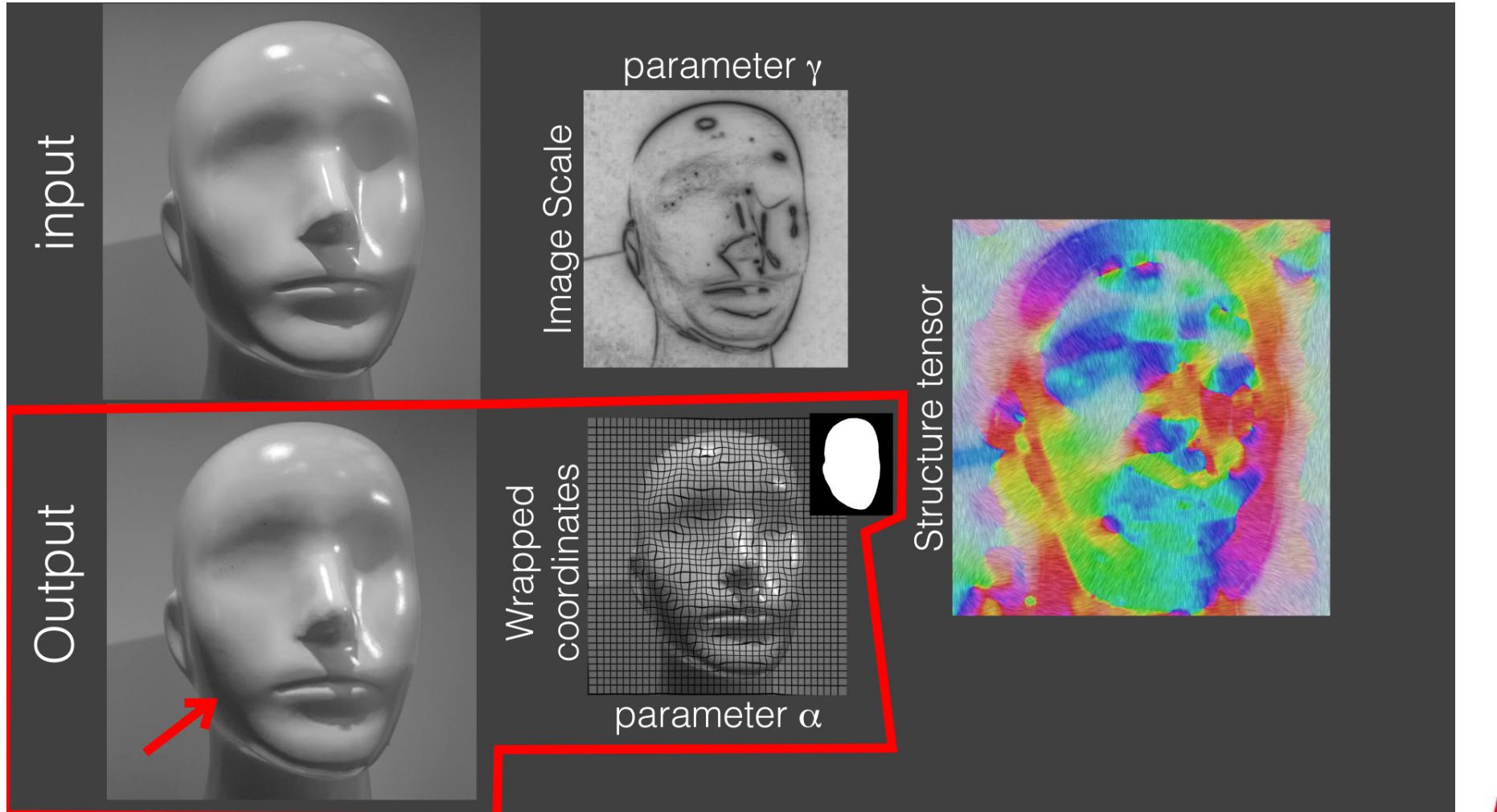
# Warping pipeline



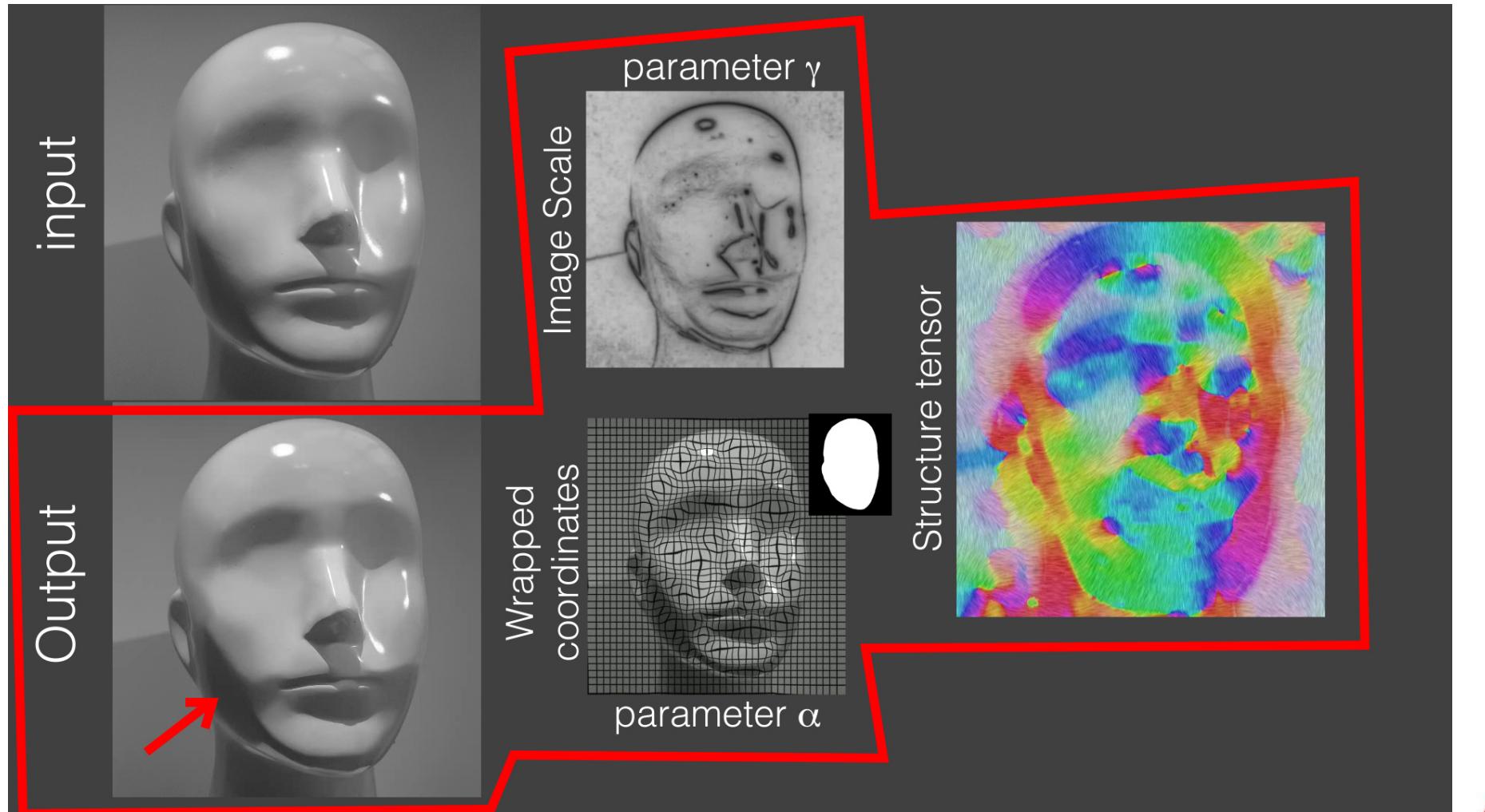
# Sharpening: $\alpha = -1$



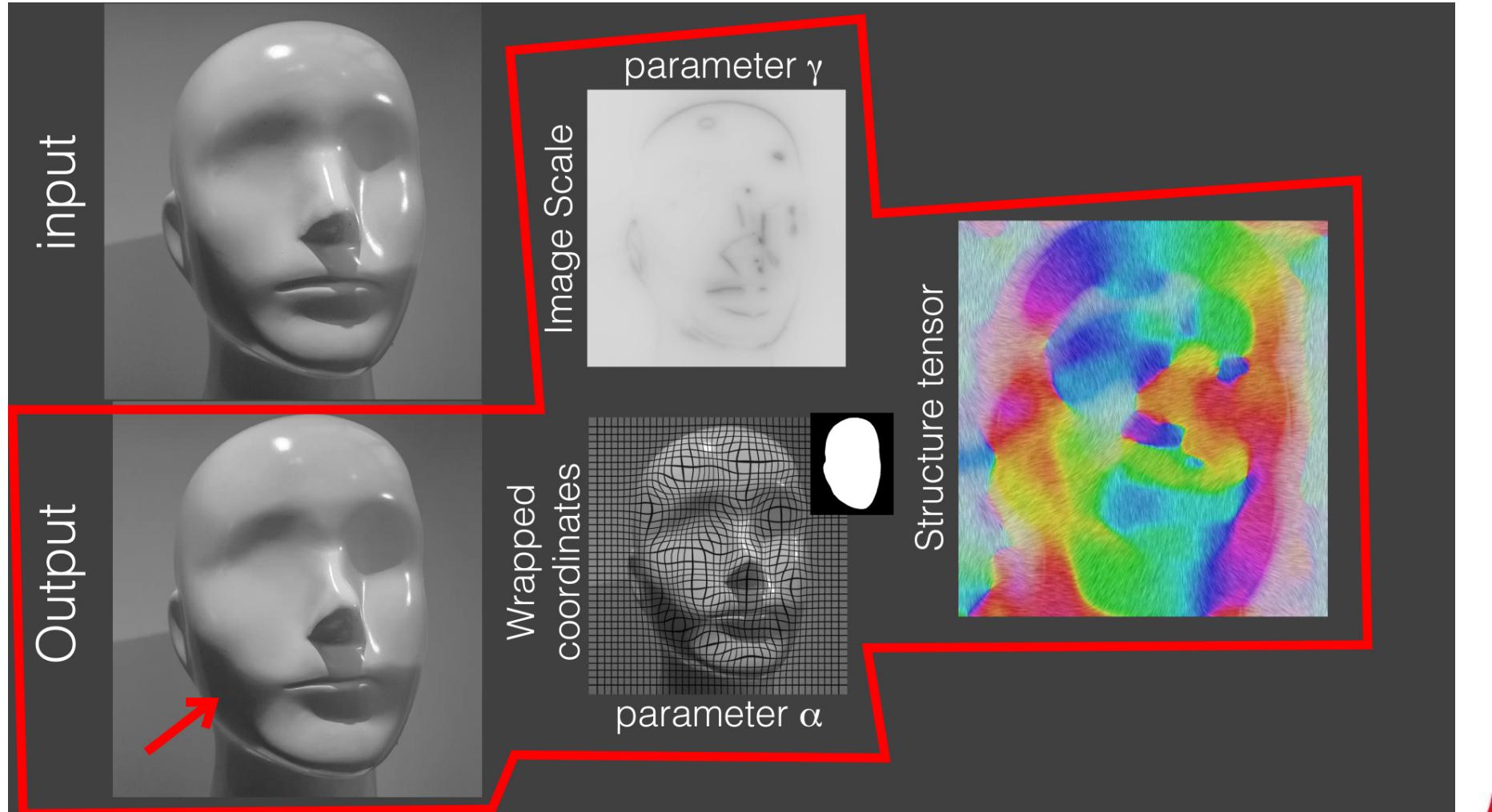
# Rounding: $\alpha = 1$



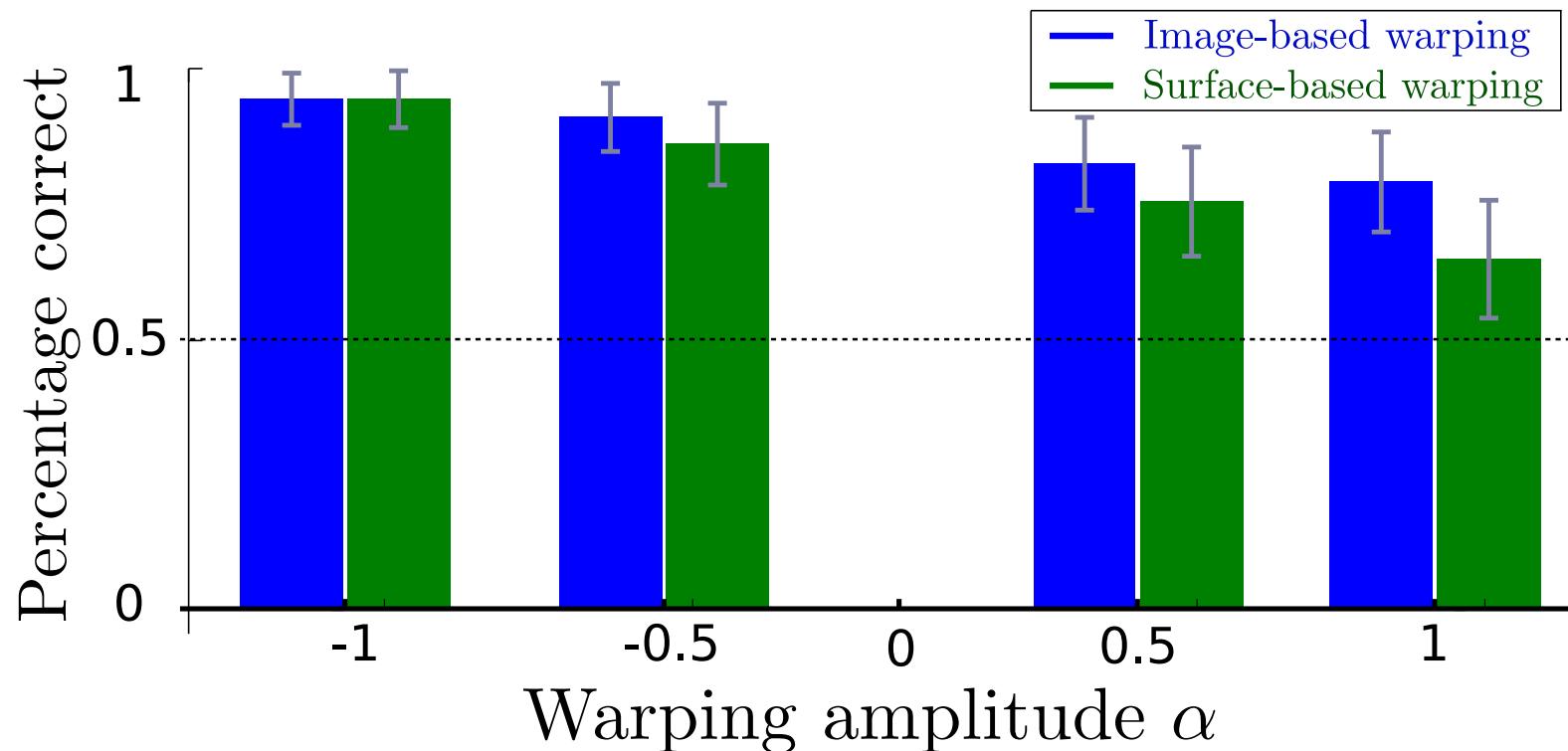
# Sharpening, small scale



# Sharpening, large scale



# User study



# Results

# Conclusion

Perception issues in Visualization

Perception of Depth – Application to DVR evaluation

Perception of Noise – Application to Uncertainty Vis.

Perception of Shape – Application to Image manipulation

In this talk: low level visual perception issues

cognitive levels much more difficult

Take away message: perception matters!



UNIVERSITÉ  
Grenoble  
Alpes



, 22nd May 2017, Visual perception in Visualization, G.-P. Bonneau