Curvature and the Rerceptual Organization of Texture Flow

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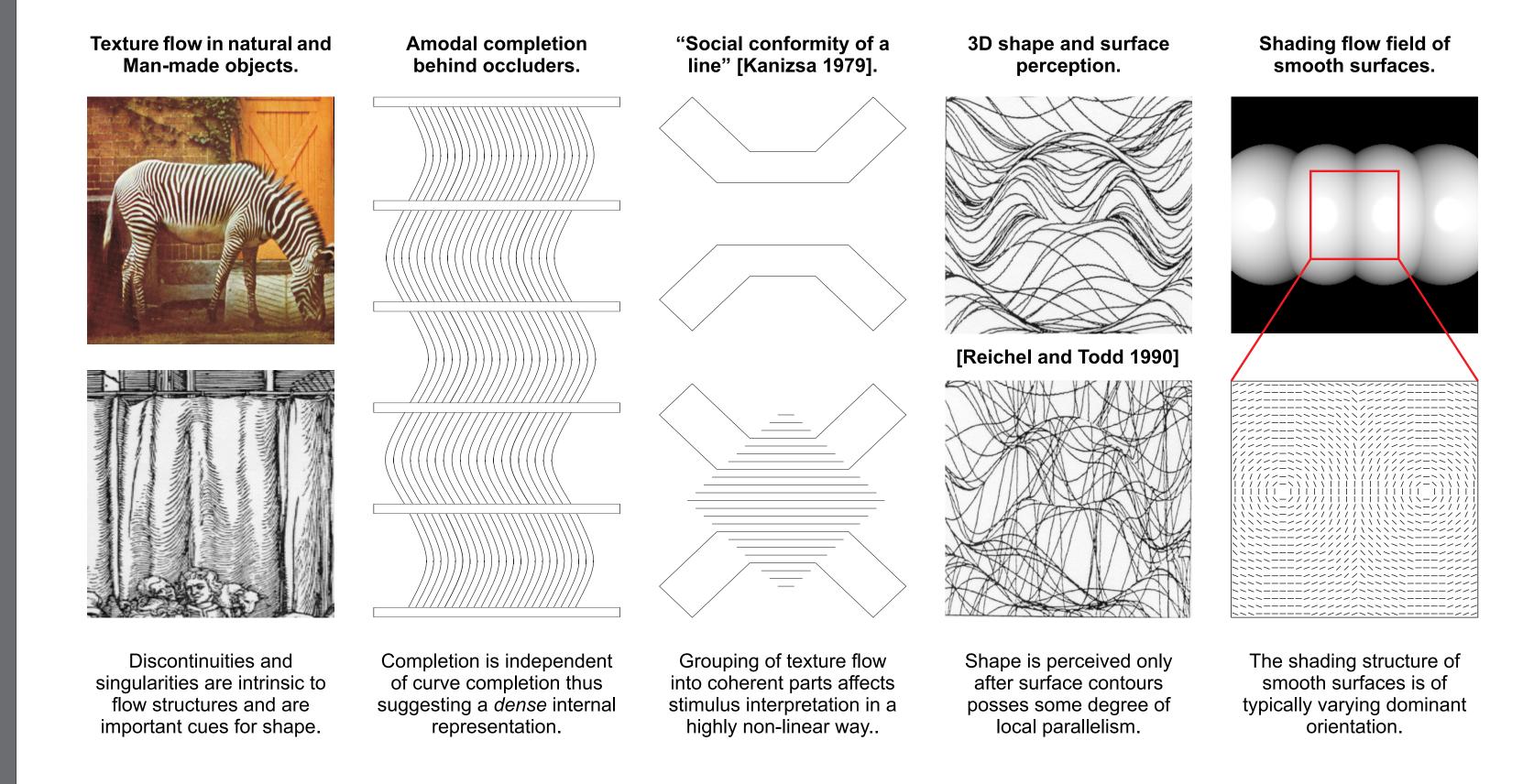
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1: Introduction

Texture flow (or orientation-defined texture) is a dense visual percept characterized by local (almost) parallelism and (typically varying) dominant local orientation. Its organization into coherent parts is fundamental to perceptual organization, shape interpretation, and shading analysis.

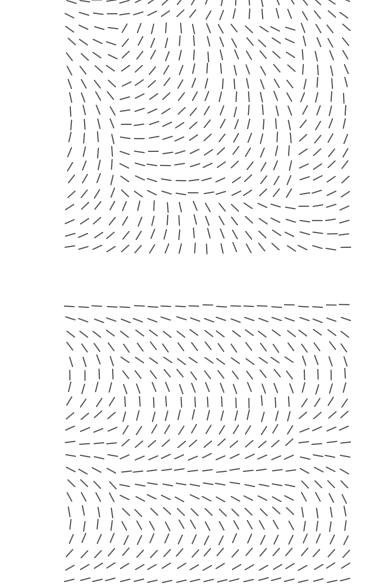


Which aspects of the *orientation content* influence orientation-based segmentation? Following texton theory and research into orientation gradients [e.g., Nothdurft 1985, Landy & Bergen 1991], current models for orientation-based segmentation depend only on the relationship between two scalars - the change of orientation between regions ($\Delta\theta$) and the change of orientation within regions ($\Delta\theta$) [e.g., Nothdurft 1991, Mussap & Levi 1999].

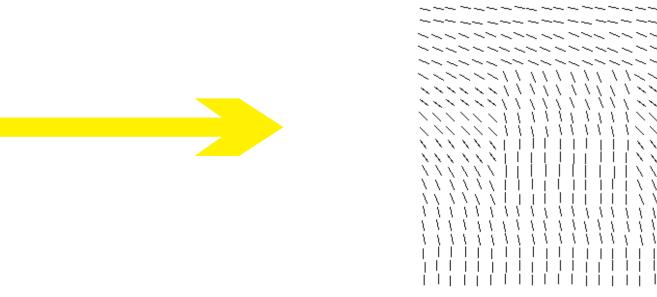
But:

Orientation gradients, like any other gradient, are vector quantities whose direction (as opposed to magnitude) may have strong perceptual consequences.

> The two patterns differ only in the direction of $\Delta\theta$. Its magnitude, as well as the value of $\Delta\theta$, are otherwise identical. Note the perceptual differences in the boundaries of the square.



The same $\Delta\theta_{\text{min}}$ and $\Delta\theta_{\text{min}}$ can give rise to boundaries of very different perceptual saliency.



In this pattern $\Delta\theta$ is constant across the image and $\Delta\theta$ is constant along the boundaries (a square). And yet, the top edge is much more salient than the bottom one. Why?

To understand what might cause these phenomena, a deeper understanding of the behavior of orientation in texture flows, and the parameters that govern is required. Geometrical analysis identifies two curvatures and psychophysical experimentation indicates that these curvatures are a critical factor in orientation-based segmentation, beyond $\Delta \theta_{\text{matrice}}$ and $\Delta \theta_{\text{within}}$!

2; Geometry and Representation

Any texture flow can be represented as a scalar function $\theta(x,y)$ over the image. Its 3D surface representation (where height=orientation) makes explicit the discontinuities (depicted as abrupt height change) and support the importance of the gradient. However, it provides no insight into the intrinsic geometry. For that we utilize the frame field representation [O'Neill 1966] which leads to the texture flow connection equation and its covariant derivatives:

 $\nabla_V E_N = -w_{12}(V)E_T$

Natural basis leads to two texture flow curvatures Tangential curvature : $\kappa_T \stackrel{\triangle}{=} w_{12}(E_T)$

which can be represented in terms of vector projection

Consequences:

 $\kappa_T = \nabla \theta \cdot E_T = \nabla \theta \cdot (\cos \theta, \sin \theta)$ $\kappa_N = \nabla \theta \cdot E_N = \nabla \theta \cdot (-\sin \theta, \cos \theta)$

► The same orientation gradient can give rise to different curvatures!!

Normal curvature : $\kappa_N \stackrel{\triangle}{=} w_{12}(E_N)$

(in red) to each point along the he covariant derivative The frame field representation

orientation as the frame is translated on the texture flow along an arbitrary direction *V.* Consequently, the tangential and normal curvatures are t covariant derivatives in the tangential and normal directions They quantify how much the frame rotates as we move parallel or *perpendicular* to the flow,

More importantly...a discontinuity with $\Delta \kappa = \Delta \kappa$ typically produces the strongest perceptual effect. In particular, with intermediate-valued gradients (10< $\nabla\theta$ <25), such discontinuities are the only ones, on average, to result in reliable segmentation (accuracy above 75%). curvatures cannot be simultaneously constant within a texture patch (however small) unless

performance.

Ignoring the curvature dimension, performance

is comparable to previous findings. Accuracy

However, significant differences are revealed

within each combination of $\nabla \theta$ and $\Delta \theta$ when

For small and intermediate $\nabla \theta$ values, a

discontinuity with $\Delta \kappa = 0$ significantly reduced the

"pop-out" effect and lowers segmentation

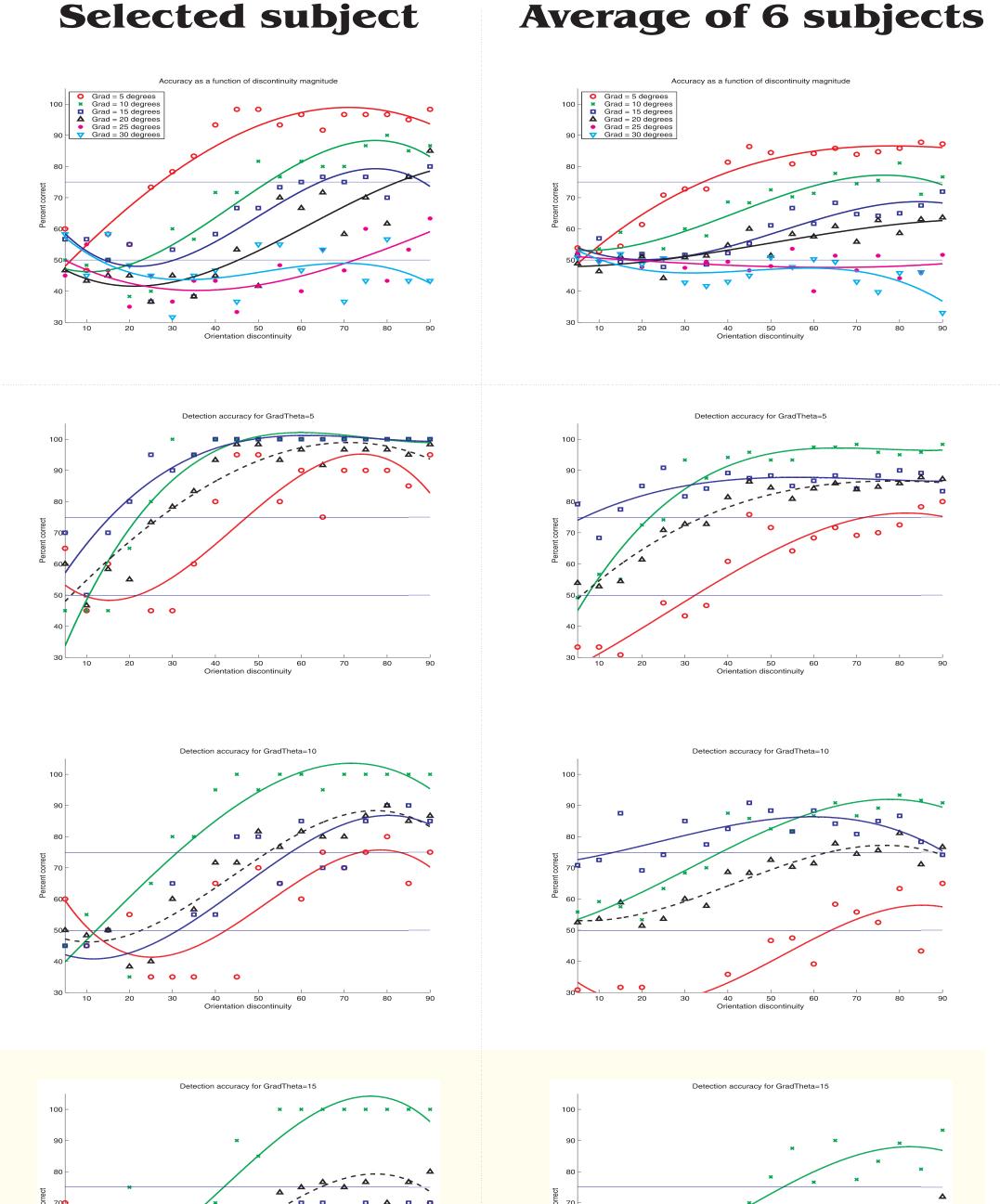
curvature discontinuities are taken into account.

decreases with larger $\nabla \theta$ and smaller $\Delta \theta$.

For $\nabla\theta$ higher than 25 degrees (graphs omitted here), no type of discontinuity surpasses the others and all of them, on average, lie around chance level.

<u>Legend</u>

Results



 $\nabla\theta$ =15 degs Discontinuities with $\Delta \kappa = \Delta \kappa$ are the only ones which cross the detection threshold

30 40 50 60 70 80 90 Orientation discontinuity

 $\nabla\theta$ =20 degs Discontinuities with $\Delta \kappa = \Delta \kappa$ are the only ones which cross the detection threshold

Although curvatures

are still small

discontinuities with

 $\Delta \kappa = 0$ are nevertheless

significanly inferior to

the other two types of

orientation boundaries

Discontinuities with

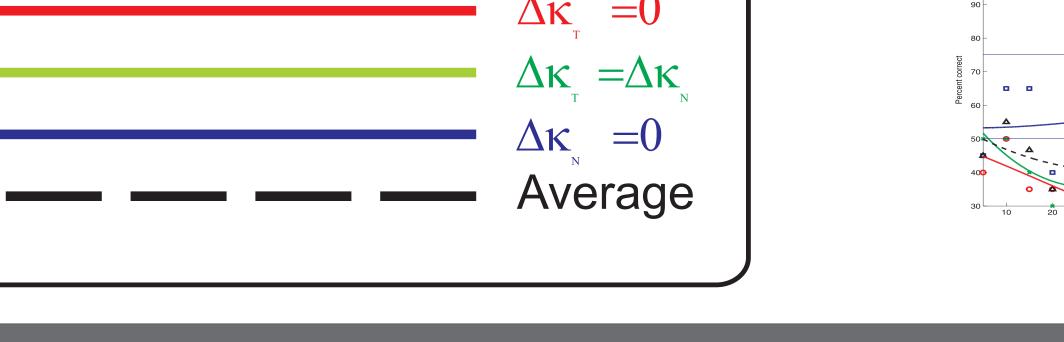
 $\Delta \kappa = 0$ are significantly

inferior to the other two

types of orientation

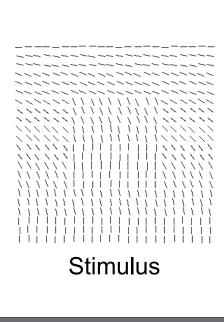
boundaries

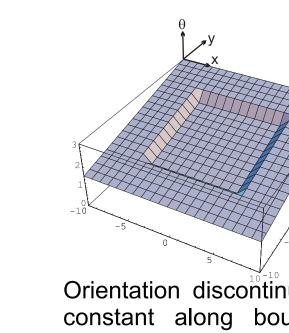
 $\nabla\theta$ =25 degs Performance on all types of discontinuities collapse to similar chance-level accuracy

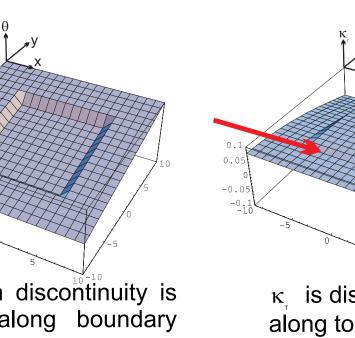


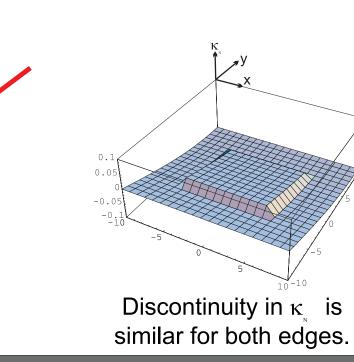
5: Implications

- Contrary to existing models, performance in orientation-based segmentation depends on more than the relationship between the (scalar values of the) rate of change of orientation within regions ($\Delta \theta$) and between regions ($\Delta\theta$). Since these notions are fundamental to preattentive vision in general, our findings justify a reexamination of the role of features, and feature gradients, in preattentive vision.
- ► Performance in orientation-based segmentation depends on discontinuities in flow curvatures.
- ► Psychophysics suggests the existence of a representation for curvature(s) in the human visual system. In particular, these experiments imply, for the first time, that the human visual system maintains a representation for a normal curvature, which demands an extension of current "association field" models dominated by curve integration [e.g., Field, Hayes, & Hess, 1993].
- ► Since flow curvatures are basic geometrical properties, an inquiry into the role of intrinsic geometry of other perceptual features (e.g., shading, color, motion) may be in place.
- Epilogue: differences in curvature discontinuities explain the asymmetry in the saliency of the top and bottom edges of the Nothdurft square in the Introduction!!.

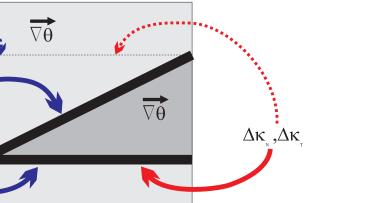




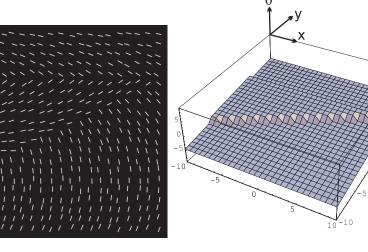




3: Methods **Stimuli:** figure/ground-style texture flows with two possible configurations (for the figure)

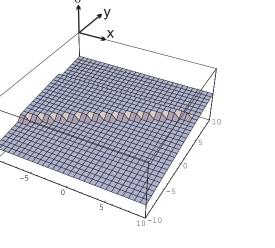


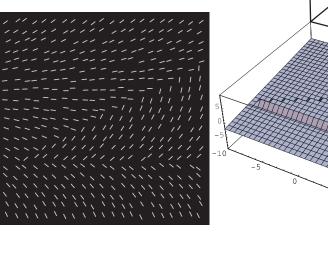
both are zero [Ben-Shahar & Zucker 2001]!!

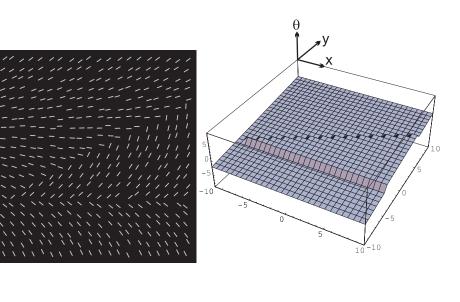


► Could texture flow curvatures be involved in orientation-based segmentation? Special care

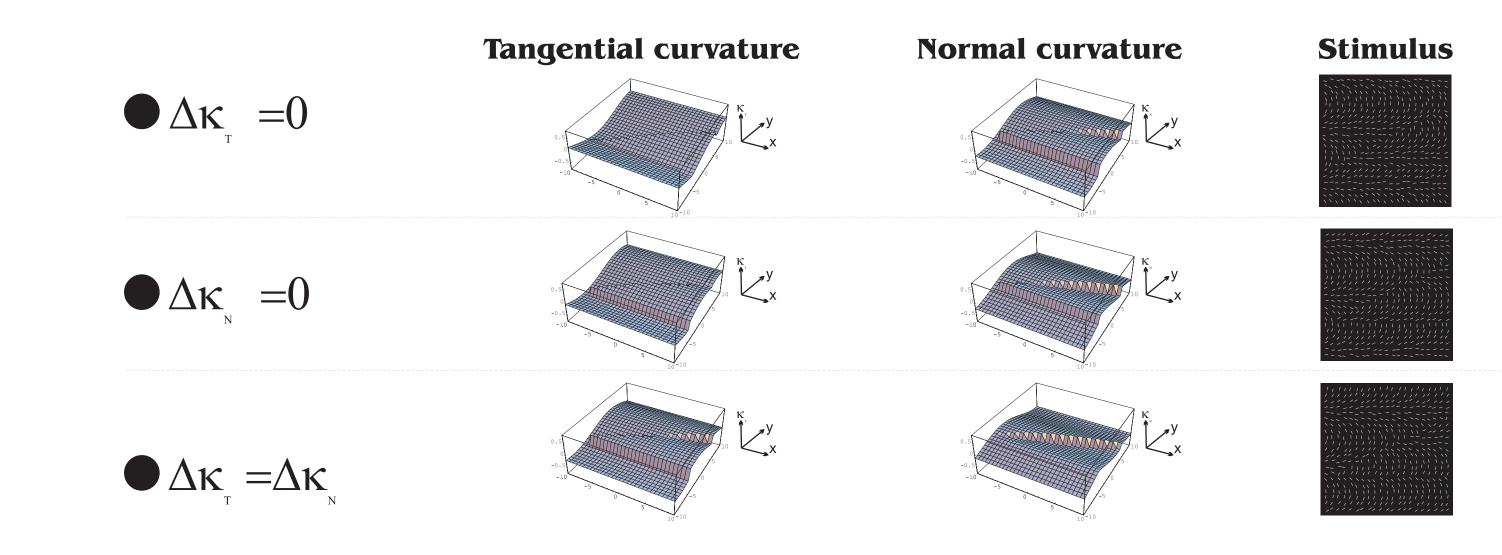
is needed when testing such a hypothesis because unlike orientation gradients, texture flow







- **constant** orientation gradient $\nabla \theta$ (= $\Delta \theta$), either {5, 10, 15, 20, 25, 30} degrees. **constant** orientation discontinuity $\Delta\theta$ (= $\Delta\theta_{\text{between}}$), either {5,10,15,20,...80,85,90} degrees.
- \blacktriangleright constant curvature discontinuities $\Delta \kappa$ and $\Delta \kappa$, either one of three possibilities Note: Given $\nabla \theta$, $\Delta \theta$ and one $\Delta \kappa$, the other $\Delta \kappa$ is fully determined through Eq. (1).



Procedure: Two-Alternative Forced-Choice task. 20 presentations of each combination of $\nabla \theta, \Delta \theta$, and curvature discontinuities (i.e., 60 presentations of each combination of $\nabla\theta$, $\Delta\theta$). Total of 6480 trials in six sessions of 1080 trials each.

