

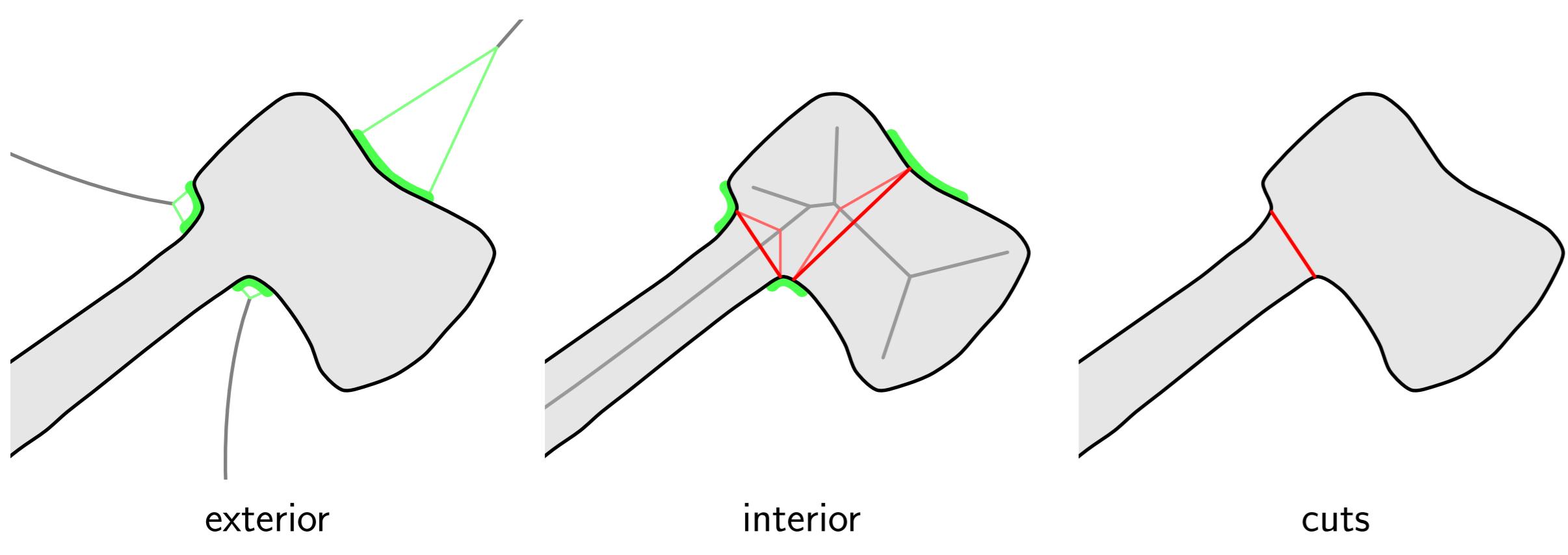
# Planar Shape Decomposition Made Simple

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## Motivation

- ▶ Planar shape decomposition without global optimization or differentiation
- ▶ All information available from (exterior and interior) medial axis representation
- ▶ Most rules and salience measure from psychophysical studies accommodated in a simple computational model



## Shape Representation

- ▶ A planar shape is a set  $X \subset \mathbb{R}^2$ ; its boundary  $\partial X$  is a finite union of mutually disjoint simple closed curves

### Medial axis [1]

- ▶ The distance map  $D(X) : X \rightarrow \mathbb{R}$  is a function mapping each point  $y \in X$  to

$$D(X)(y) = \inf_{x \in \partial X} \|y - x\|$$

- ▶ For  $y \in \mathbb{R}^2$ , the projection or contact set

$$\pi(y) = \{x \in \partial X : \|y - x\| = D(X)(x)\}$$

is the set of points on the boundary at minimal distance to  $y$ ; each  $x \in \pi(y)$  is a projection or contact point of  $y$

### (interior) medial axis

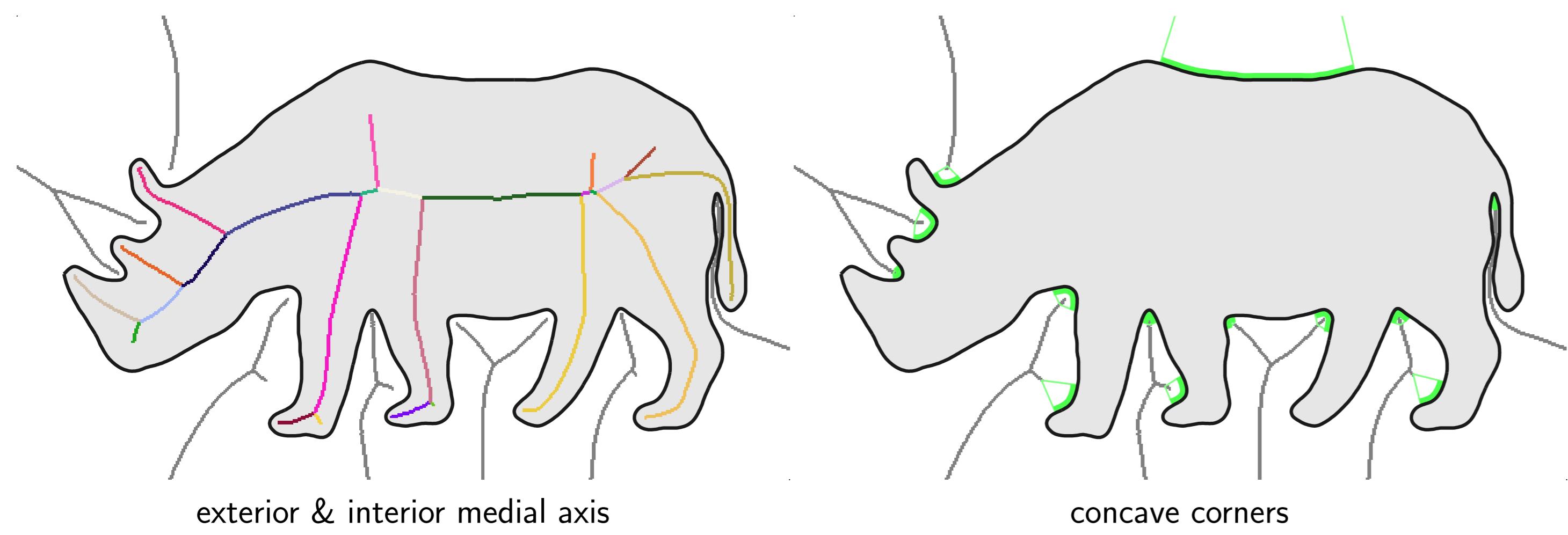
$$\mathcal{M}(X) = \{x \in \mathbb{R}^2 : |\pi(x)| > 1\}$$

is the set of points with more than one projection points

- ▶ The exterior medial axis of  $X$  is the medial axis of its complement  $\mathbb{R}^2 \setminus X$

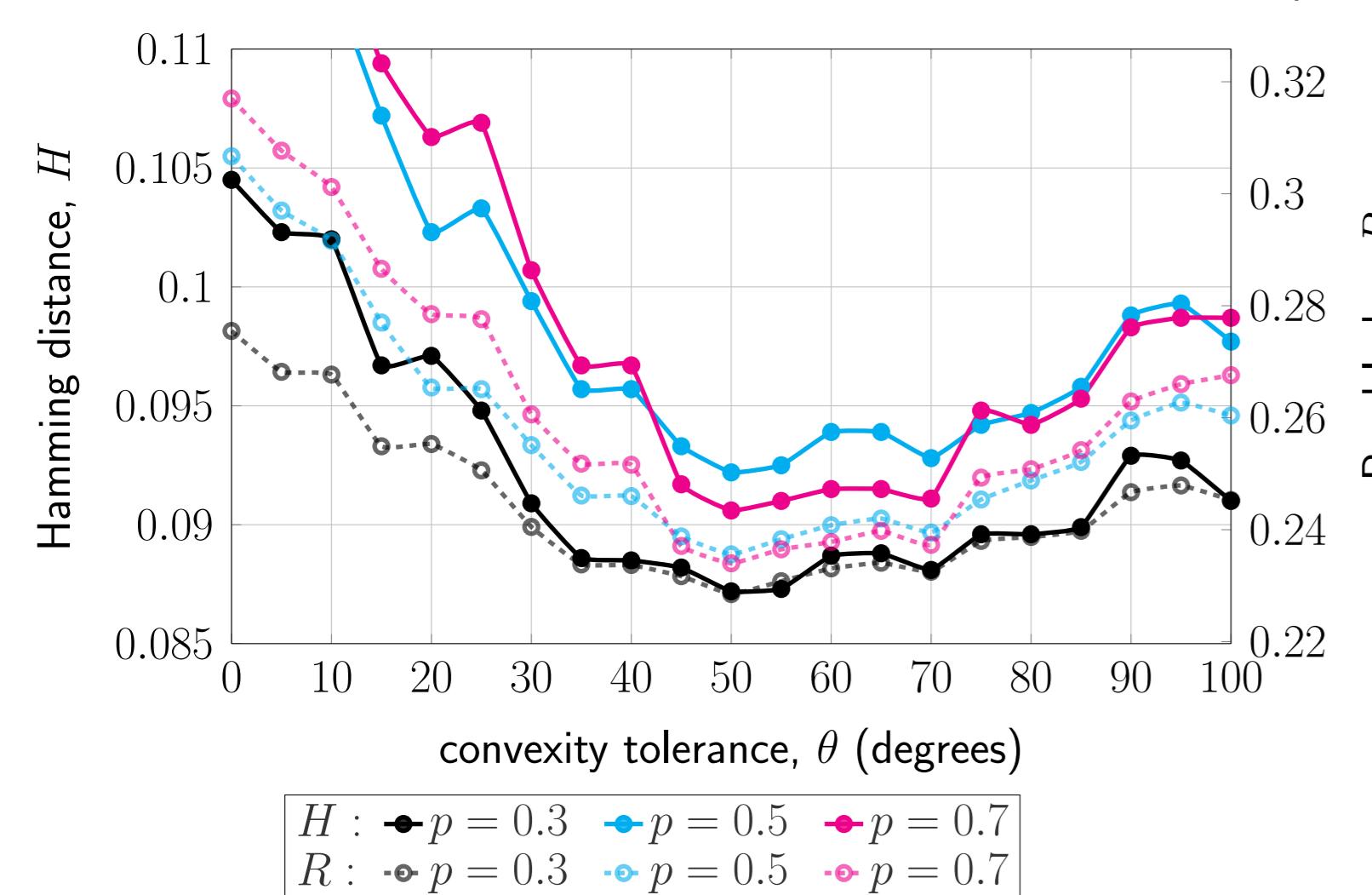
### Construction [2]

- ▶ Given two points  $x, y \in \partial X$ , the arc length  $\ell(x, y)$  is the length of the minimal arc of  $\partial X$  having  $x, y$  as endpoints or  $\infty$  if no such arc exists
- ▶ Given a point  $z$ , its chord residue  $r(z) = \sup_{x, y \in \pi(z)} \ell(x, y) - \|x - y\|$  is the maximal difference between arc length and chord length over all pairs of points in its projection
- ▶ Construction begins at local maxima of distance map and propagates as long as the residue is higher than a given threshold  $\sigma > 0$



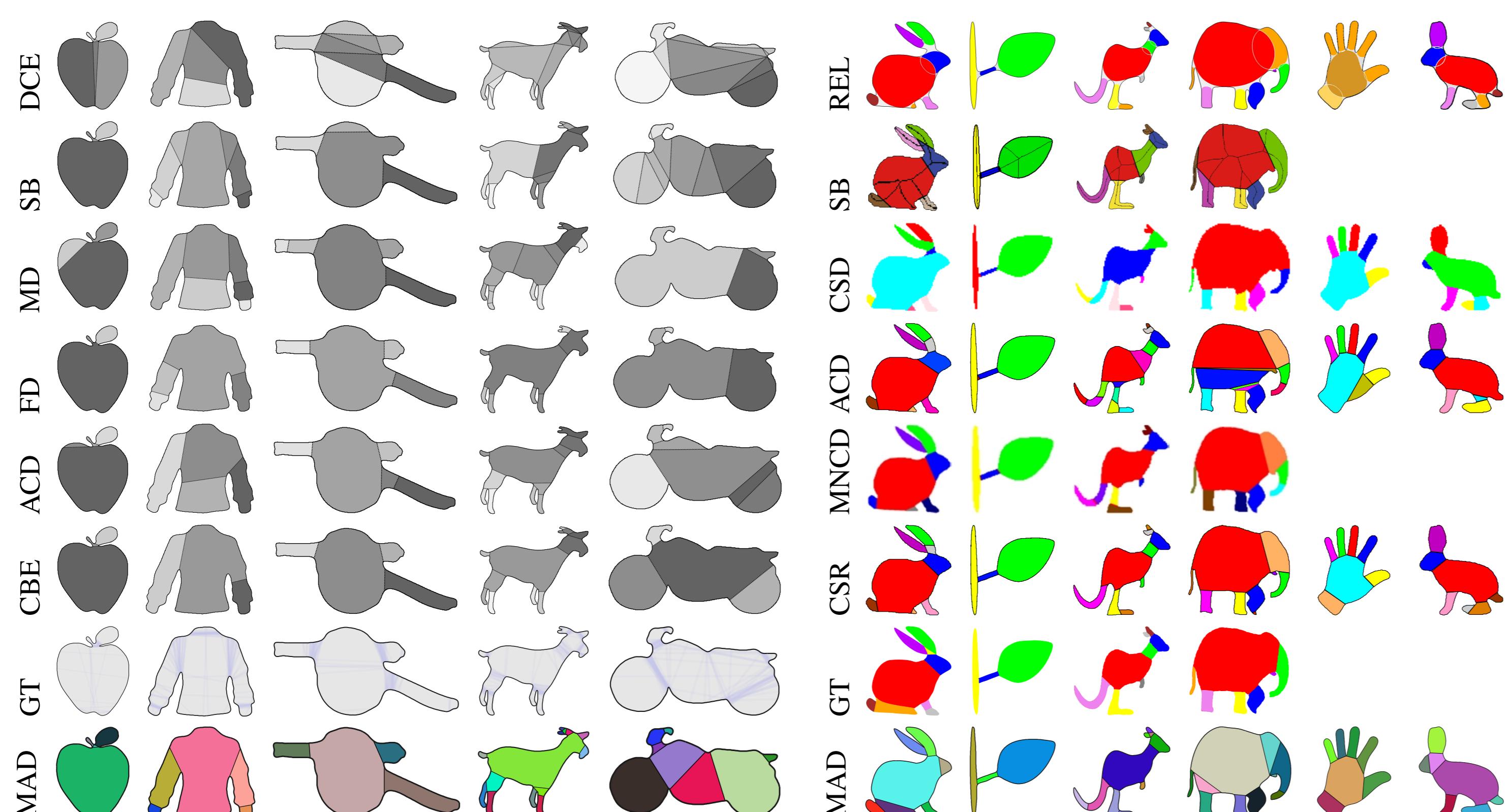
## Quantitative Evaluation

- ▶ Evaluation measures: Hamming distance and Rand Index (Jaccard distance)



$H$ : ●  $p = 0.3$  ○  $p = 0.5$  ■  $p = 0.7$   
 $R$ : ○  $p = 0.3$  □  $p = 0.5$  ▲  $p = 0.7$

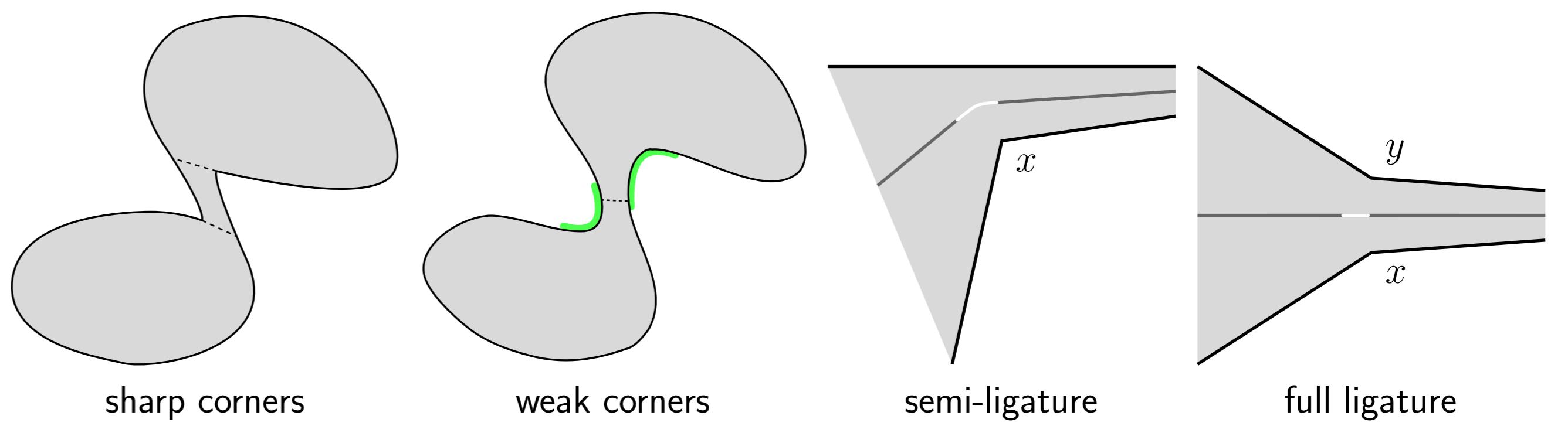
## Qualitative Evaluation



## Shape Decomposition

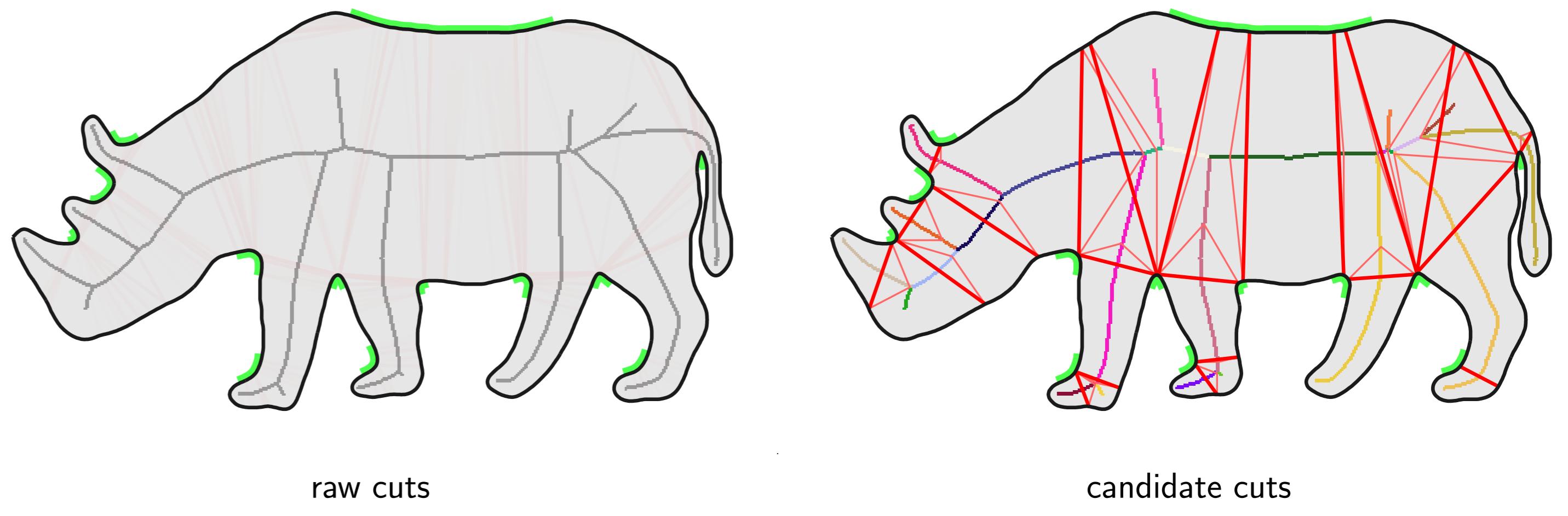
### Minima rule

- ▶ A shape should be cut at points of negative minima of curvature [3]
- ▶ But these are exactly projection points of end vertices of the exterior medial axis [1]
- ▶ Moreover, one may get not just one boundary point but an entire arc, called a (concave) corner
- ▶ Without differentiation, an end-vertex with its two projection points determine the position, spatial extent, orientation and strength of each concavity



### Symmetry

- ▶ A cut of a shape  $X$  is a line segment connecting two points of  $\partial X$
- ▶ All prior work examines all possible pairs of points on  $\partial X$  as candidate cut endpoints; we only consider pairs of points that are projection points of the same point of the interior medial axis
- ▶ A cut may have one or two corner points as endpoints, called single or double cut respectively
- ▶ Raw cuts: traverse interior medial axis collecting all pairs of projection points such that at least one lies on a corner; this is stronger than requiring cuts to cross an axis of local symmetry [3]

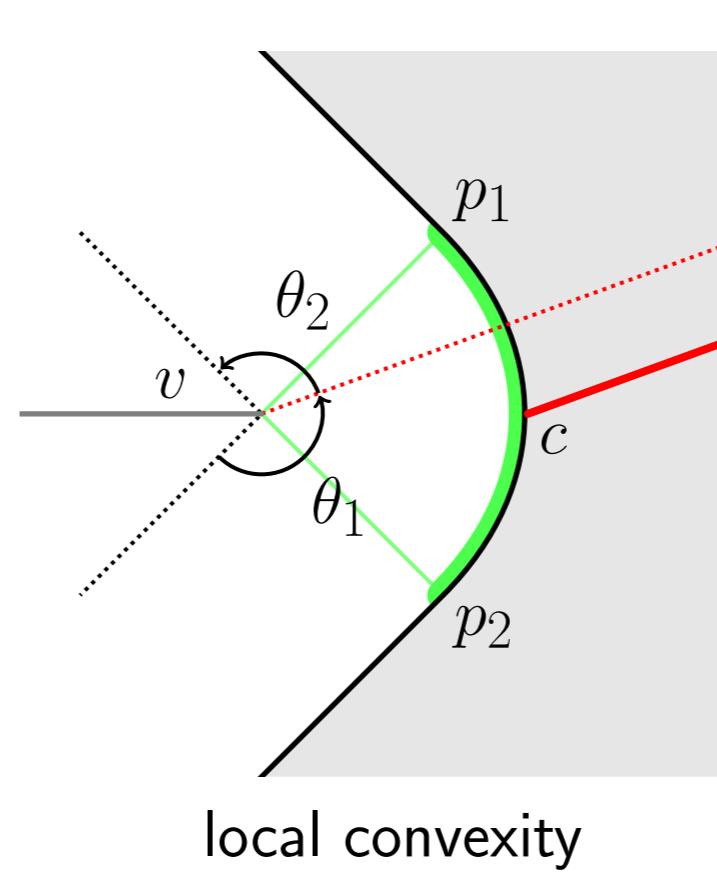


### Equivalence

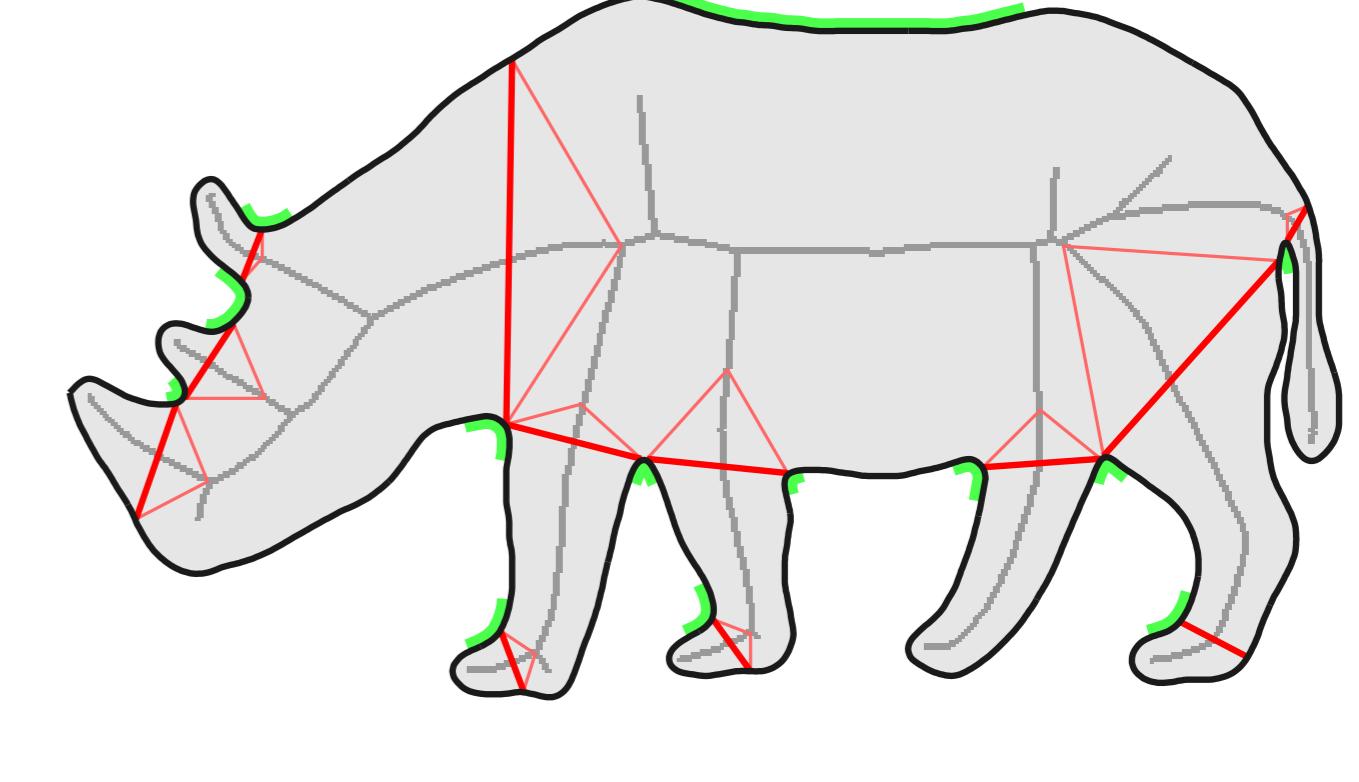
- ▶ Select candidate cuts by applying equivalence rules on raw cuts
- ▶ Branch equivalence: two cuts on the same branch whose endpoints share at least one corner; double cuts have priority over single cuts
- ▶ Corner equivalence: two (double) cuts whose endpoints lie on the same pair of corners; the cut with the maximal protrusion strength is selected

### Salience measures

- ▶ Protrusion strength: ratio of cut length to arc length; select cuts with protrusion less than  $p$



local convexity



selected cuts

### Local convexity & short-cut rule

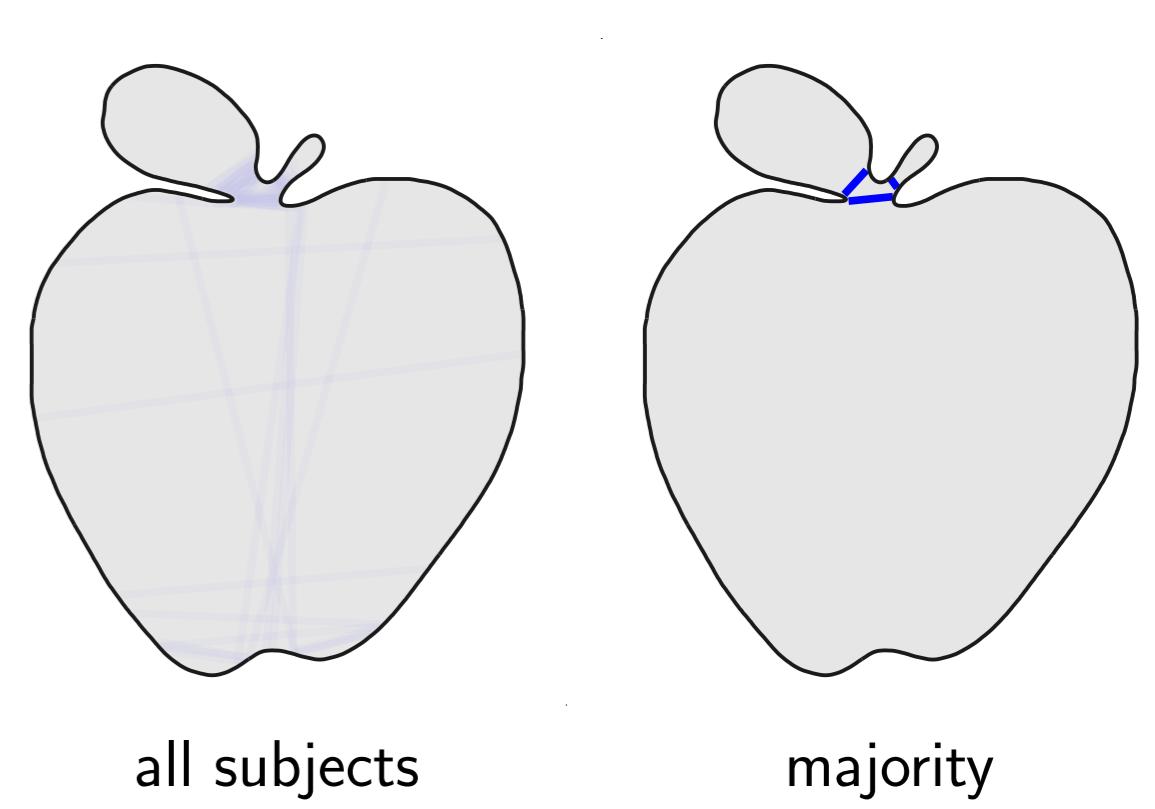
- ▶ Most approaches seek the minimal number of cuts such that each shape part is approximately convex
- ▶ But negative minima of curvature are exactly points where the shape is locally maximally concave
- ▶ For each corner, we select independently the minimal number of cuts such that the interior angle of each part is less than  $\pi + \theta$ , where  $\theta$  is a tolerance
- ▶ Priority given according to short-cut rule [4], but arbitrary salience measures apply

## Dataset

- ▶ Snodgrass and Vanderwart (S&V) everyday object dataset contains 260 line drawings
- ▶ De Winter and Wagemans dataset [5] evaluates exactly segmentation of 88 object outlines
- ▶ The subset has been converted to smooth outlines and each segmented by 39.5 subjects on average
- ▶ For each shape there are 122.4 part-cuts, that is 3.1 cuts per subject on average

## Majority Voting

- ▶ Part-cuts of human subjects are typically inconsistent: evaluate on majority cuts
- ▶ Apply agglomerative clustering on all human cuts according to arc distance
- ▶ Select cluster representatives by averaging endpoints on the parametrization of the boundary curve
- ▶ Discard cluster with less than  $t$  votes



## References

- [1] Choi et al. Mathematical theory of medial axis transform. *Pacific Journal of Mathematics*, 1997.
- [2] Avrithis & Rapantzikos. The medial feature detector: stable regions from image boundaries. *ICCV*, 2011.
- [3] Hoffman & Richards. Parts of recognition. *Cognition*, 1984.
- [4] Singh et al. Parsing silhouettes: the short-cut rule. *Perception and Psychophysics*, 1999.
- [5] De Winter & Wagemans. Segmentation of object outlines into parts. *Cognition*, 2006.