

# Evaluating Superpixels in Video: Metrics Beyond Figure-Ground Segmentation

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process  
automation

# Outline

- What are **superpixels** and what are they good for?
- Introduce two **new criteria** for evaluation based on ground truth **optical flow**
- Results of various open source **algorithms**

# What are Superpixels?



Pixel



Superpixel



Segments /  
Objects



- Increasing size
- Increasing meaning

# What are Superpixels good for?

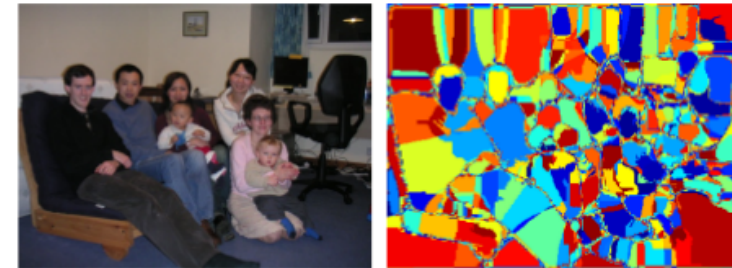
## Segmentation



(a) original image (b) superpixels (c) confidences (d) classification (e) GC refinement

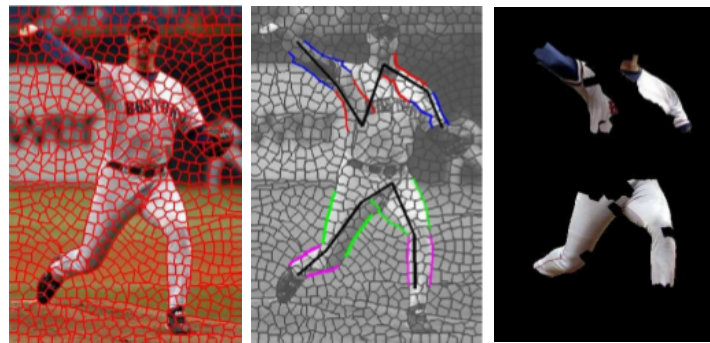
Mehrani and Veksler, 2010

## Object Recognition



Pantofaru et al., 2008

## Body model estimation



Mori, 2005

## Depth estimation



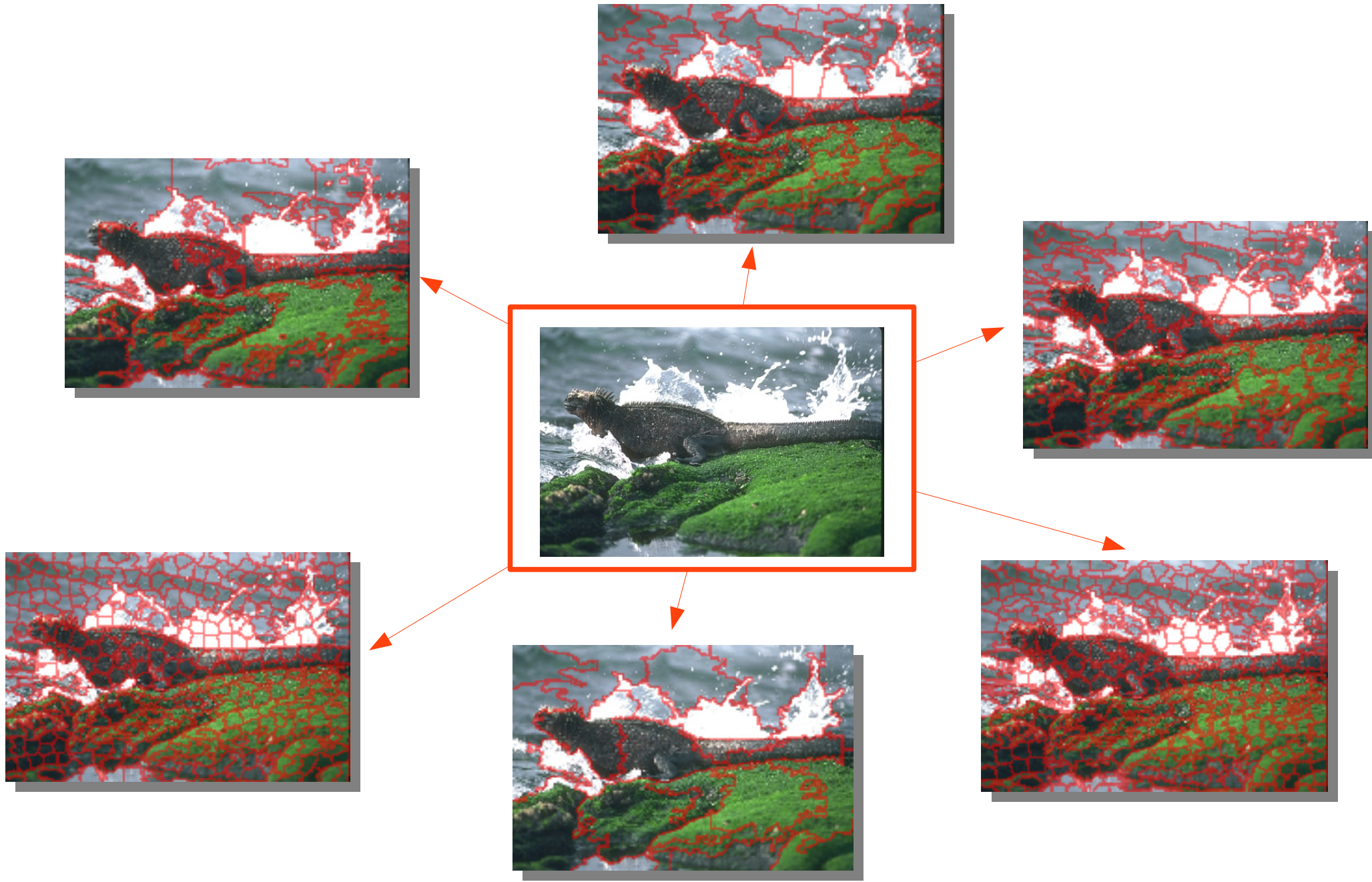
Zitnick and Kang, 2007

## Multi-class object segmentation

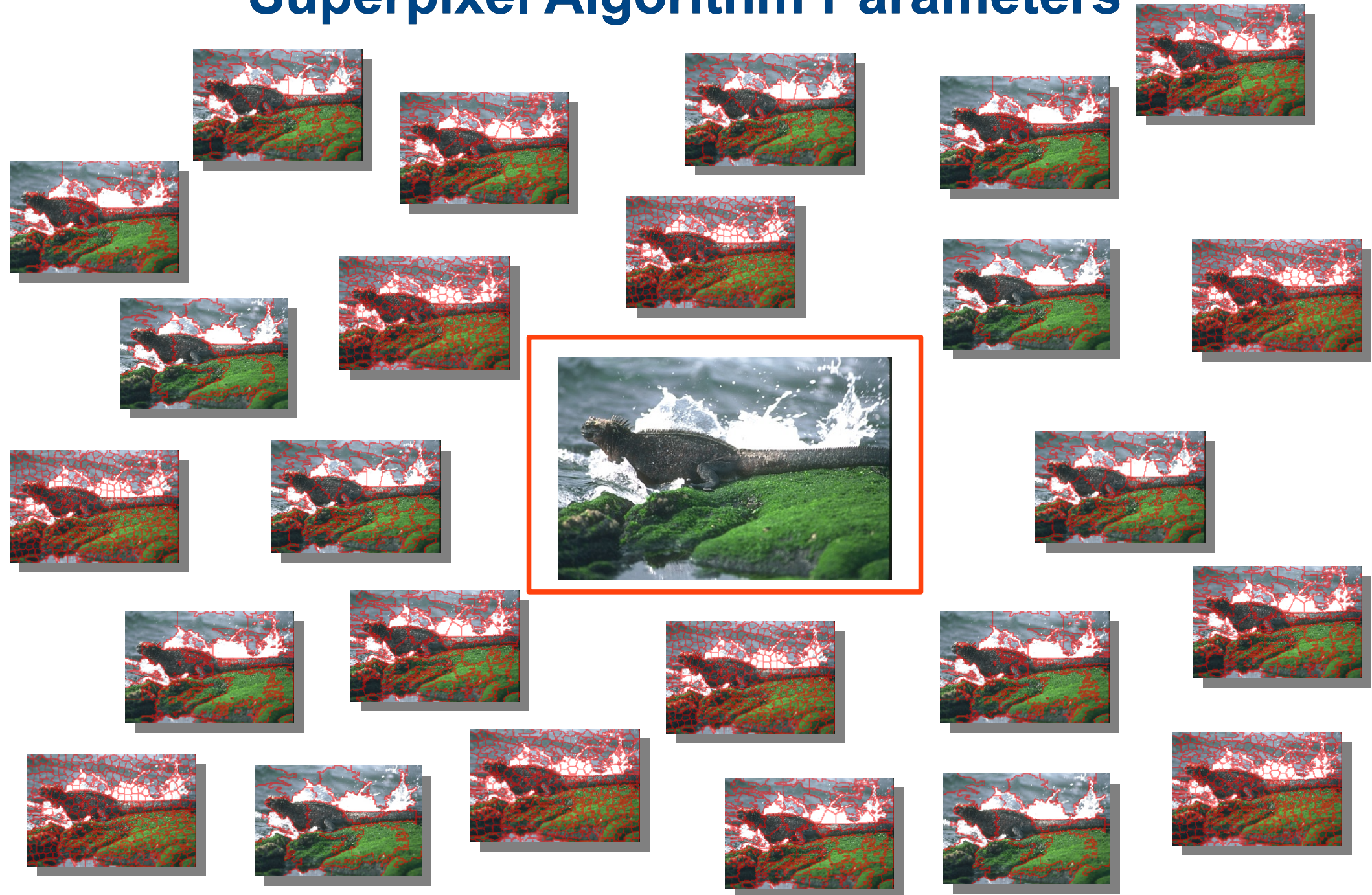


Yang et al., 2010

# Superspixel Algorithms



# Superspixel Algorithm Parameters



# What are desired properties of superpixel algorithms?

- Segmentation quality
  - Boundary recall
  - Undersegmentation error

→ Existing metrics based on figure ground segmentations
- Runtime
- Availability (implementation)
- So ... **Why am I standing here?**

# Stability-Criteria

Does the segmentation algorithm find the same regions or object boundaries robust to changes in the image?



**Unstable  
Segmentation**

**vs**



**Reasonably Stable  
Segmentation**



# Discontinuity-Criteria

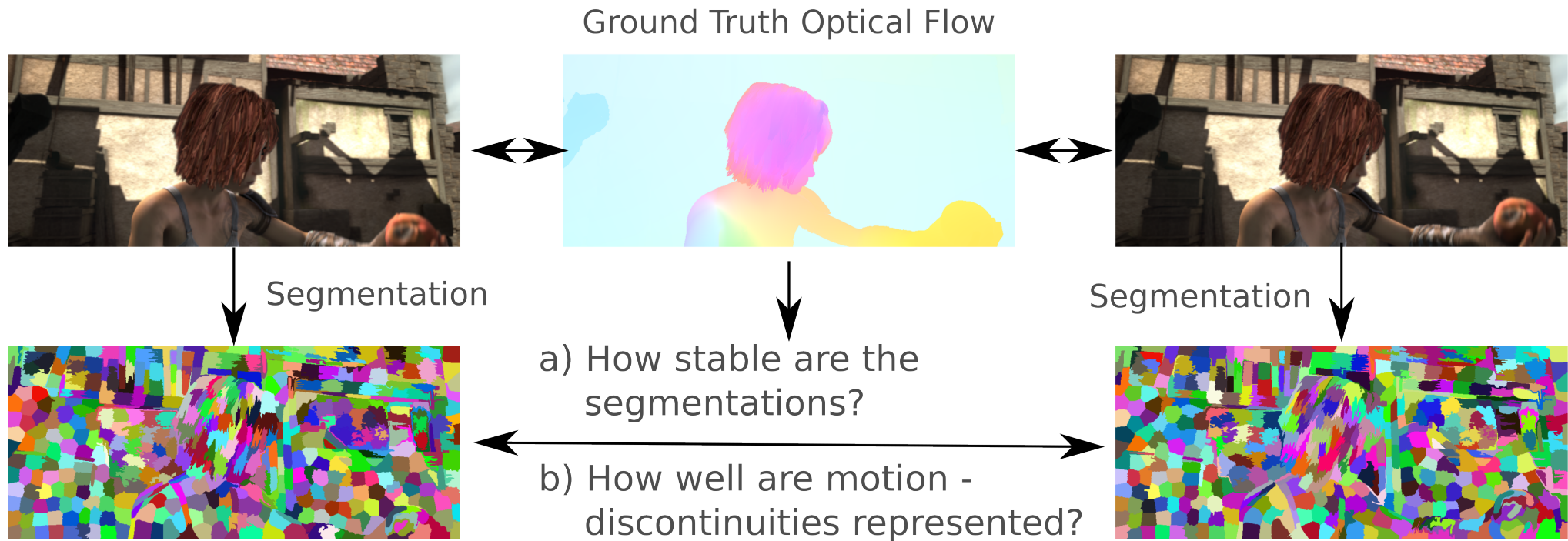
How well are motion discontinuities in the image sequence represented by the algorithms segment boundaries?

E.g.:

- Motion gradient between a moving foreground object and the background
- The interior of a non-rigid object

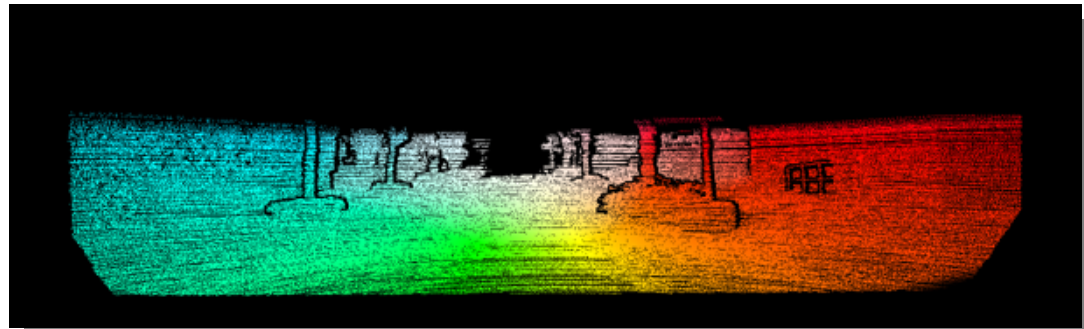


# Idea: Exploit Ground Truth Optical Flow



# Optical Flow Datasets: KITTI

- 194 image pairs of street scenes
- Ground truth optical flow from 3d laser data
- Geiger et al., 2012



# Optical Flow Datasets: SINTEL

- Based on **animated** short movie Sintel
- 23 sequences with 20-50 frames
- Almost **perfect ground truth** optical flow computed from rendering information
- Butler et al., 2012



## So far:

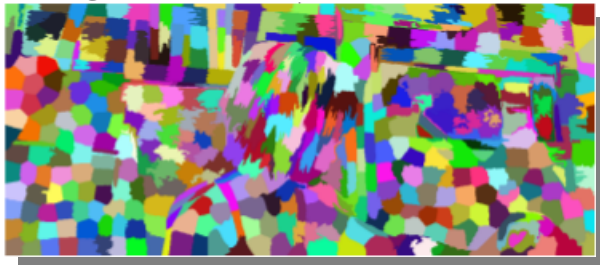
- Introduced two **new criteria** for a good superpixel algorithm
    - Stability
    - Representation of motion discontinuities
  - Proposed to use **ground truth optical flow** for evaluation of superpixel algorithms
  - Introduced the two used **datasets** (KITTI & Sintel)
- Now the more technical part: **Metrics**
- Followed by **results** on several available algorithms

# Measuring the Segmentation Stability

Image 1



Segmentation L1



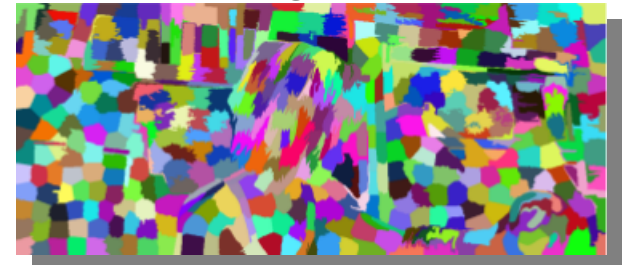
Apply ground truth optical flow



Image 2



Segmentation L2



Undersegmentation Error



# Measuring the Segmentation Stability: MUSE

How well is the segmentation L2 reconstructed by the transformed L1?

$$MUSE = \frac{1}{N} \left[ \sum_{a \in L_1^F} \left( \sum_{b \in L_2: a \cap b \neq \emptyset} \min(b_{in}, b_{out}) \right) \right]$$

- Motion undersegmentation error: MUSE
- Reconstruct each segment of L2 with segments of L1
- Pixels without valid flow information are ignored
- Since undersegmentation error is not a symmetric metric we compute the average of both cases

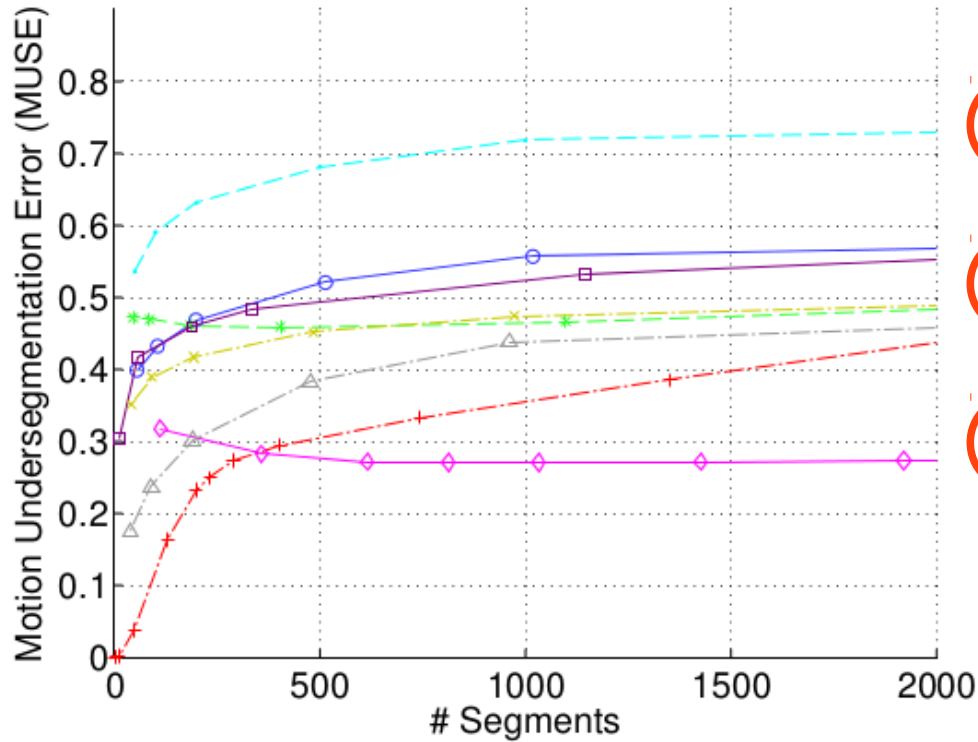
# Results: Compared Algorithms

- Requirements:
  - Available open source **implementation**
  - Reasonable **runtime** on images of size  $1024 \times 436$  for application on videos
- Compared Algorithms:
  - Felzenszwalb-Huttenlocher Segmentation (FH)
  - Edge Augmented Mean Shift (EAMS)
  - Quickshift (vIQS)
  - Marker-Controlled Watershed (WS)
  - Entropy Rate Superpixel Segmentation (ERS)
  - Two implementations of Simple Linear Iterative Clustering
    - Original implementation (oriSLIC)
    - Implementation of VIFeat (vISLIC)
  - Baseline: Regular grid (BOX)

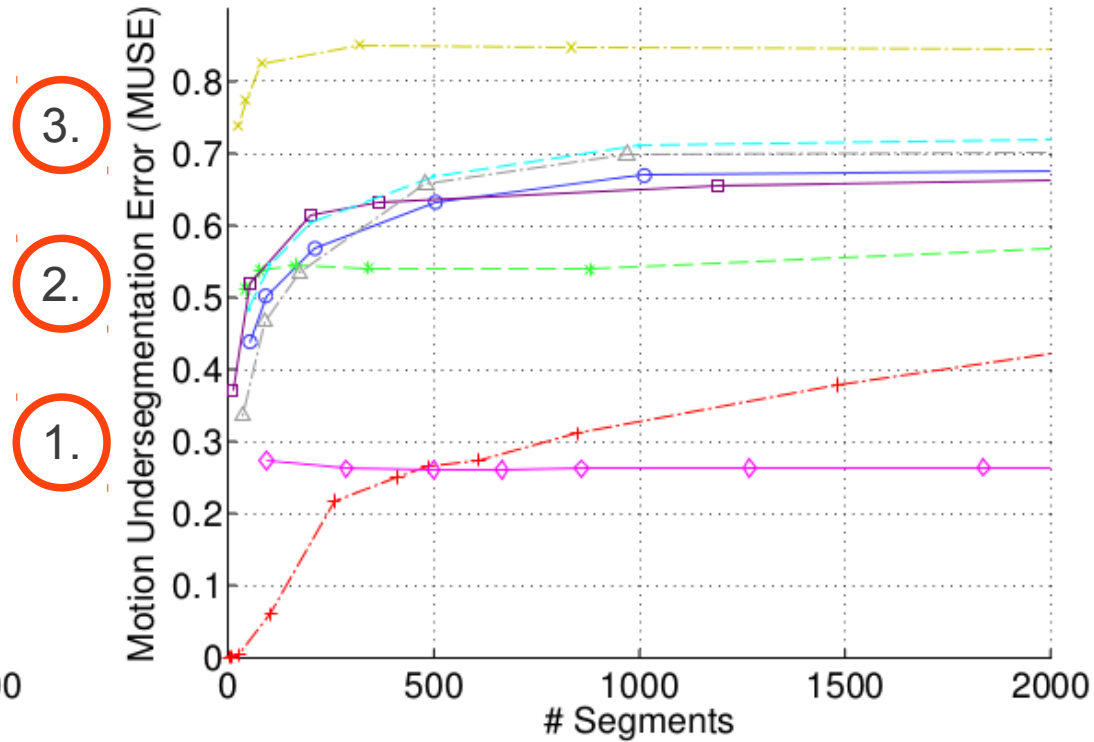


# Results: MUSE

a) Sintel



b) KITTI



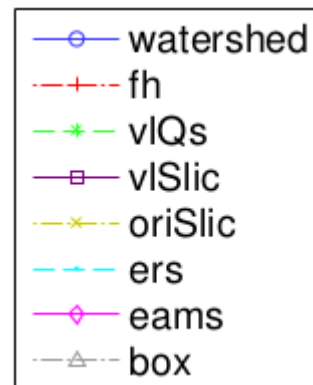
3.

2.

1.

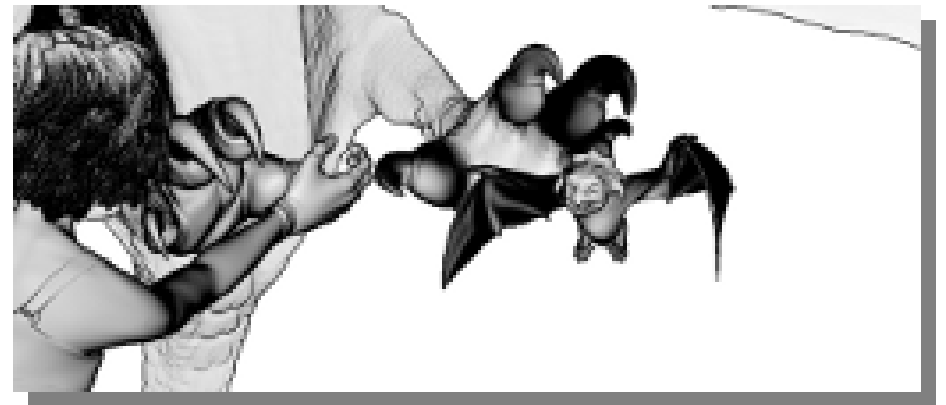
We can distinguish **three groups** of algorithms:

1. Gradient affine algorithms without regulation of size or distribution
2. Algorithms with additional stable boundaries (compactness constraints)
3. Unstable Algorithms (parameter sets)



# Measuring the Accordance with Motion Discontinuities

- Motion discontinuities result in high gradients in the optical flow field



- High motion gradient indicates **objects or object parts that can move differently** and thus should probably be handled individually in the application.
- Motion Discontinuity Error: MDE

# Measuring the Accordance with Motion Discontinuities: MDE

Summarize over all image pixels (i,j)

Strength of motion discontinuity at this pixel

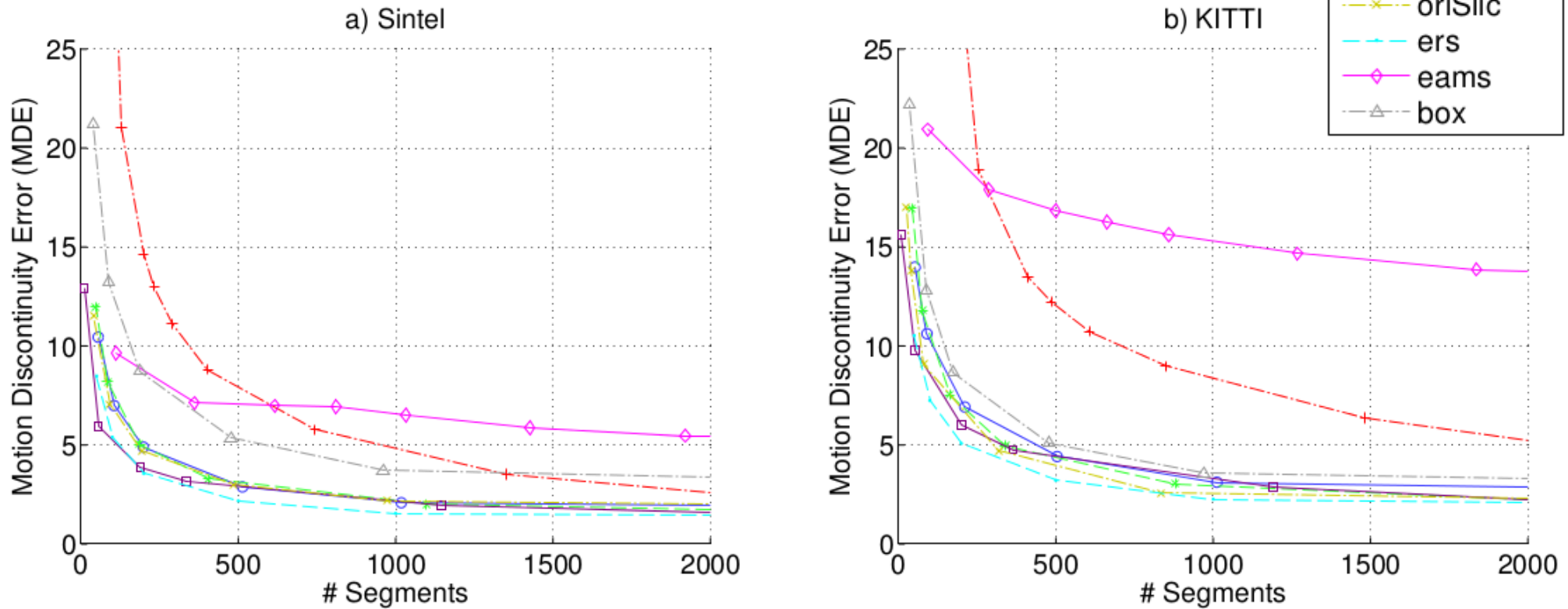
$$MDE = \frac{1}{\sum_i \sum_j \|\nabla F(i, j)\|_2} \sum_i \sum_j \|\nabla F(i, j)\|_2 \cdot D(B(i, j))$$

Total amount of motion

How far is the pixel from a segment border?

- F: A ground truth optical flow field from an image I to another image
- B: The boundary image of a segmentation of image I
- D(B): The distance transform of B containing for each pixel the distance to the nearest segment boundary

# Results: MDE



Results are inverted:

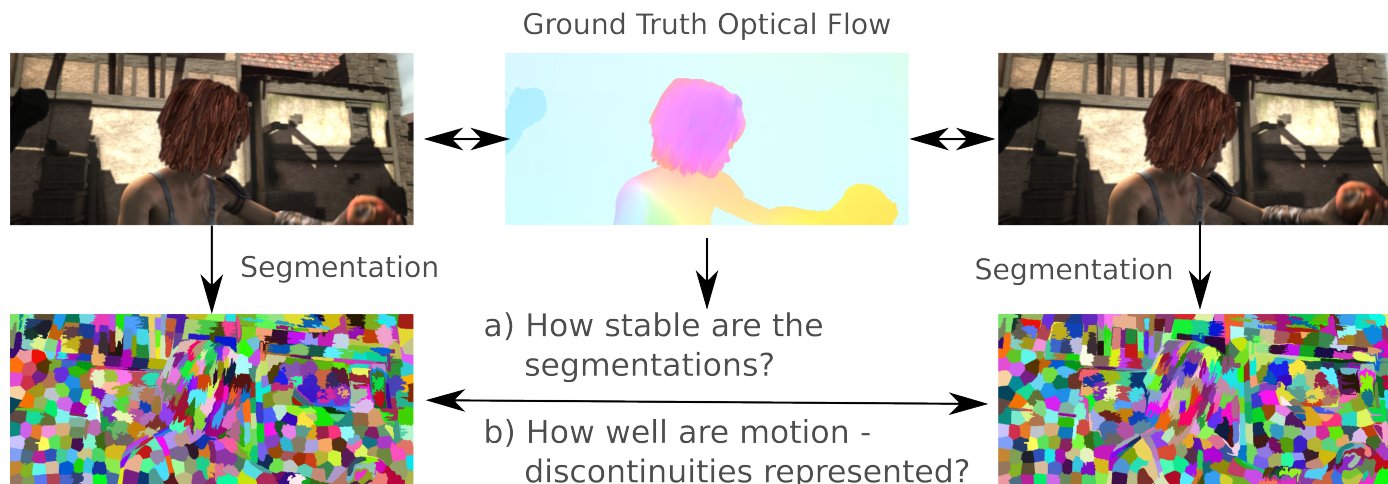
- Compactness constraints help a lot
  - They introduce additional borders that decrease the average values of the distance transform of the boundary image
- Exclusively relying upon image gradients is not enough

## Discussion of the metrics

- MDE favors segmentations with many boundary pixels
  - Similar to measures like boundary recall
  - Should be used together with complementing measurements
- MDE and MUSE are somehow complementary
  - Look for well balanced algorithms/parameters
- MDE and MUSE do not replace figure-ground metrics!

# Conclusions

- Proposed two novel criteria for evaluating superpixel segmentations
  - Segmentation stability
  - Representation of motion discontinuities
- Exploit ground truth optical flow data to measure them
  - **Motion Undersegmentation Error (MUSE)** evaluates the stability of segmentations
  - **Motion Discontinuity Error (MDE)** evaluates how well differently moving image parts are separated by the segmentation algorithms.
  - Matlab Toolbox: <http://www.tu-chemnitz.de/etit/proaut/forschung/superpixel.html>
- (vISLIC and Quickshift algorithms showed best balanced results)



# Questions?

- E.g.:
  - How are the **parameters** of the algorithms chosen?
  - Stable superpixels for video? Why not **supervoxels!**?
  - What should the **perfect superpixel algorithm** look like?
  - What about **other algorithms** like Normalized Cuts, Turbopixels, Superpixel Lattice, gPb-OWT-UCM, ...?

Or let's discuss at the coffee break