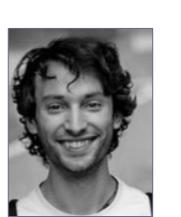


Improved Image Boundaries for Better Video Segmentation

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¹ Max Planck Institute for Informatics, Germany ² OSRAM Corporate Technology, Germany ³ Saarland University, Germany











Video Segmentation

Goal: Find segments which describe the video in a way that mimics human annotations

Video







time

Human annotations:





Appearance and motion cues

Spatio-temporal neighborhood

[Fragkiadaki & Shi CVPR'12], [Ochs et al. PAMI'14], [Galasso et al. CVPR'14], [Yi & Pavlovic ICCV'15]

Video

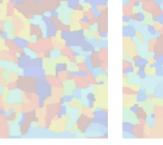






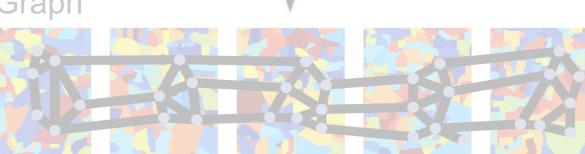
Superpixels







Graph



Segmentation

















3. Graph partitioning

1. Superpixels per frame

2. Graph construction

Spectral clustering methods



Appearance and motion cues

Spatio-temporal neighborhood

Spectral clustering methods

1. Superpixels per frame

2. Graph construction

3. Graph partitioning

[Fragkiadaki & Shi CVPR'12], [Ochs et al. PAMI'14], [Galasso et al. CVPR'14], [Yi & Pavlovic ICCV'15]

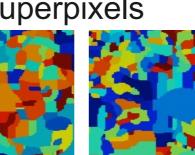
Video

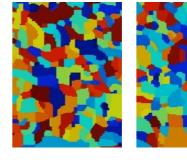


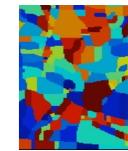




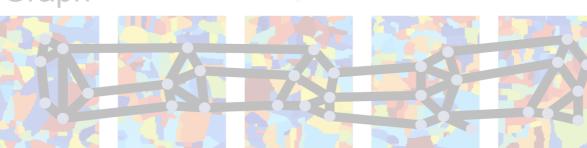
Superpixels







Graph



Segmentation







Appearance and motion cues

Spatio-temporal neighborhood

Spectral clustering methods

1. Superpixels per frame

2. Graph construction

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[Fragkiadaki & Shi CVPR'12], [Ochs et al. PAMI'14], [Galasso et al. CVPR'14], [Yi & Pavlovic ICCV'15]

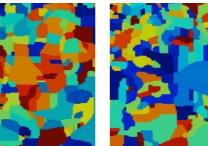
Video

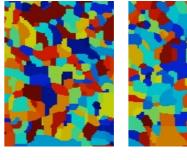


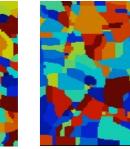


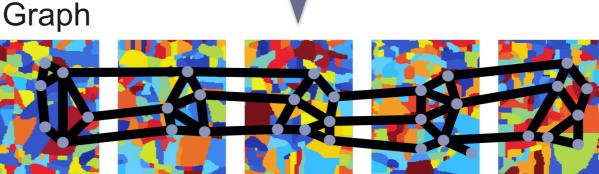


Superpixels









Segmentation



[Fragkiadaki & Shi CVPR'12], [Ochs et al. PAMI'14], [Galasso et al. CVPR'14], [Yi & Pavlovic ICCV'15]

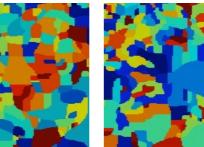
Video

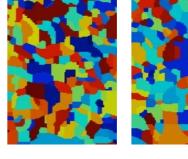


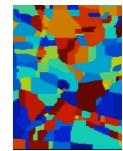




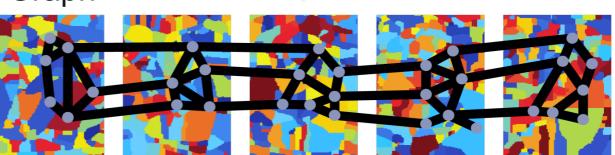
Superpixels







Graph





- 1. Superpixels per frame
 - Appearance and motion cues

- 2. Graph construction
 - Spatio-temporal neighborhood

- 3. Graph partitioning
 - Spectral clustering methods



Appearance and motion cues

[Fragkiadaki & Shi CVPR'12], [Ochs et al. PAMI'14], [Galasso et al. CVPR'14], [Yi & Pavlovic ICCV'15]

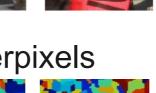
Video

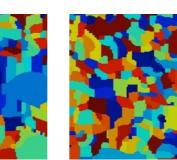


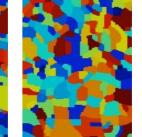


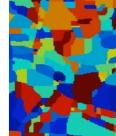


Superpixels







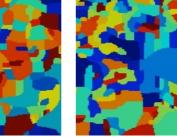


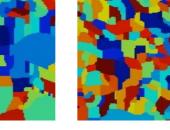
Graph



Segmentation













































































3. Graph partitioning

2. Graph construction

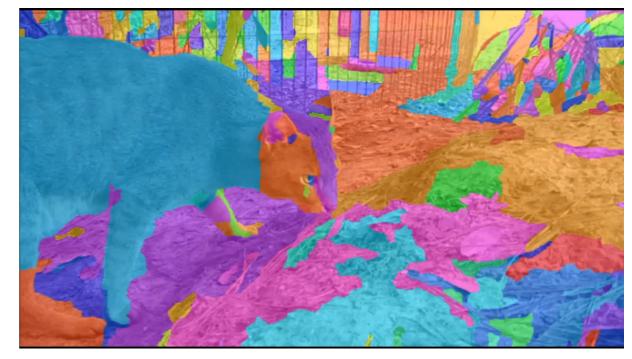
1. Superpixels per frame

Spectral clustering methods

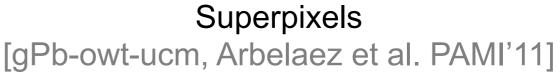


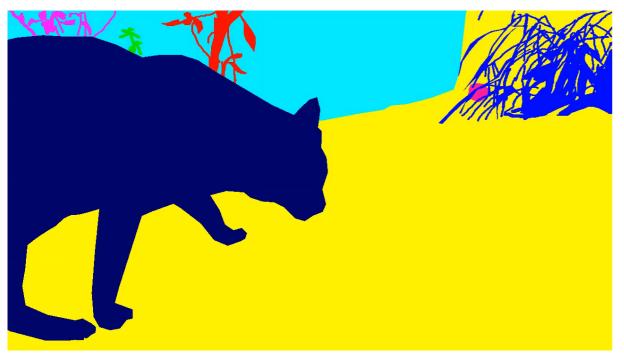
Motivation





Video





- Volatile and flickering superpixels •
- Label leakage •

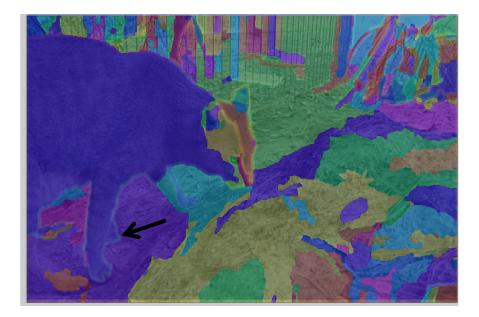
Ground truth



Motivation

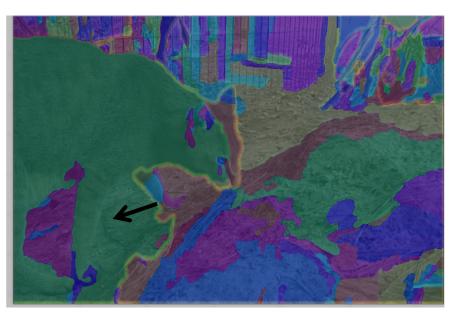


Video: frame t





Video: frame *t*+1



Superpixels



Motivation

Properties of good superpixels:

- high boundary recall
- good temporal consistency
- as few as possible



Video



Our superpixels



Better Superpixels for Video Segmentation



Video



Superpixels





Superpixel/voxel Methods

- 1. Classical superpixel/voxel methods
 - homogeneous shape and size
 - regular topology



Video



Sticky superpixels [Dollar et al. PAMI'15]



TSP [Chang et al. CVPR'13]



SLIC 2D [Achanta et al. PAMI'12]



SLIC 3D [Achanta et al. PAMI'12]



Video SEEDS [Bergh et al. ICCV'13]



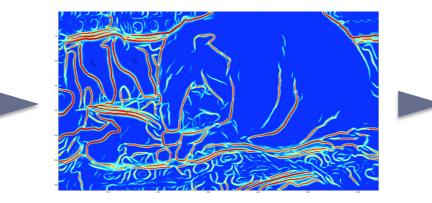
Superpixel/voxel Methods

- 2. Boundary based superpixel/voxel methods
 - heterogeneous shape and size
 - semantic regions

gPb-owt-ucm [Arbelaez et al. PAMI'11]



Video



Boundaries

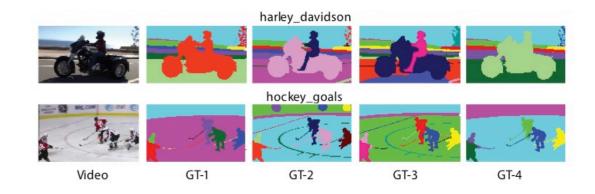


Hierarchical image segmentation

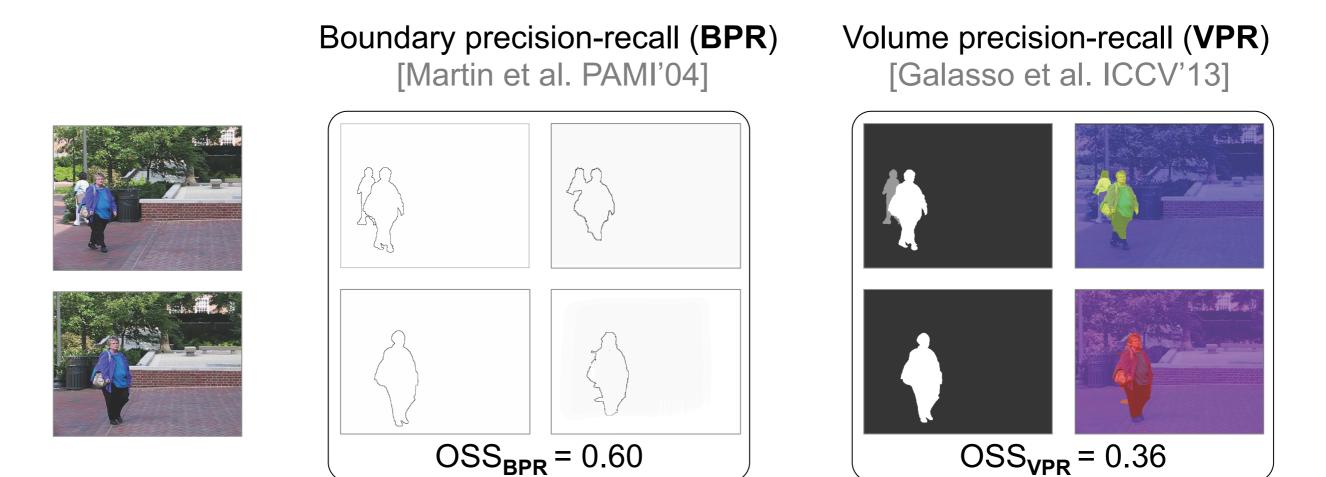


VSB100 [Galasso et al. ICCV'13]

- 100 HD-quality video sequences
- 4 sets of human annotations
- Training, validation and test sets [24 + 16 + 60]



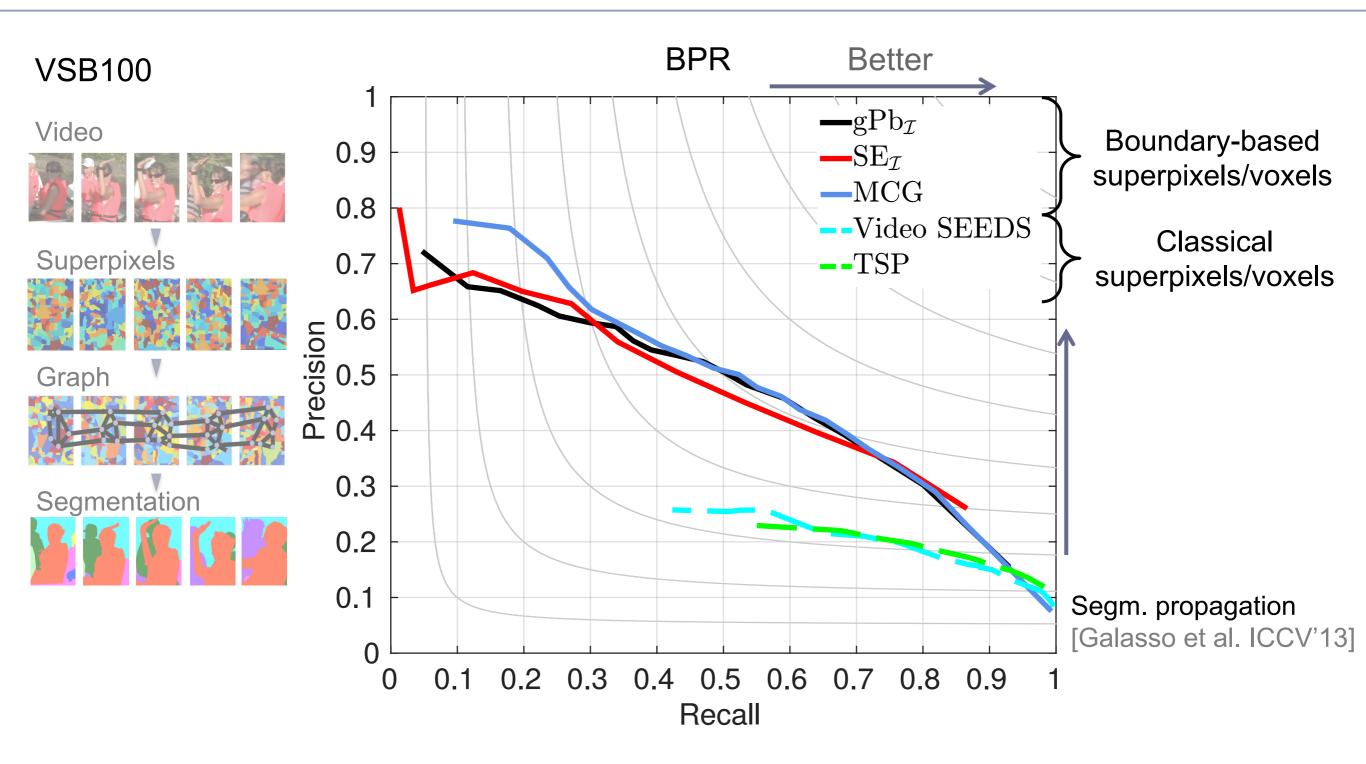
Benchmark metrics:



OSS: F-measure at optimal segmentation scale



Superpixel/voxel Methods for Video Segmentation



Superpixels built from boundaries are more effective for graph-based video segmentation.



Better Boundaries for Superpixels

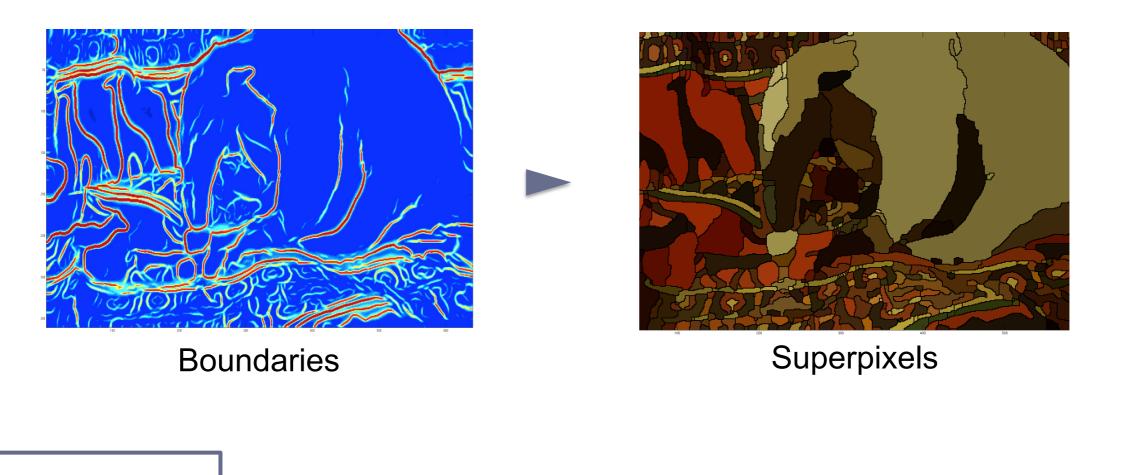






Image domain cues

- Image boundaries
- Higher-level cues via object proposals

Time domain cues

- Temporal smoothing
- Motion boundaries

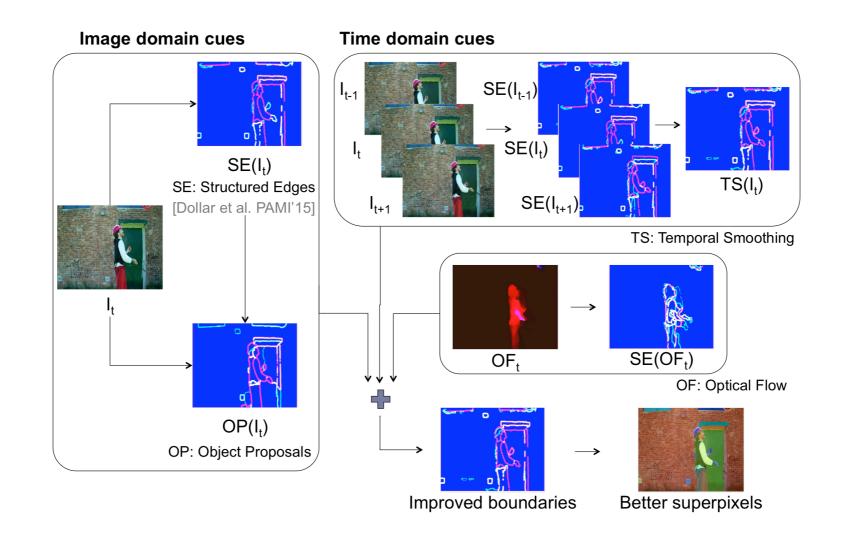




Image domain cues

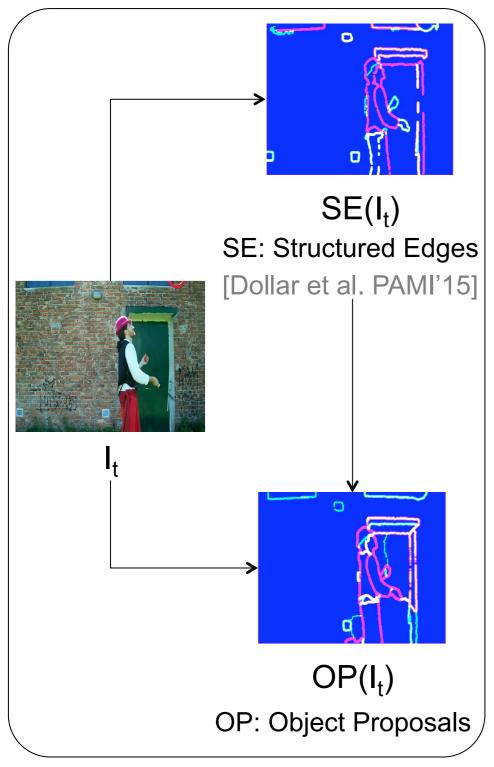
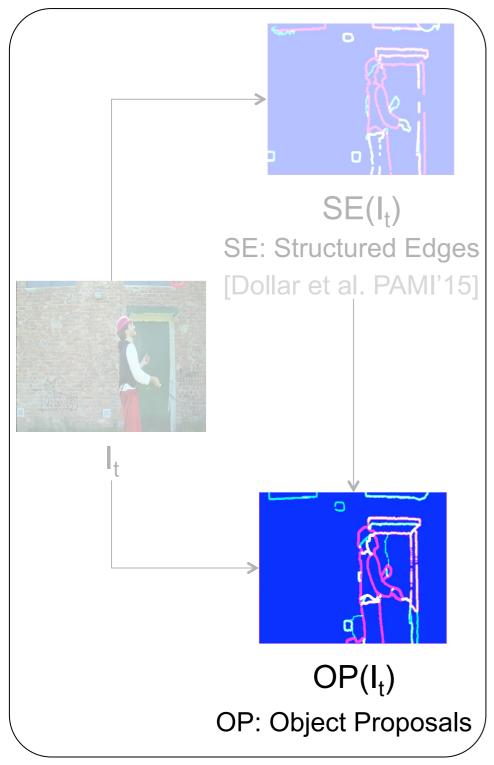




Image domain cues

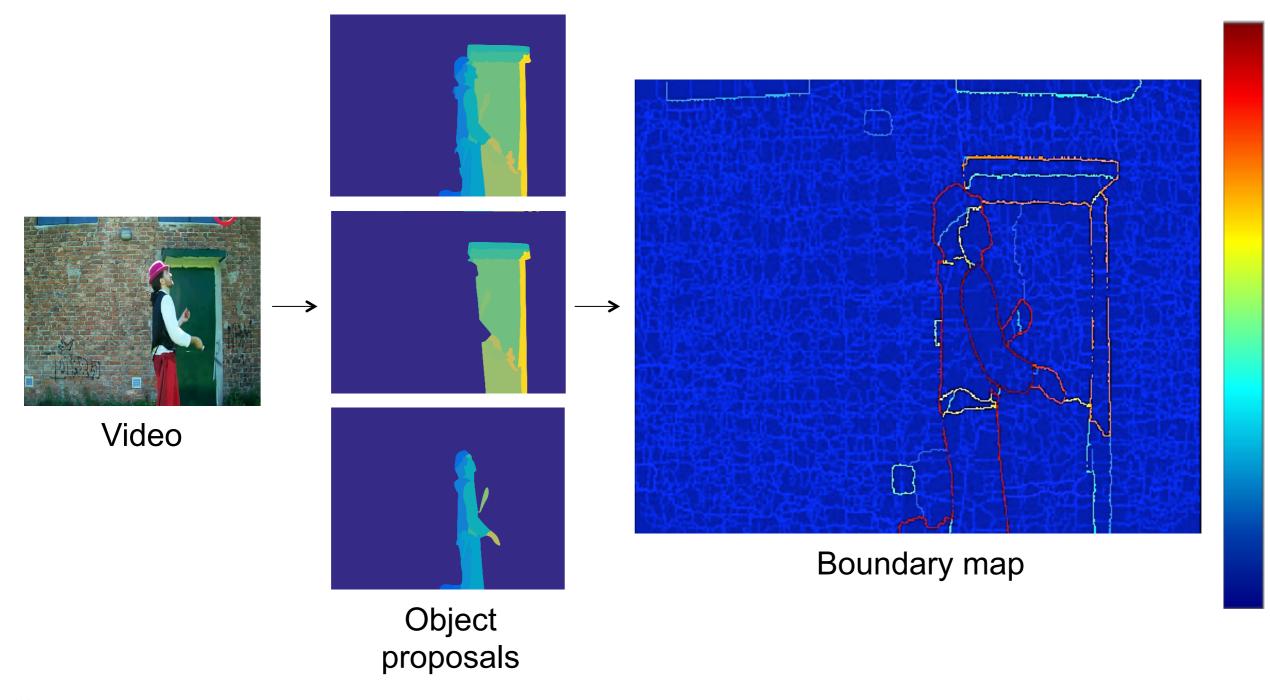




Integration of Higher-Level Object Cues

Object Proposals [Humayun et al. CVPR'14]

• average the contours of each object proposal segment





times edge

is used to separate fg/bg

Image domain cues

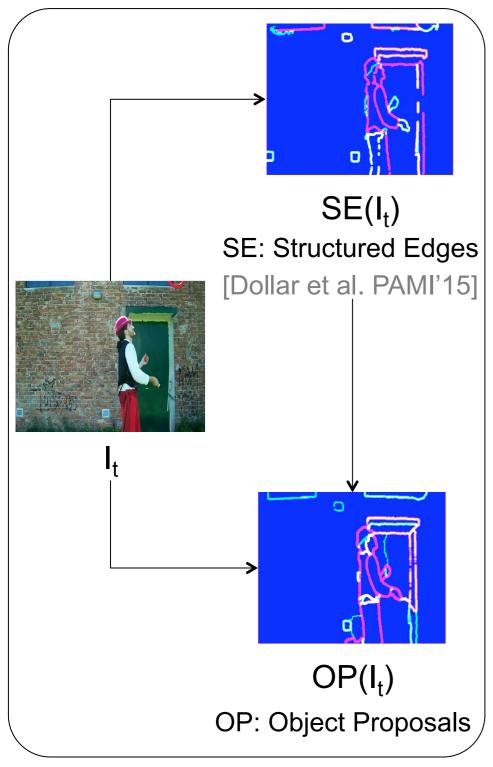
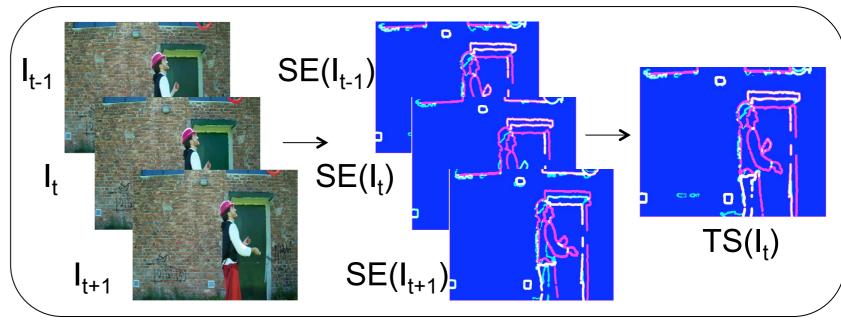


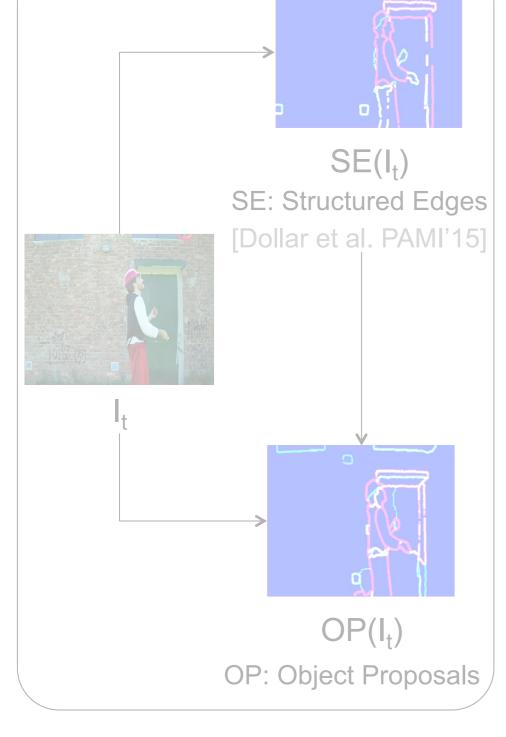


Image domain cues

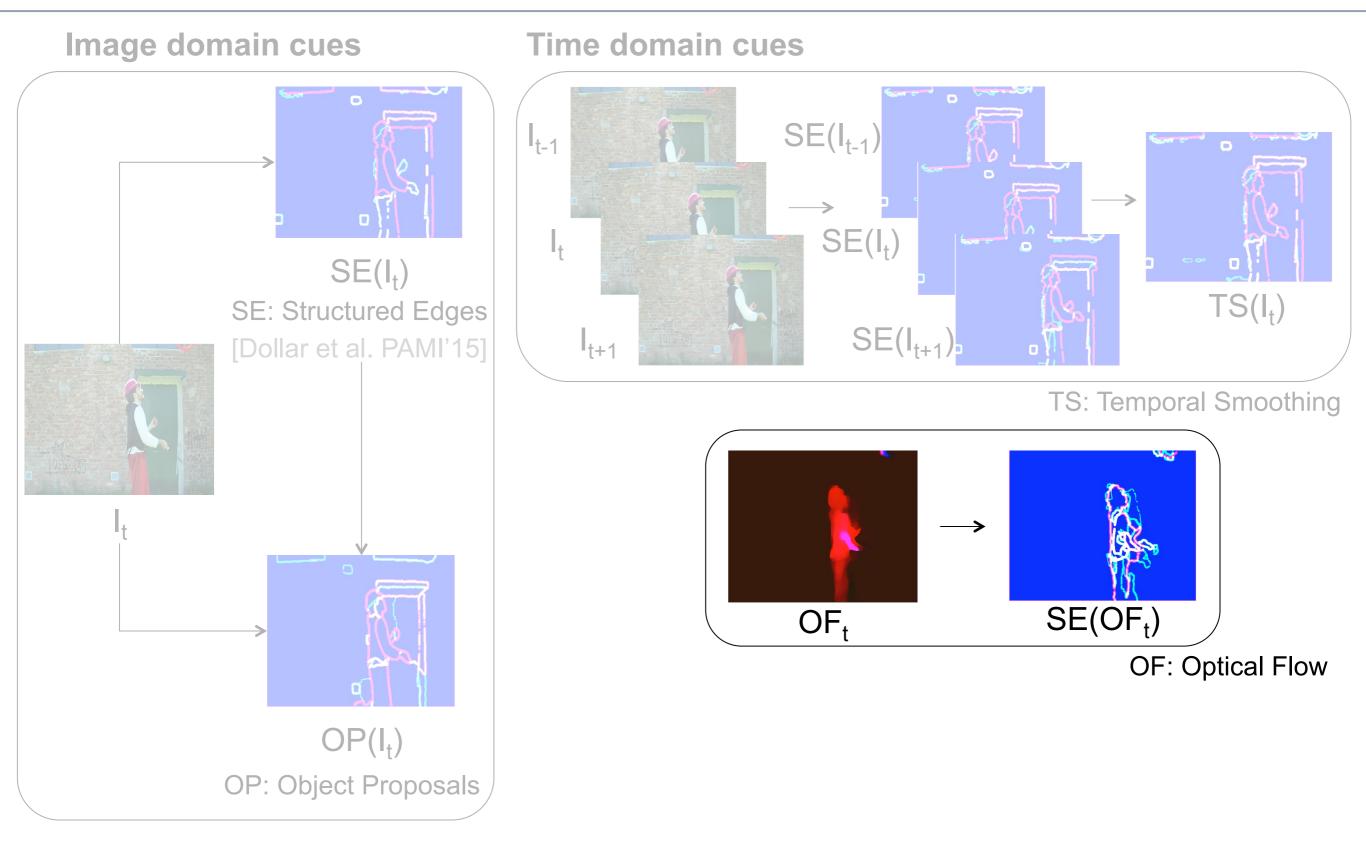
Time domain cues



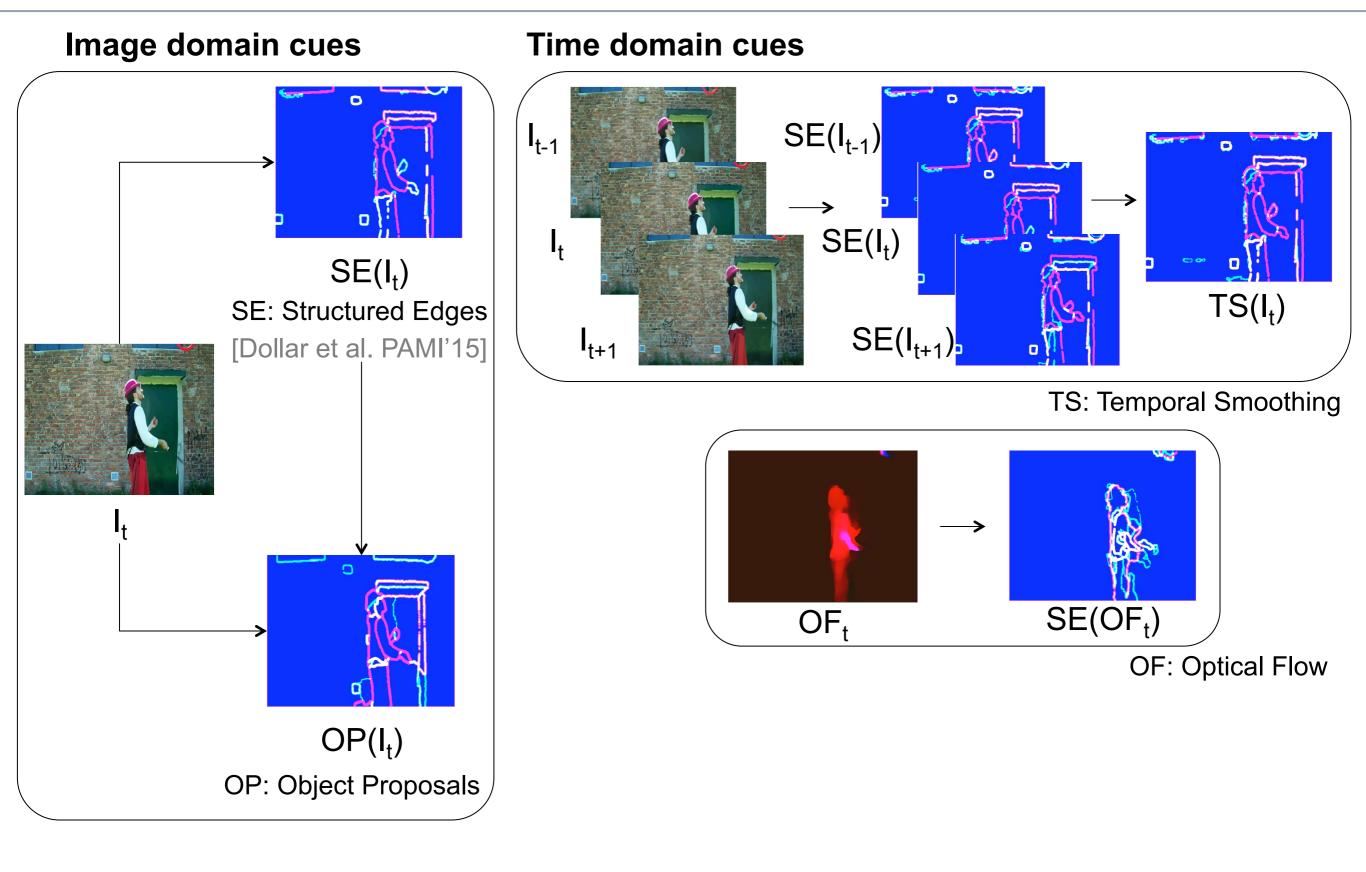
TS: Temporal Smoothing



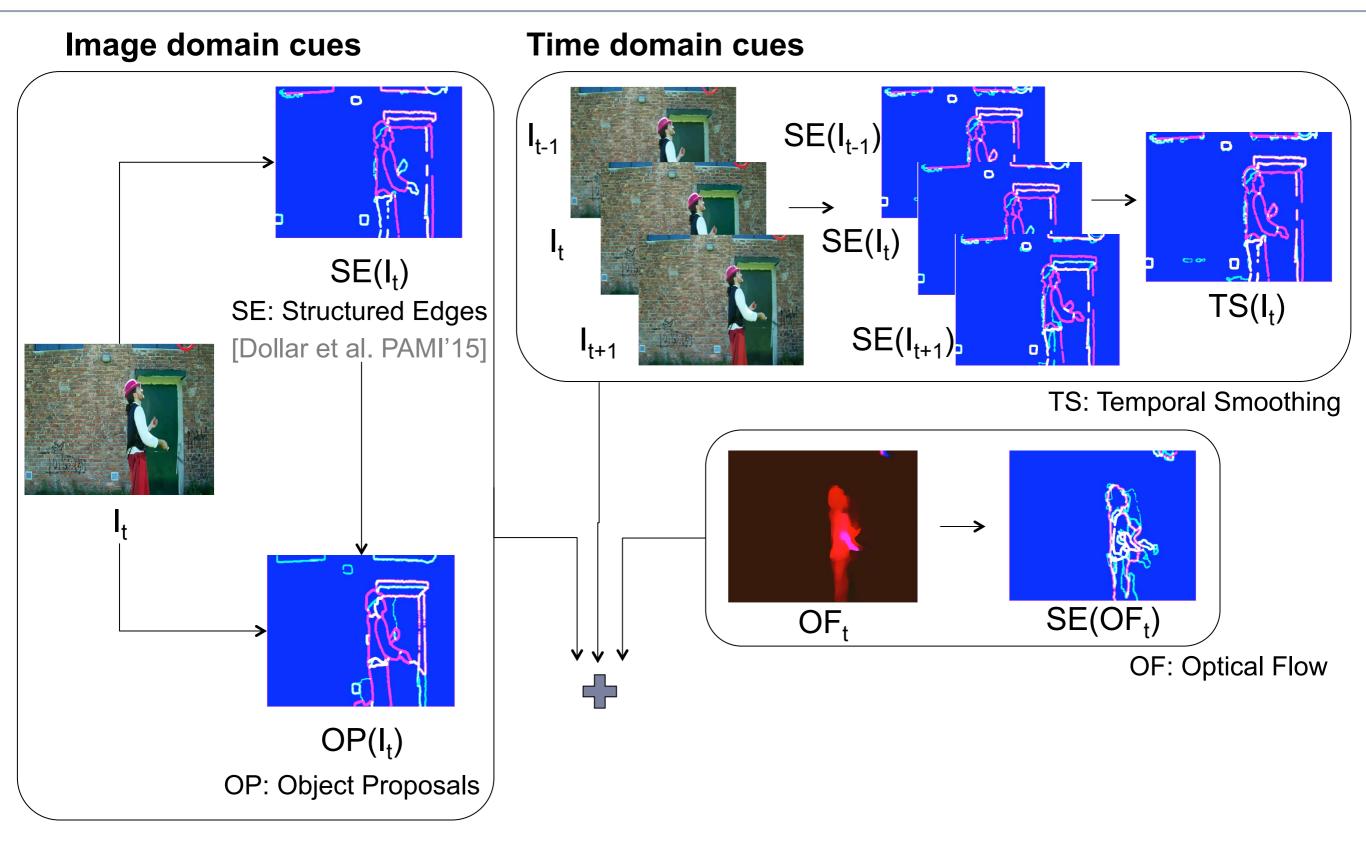




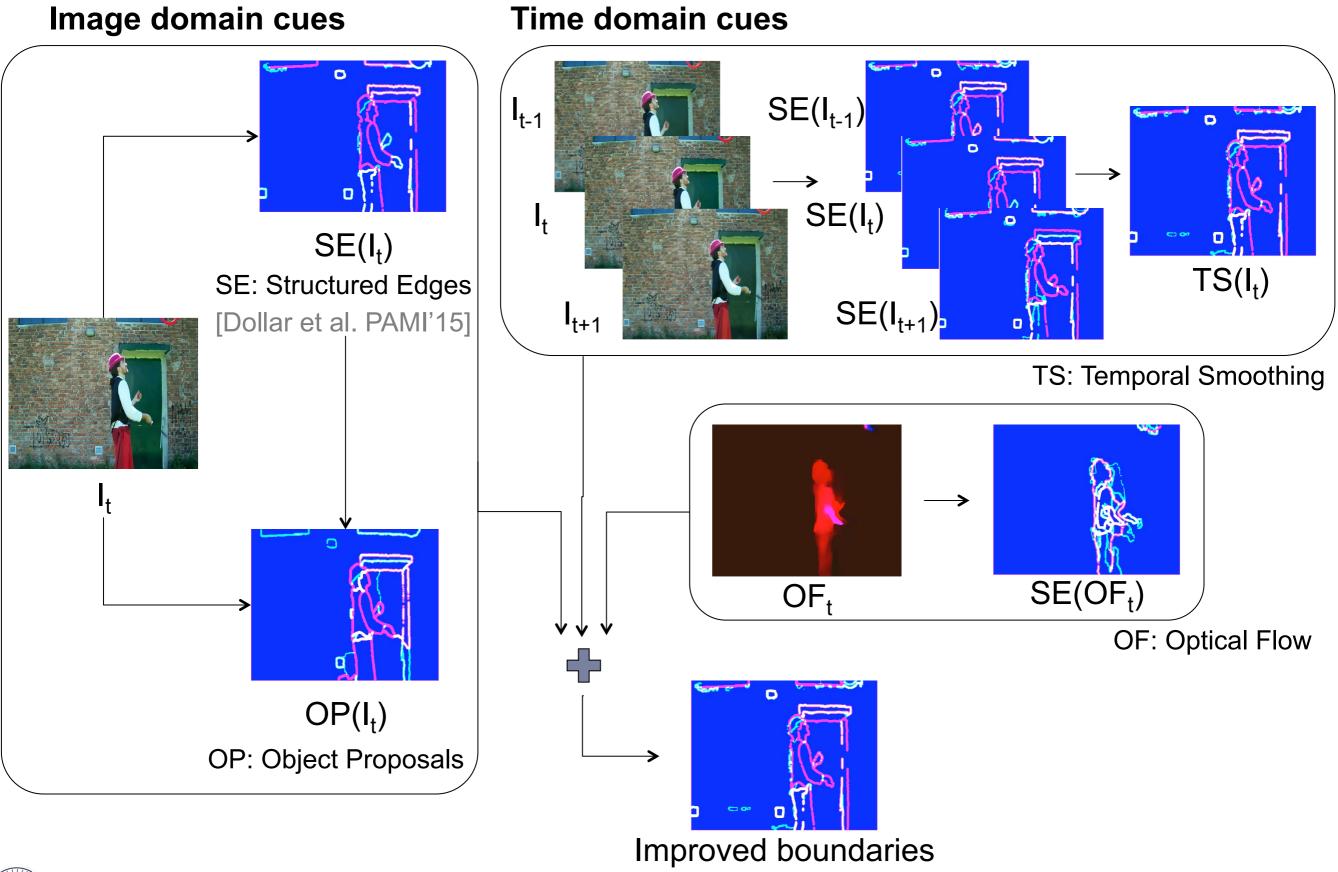




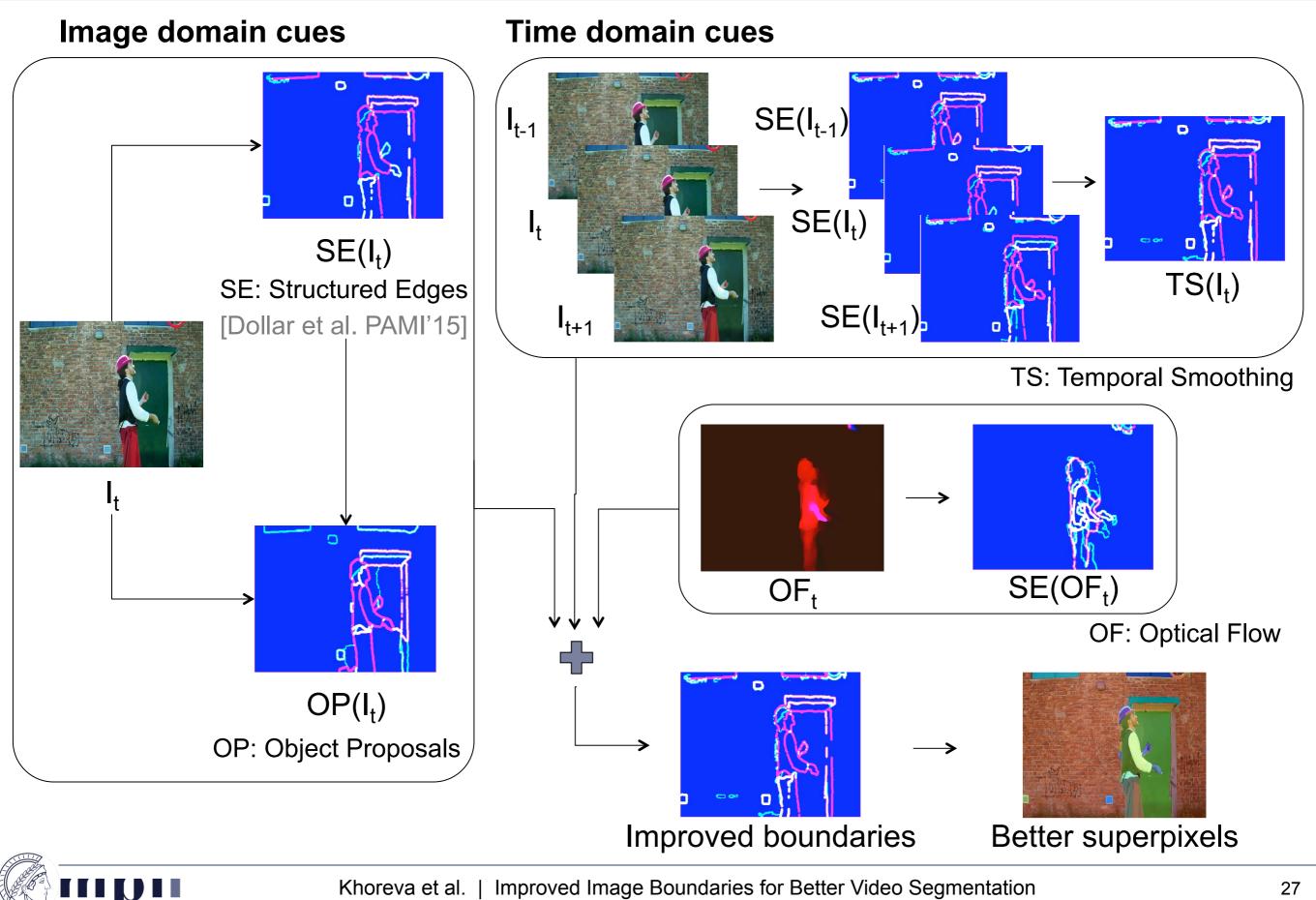




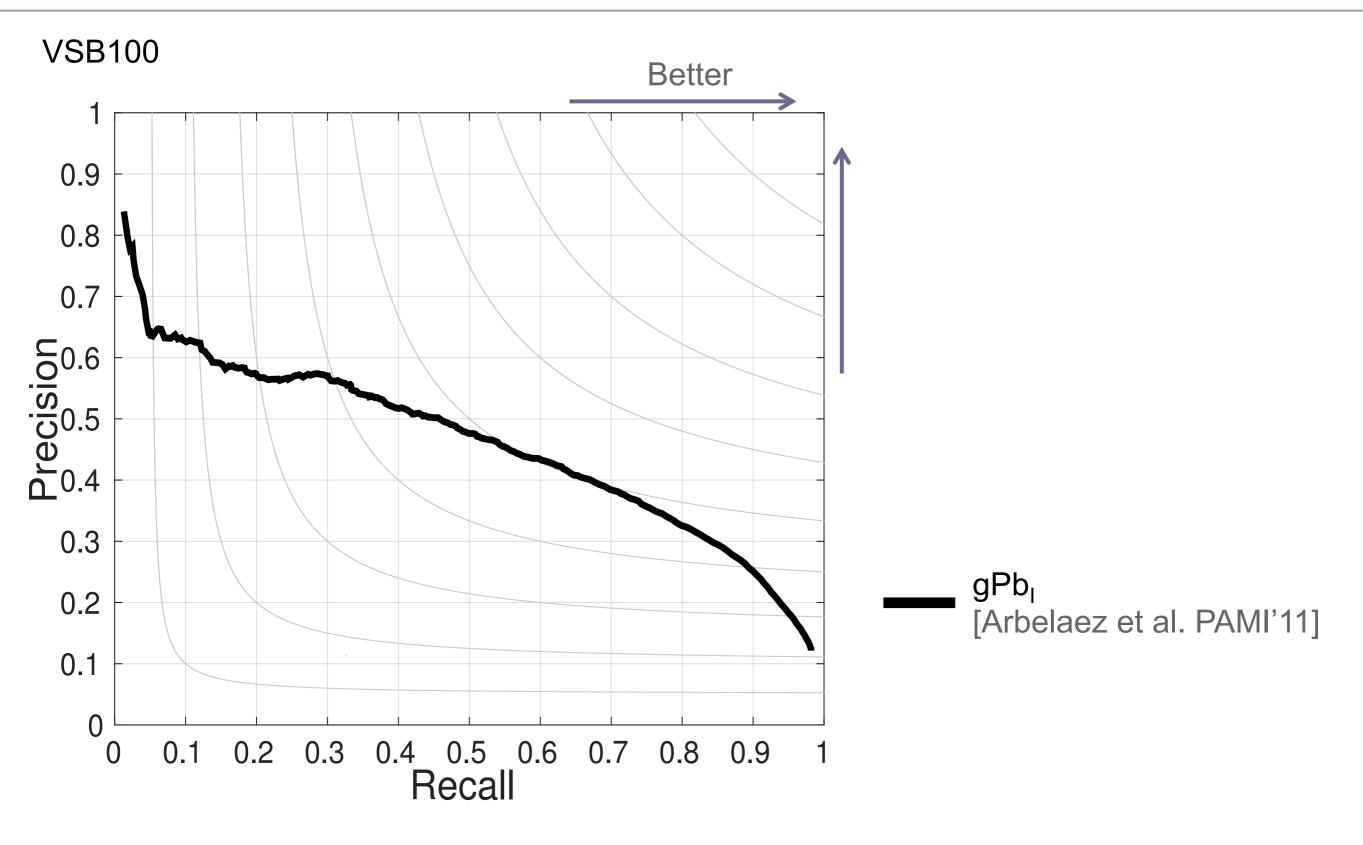




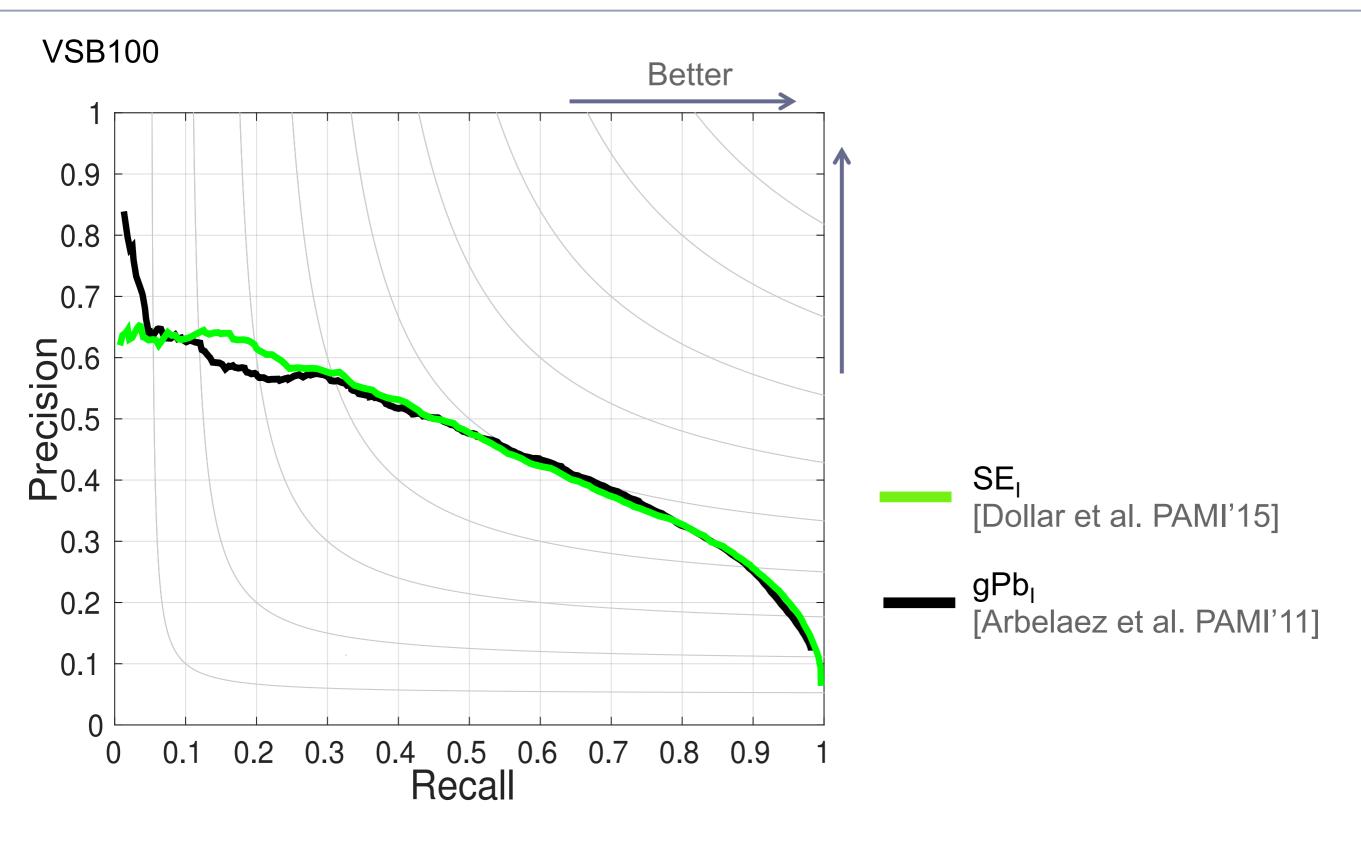




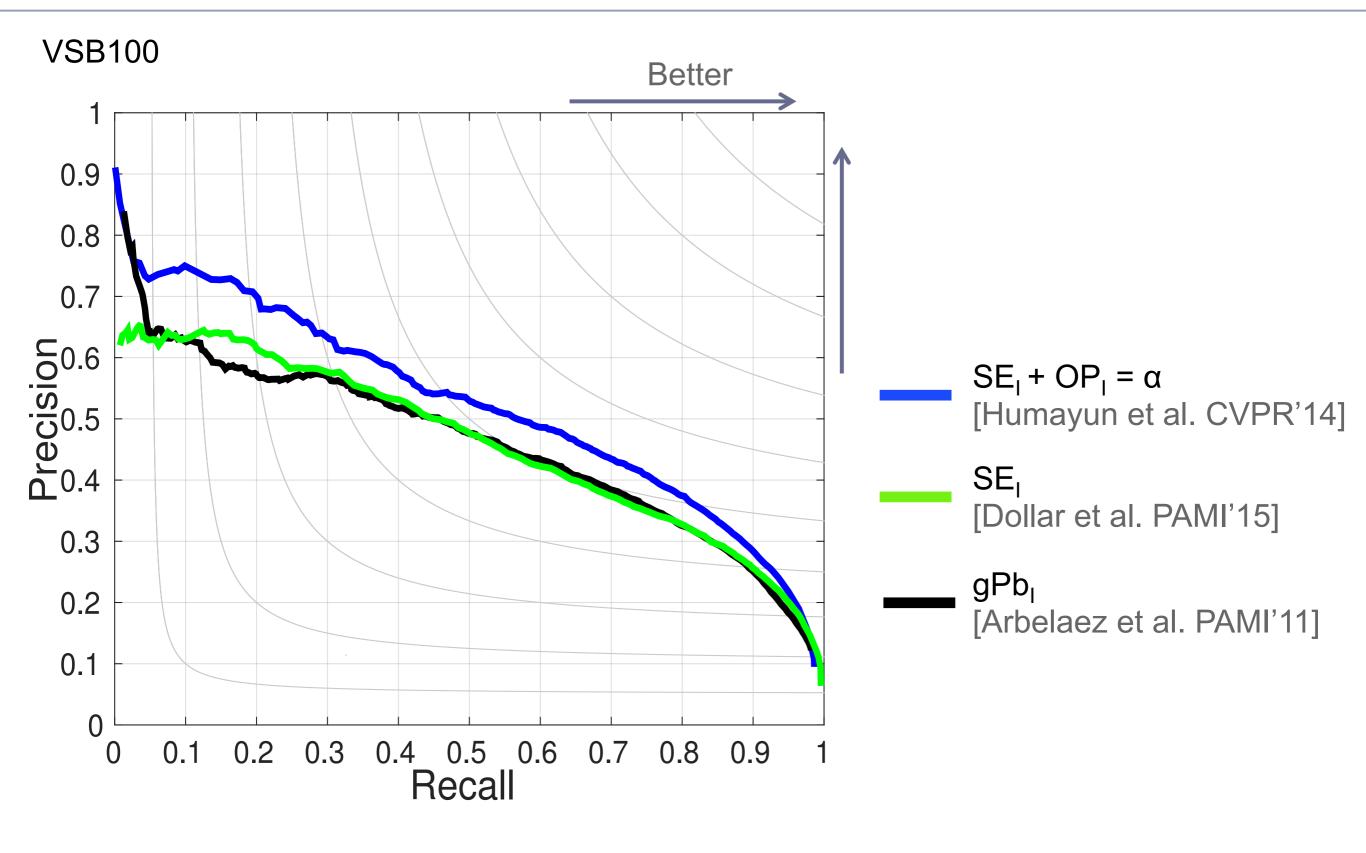
Khoreva et al. | Improved Image Boundaries for Better Video Segmentation



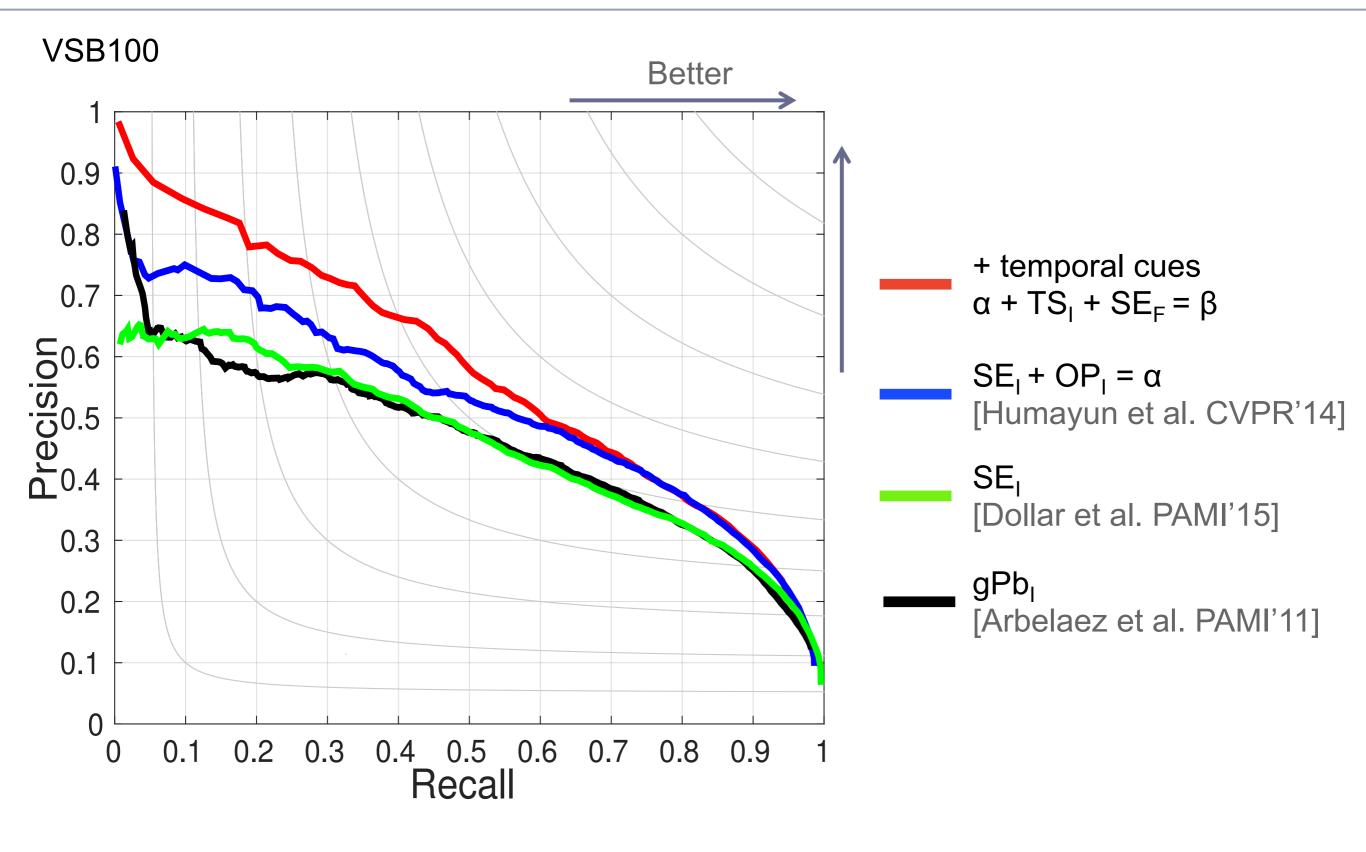




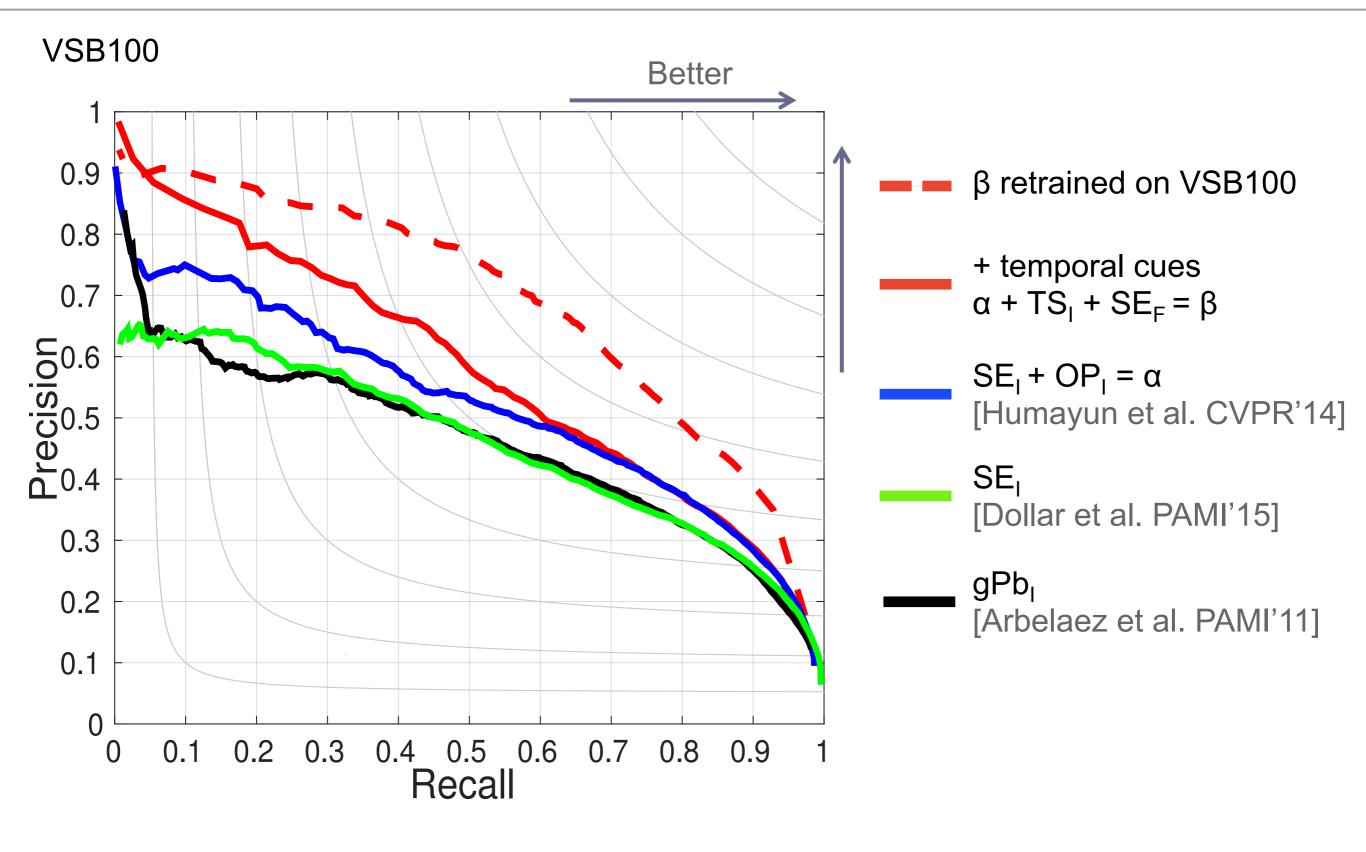






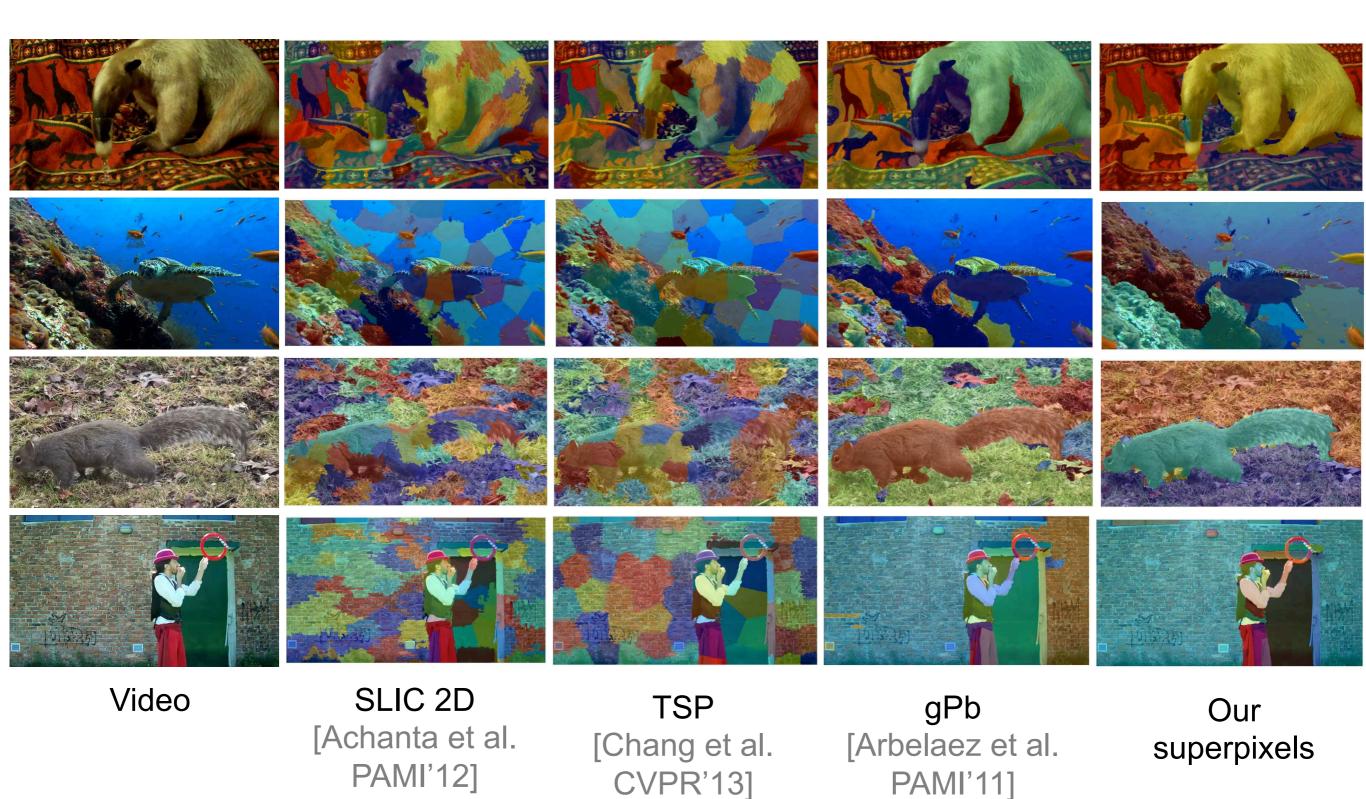






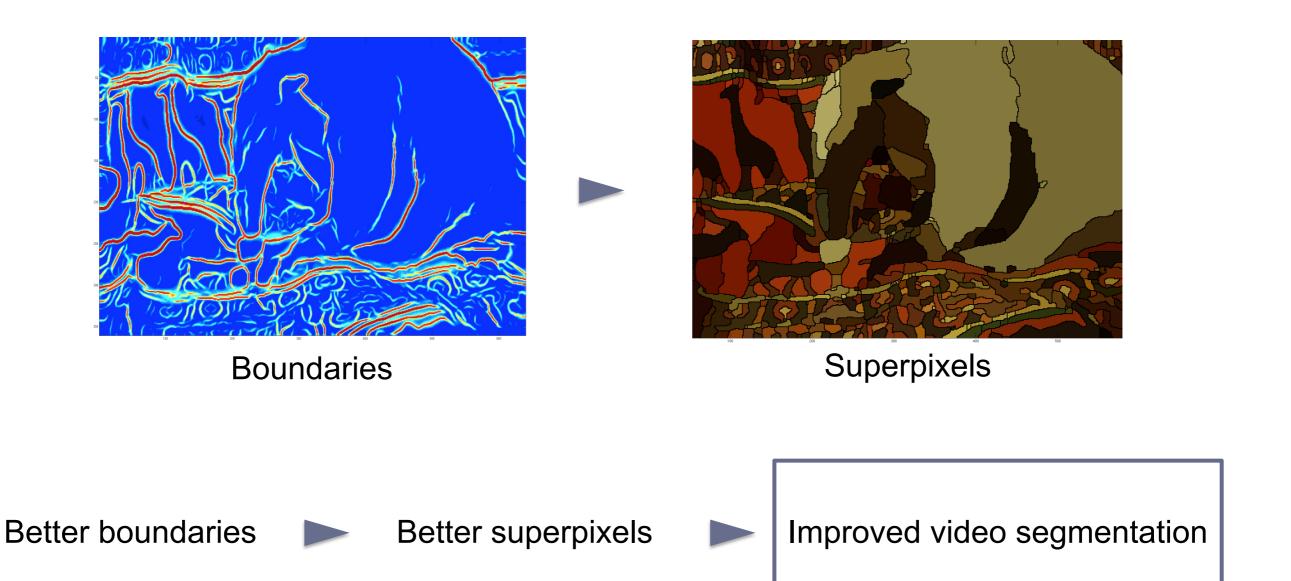


Comparison of Superpixel/voxel Methods

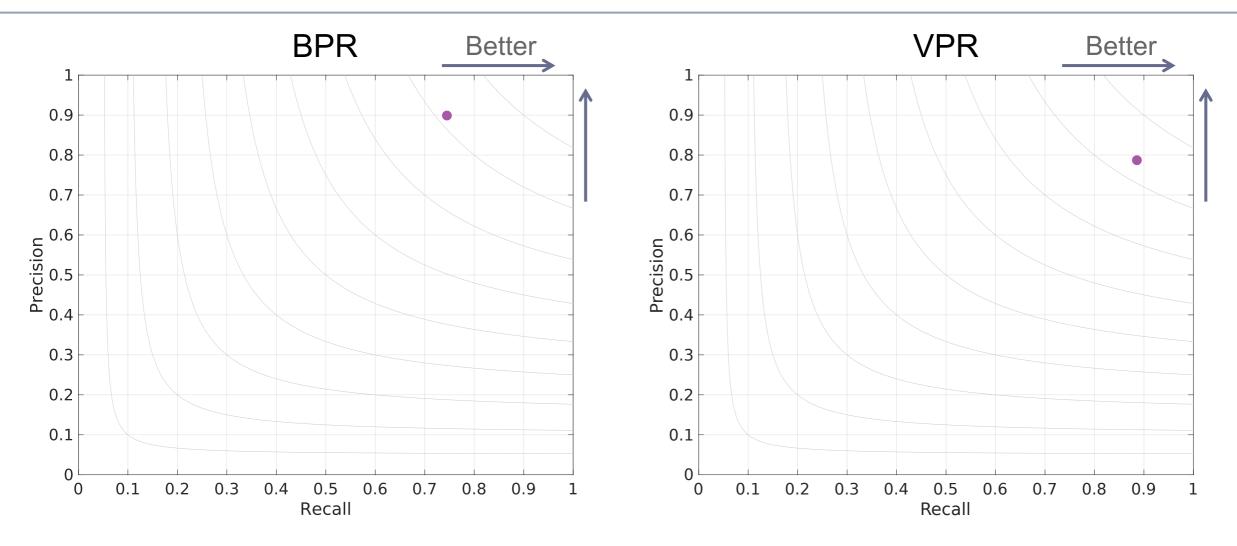




Better Boundaries for Superpixels

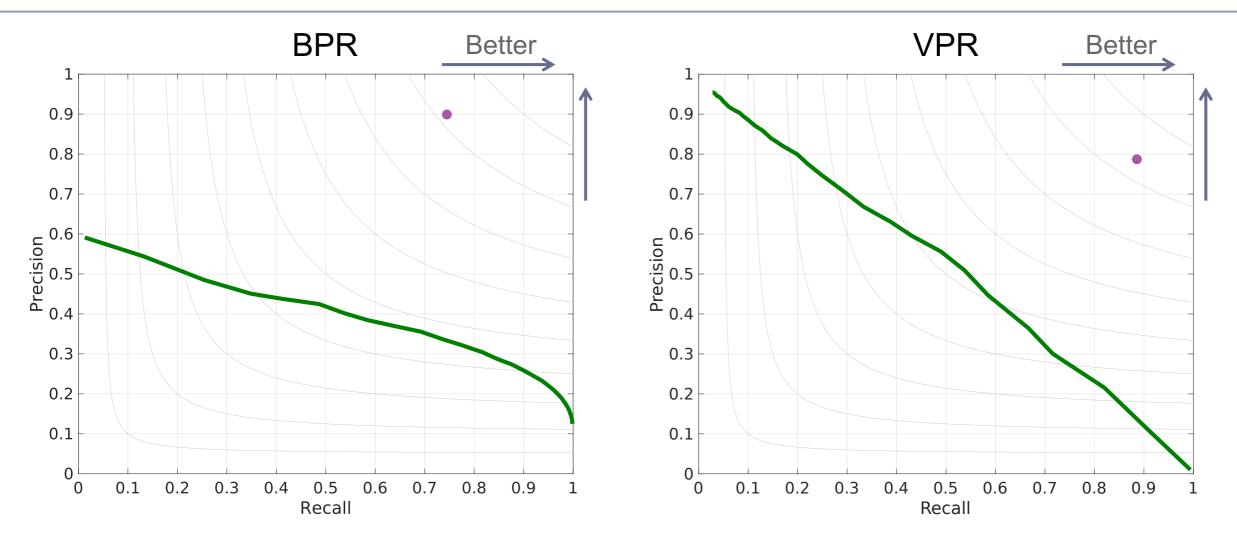






Human

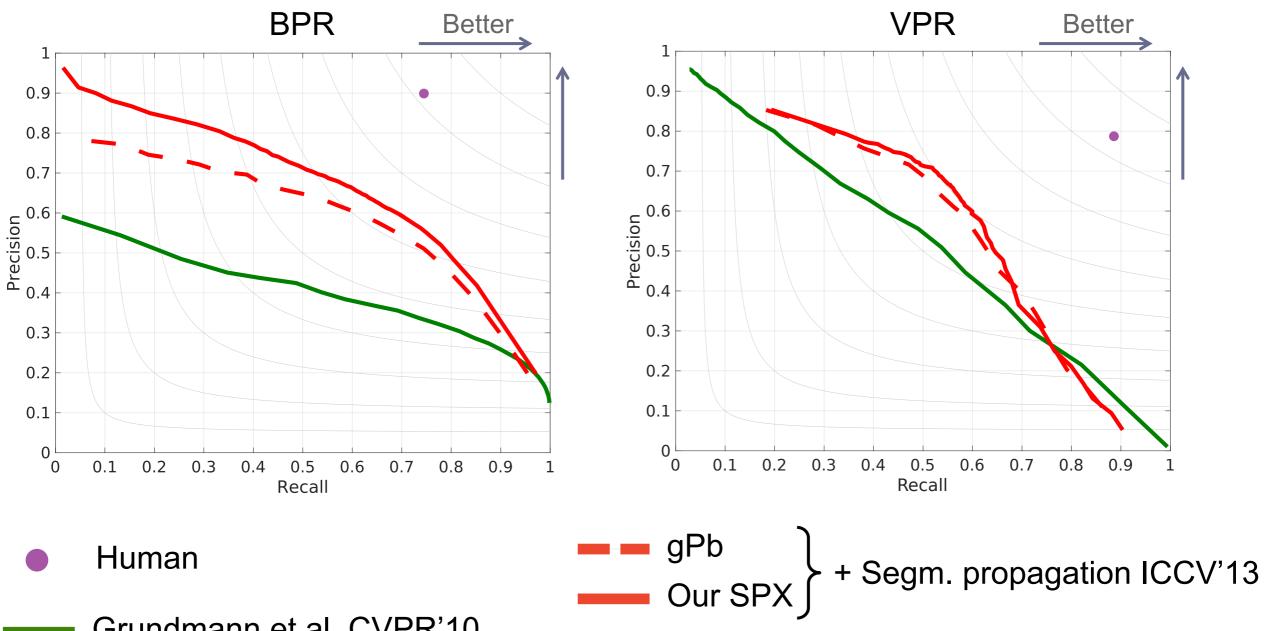




Human

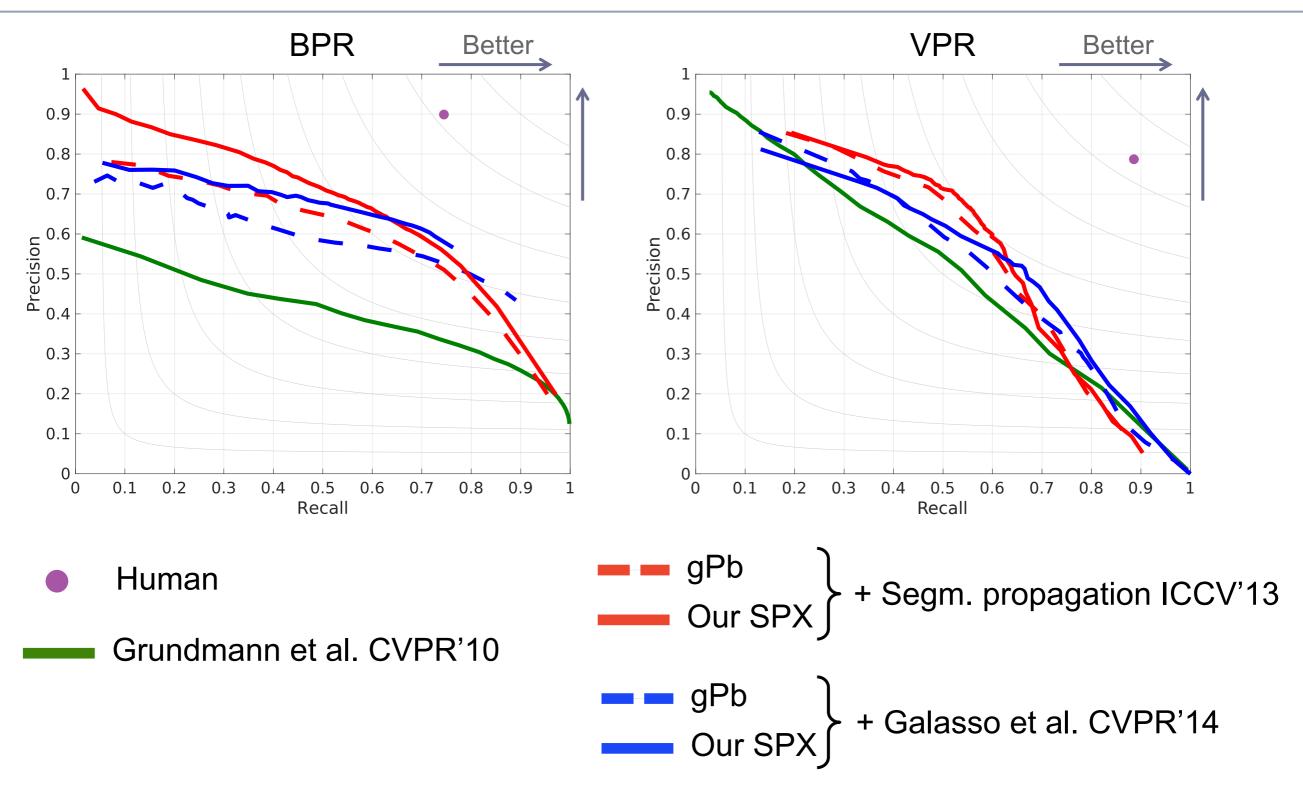
Grundmann et al. CVPR'10



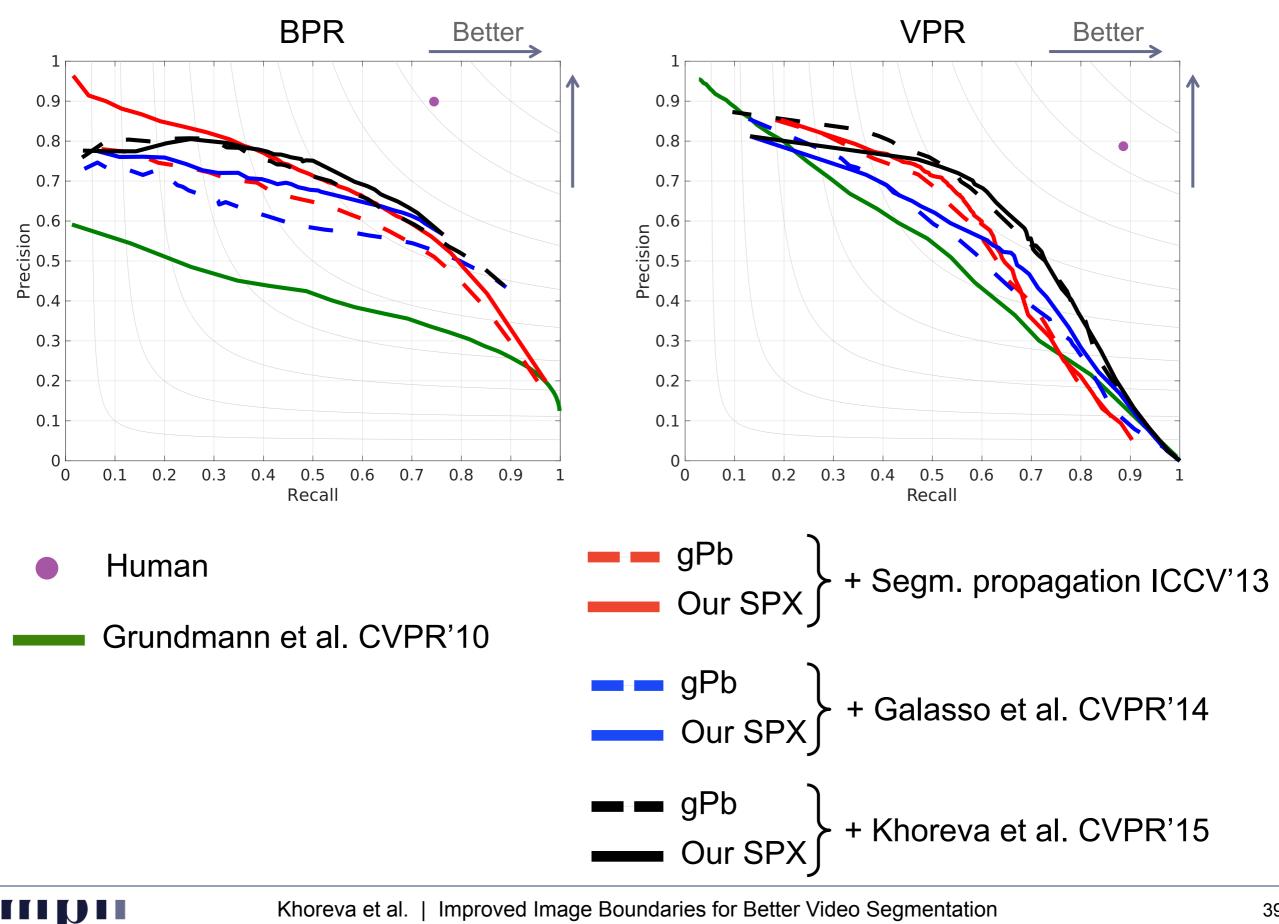




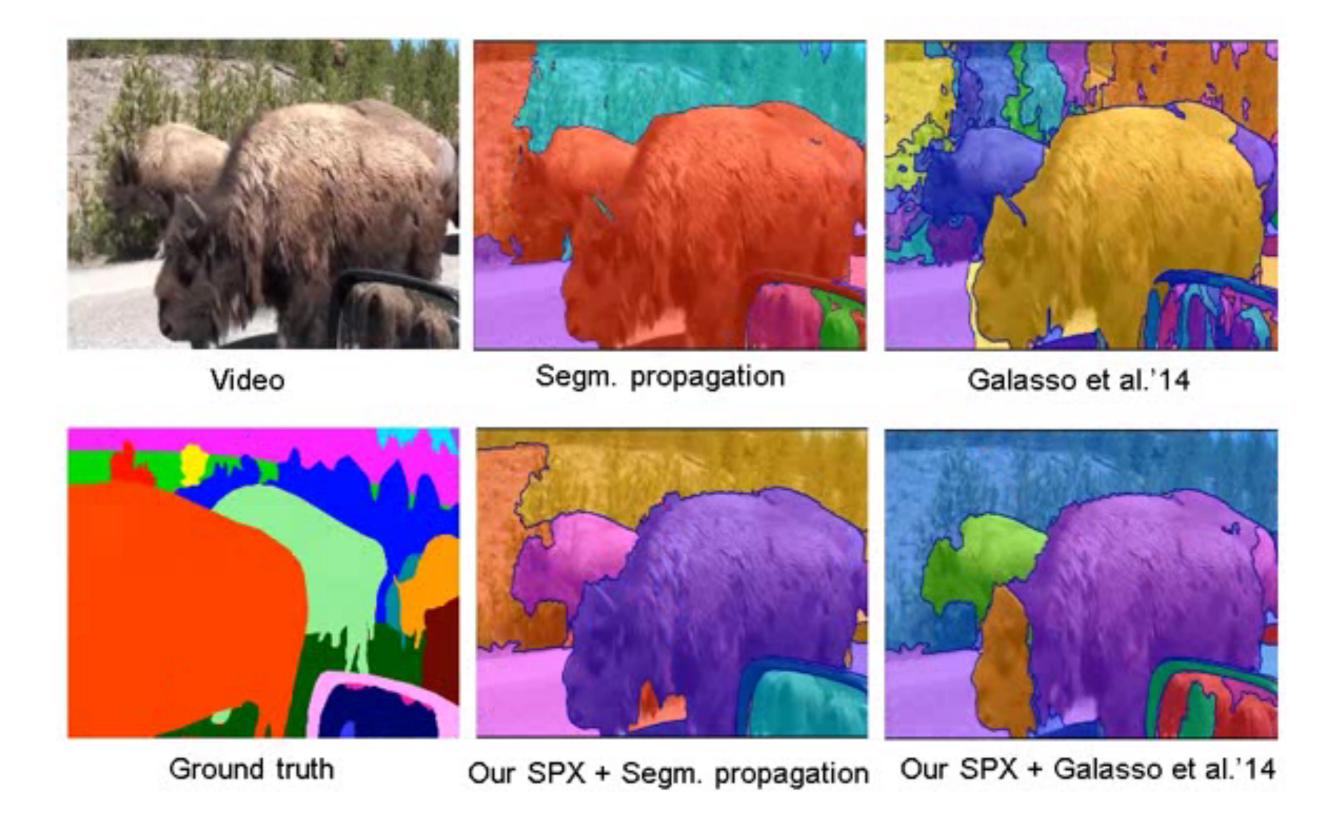






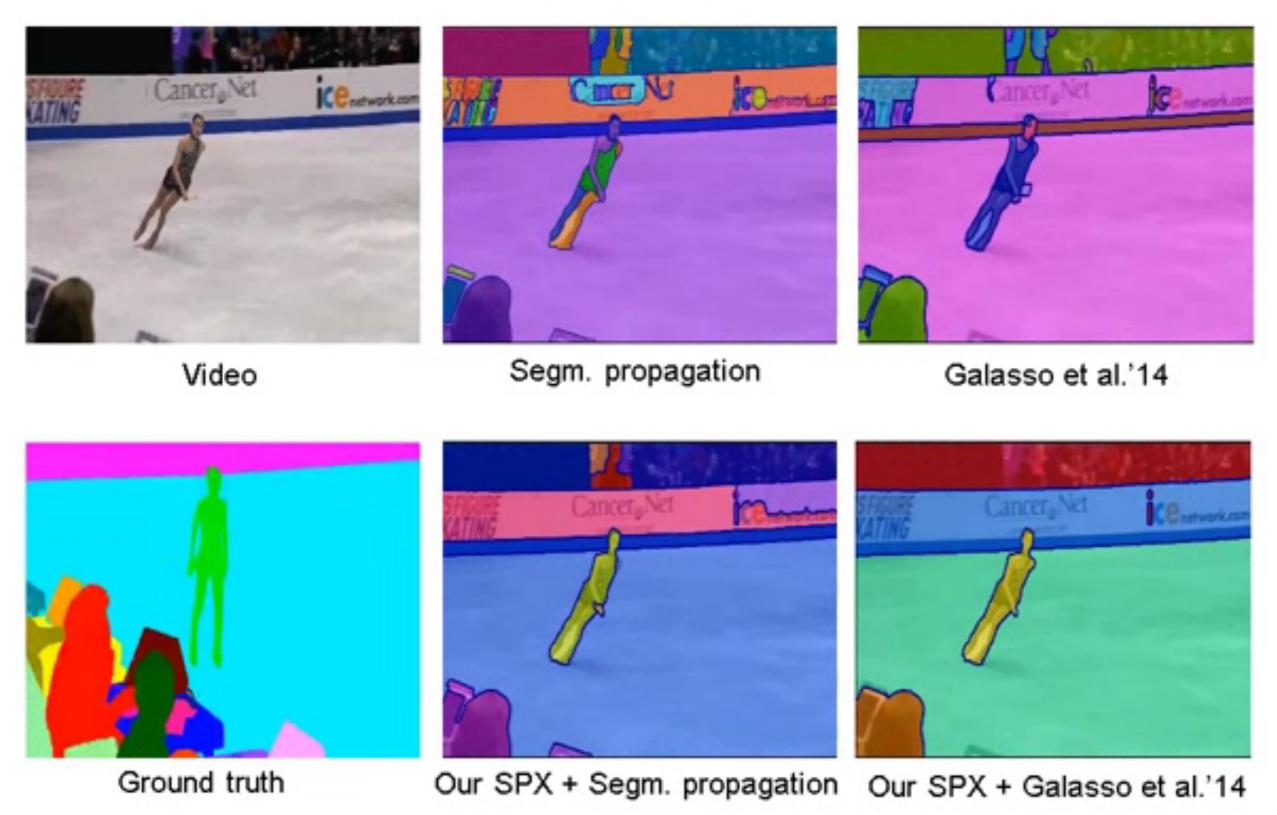


buffalos





kim yu na





Contributions

- a comparative evaluation of the superpixels/voxels methods
- improved boundary estimates (and thus superpixels) by the fusion of image and time domain cues
- integration of high-level object-related cues into the local image segmentation
- state-of-the-art video segmentation results on the VSB100 and BMDS datasets

