

Universal Weakly Supervised Segmentation by Pixel-to-Segment Contrastive Learning



Tsung-Wei Ke



Jyh-Jing Hwang



Stella X. Yu



Berkeley
UNIVERSITY OF CALIFORNIA



Semantic Segmentation: Classify Pixels into Semantic Categories

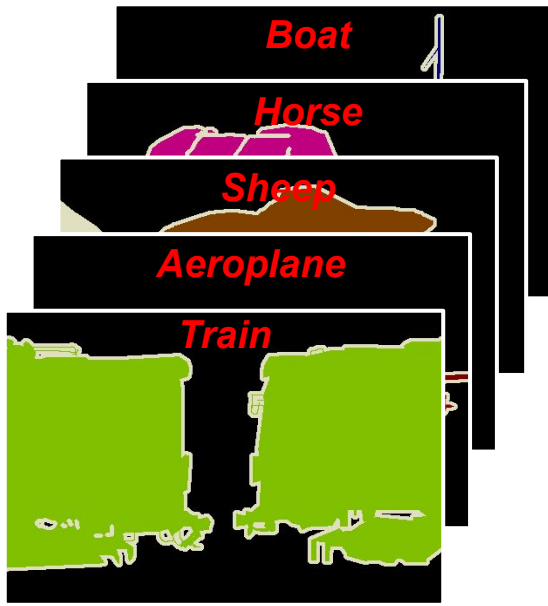


Images

Segmentation CNN

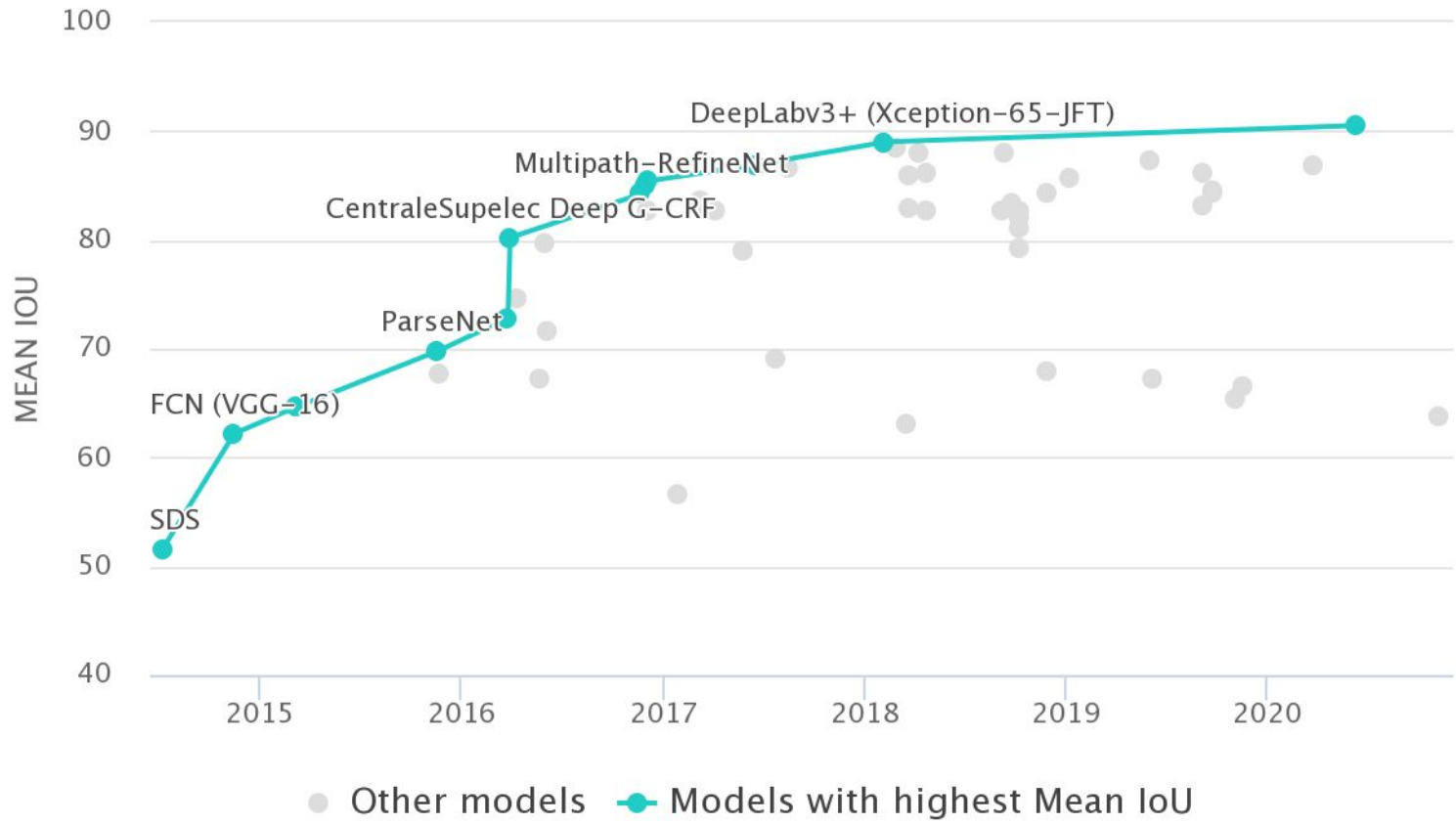


Predictions



Pixel-wise Annotations

State-of-the-art Methods Require Pixel-wise Annotations



<https://paperswithcode.com/sota/semantic-segmentation-on-pascal-voc-2012>


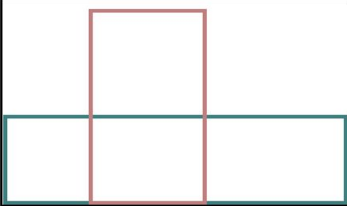

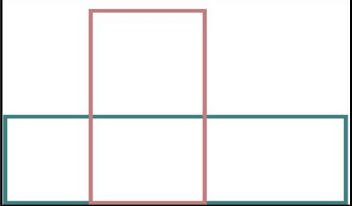
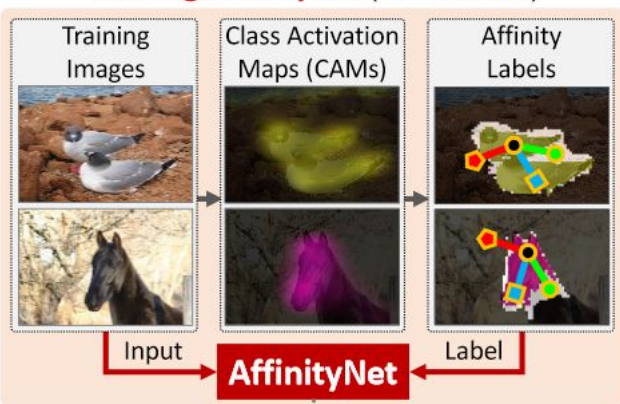
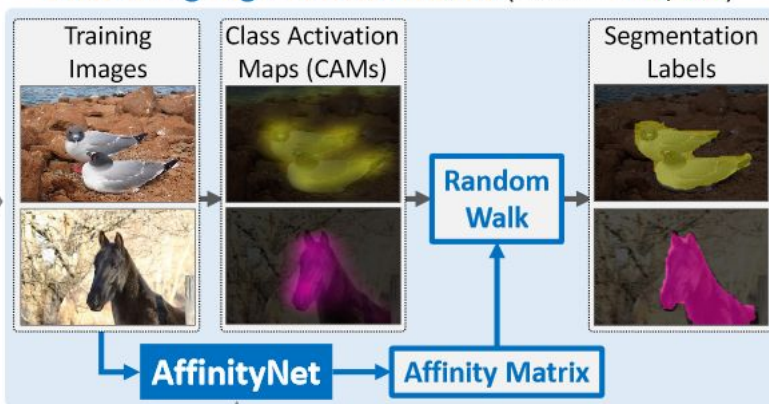
Image	Image Tags	Boxes
	<p>Person Motorbike</p>	
Supervision	Coarse	
Current Methods	Class Activation Map	

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	<p>Person</p> <p>Motorbike</p>	
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Training AffinityNet (Section 3.2)



Generating Segmentation Labels (Section 3.3, 3.4)



Learning Segmentation Net

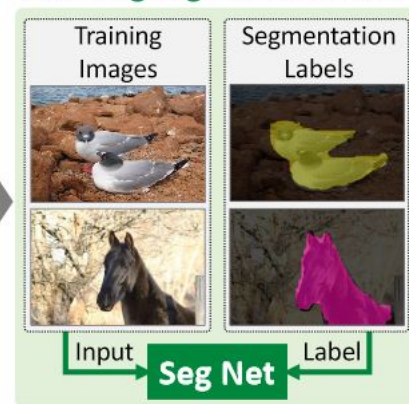



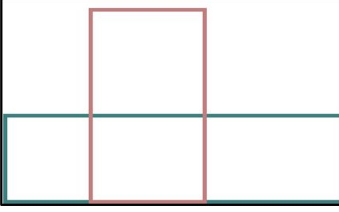
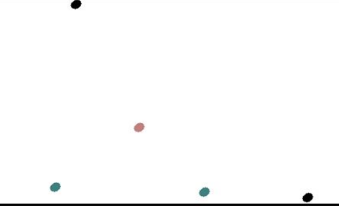


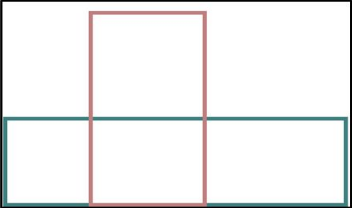
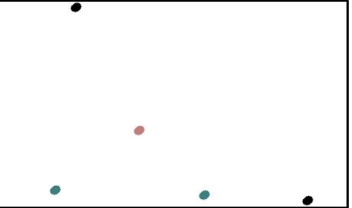
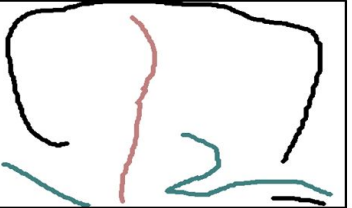
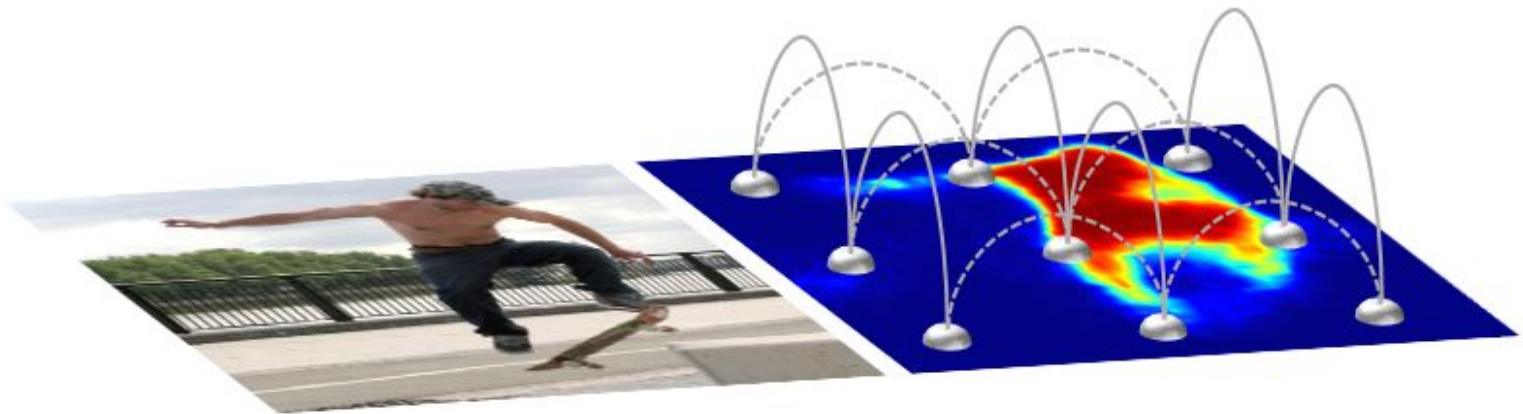

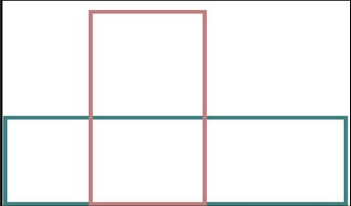
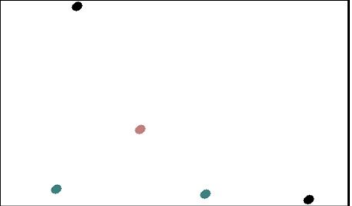
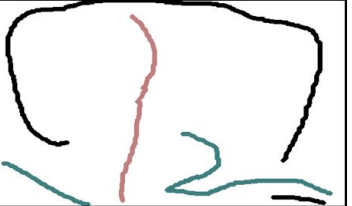
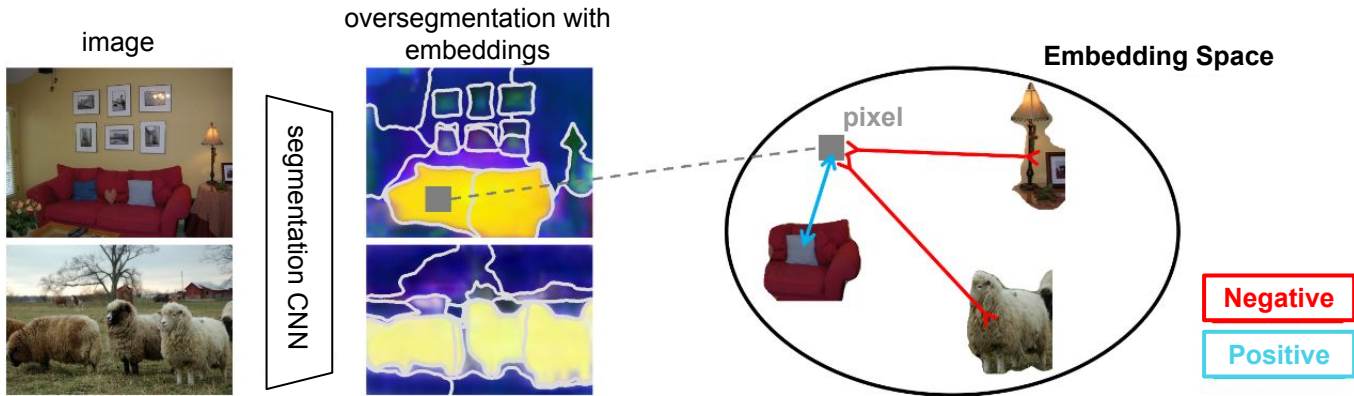
Image	Image Tags	Boxes	Points	Scribbles
	<p>Person Motorbike</p>			
<p>Supervision</p>	<p>Coarse</p>		<p>Sparse</p>	
<p>Current Methods</p>	<p>Class Activation Map</p>		<p>Conditional Random Fields</p>	

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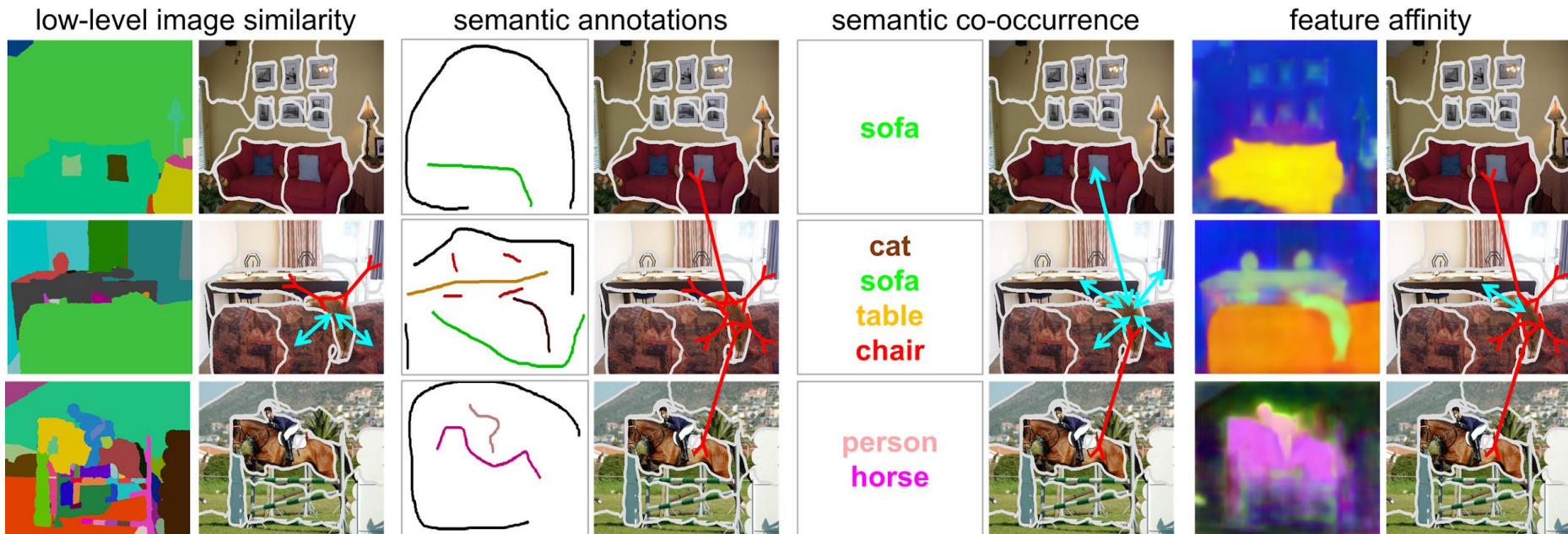


Normalized Cut Loss for Weakly-supervised CNN Segmentation. Tang et al. CVPR 2018.

Image	Image Tags	Boxes	Points	Scribbles
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Our Method	single pixel-to-segment contrastive learning loss formulation			



Our SPML: Contrasts Pixels with Segments on 4 Types of Relationships



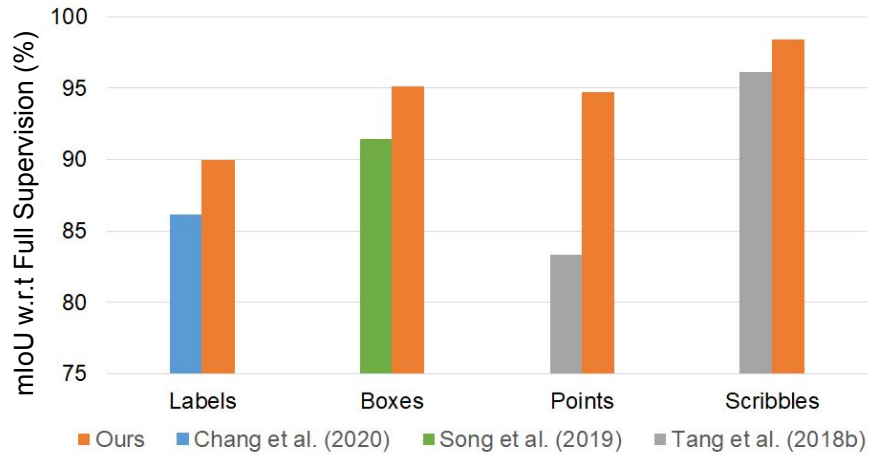
$$L(i) = \lambda_I L_{\text{SegSort}^+}(i, \mathcal{V}^+, \mathcal{V}^-) + \lambda_C L_{\text{SegSort}^+}(i, \mathcal{C}^+, \mathcal{C}^-) + \lambda_O L_{\text{SegSort}^+}(i, \mathcal{O}^+, \mathcal{O}^-) + \lambda_A L_{\text{SegSort}^+}(i, \hat{\mathcal{C}}^+, \hat{\mathcal{C}}^-)$$

Contrastive loss for pixel i with positive segments \mathcal{C}^+ , negative segments \mathcal{C}^- :

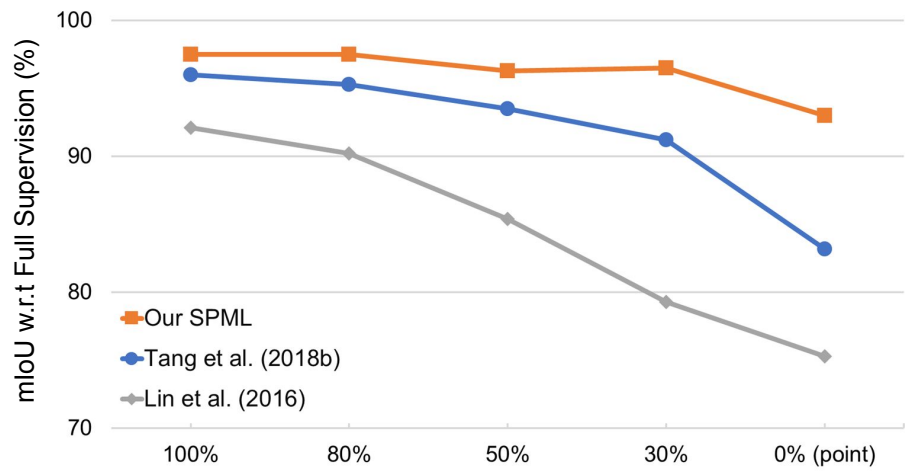
$$L_{\text{SegSort}^+}^i = -\log \frac{\exp(\kappa \mathbf{p}_s^\top \mathbf{e}_i)}{\sum_{l \in \Omega} \exp(\kappa \mathbf{p}_l^\top \mathbf{e}_i)}$$

Beats All Weak Supervision SOTA's on Pascal VOC & DensePose

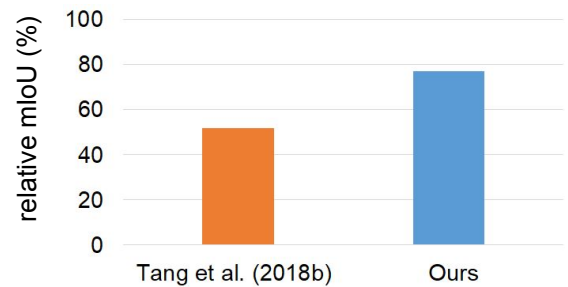
VOC 2012



VOC 2012 on Varying Scribble Length



DensePose



Pascal: Varying sparsity of scribbles and point annotation



(a) Image

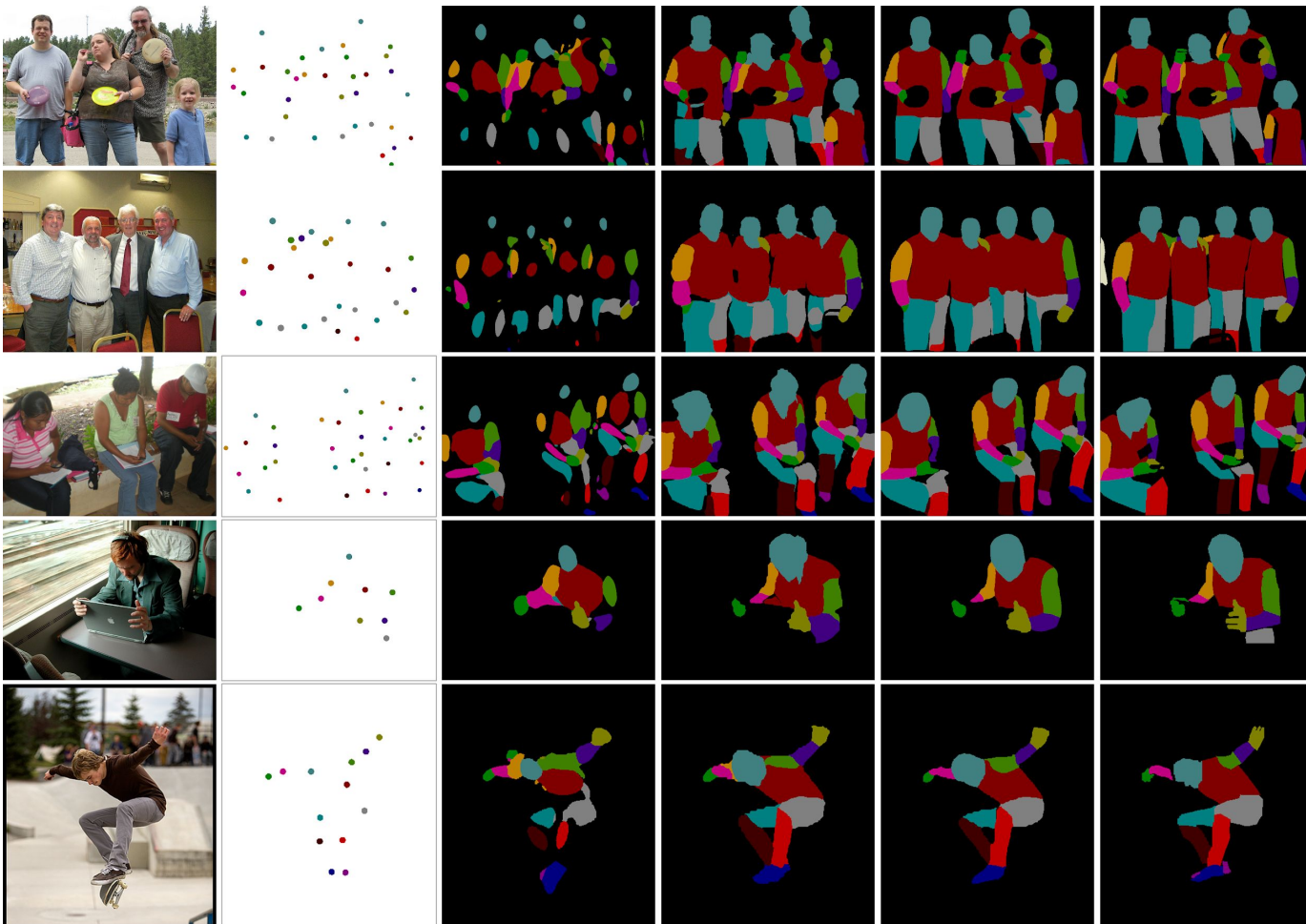
(b) Annotation

(c) Baseline

(d) Our SPML

(e) Full Supervision

(f) Ground Truth



(a) Image

(b) Annotation

(c) Baseline

(d) Our SPML

(e) Full Supervision

(f) Ground Truth

Context-Aware Segment Retrieval via Learned Pixel-wise Feature



Code available at <https://github.com/twke18/SPML>

