

Annotation-Free and One-Shot Learning for Instance Segmentation of Homogeneous Object Clusters

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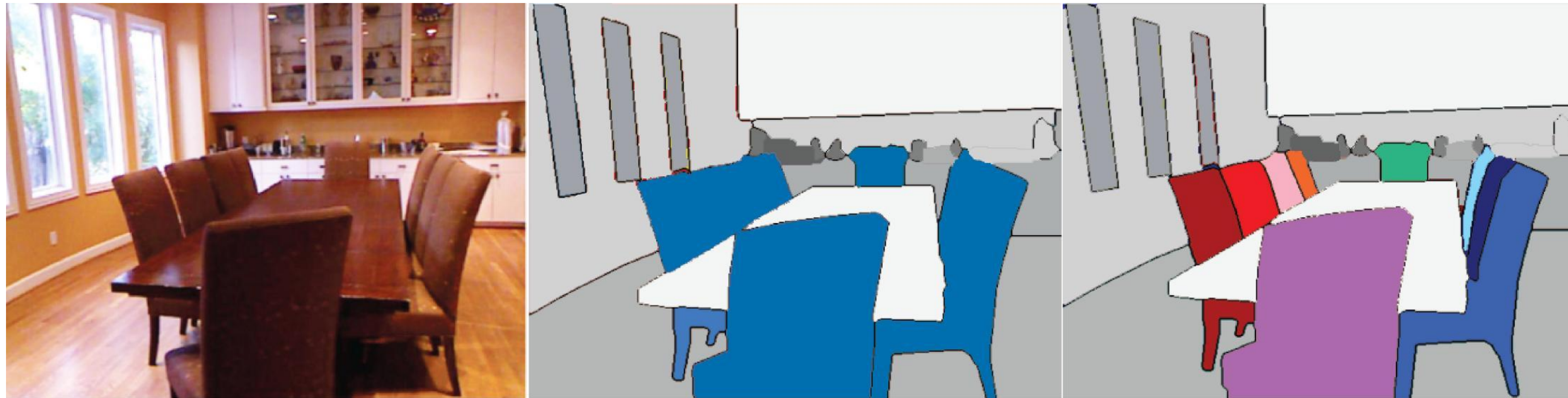
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Jul.18th 2018 @ IJCAI

Background



Input Image

Semantic Segmentation

Semantic Instance Segmentation

Background

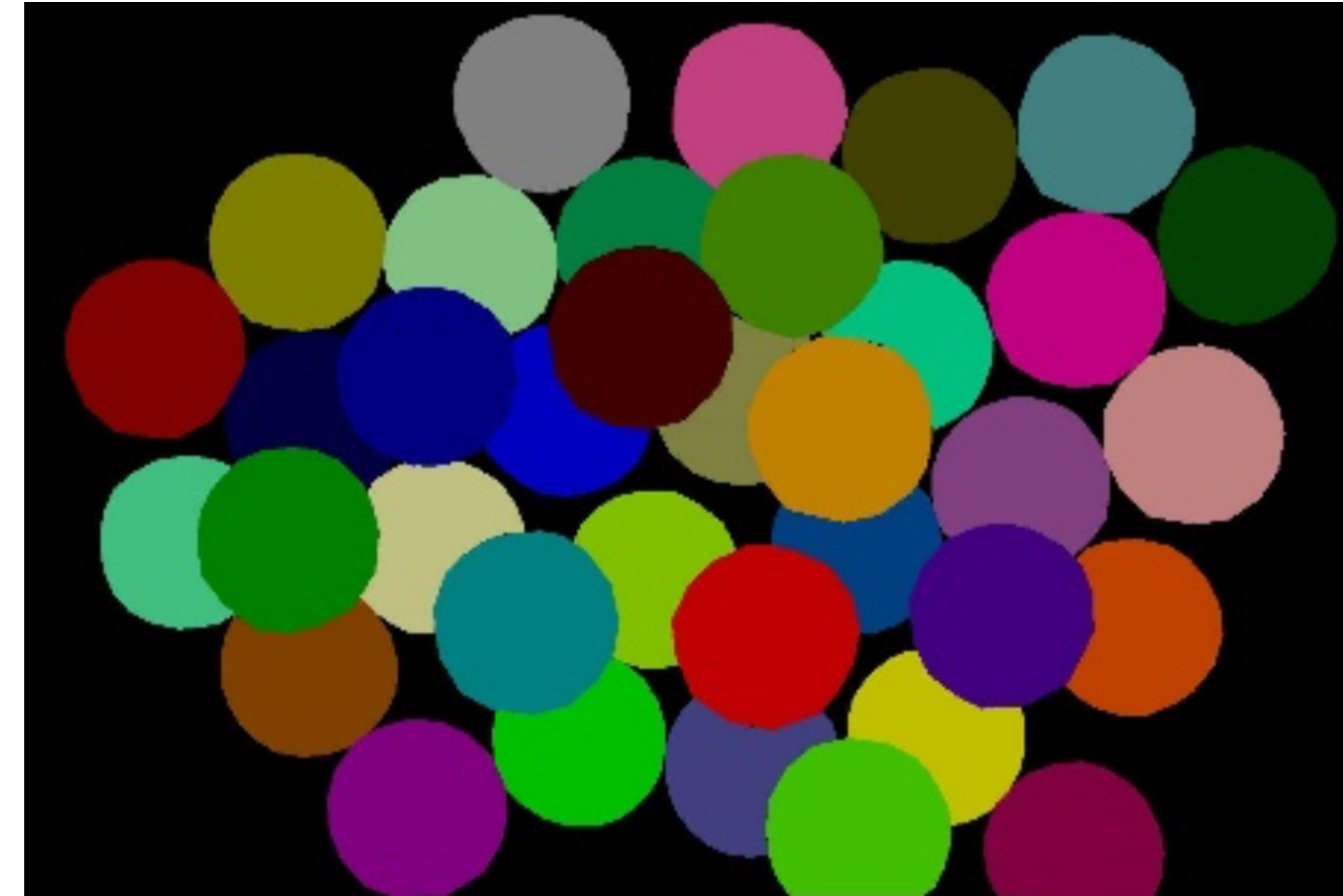


homogeneous object clusters (HOC)

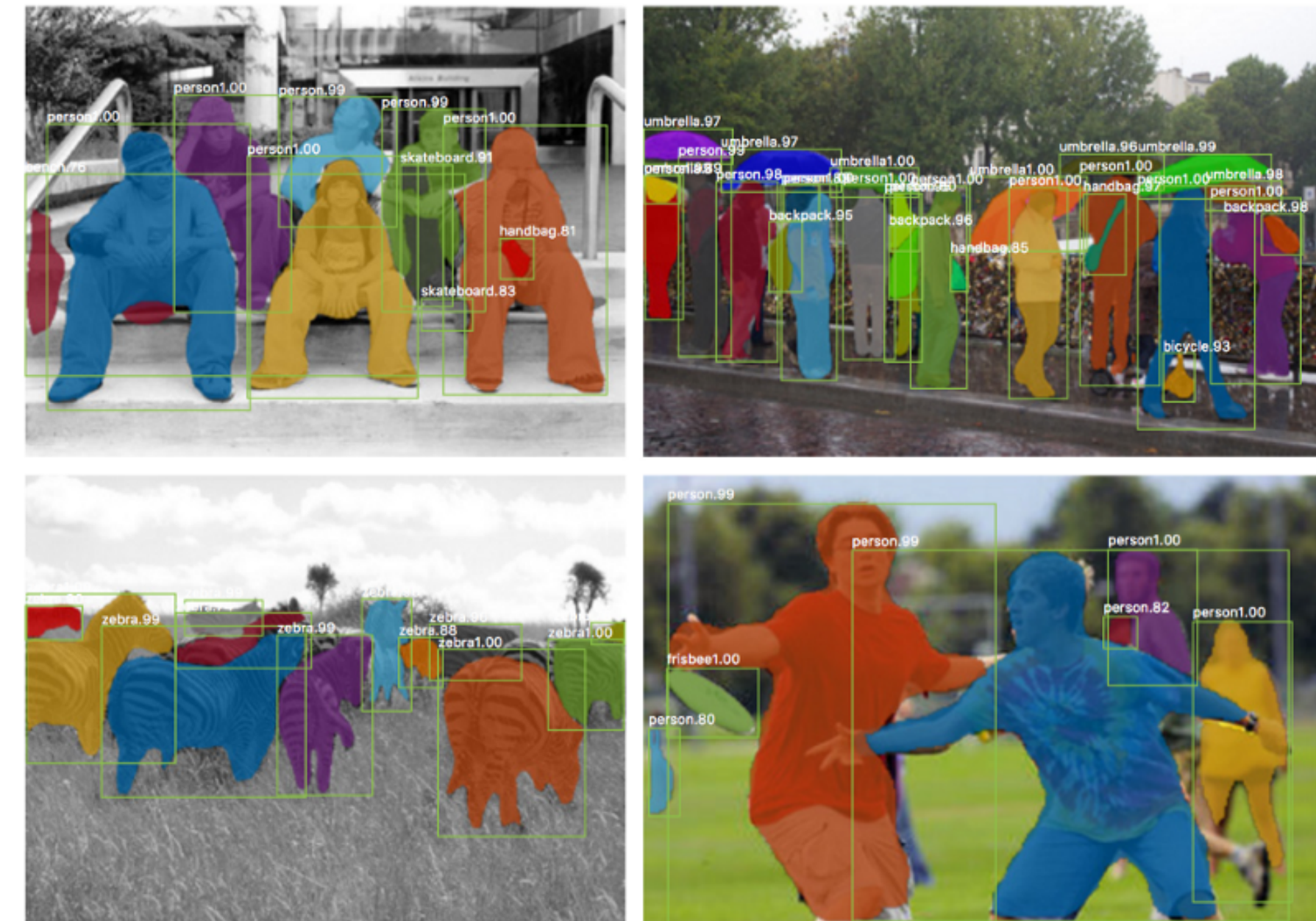
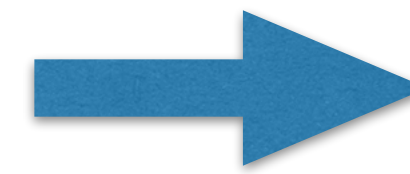
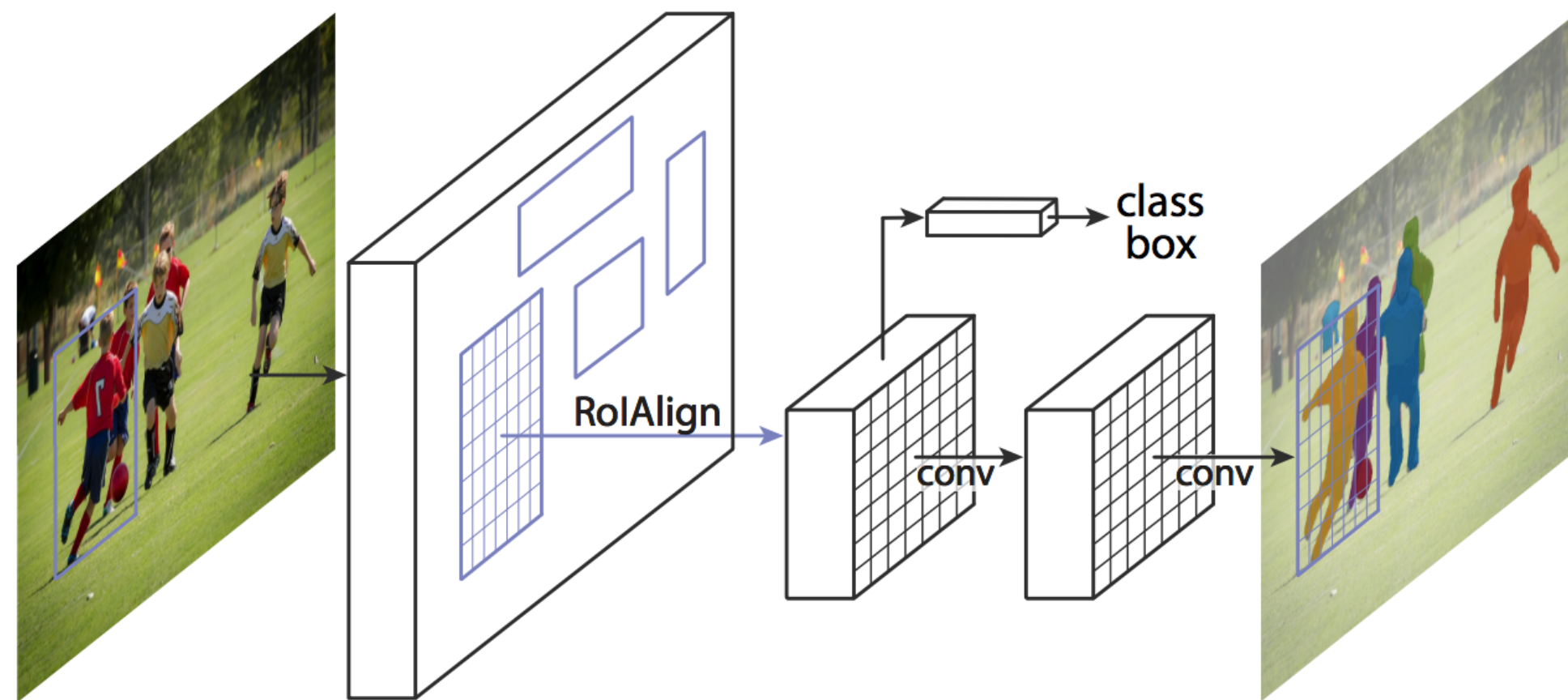


- objects of the same class
- densely distributed, highly occluded

Background

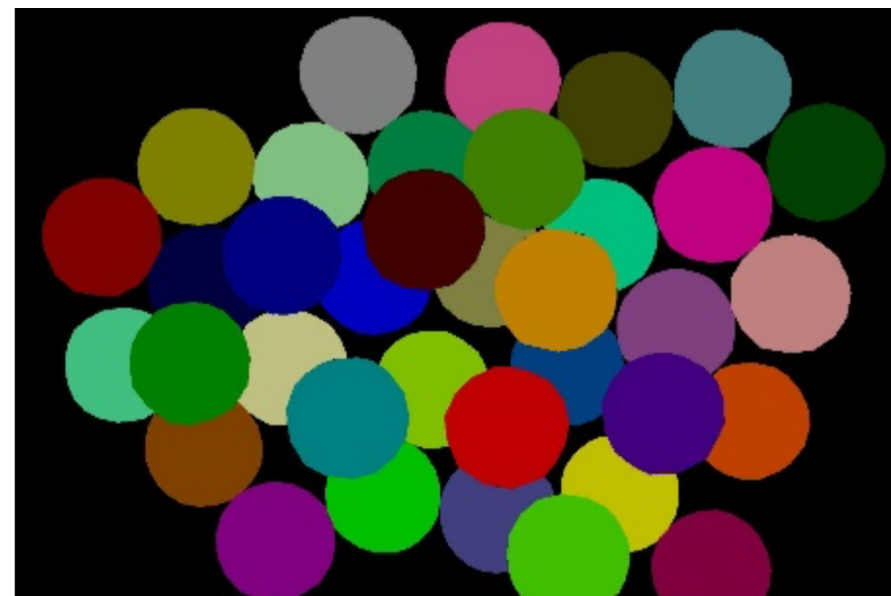


Mask R-CNN

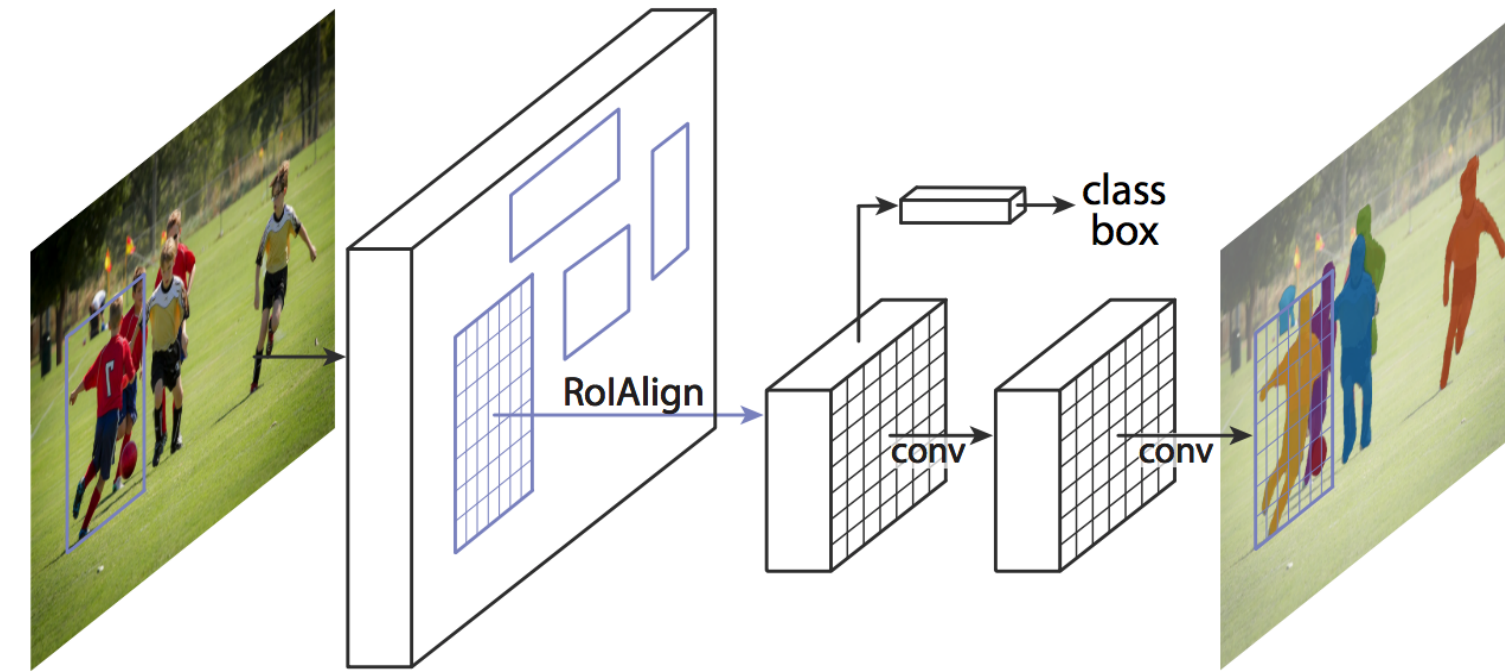


COCO: ~ 900k instances

Mask R-CNN



=

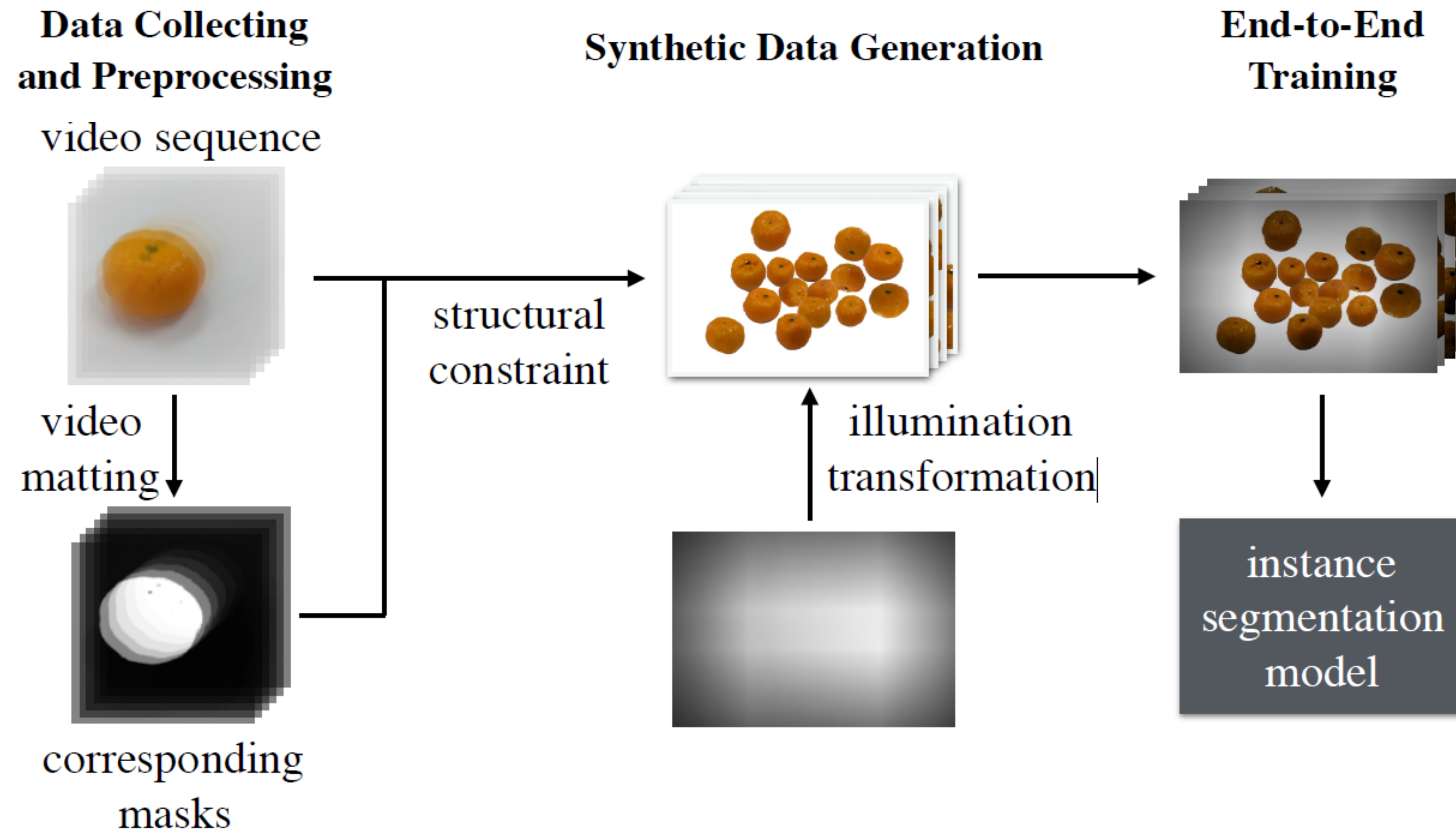


+

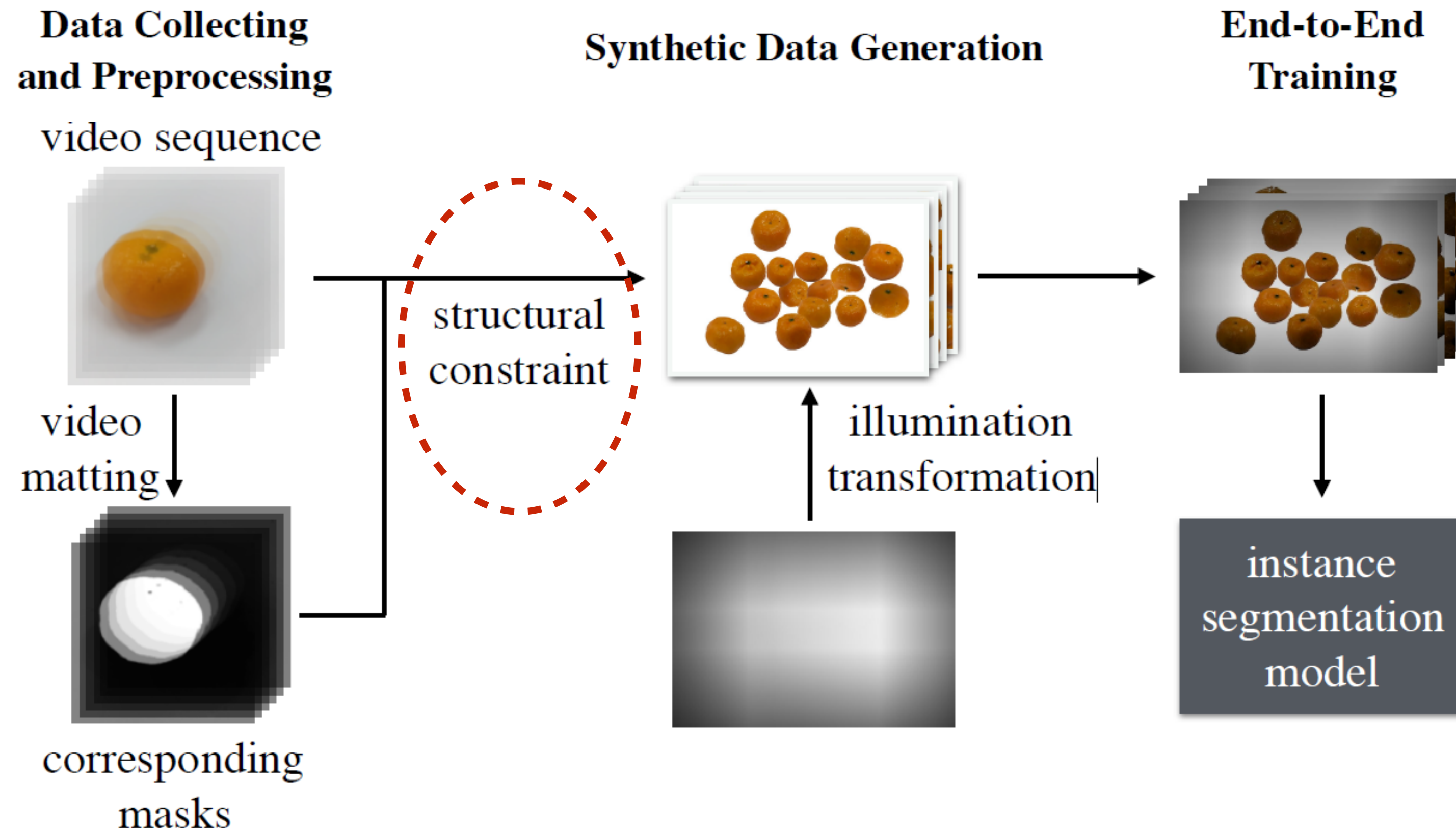
thousands of **annotated**
images of pingpong balls

Can we do it cheaply?

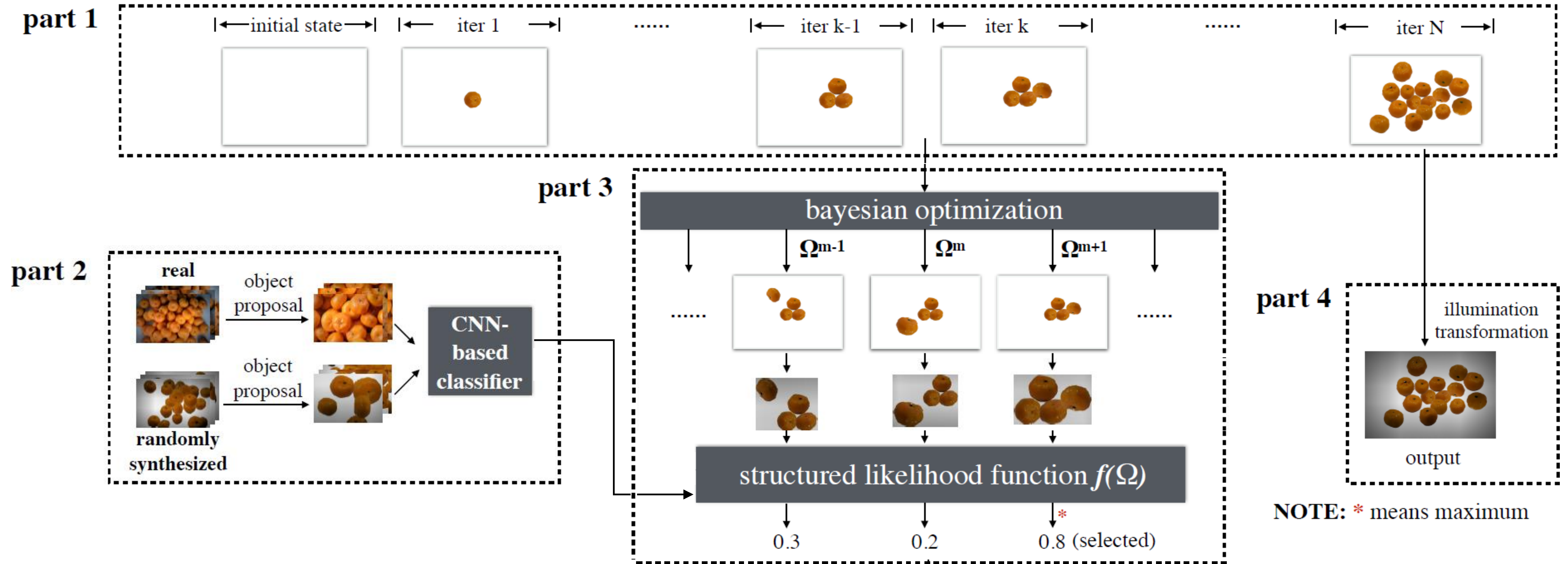
Our Pipeline



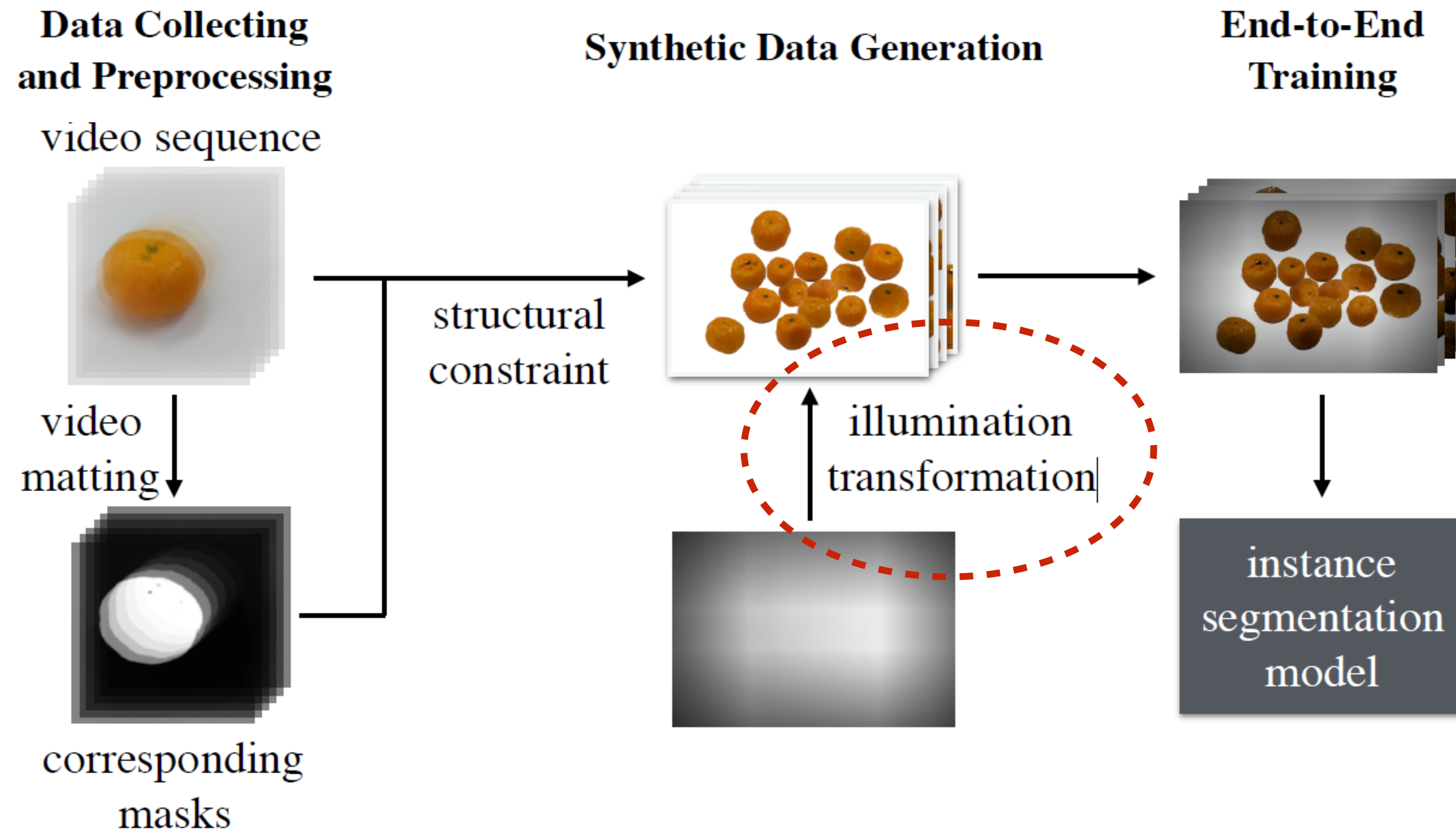
Structural Constraint



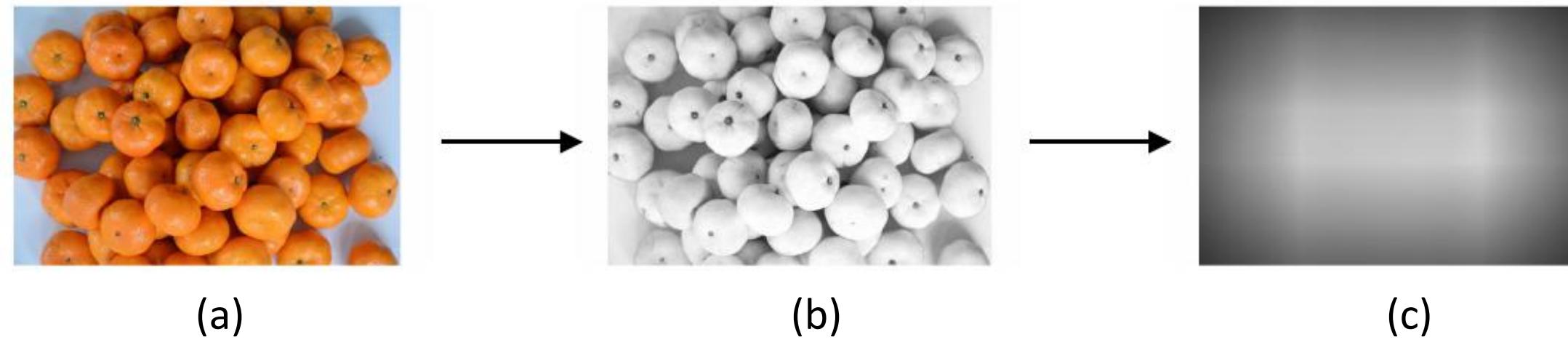
Structural Constraint



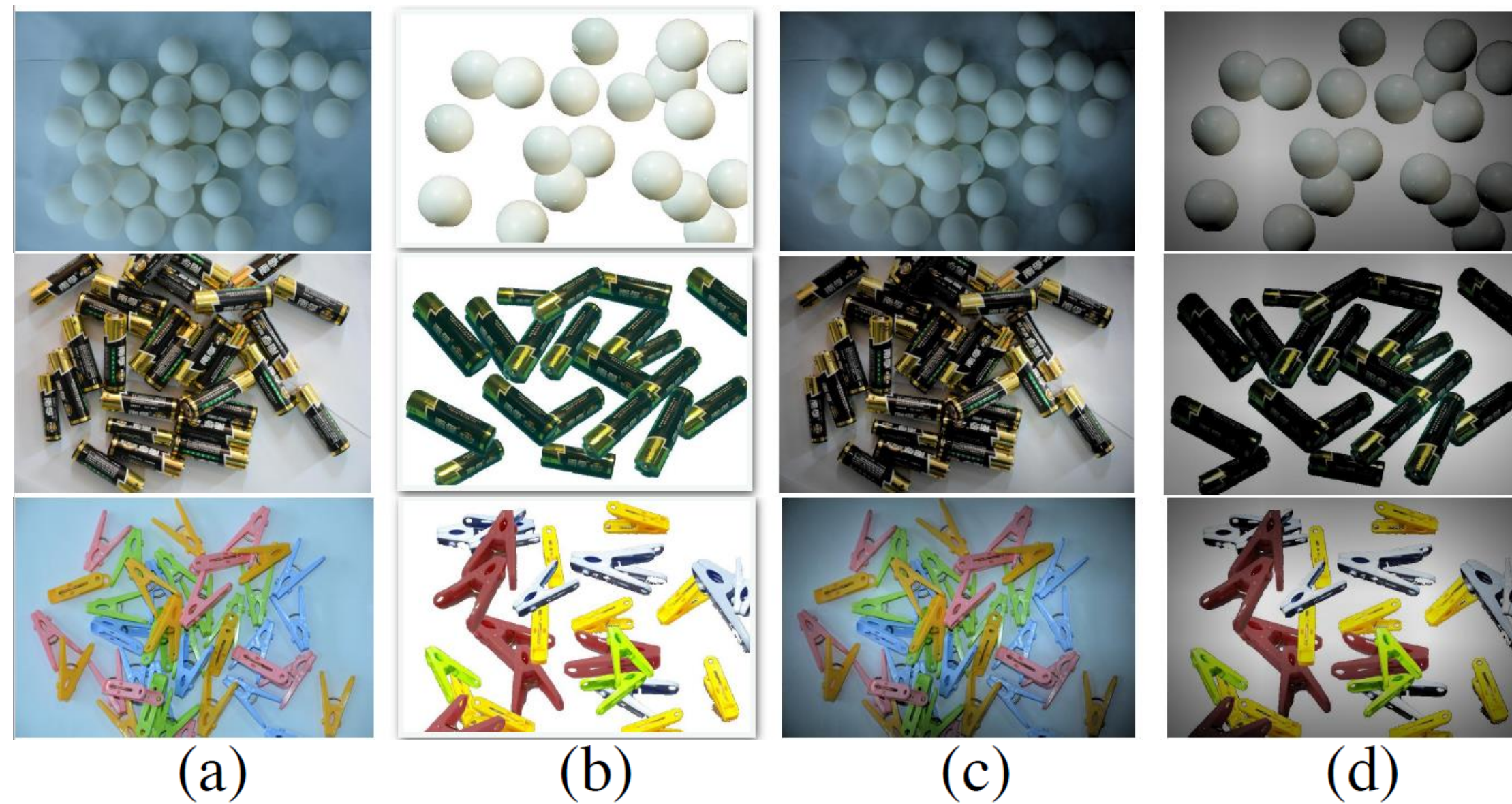
Illumination Transformation



Illumination Transformation

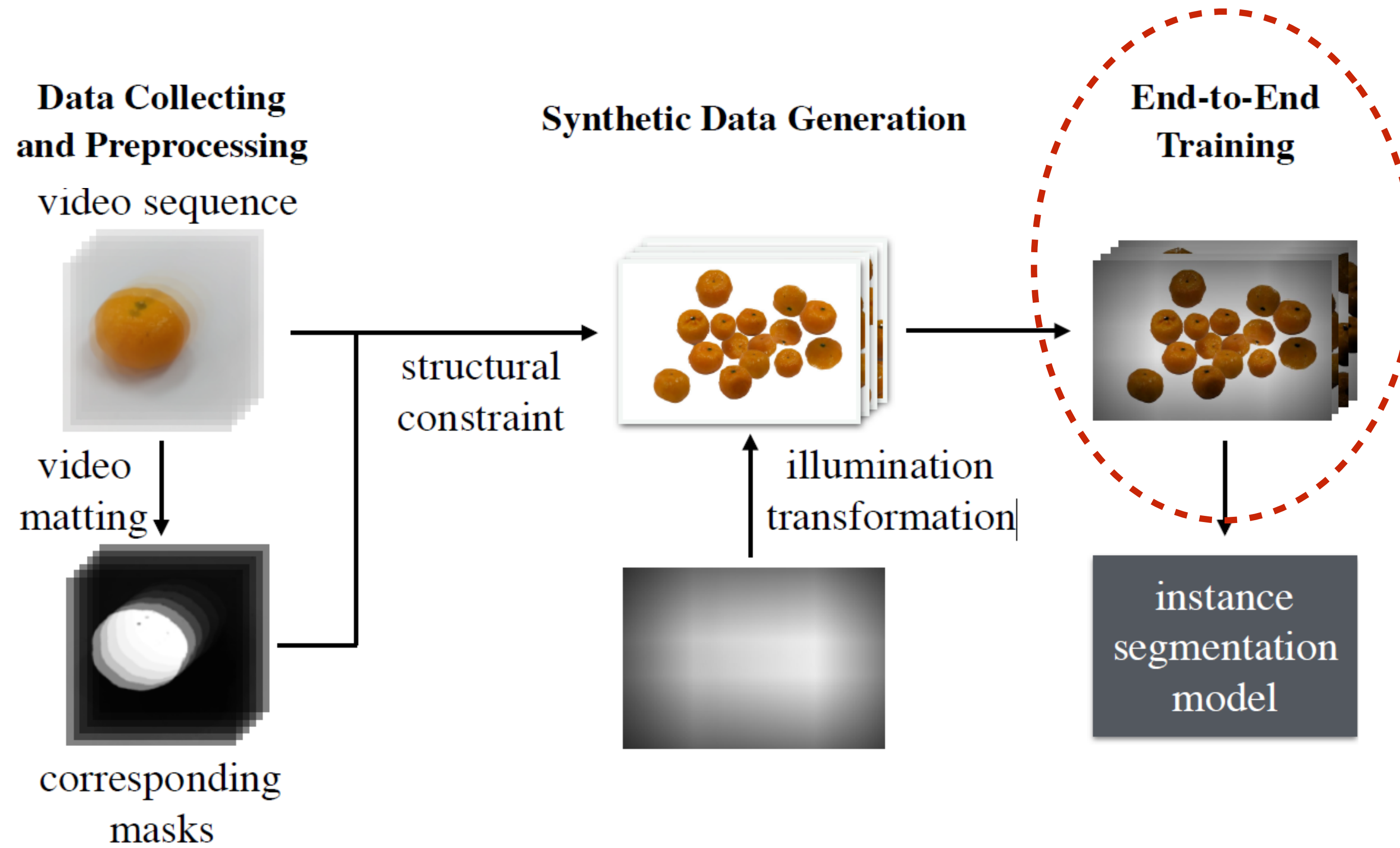


detail removing: apply large kernel Gaussian smoothing



$$V_{syn} = V_{syn} - \text{mean}(V_{syn}) + \text{blur}(V_{real})$$
$$V_{real} = V_{real} - \text{mean}(V_{real}) + \text{blur}(V_{real})$$

End-to-End Training



Dataset

- 10 classes
- 200 images in total
- 18.3 instances per image
- different backgrounds and lighting condition



Baselines

Single: images containing single object as training data

Random: randomly synthesized images as training data

Random+illumination: applies illumination transformation method to transform the synthetic data generated in **Random**

Random+structure: uses the method proposed in “Structural Constraint” section to generate structurally realistic training data

Experiments

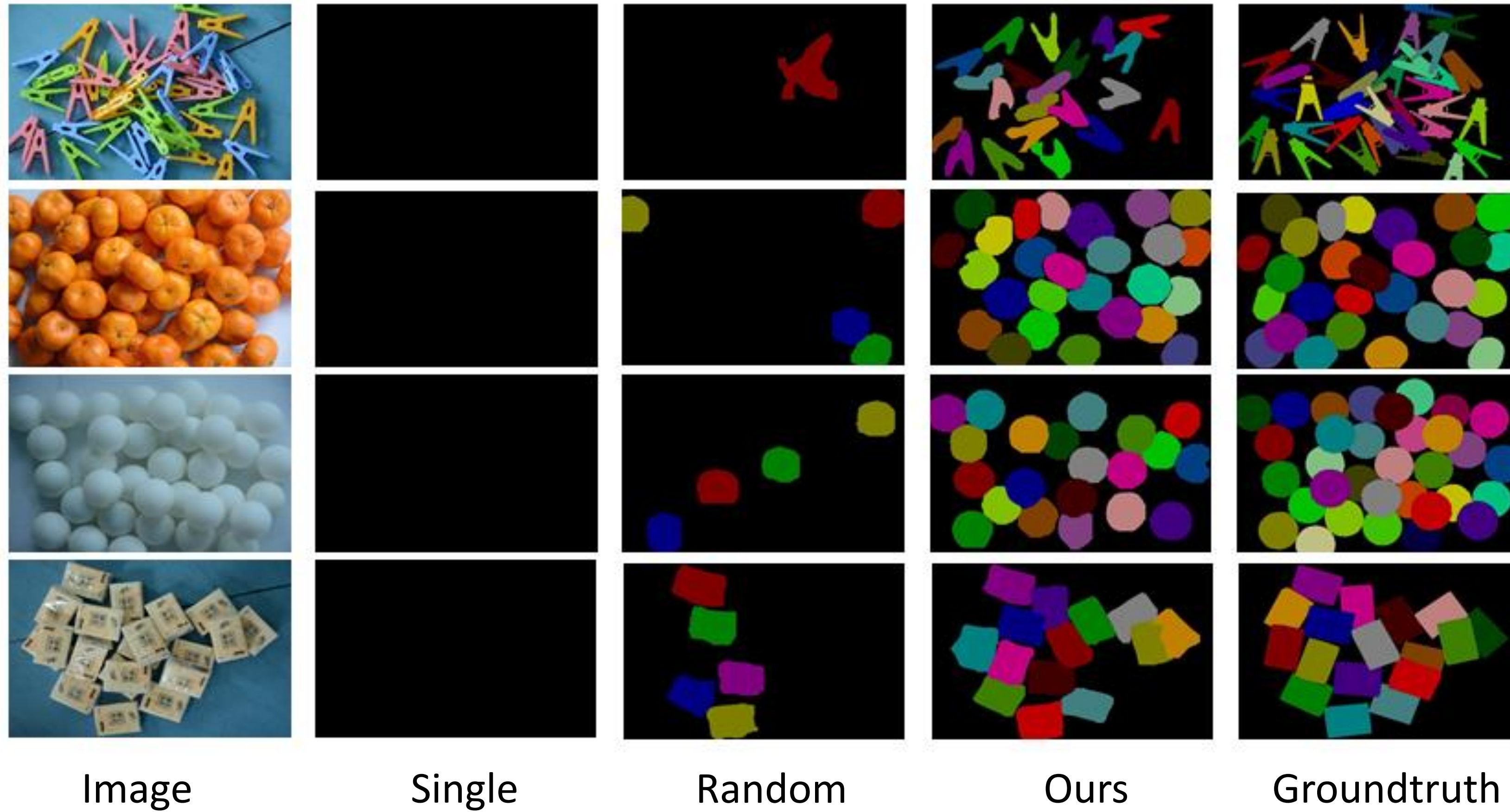
Table 1: Results on $mAP^r @0.5$ on our dataset. All numbers are percentages %.

	badminton	battery	clothespin	grape	milk	hexagon nut	orange	ping pong	tissue	wing nut	mAP
Single	12.1	23.2	4.4	19.3	8.2	17.5	17.6	14.2	13.1	21.1	15.1
Random	40.6	50.9	38.4	50.8	26.7	52.9	63.8	67.2	83.9	32.9	50.8
Random+illumination	44.6	48.7	34.3	41.6	26.0	46.3	54.9	64.2	68.9	39.4	46.9
Random+structure	34.2	39.3	52.6	72.7	31.3	62.8	90.3	81.7	90.7	23.7	57.9
Ours	53.0	69.5	67.7	72.5	52.6	73.6	90.0	81.2	90.4	48.4	69.9

Table 2: Results on $mAP^r @[0.5:0.95]$ on our dataset. All numbers are percentages %.

	badminton	battery	clothespin	grape	milk	hexagon nut	orange	ping pong	tissue	wing nut	mAP
Single	8.7	21.2	2.0	16.8	5.0	15.2	16.0	11.8	9.9	18.8	12.5
Random	33.1	44.7	29.3	44.8	20.9	43.4	56.3	59.5	68.4	27.7	42.8
Random+illumination	36.5	43.2	28.4	37.2	20.8	38.3	50.4	56.4	56.8	35.5	40.4
Random+structure	29.8	36.5	36.7	66.5	22.0	45.2	83.8	76.4	80.4	20.7	49.8
Ours	44.2	60.3	46.2	65.4	41.3	54.8	84.0	75.6	80.4	39.3	59.2

Qualitative Results



Thanks

Q & A