

# Cell Segmentation in the Presence of Dynamically Localized Proteins

Özgün Çiçek\*  
cicek@cs.uni-freiburg.de

Yassine Marrakchi\*  
marrakch@cs.uni-freiburg.de

Enoch B. Antwi  
enoch.antwi@bioss.uni-freiburg.de

Barbara Di Ventura  
barbara.diventura@biologie.uni-freiburg.de

Thomas Brox  
brox@cs.uni-freiburg.de

\* Equal Contribution

## Motivation

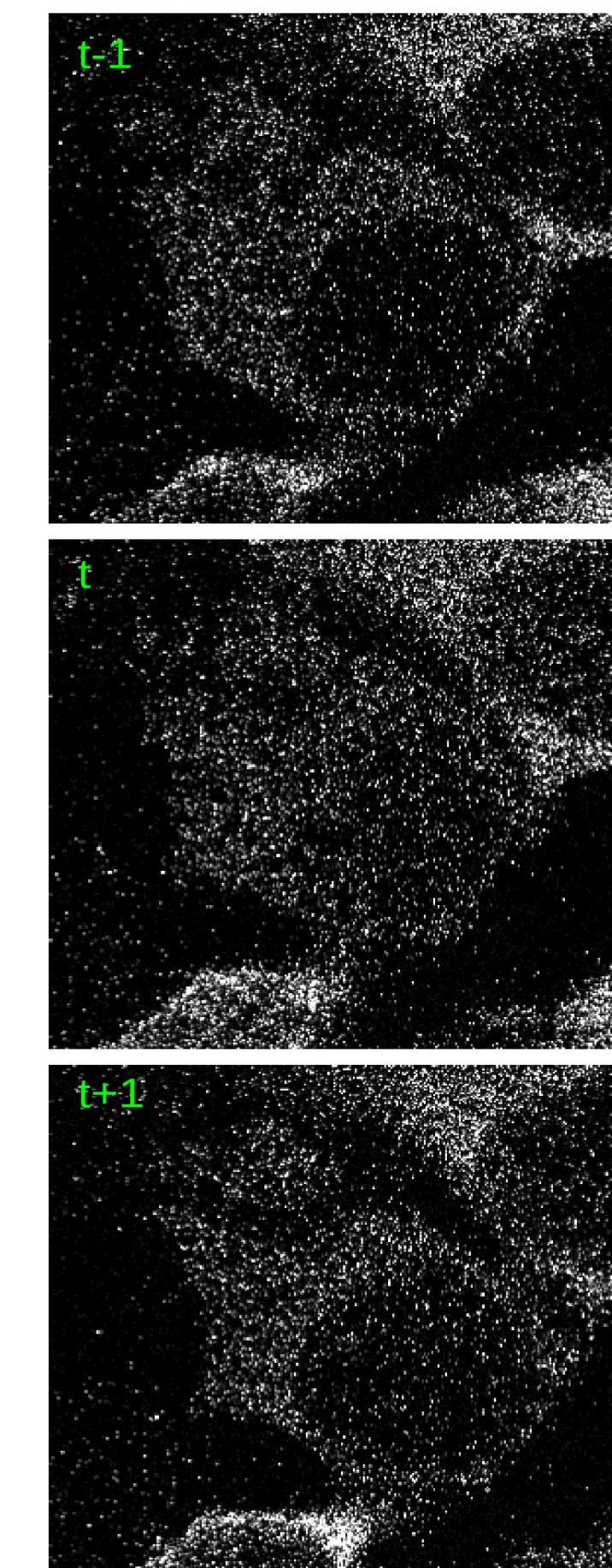
The dynamic localization patterns of proteins dictate their function.

Biologists use optogenetics to control protein localization via light exposure.

Repeatedly giving light creates oscillations of the protein in and out of the nucleus.

These oscillations cause regular and temporary deterioration in visibility.

Deep learning methods become unreliable when the visibility is drastically deteriorated.

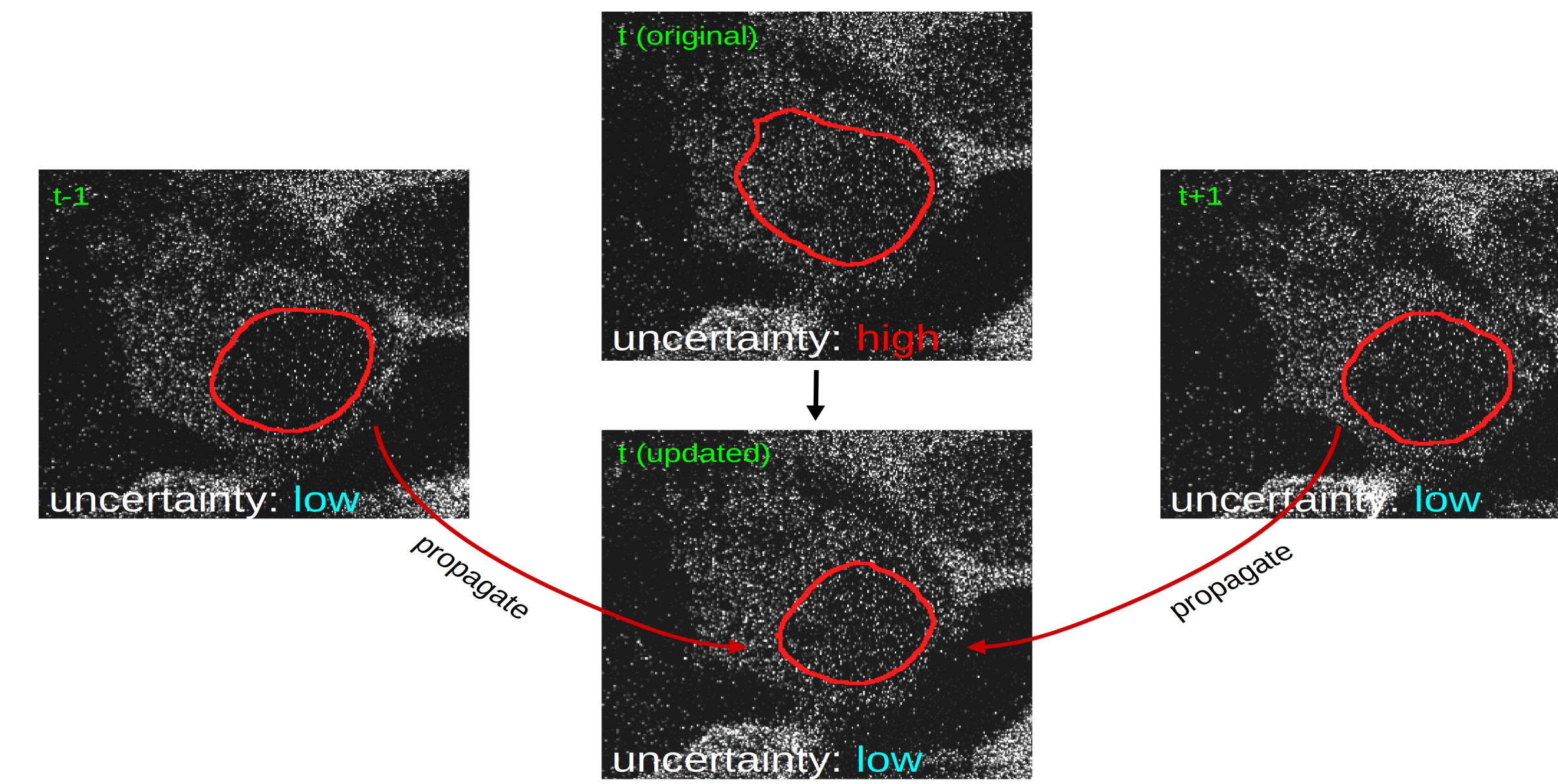


Time-lapse confocal images of HEK293T cells with fusion protein mCherry-LINuS

## Our Contributions

Mask R-CNN (He et al., 2017) with uncertainty to detect erroneous predictions.

Optical flow to refine detected errors by propagating certain predictions from neighboring frames.



Oscillation at time t causing bad nuclei segmentation (up) and the corrected segmentation of it using our propagation method (down).

## Results

### Uncertainty Evaluation in mAP (@0.5/@0.75 IoU)

	model uncertainty		combined uncertainty	
	mAP (sm)	mAP (ent)	mAP (sm)	mAP (ent)
Single	0.77/0.48	0.80/0.49	0.74/0.60	0.83/0.69
Dropout	0.74/0.61	0.78/0.65	0.77/0.61	0.83/0.67
Ensemble	0.82/0.64	0.78/0.61	0.78/0.63	0.83/0.70
SGDR Ensemble	0.75/0.54	0.72/0.51	0.71/0.49	0.63/0.44
WTA Merged	0.74/0.47	0.82/0.49	0.83/0.56	0.85/0.64
EWTA Merged	0.64/0.51	0.73/0.58	0.80/0.64	0.77/0.59

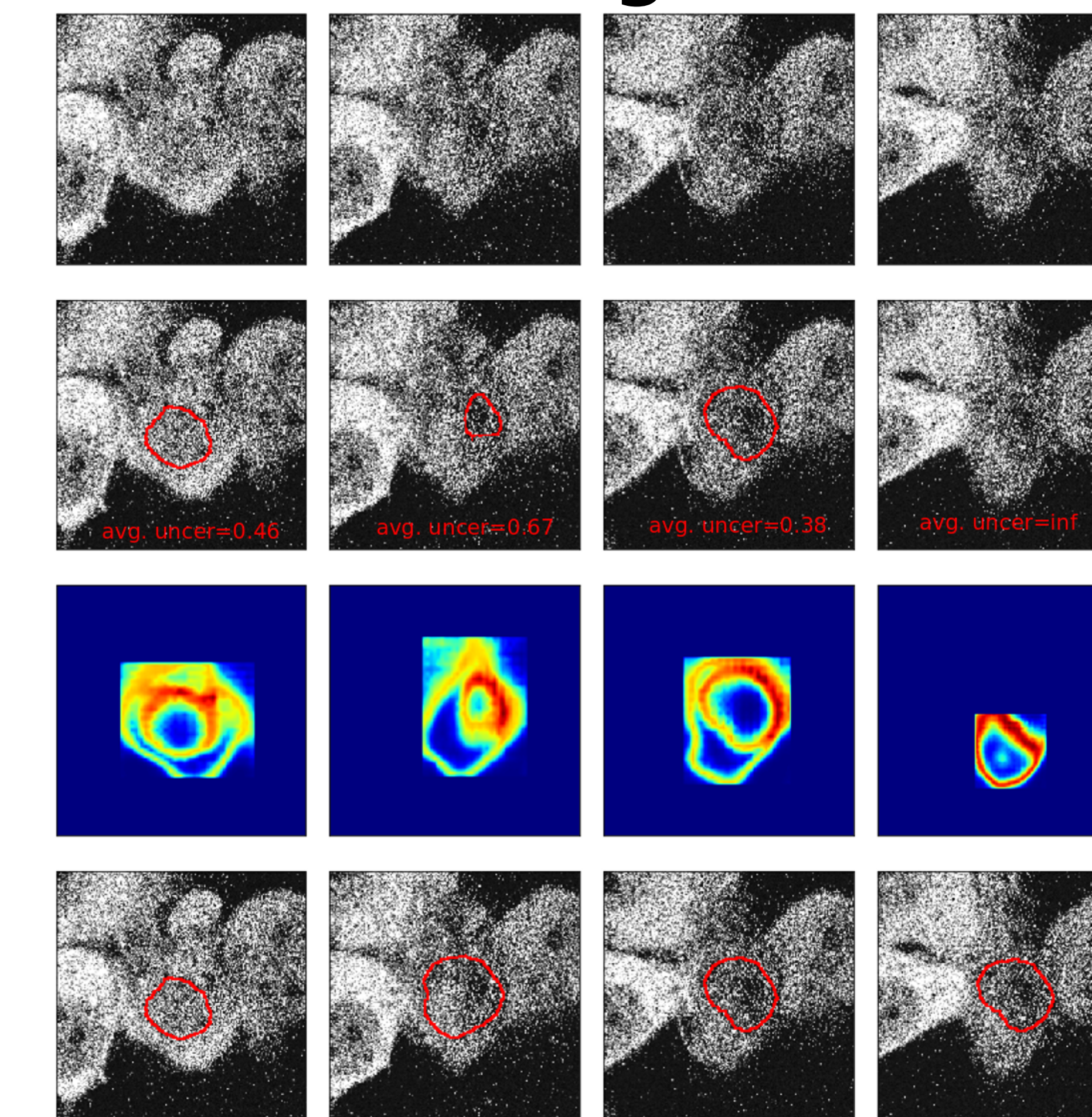
WTA merged is the best at 0.5 IoU threshold and at 0.75 ensemble is the best. We use the WTA merged with data uncertainty since it is computationally more efficient than ensemble.

### Propagation Evaluation in mean IoU

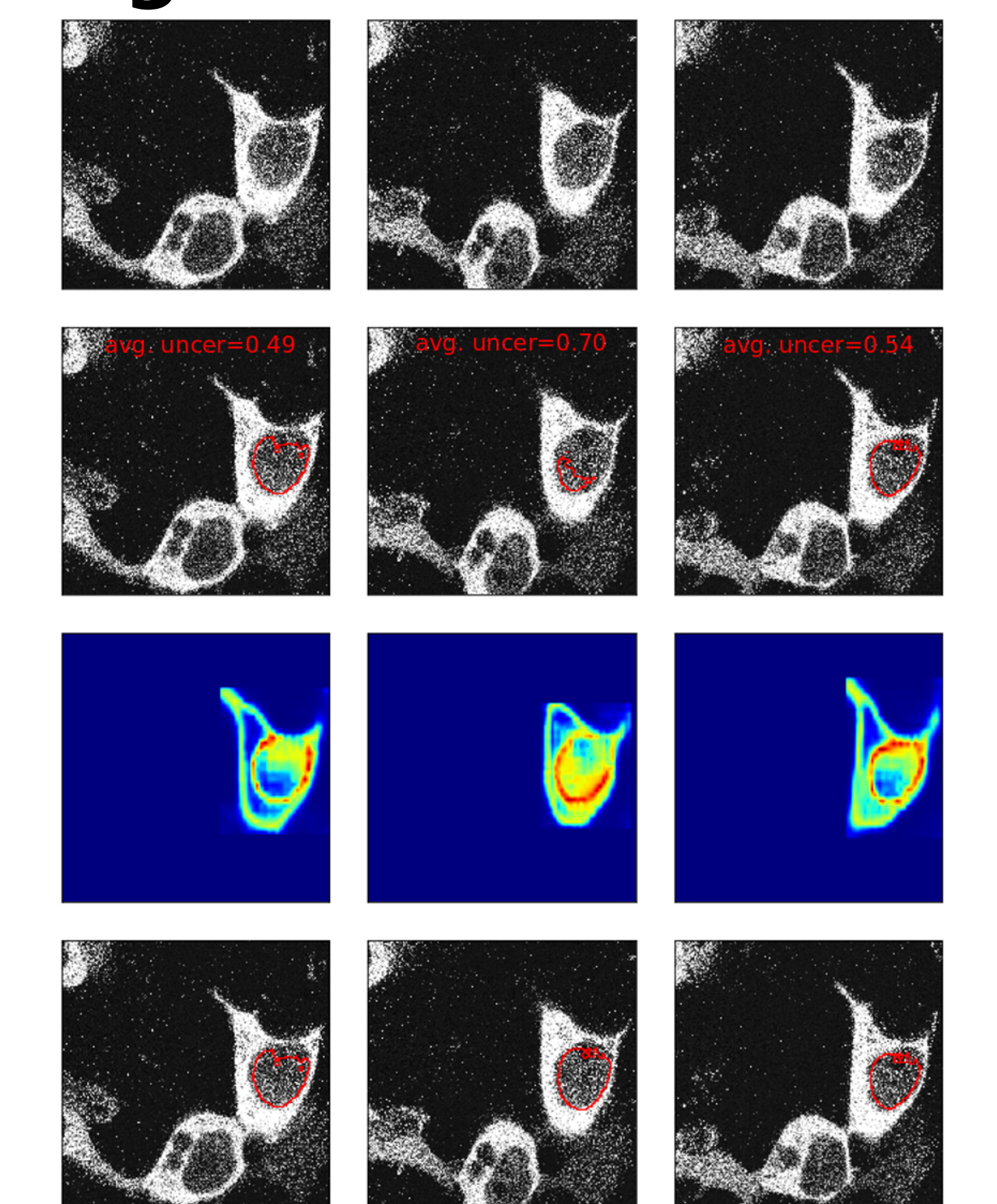
update	warp with	mask fusion	all(117)	updated(51)	extrapolated(11)	non-updated(55)
none	none	no	0.62	0.55	0.00	0.80
uncertain	shift+scale	no	0.71	0.68	0.39	0.80
uncertain	mean nuclei flow	no	0.73	0.71	0.45	0.80
all	mean nuclei flow	no	0.69	0.70	0.40	0.74
uncertain	pixel-wise flow	no	0.73	0.72	0.44	0.80
uncertain	pixel-wise flow	yes	0.72	0.70	0.40	0.80

Our method can effectively improve erroneous nuclei predictions.

### Extreme Signal Loss

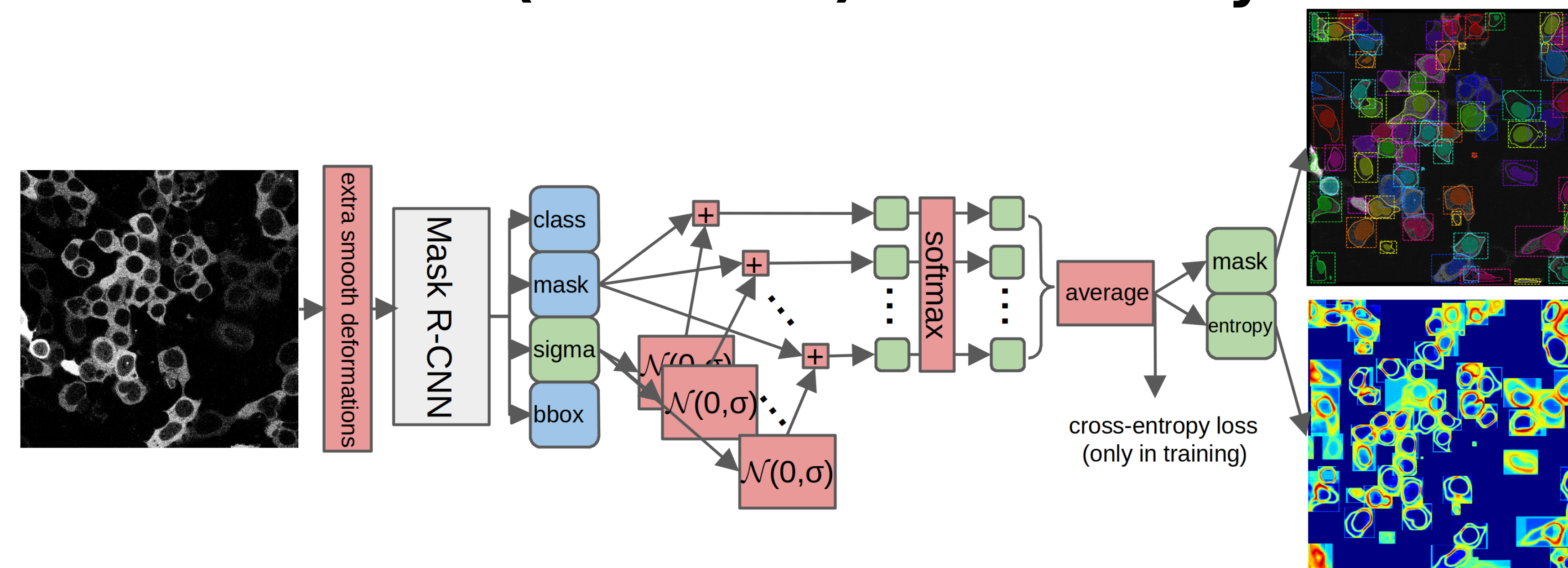


### Segmentation Failure



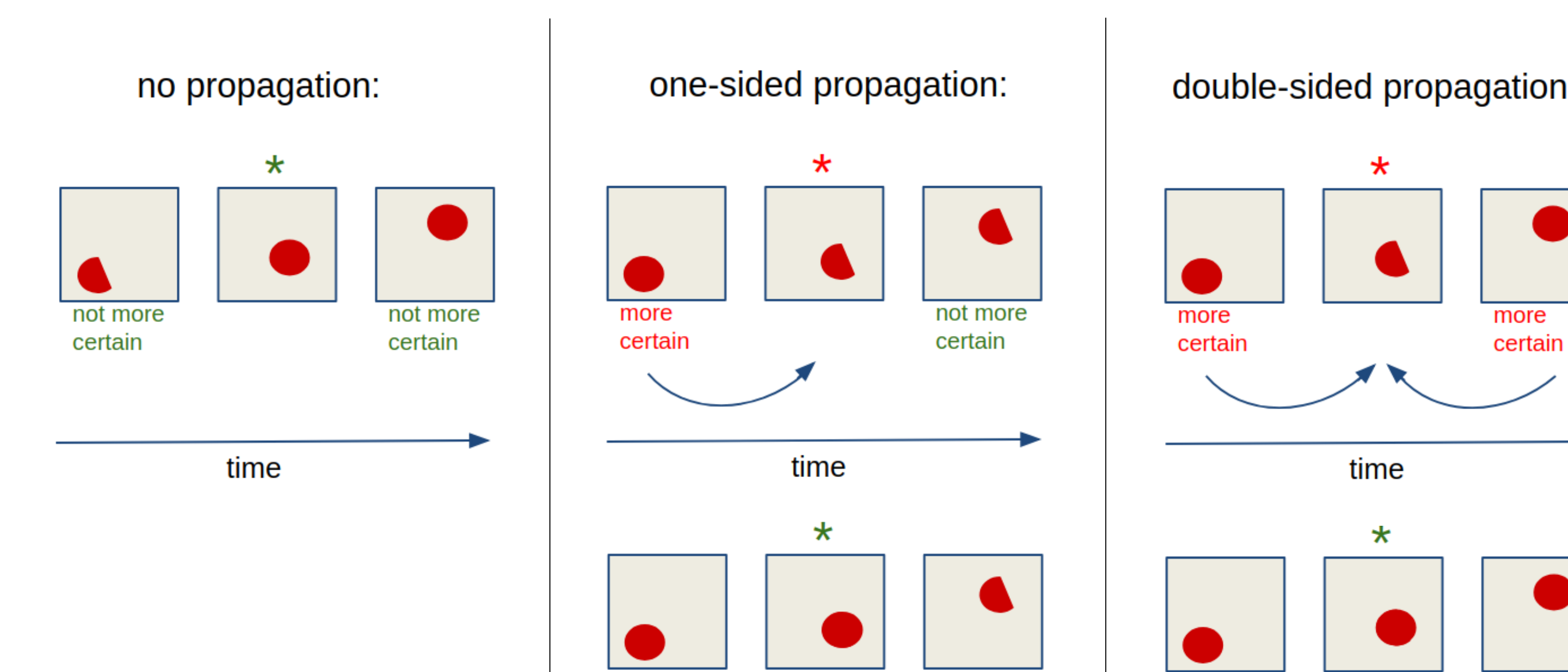
## Overview of Approaches

### Data (Aleatoric) Uncertainty



### Uncertainty-Based Nuclei Mask Propagation

Traverse the video in increasing average uncertainty order:



Traversing continues with the updated masks and uncertainties to facilitate long propagation horizon.

## Conclusions

We solve a real task commonly experienced in signalling studies which is not yet addressed.

Our method improves nuclei segmentation over several baselines.

Our method can facilitate automated analysis of dynamically localized proteins without additional markers.

## References

- [1] He, K., Gkioxari, G., Dollár, P., Girshick, R.B.: Mask R-CNN. In: ICCV (2017)
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- [3] Makansi, O., Ilg, E., Cicek, O., Brox, T.: Overcoming limitations of mixture density networks: A sampling and fitting framework for multimodal future prediction. In: CVPR (2019)

## Acknowledgements

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