



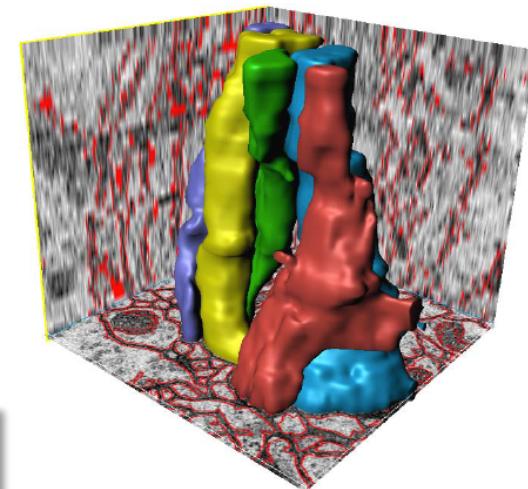
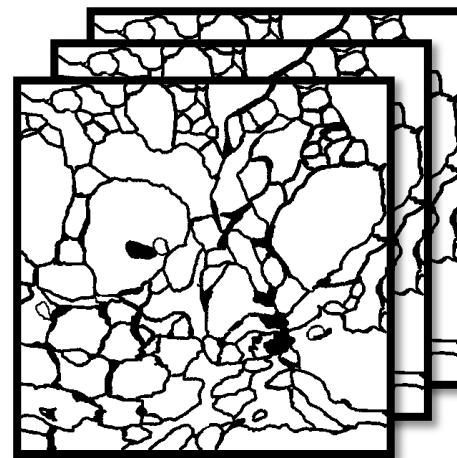
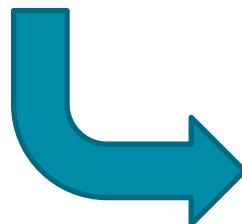
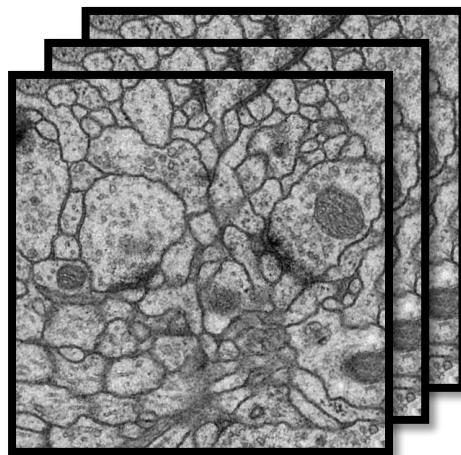
# SuperSlicing Frame Restoration for Anisotropic ssTEM and Video Data

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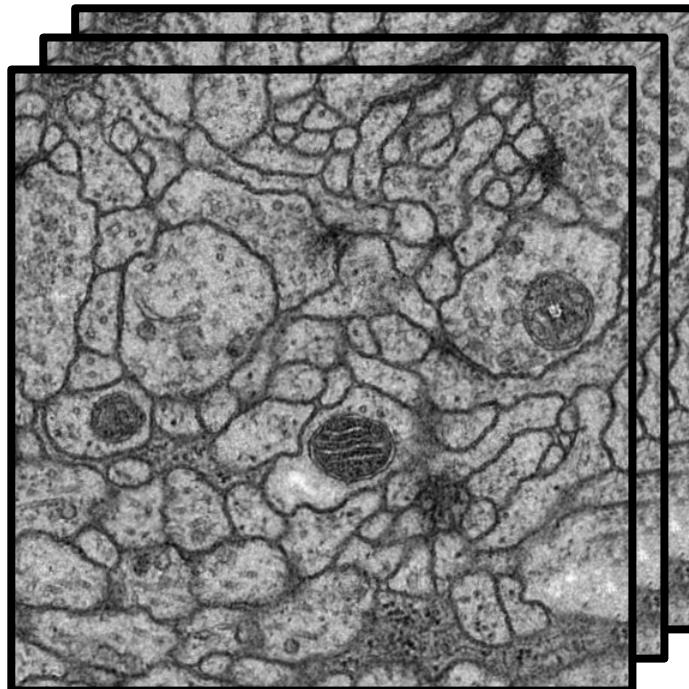
<http://dlaptev.org>  
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# 3d topology reconstruction from neuroimaging\*



\*only one of many branches in Connectomics research

# Connectomics



Sections from a serial section  
Transmission Electron  
Microscopy (ssTEM) data set of  
the Drosophila first instar larva  
ventral nerve cord

## Imaging

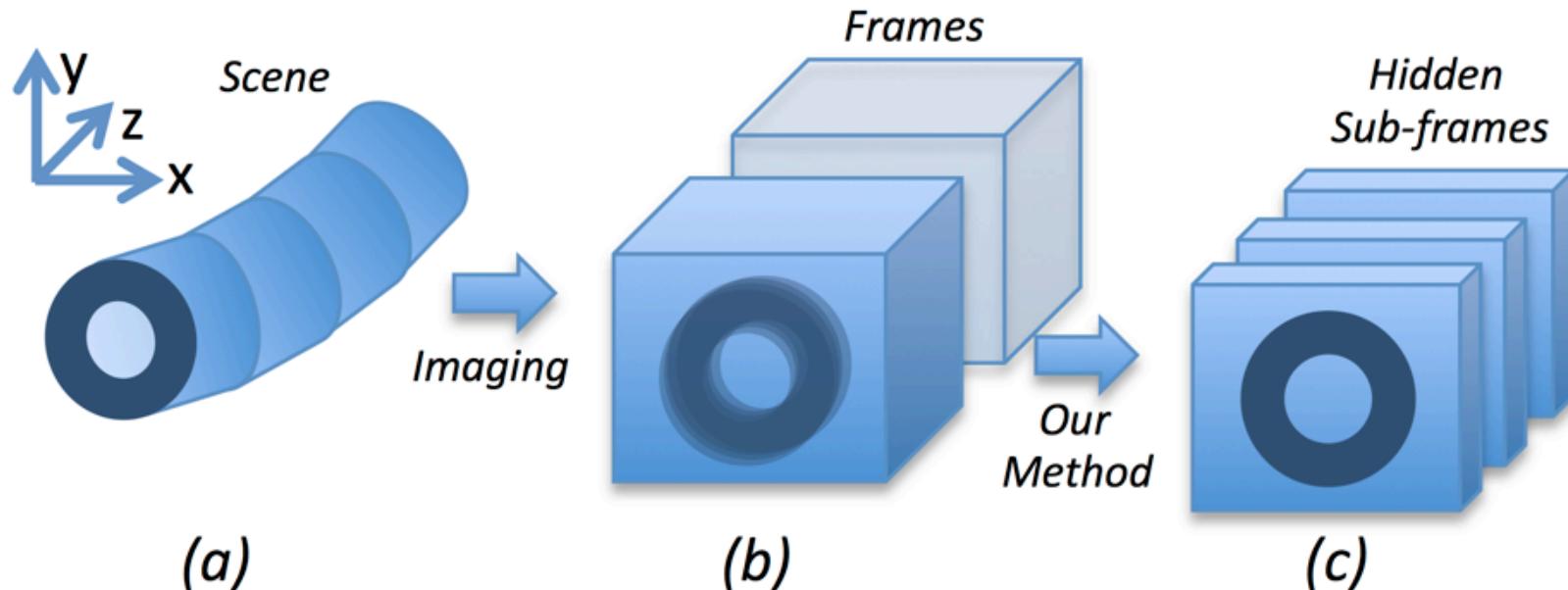
- Slices  $2\mu\text{m} \times 2\mu\text{m} \times 50\text{nm}$
- Slice projection with EM
- Resolution:  $4 \times 4 \times 50\text{nm}$   
 $\Rightarrow$  highly anisotropic

## Problems

- Blurred membranes  
 $\Rightarrow$  It is difficult to detect
- Slices sufficiently different  
 $\Rightarrow$  We cannot use 3d

# Idea of SuperSlicing

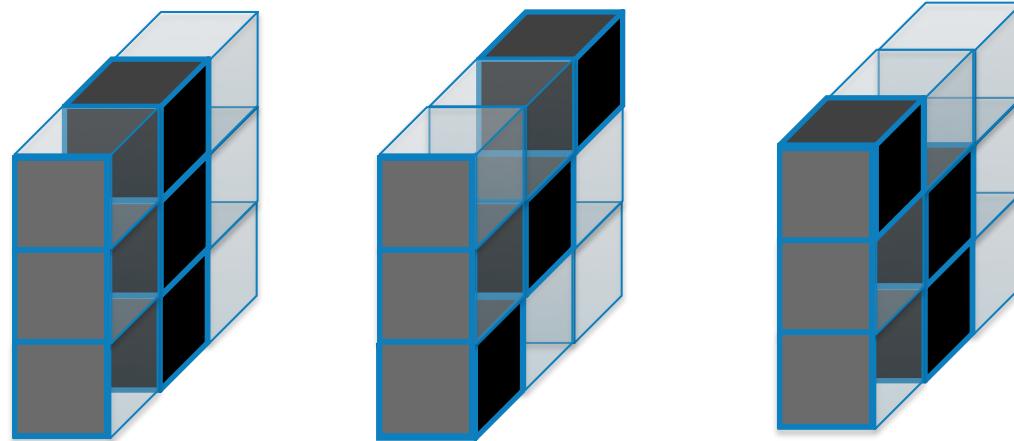
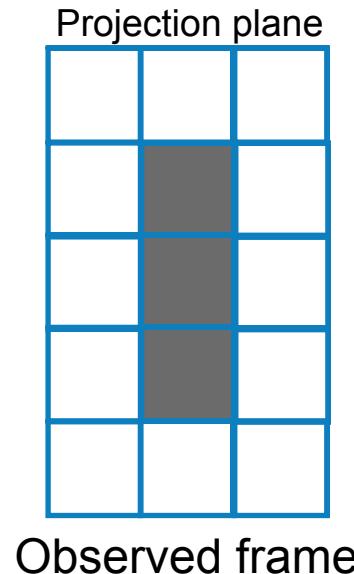
Decompose the observed frame into “hidden sub-frames”



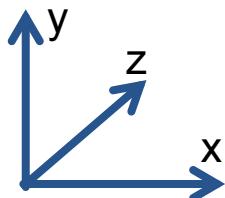
Very ill-posed problem in general

# Idea of SuperSlicing

Decompose the observed frame into “hidden sub-frames”



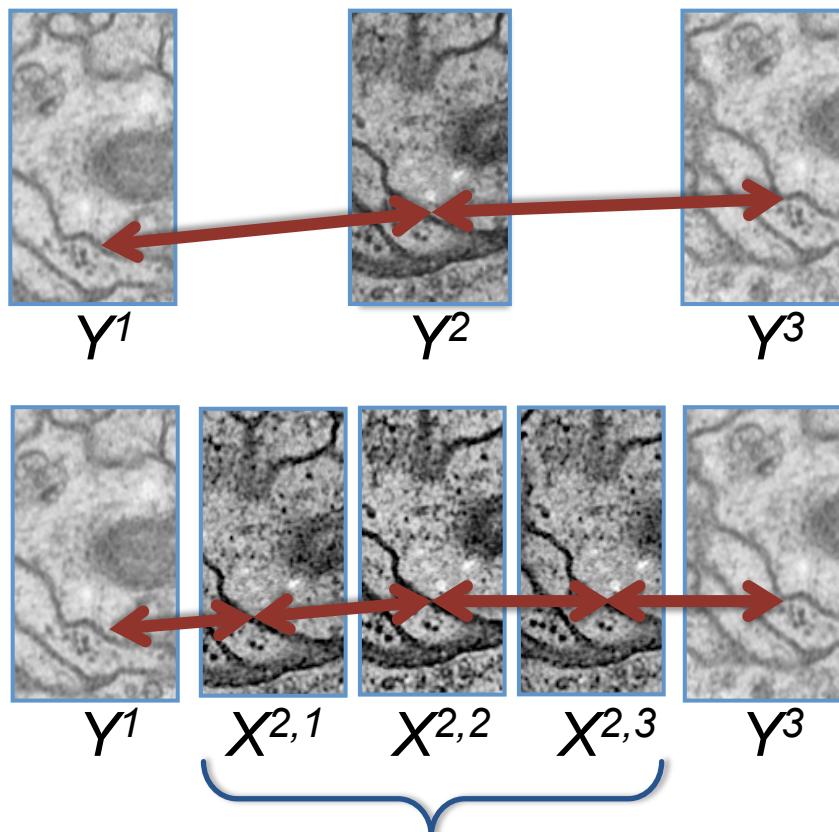
Three possible sub-frame decompositions



Need structural constraints

# Idea of SuperSlicing

$Y^1, Y^2, Y^3$  – observed neighboring frames



- Find corresponding pixels in anisotropic frames by solving **registration task**
- Interpolate correspondences between pixels in sub-frames
- Corresponding pixels should have similar intensities

Hidden sub-frame decomposition of  $Y^2$

# SuperSlicing Algorithm

- Notation

- $Y^n$  — observed frames,  $n \in [1, \dots, N]$ ,
- $y_p^n$  — pixel  $p$  of the frame  $Y^n$ ,
- $i(y_p^n)$  — the intensity of pixel  $y_p^n$ ,
- $\epsilon(x_p^n)$  — a set of neighbors of pixel  $x_p^n$ ,
- $\Omega$  — a set of given correspondences,
- $X^{n,l}$  — hidden sub-frames decomposition of  $Y^n$ ,  
 $l \in [1, \dots, L]$ .

# SuperSlicing Algorithm

- Energy minimization problem

An average of the hidden frames should give observed frame

$$E(X^{n,1}, \dots, X^{n,L}) = \sum_{y \in Y^n} \left( i(y) - \frac{1}{L} \sum_{l=1}^L i(x_p^{n,l}) \right)^2$$

Smoothness across all the correspondences

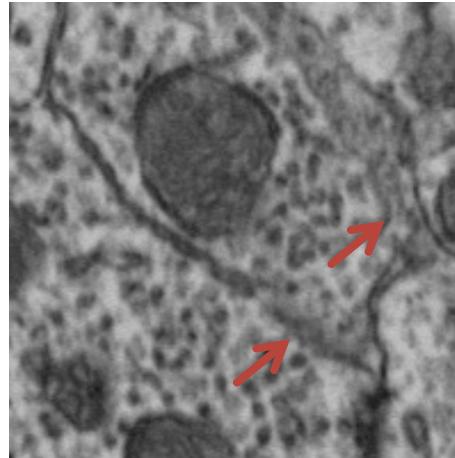
$$+ \lambda \sum_{(\hat{x}_p^{n,l}, \hat{x}_q^{n,l+1}) \in \Omega} \left( \sum_{x \in \epsilon(\hat{x}_p^{n,l})} w(x, \hat{x}_p^{n,l}) i(x) - \sum_{x \in \epsilon(\hat{x}_q^{n,l+1})} w(x, \hat{x}_q^{n,l+1}) i(x) \right)^2$$

Smoothness within the frame

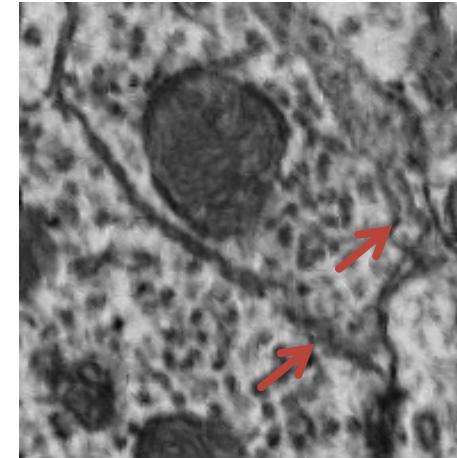
$$+ \gamma \sum_{\substack{x_p^{n,l}; x_q^{n,l} \in \epsilon(x_p^{n,l}) \\ l=1, \dots, L}} \left( i(x_p^{n,l}) - i(x_q^{n,l}) \right)^2$$

Quadratic programming: global optimum

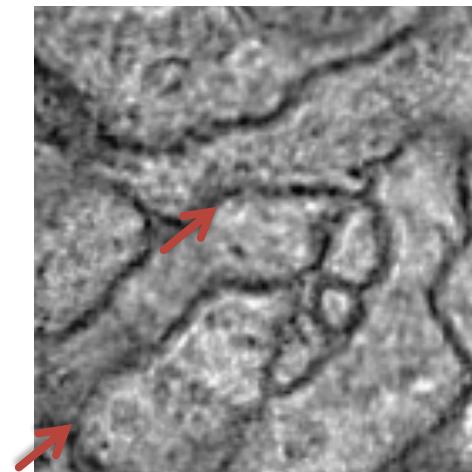
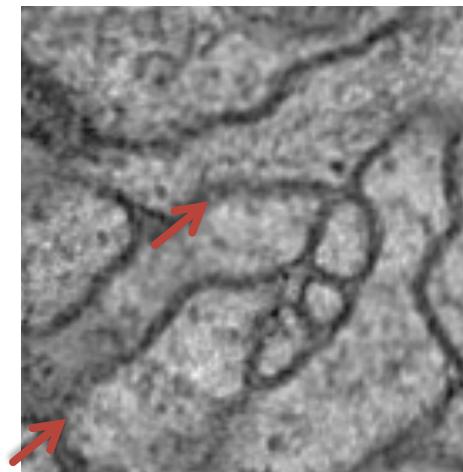
# SuperSlicing experiments: ssTEM



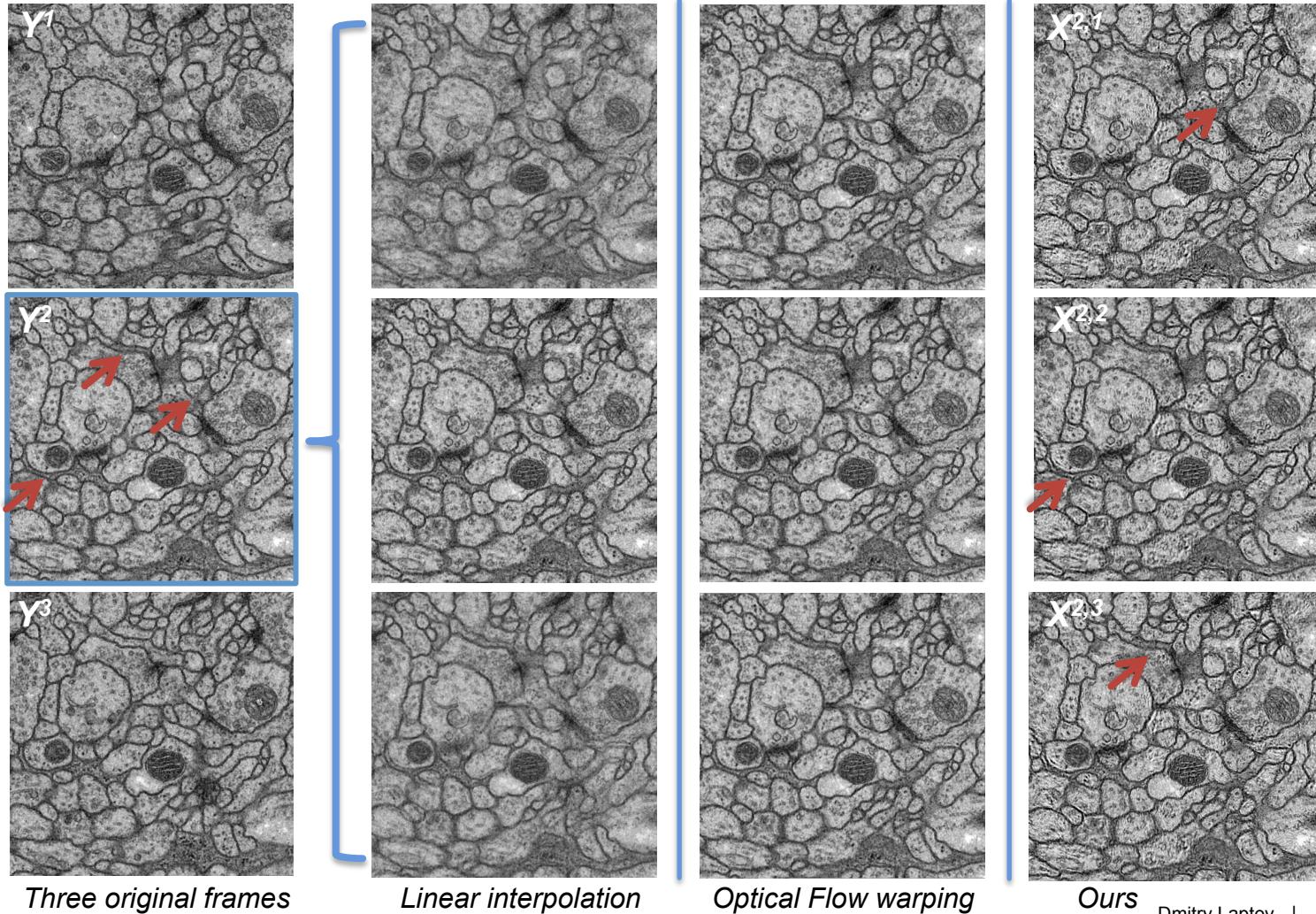
Anisotropic frames



Reconstructed sub-frame



# SuperSlicing experiments: ssTEM



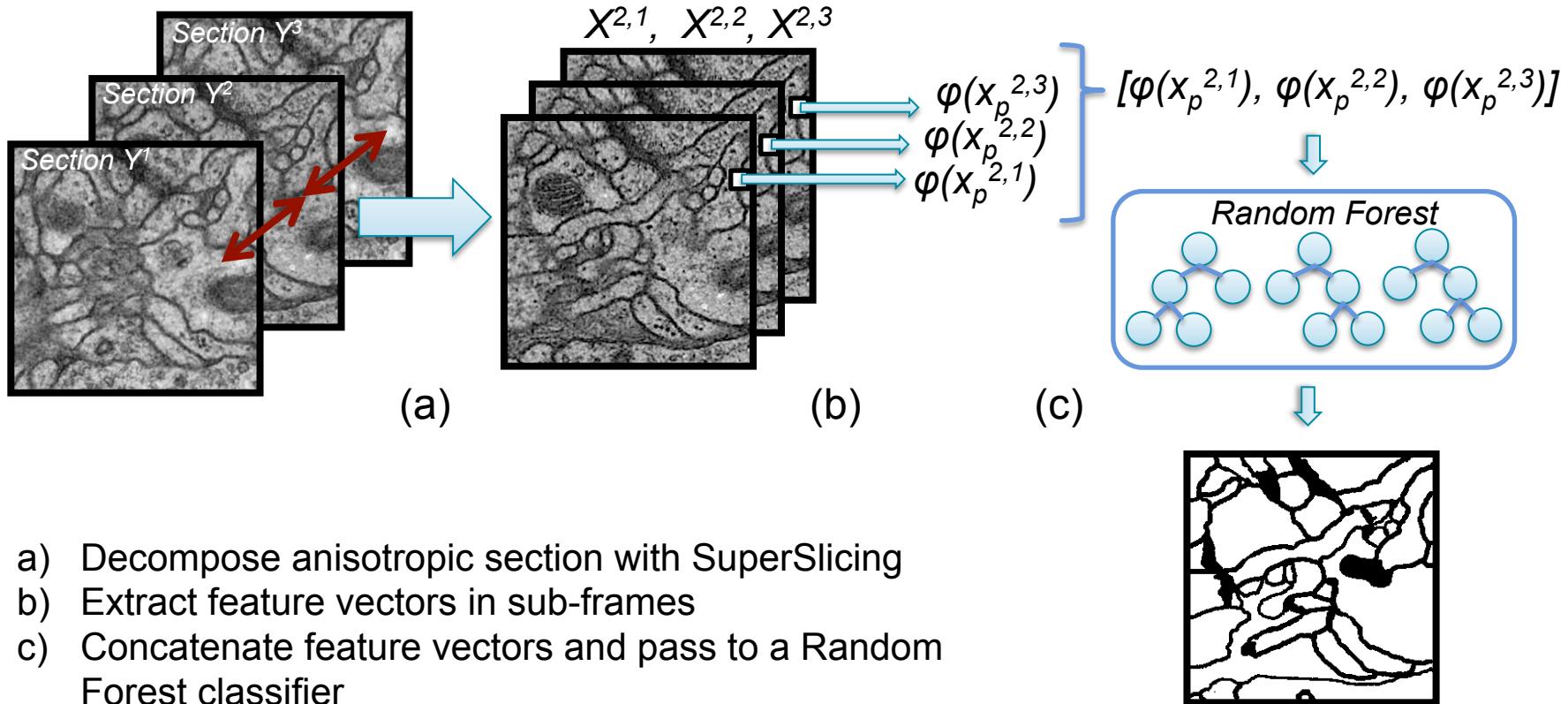
Three original frames

Linear interpolation

Optical Flow warping

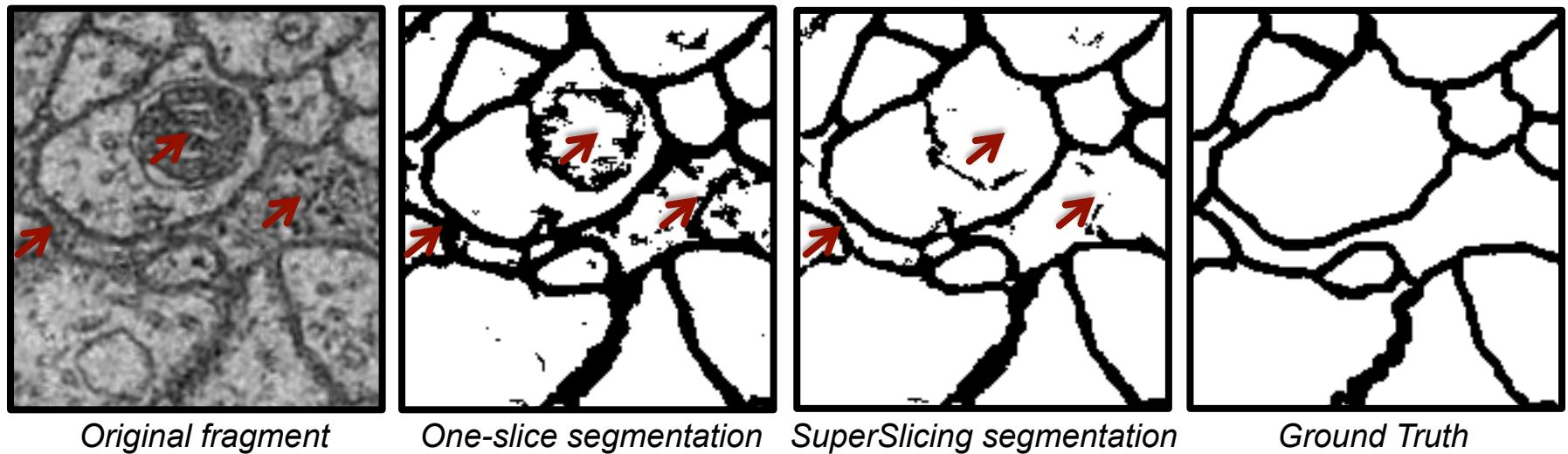
Ours

# SuperSlicing segmentation



- a) Decompose anisotropic section with SuperSlicing
- b) Extract feature vectors in sub-frames
- c) Concatenate feature vectors and pass to a Random Forest classifier

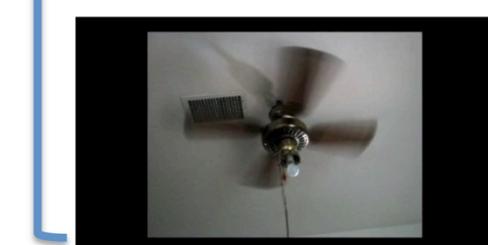
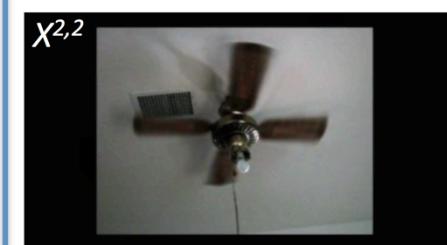
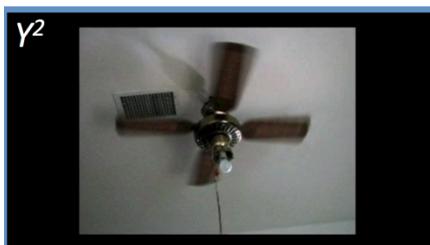
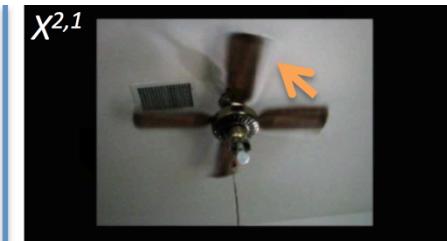
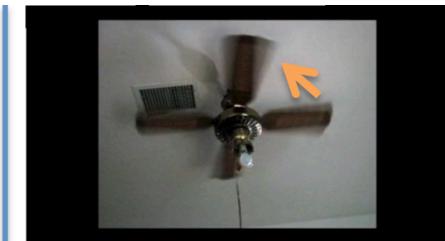
# SuperSlicing segmentation



Method	Warping error*
One section segmentation	$2.87 \times 10^{-3}$ (17%)
Three section segmentation	$2.69 \times 10^{-3}$ (11%)
SuperSlicing segmentation	$2.38 \times 10^{-3}$

\*the metric of topological inconsistency

# SuperSlicing experiments: video



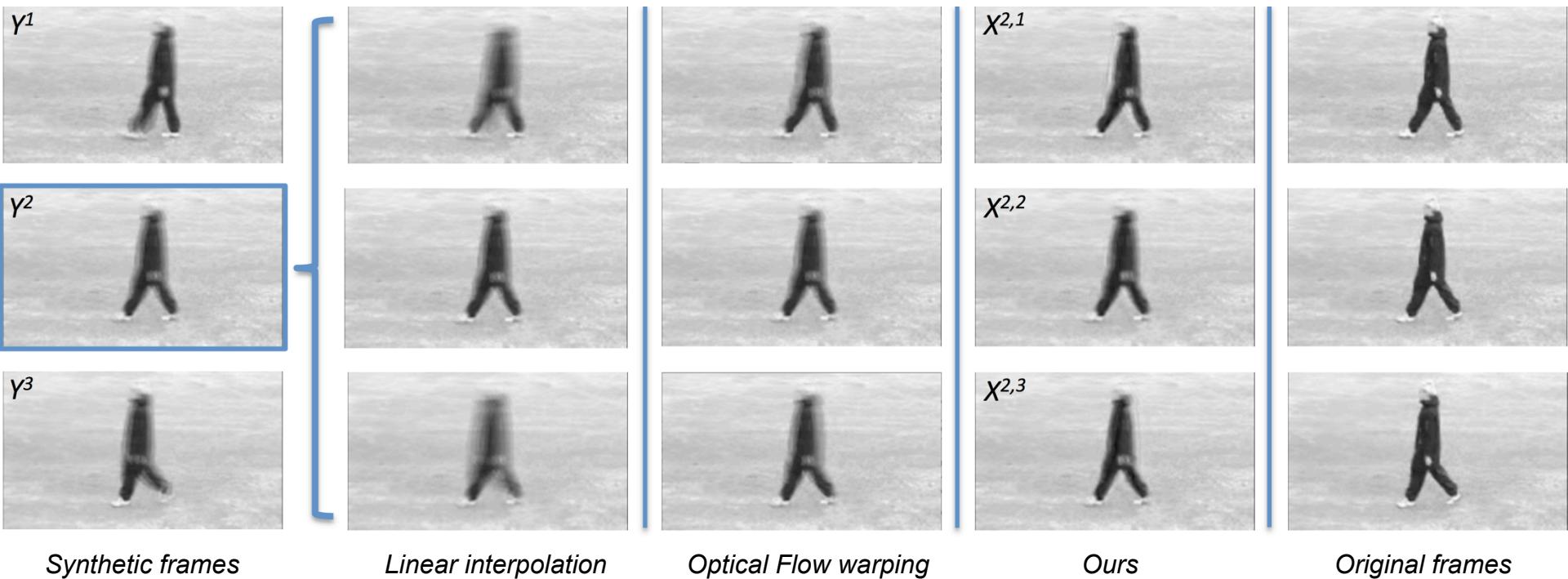
Three original frames

Linear interpolation

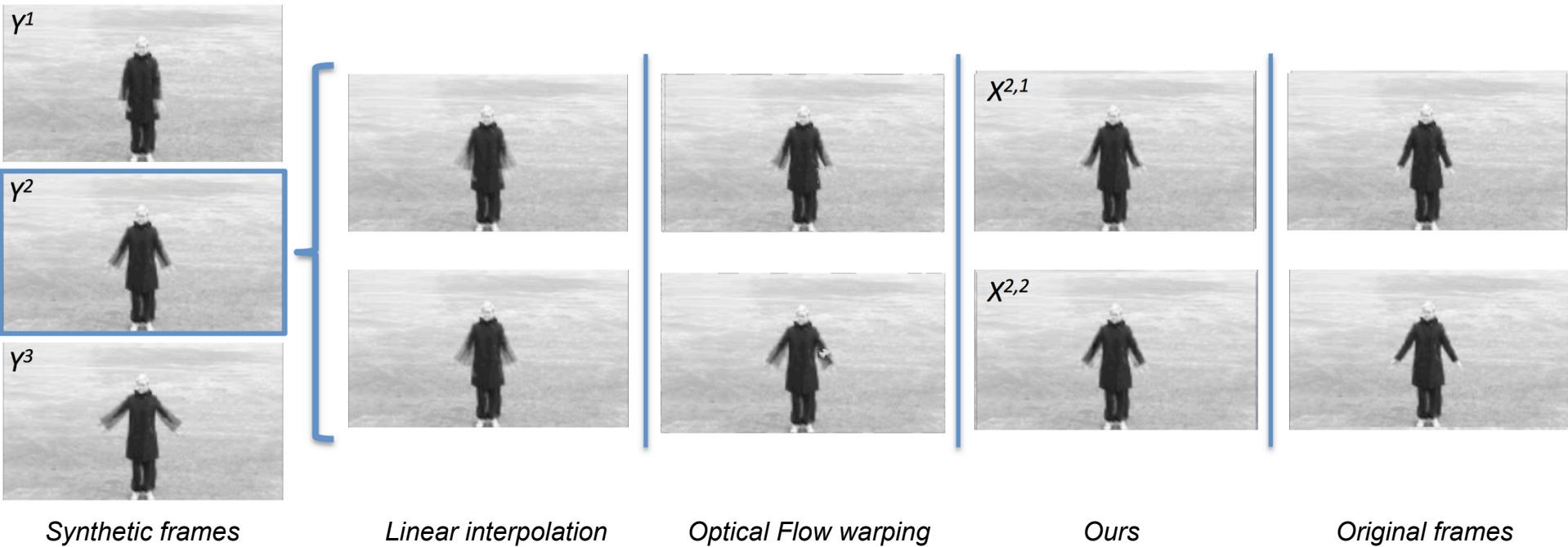
Optical Flow warping

Ours

# SuperSlicing experiments: synthetic video



# SuperSlicing experiments: synthetic video



*Synthetic frames*

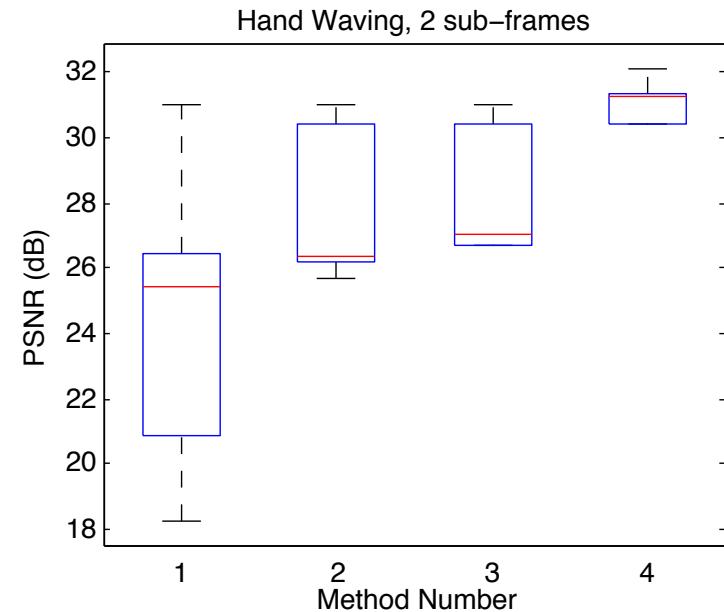
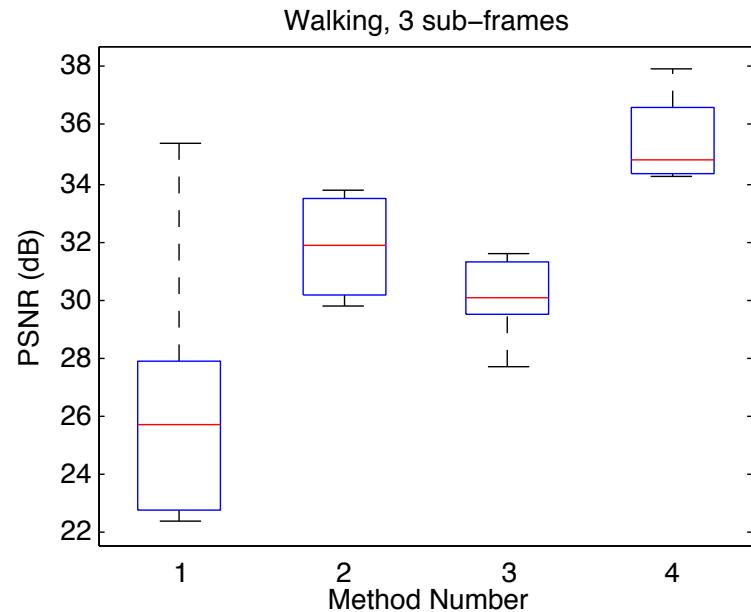
*Linear interpolation*

*Optical Flow warping*

*Ours*

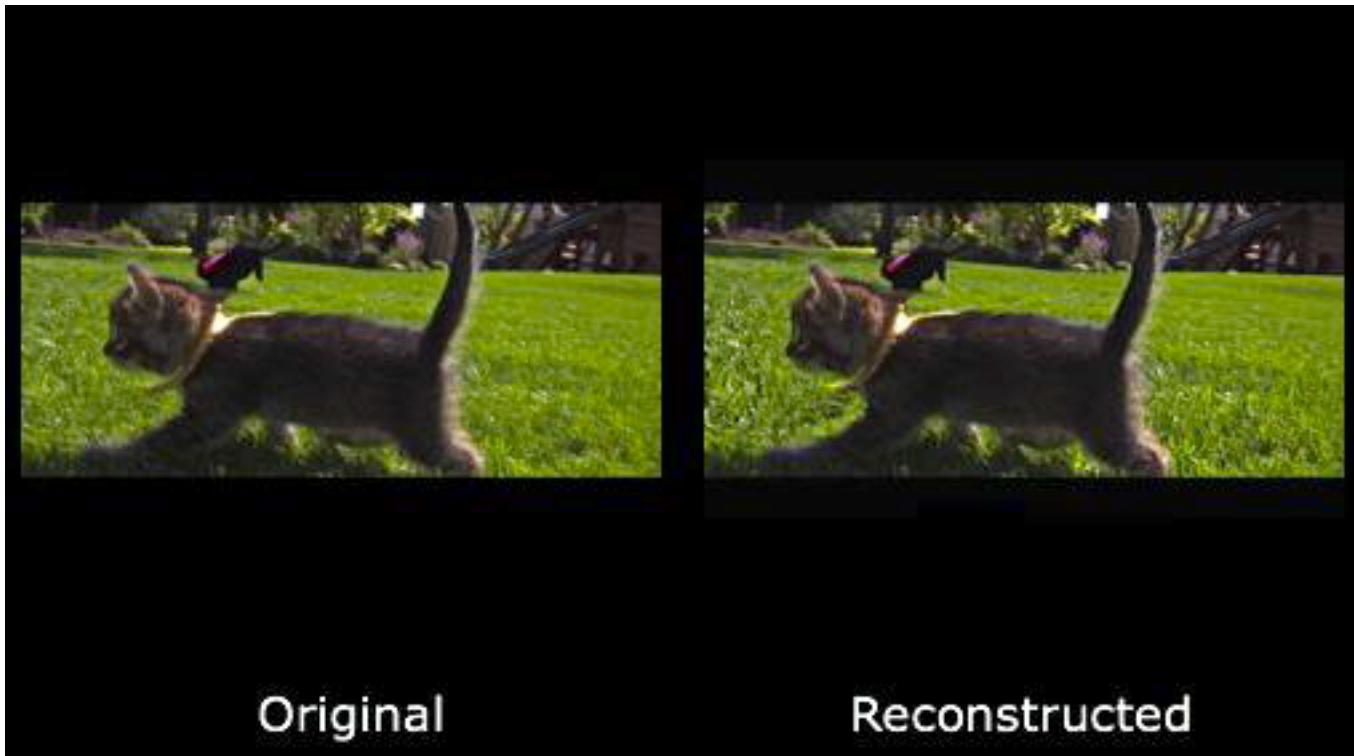
*Original frames*

# SuperSlicing experiments: synthetic video



Significantly more consistent reconstruction!

## Bonus video



See [http://dlaptev.org/other/SuperSlicing\\_demo.avi](http://dlaptev.org/other/SuperSlicing_demo.avi)

# Summary

- SuperSlicing decomposition:
  - Exploit the correspondences
  - Reconstruct, not interpolate
  - Good for visualization and processing
- SuperSlicing segmentation:
  - Apply sub-frame decomposition
  - Use concatenated feature vectors for training

**Thanks for your attention!**

Questions / ideas are welcome

# Based on previous works

1. Laptev, D.; Veznevets, A; Buhmann, J.M., "Superslicing frame restoration for anisotropic ssTEM", *2014 IEEE 11th International Symposium on Biomedical Imaging (ISBI)*, pp.1198-1201, doi: 10.1109/ISBI.2014.6868090
2. Laptev, D.; Veznevets, A; Buhmann, J.M., "Anisotropic ssTEM Image Segmentation Using Dense Correspondence across Sections", *Medical Image Computing and Computer-Assisted Intervention – MICCAI 2012*, doi: 10.1007/978-3-642-33415-3\_40