

# SENSEI: uncovering neuron structure using hard and soft approaches

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HBP Partnering Projects Meeting: Status quo & outlook

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the European Union

# SENSEI Goal

Obtaining neuronal morphology at different spatial levels from optical imaging techniques

- e.g. dendritic tree & subcellular structures/spines

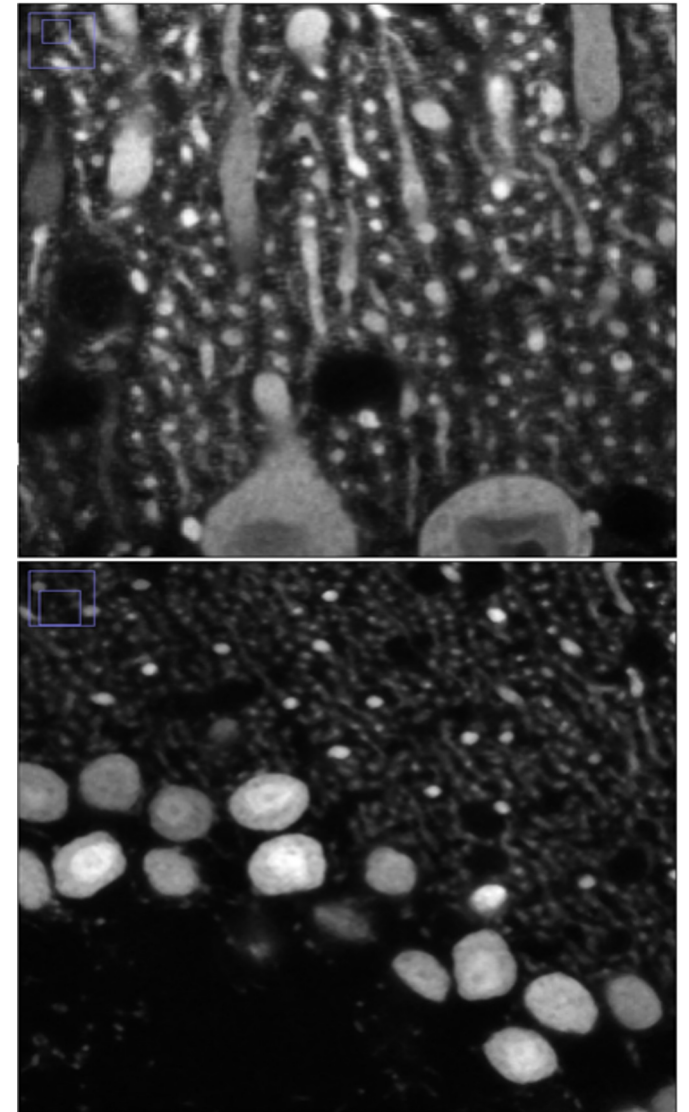
*Tools:*

Tissue Processing and Imaging

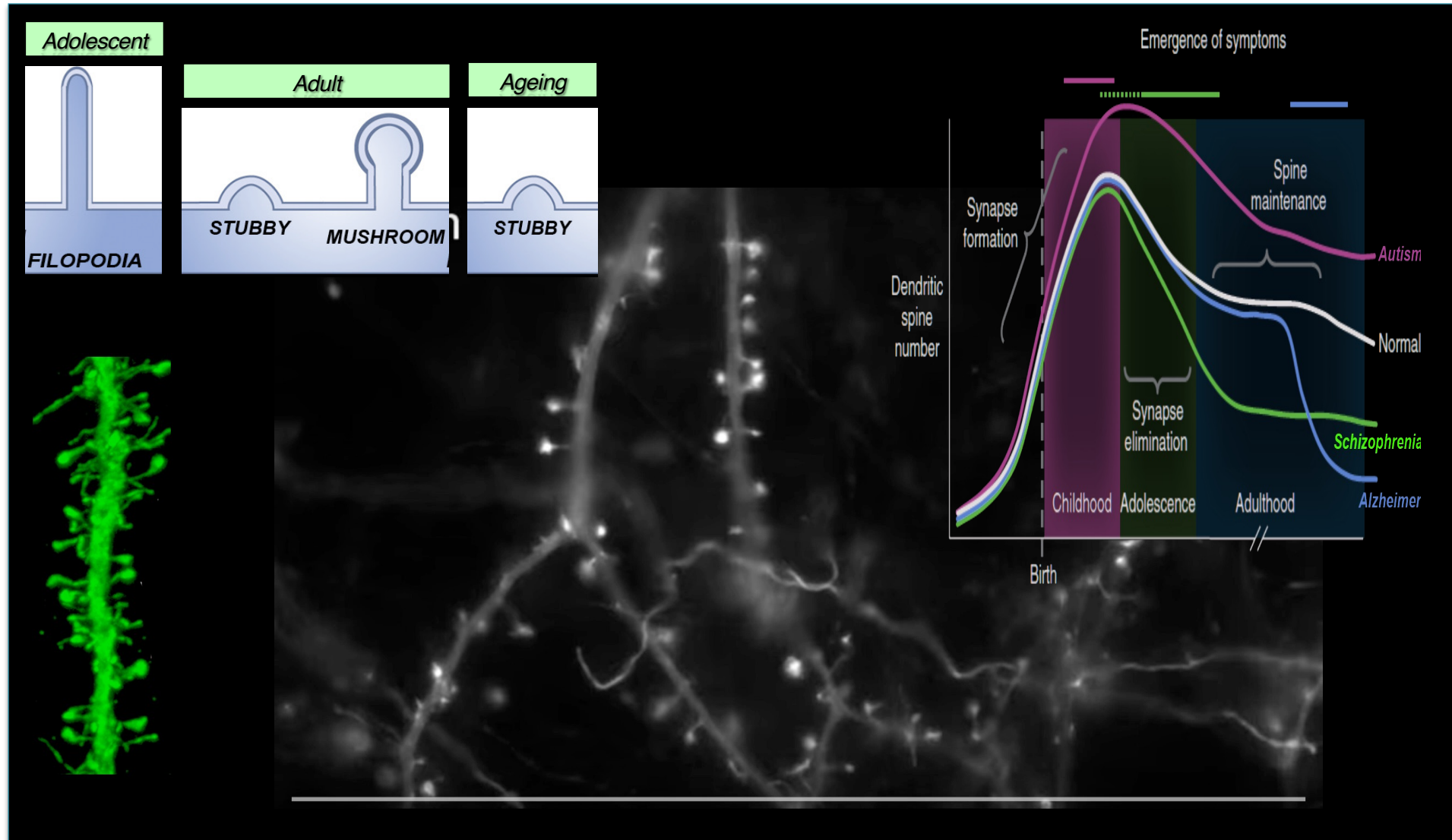
Segmentation algorithms

- model based approaches
- k-means exploiting topological information

TriScan: very fast 3D imaging fluorescence microscope



# Dendritic spine morphology is associated to learning & ageing



Maria Fischer, et al, Rapid Actin-Based Plasticity in Dendritic Spines, Neuron, Volume 20, Issue 5, 1998

Peter Penzes et al. Dendritic spine pathology in neuropsychiatric disorders Nat Neurosci. 2011 Mar; 14(3): 285–293.



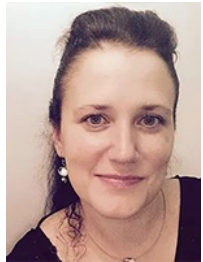
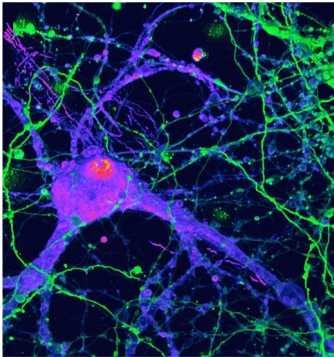
# SENSEI Team



## SEgmentation of Neurons using Standard and supEr-resolution microscopy



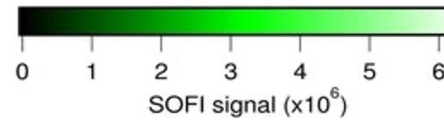
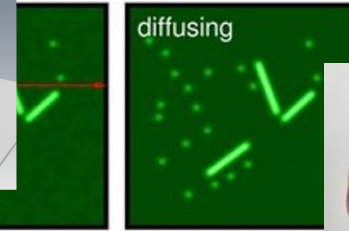
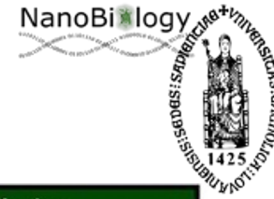
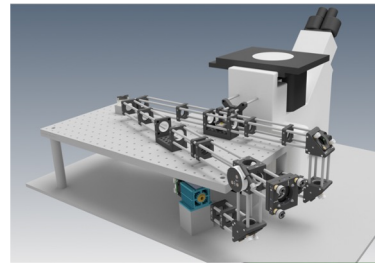
Labelling & Microscopy



Lydia Danglot



Triscan Device  
3D SOFI



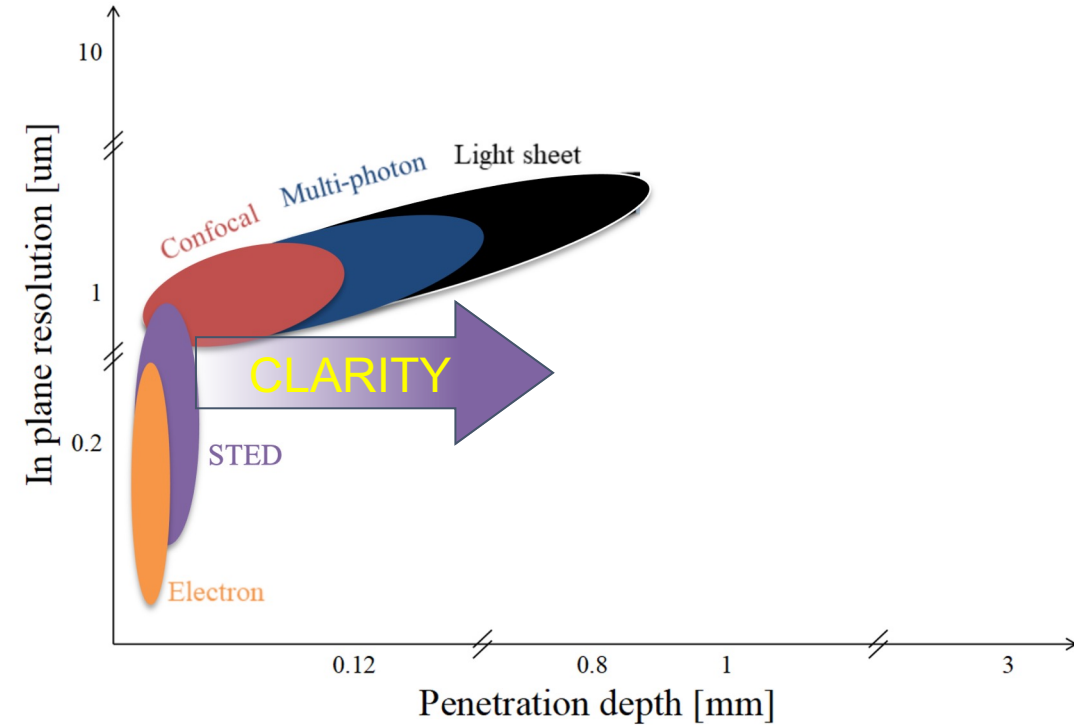
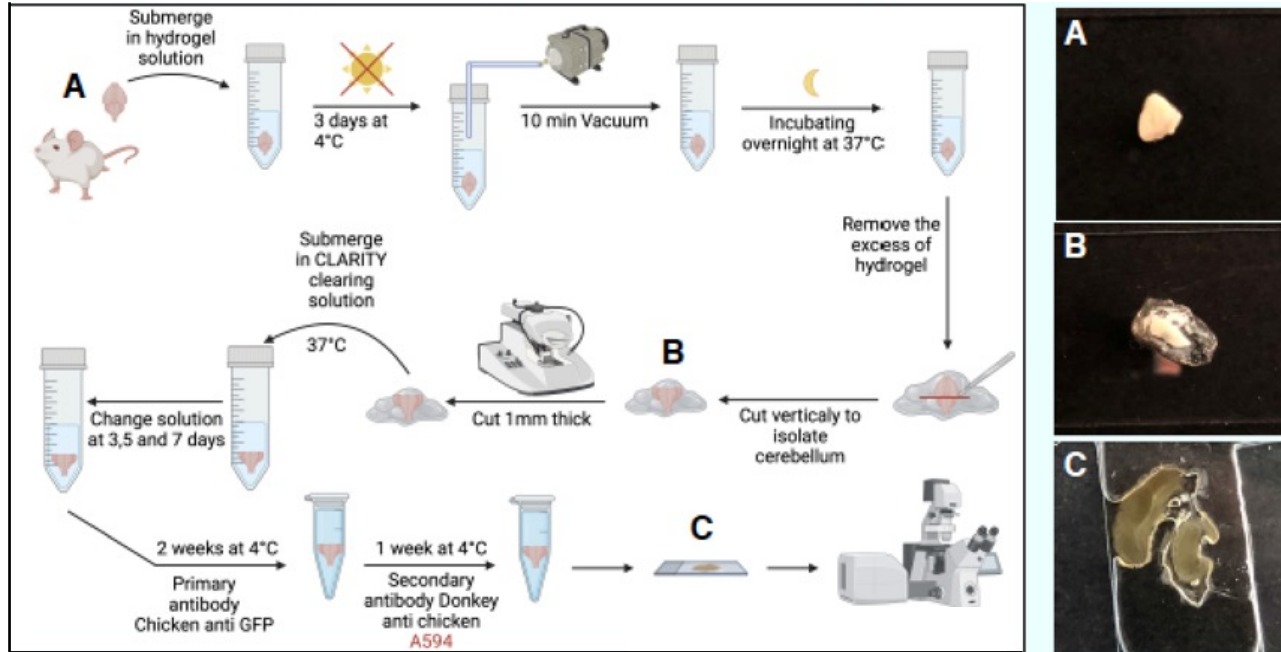
Peter Dedecker

Segmentation algorithms

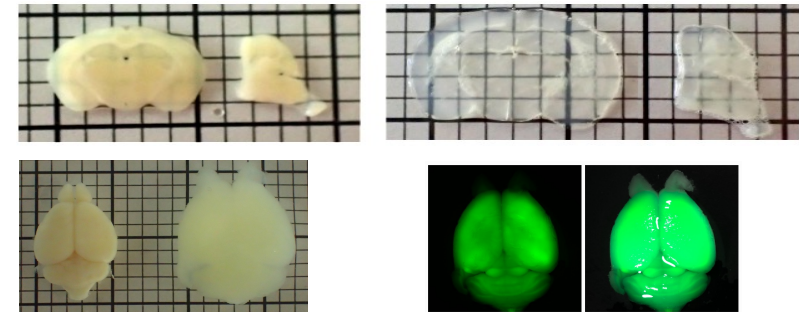
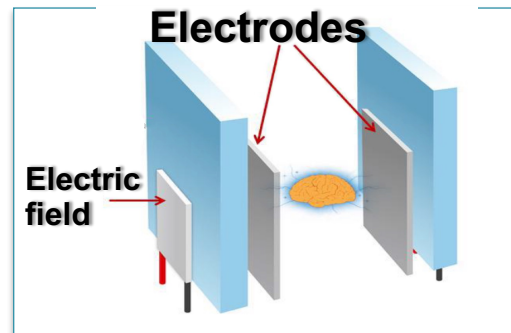
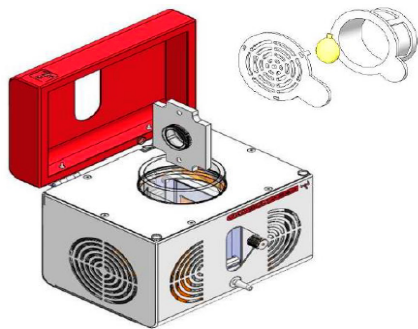




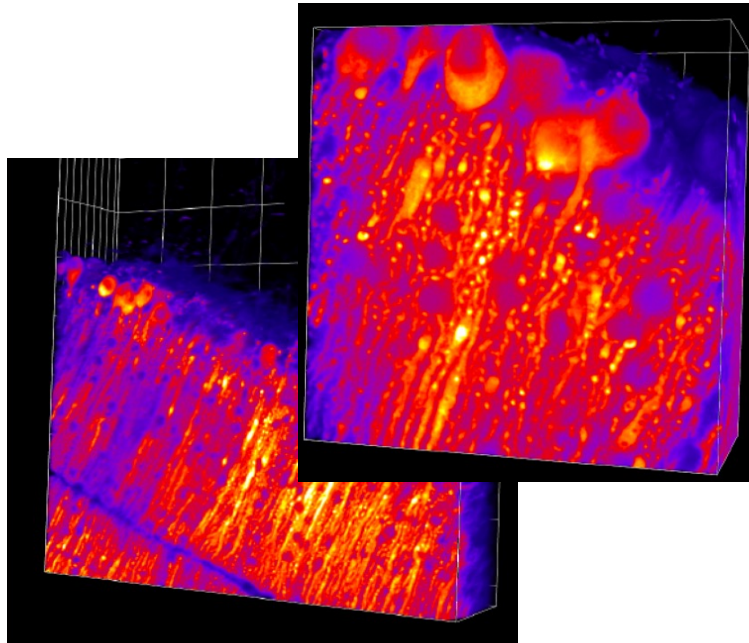
# Labeling Strategy to decipher neural shape within tissue slices



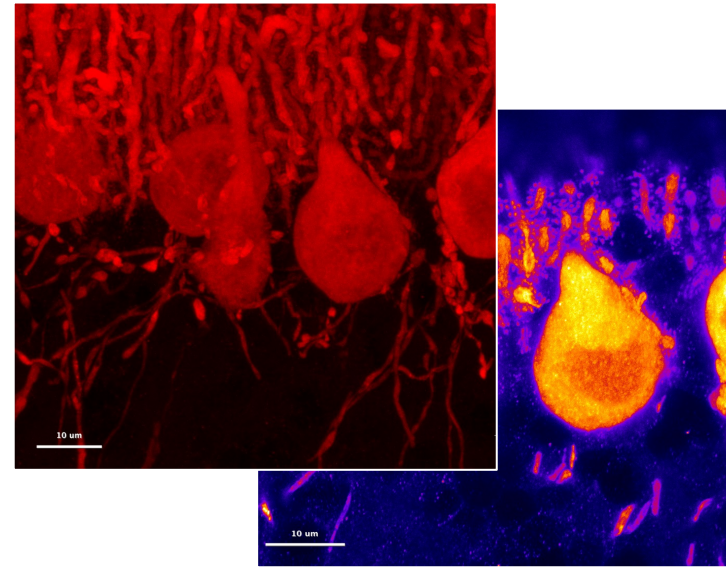
## Tissue Clarification



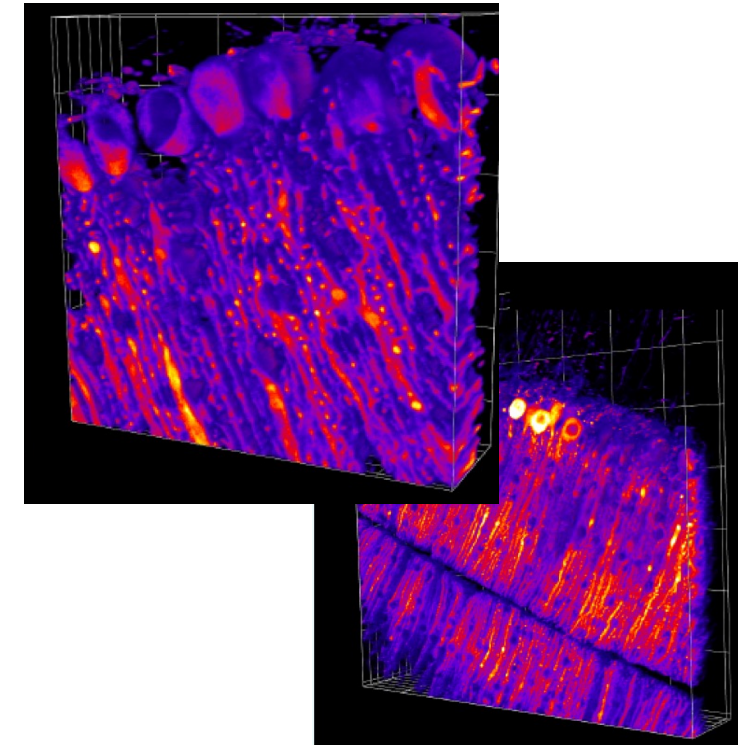
# Acquisitions with different microscopy modalities



Spinning disk  
20x 63x



Confocal Leica/3D sted  
93x

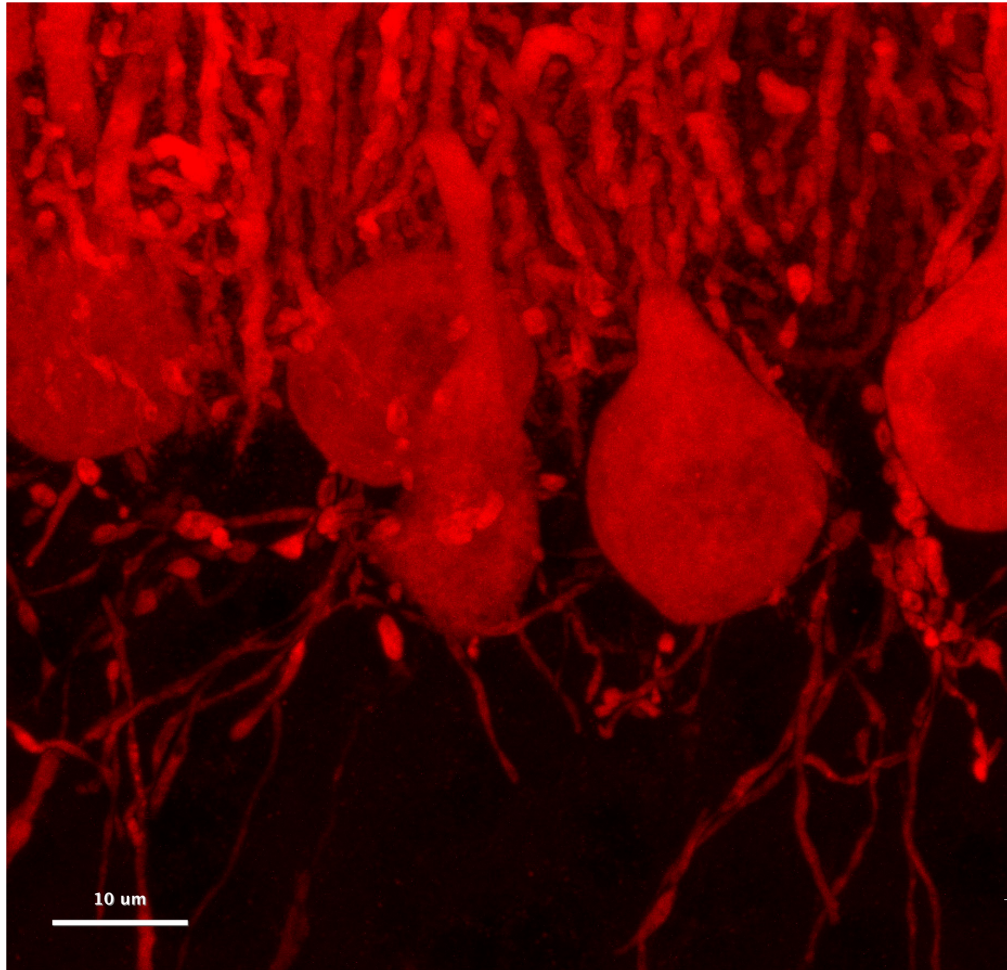


Confocal Leica  
63x/25x

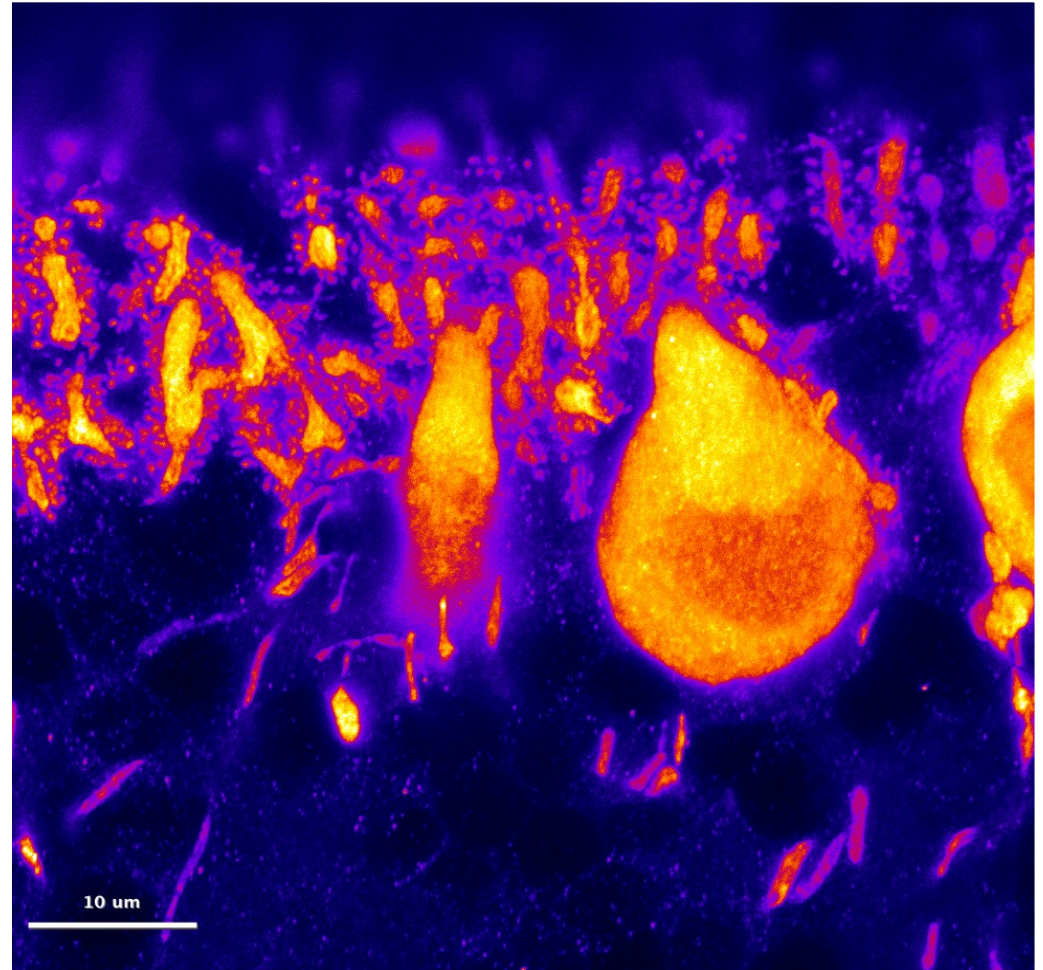
Images acquired with different modalities are acquired and will be analyzed with proposed segmentation approaches



# Acquisitions with different microscopy modalities



Confocal 93x- 3D projection



3D STED 93x – 3D projection  
spine morphology even in thick slices (500  $\mu\text{m}$ )

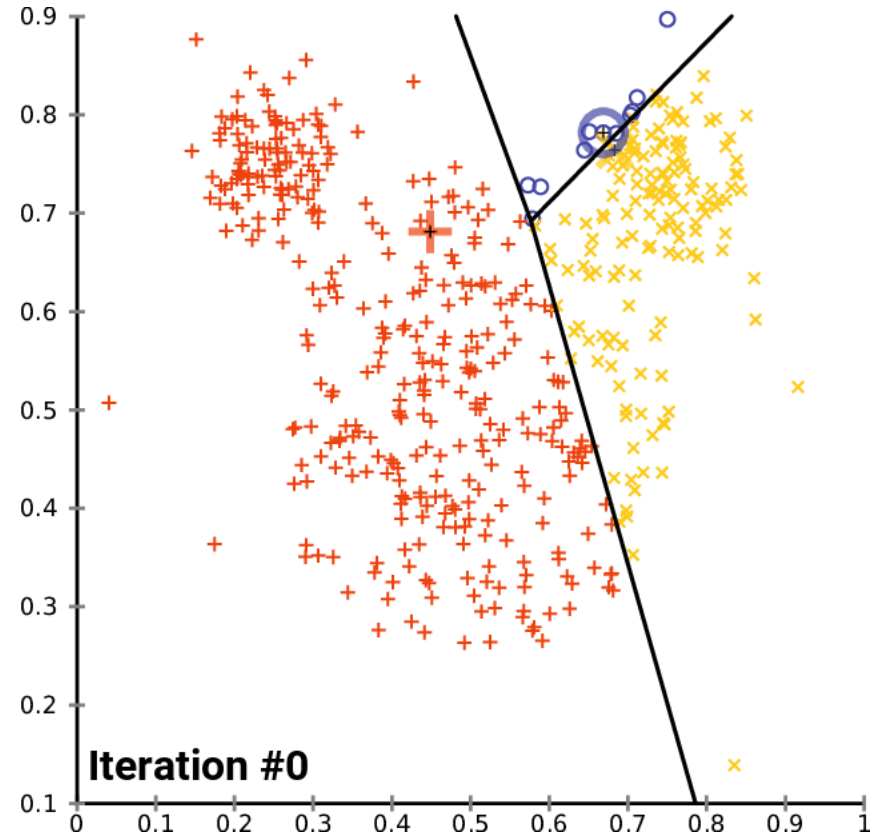


# SENPAI: a topological informed data driven approach

A k-means algorithm exploiting spatial derivatives

K-means clustering, needs

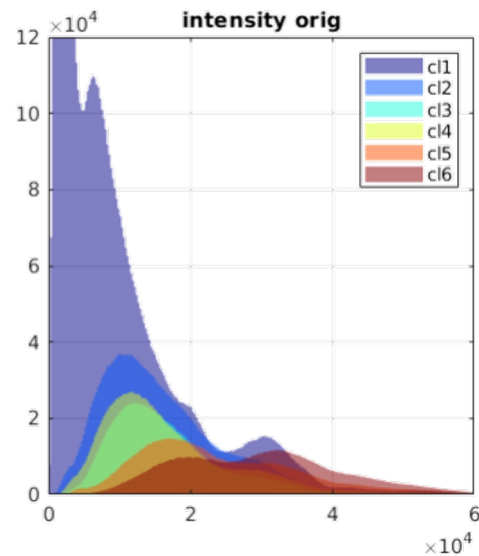
- the number of possible classes
- a definition of distance
- the criterion (i.e. minimization of within cluster variance)



[https://commons.wikimedia.org/wiki/File:K-means\\_convergence.gif](https://commons.wikimedia.org/wiki/File:K-means_convergence.gif)

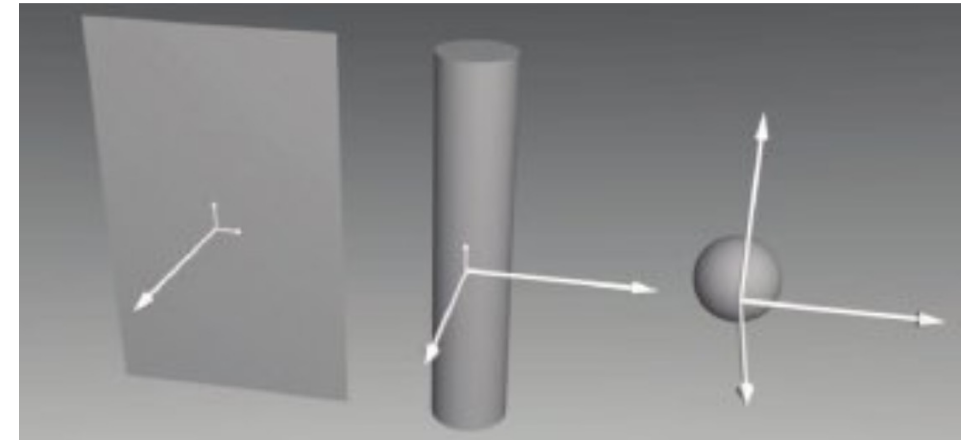
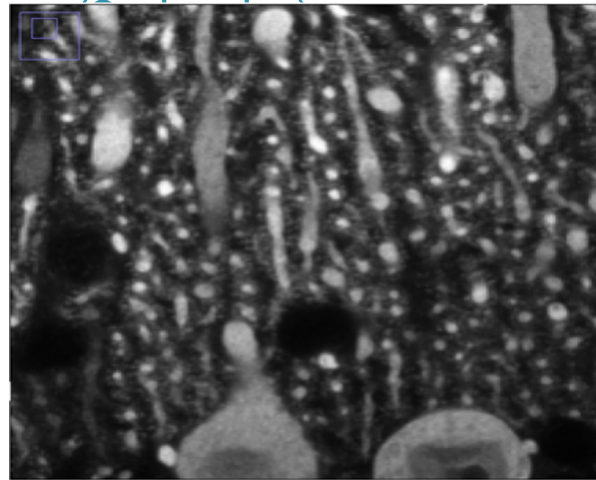
# SENPAI: a topological informed data driven approach to neuronal reconstruction

A k-means algorithm exploiting spatial derivatives



Intensity distribution of pixel classes; they overlap

If we use spatial derivatives



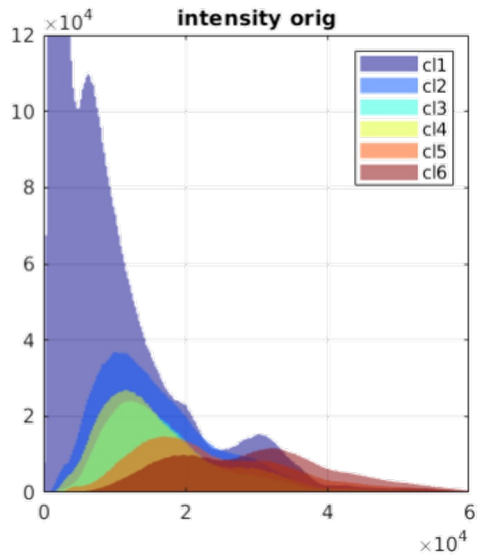
The analysis with Hessian allows to describe different spatial distributions

# SENPAI: a topological informed data driven approach to neuronal reconstruction

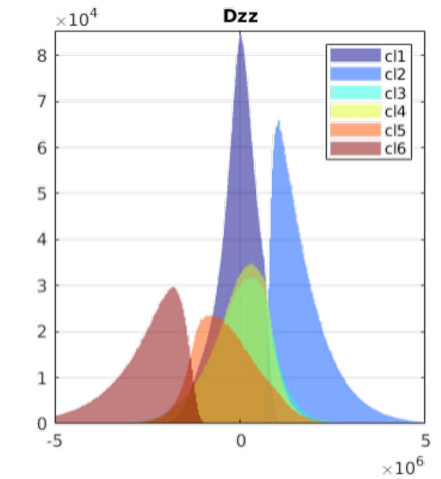
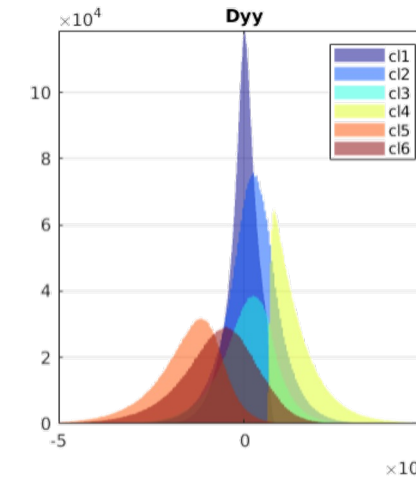
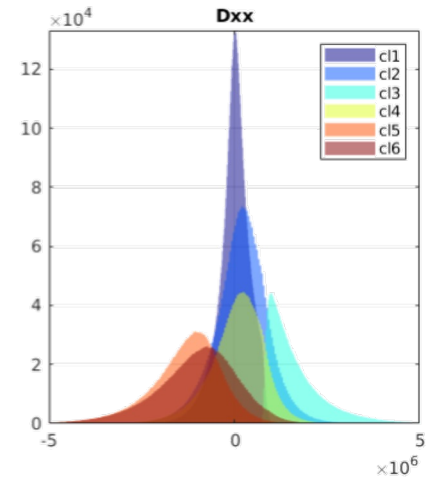
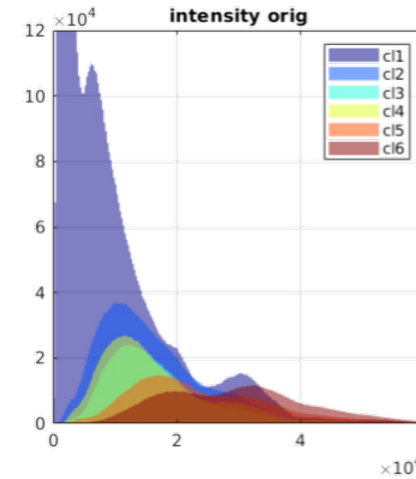
A k-means algorithm exploiting spatial derivatives

If we use spatial derivatives (2nd order)...

we can distinguish neuron-related classes



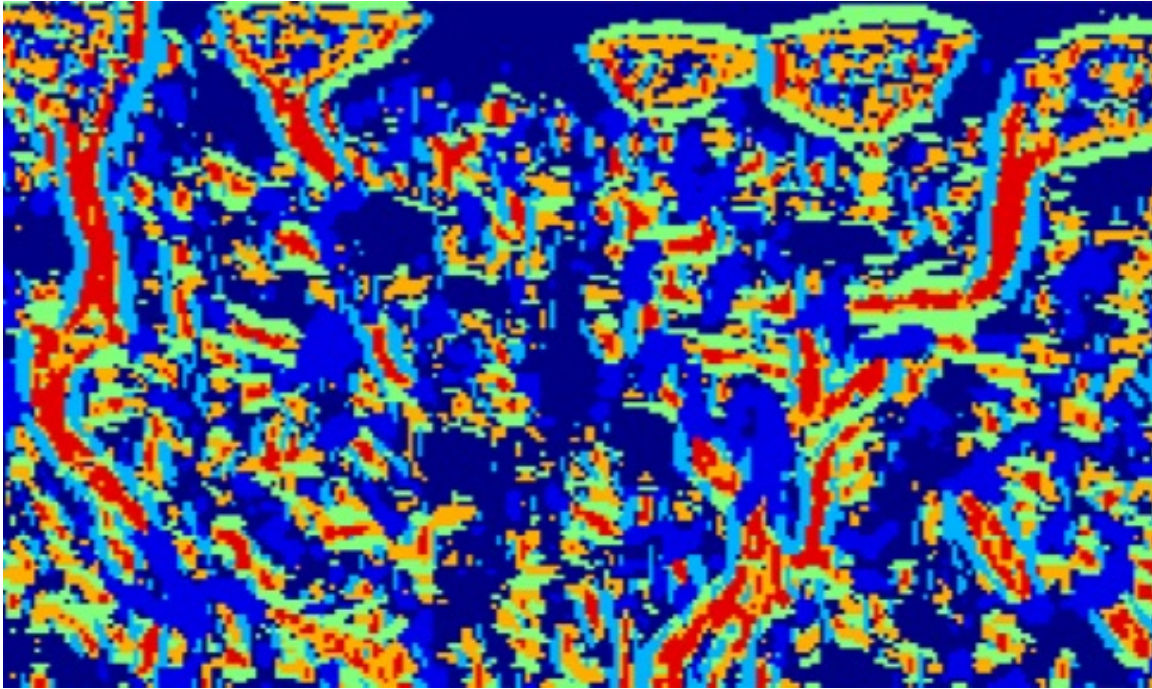
Intensity distribution of pixel classes; they overlap





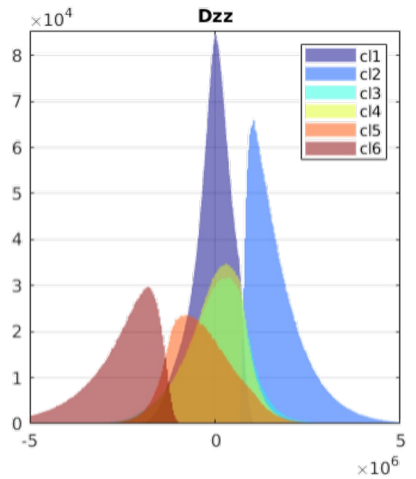
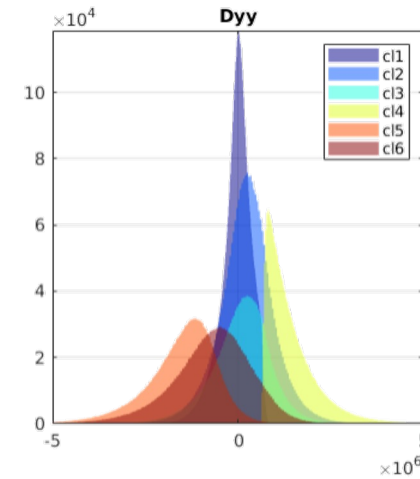
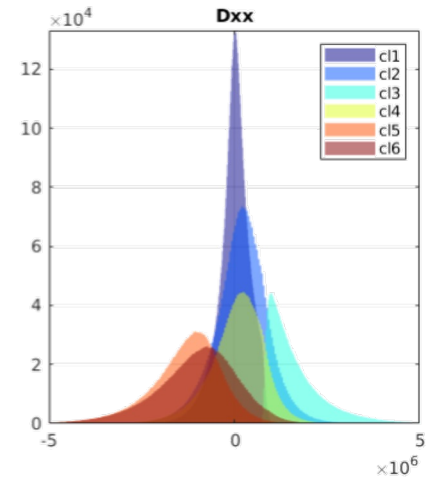
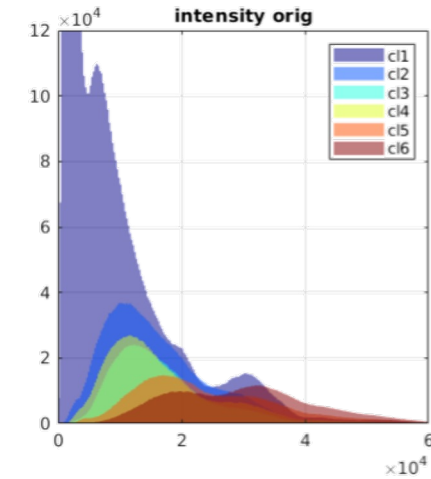
# SENPAI: a topological informed data driven approach to neuronal reconstruction

A k-means algorithm exploiting spatial derivatives



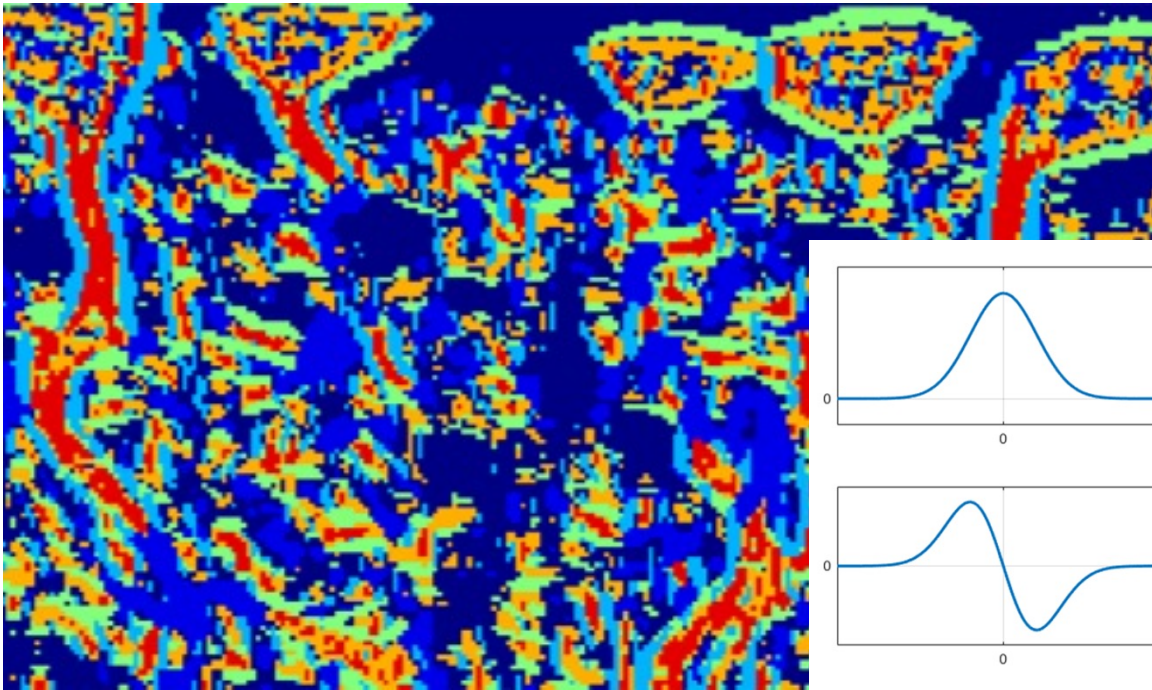
Classes are chosen according to:

- negative mean of the derivatives
- high intensity



# SENPAI: a topological informed data driven approach to neuronal reconstruction

A k-means algorithm exploiting spatial derivatives

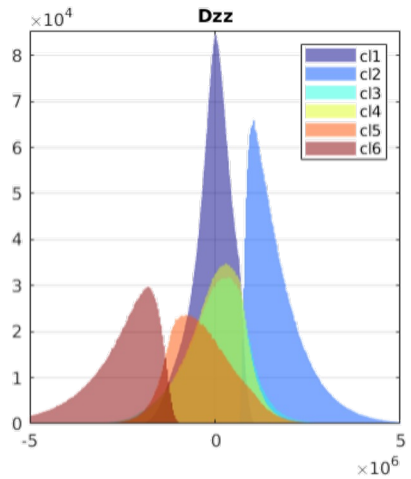
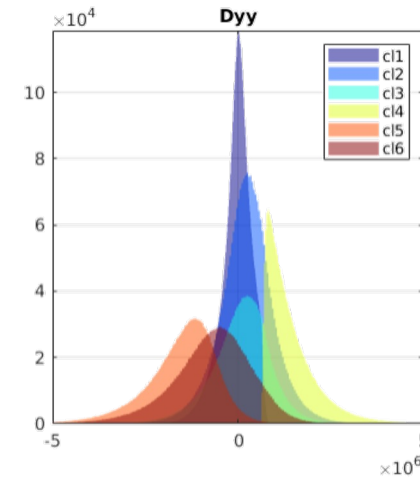
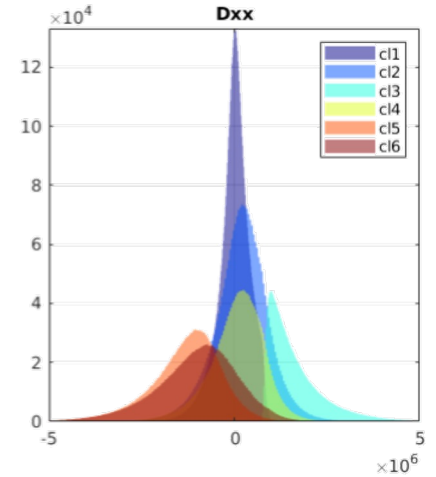
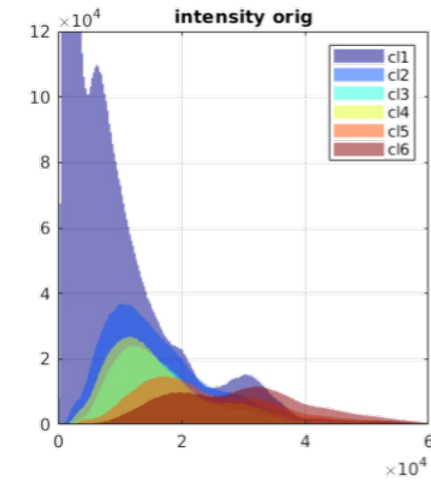


1) Profile of a dendritic branch with smooth transitions towards the background.

2) The first derivative is directionality-dependent.

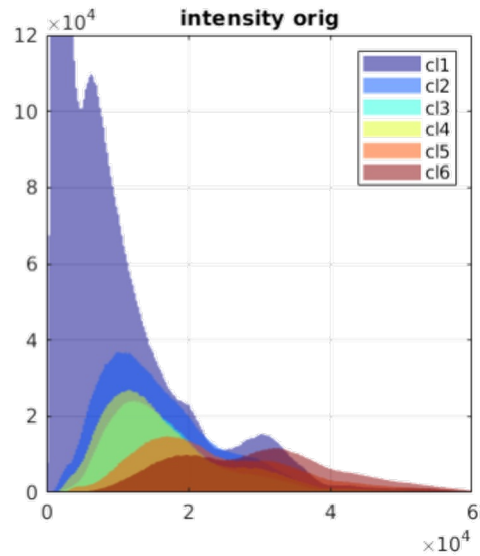
3) The second derivative encodes all the inner edges of the transition with negative values.

- Classes are chosen according to:
- negative mean of the derivatives
  - high intensity

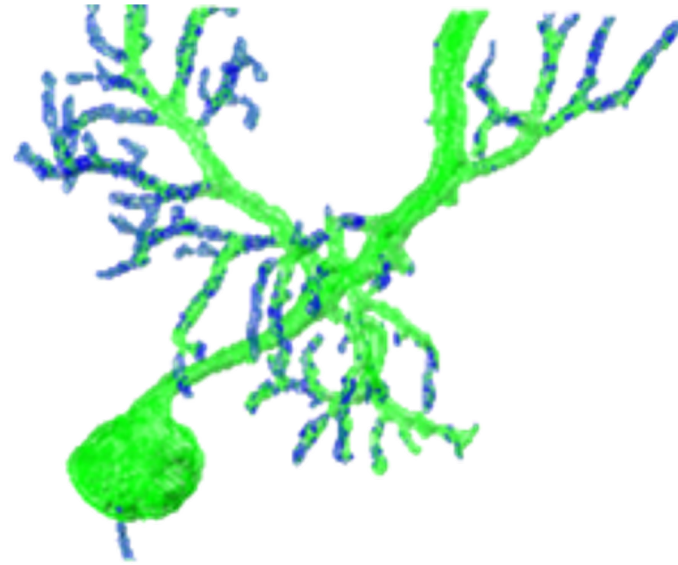


# SENPAI: a topological informed data driven approach to neuronal reconstruction

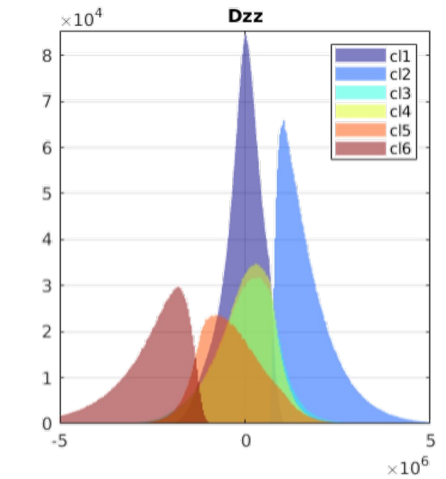
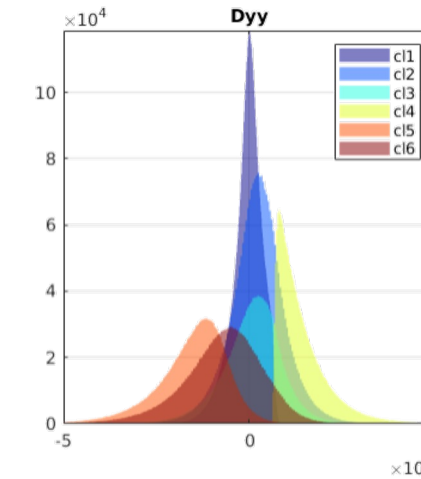
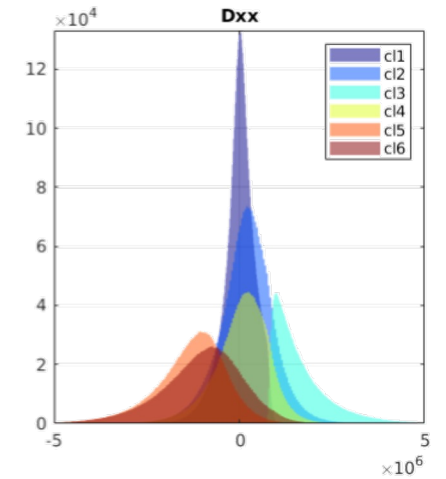
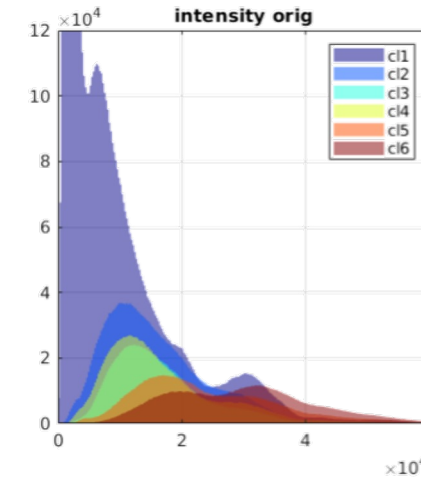
A k-means algorithm exploiting spatial derivatives



Intensity distribution of pixel classes; they overlap



In blue low intensity pixels that were detected by adding topological info

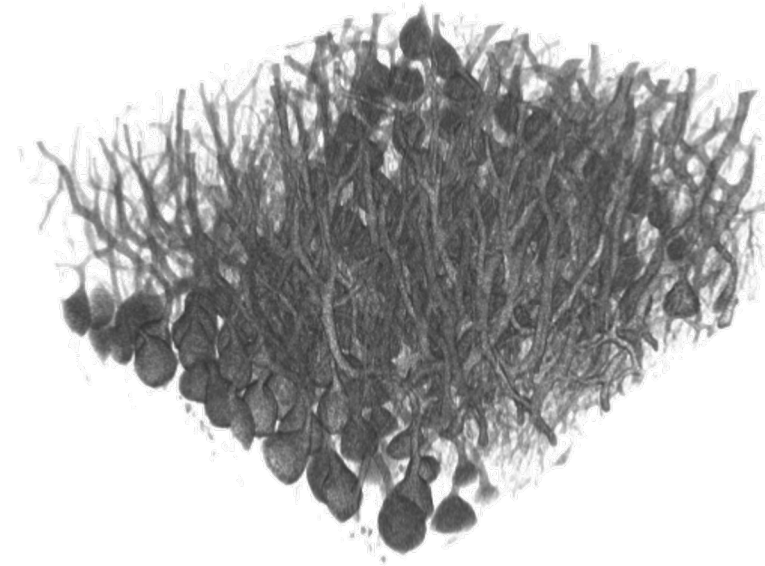
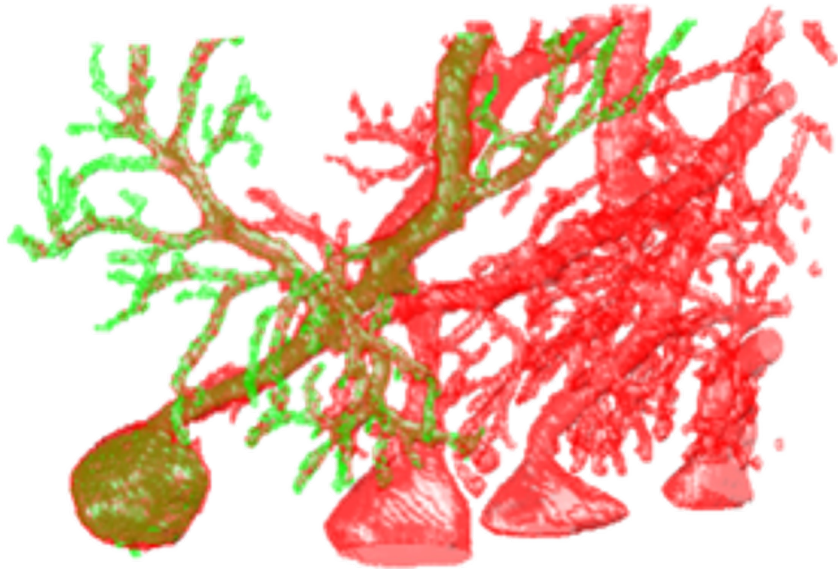




# SENPAI: a topological informed data driven approach

A k-means algorithm exploiting spatial derivatives

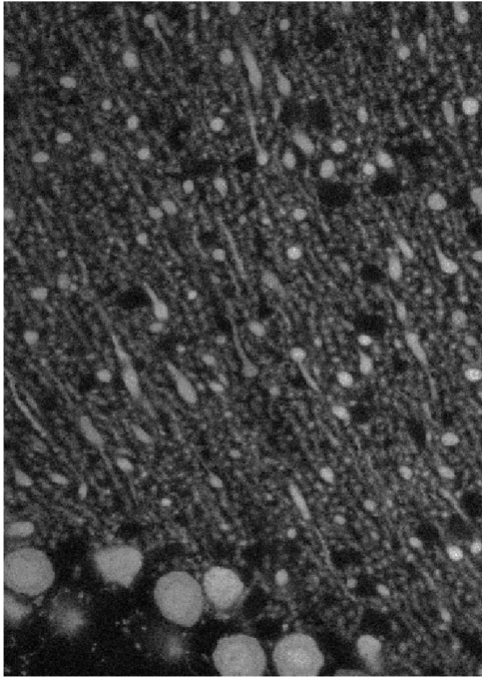
We might fail the identification of single structures in dense images



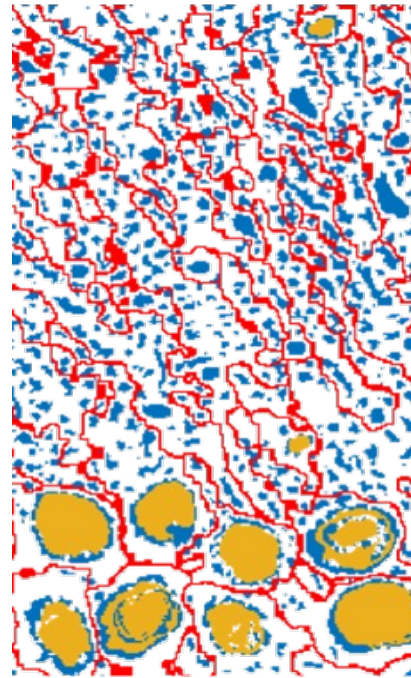
- green could not be detected using intensity
- neurons are easily merged together: single neuron identification is envisaged

# SENPAI: a topological informed data driven approach

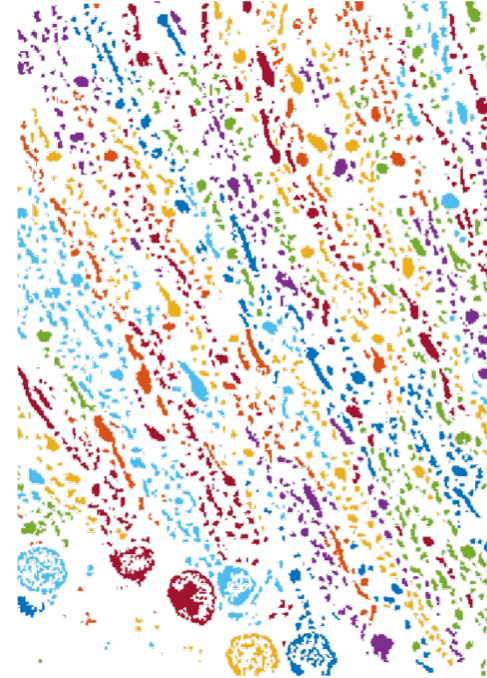
A k-means algorithm exploiting spatial derivatives... and watershed



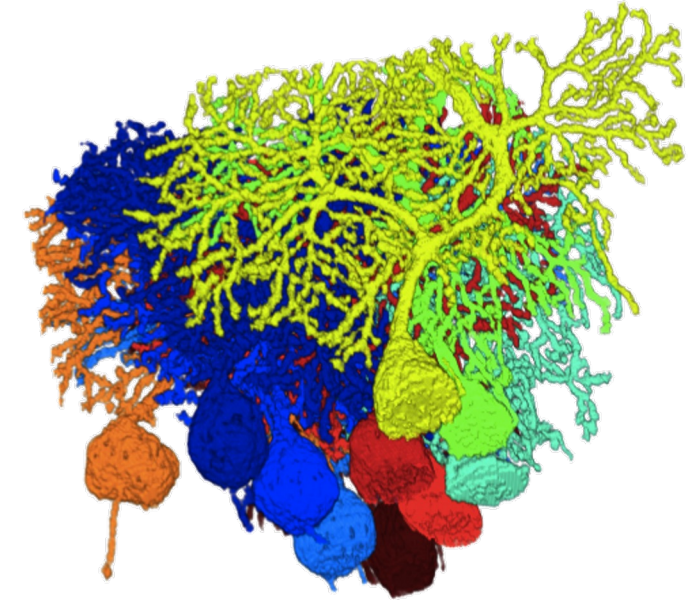
Original image  
40x



Watershed on  
segmented data



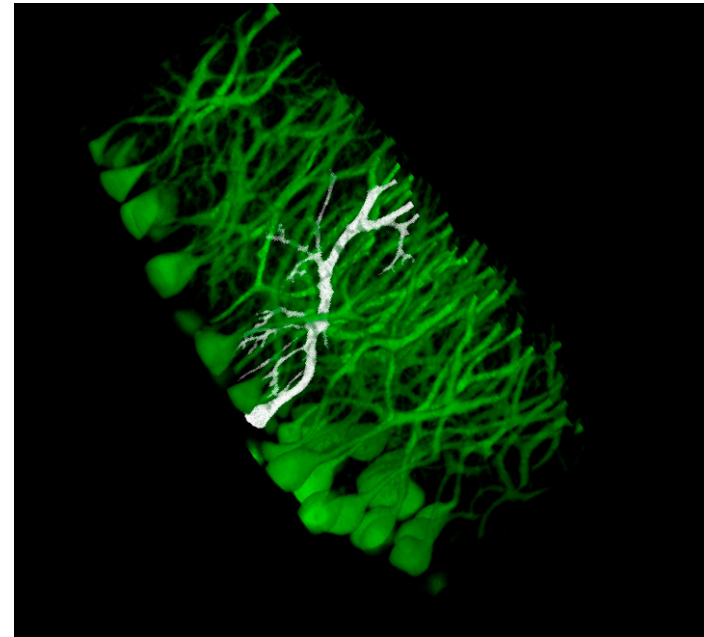
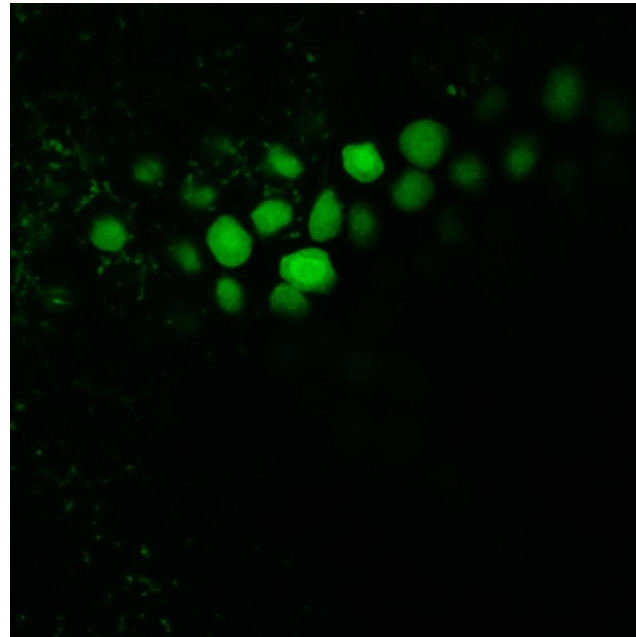
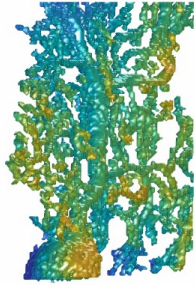
Parcellation



3D recon

# SENPAI: a topological informed data driven approach

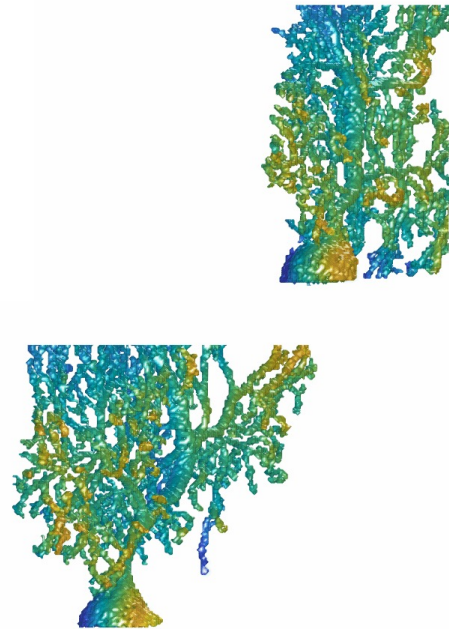
A k-means algorithm exploiting spatial derivatives and watershed



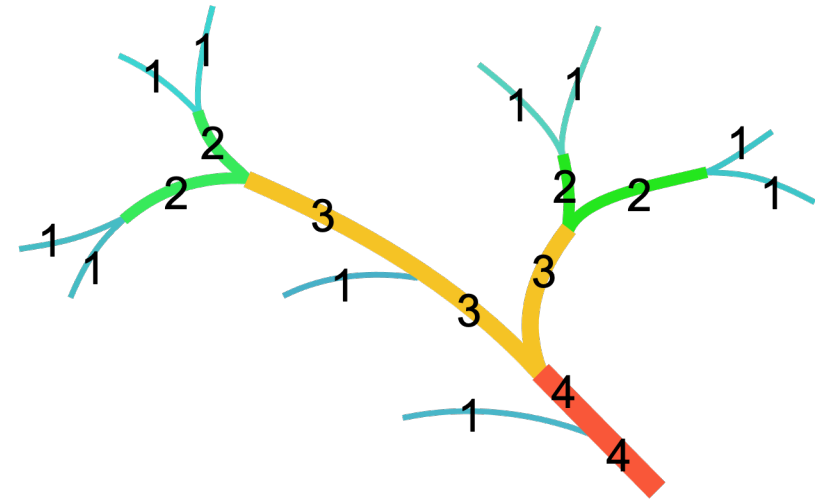


# A (possible) Validation: Strahler ordering

statistical based approach



Strahler ordering is used to describe complexity branching structures

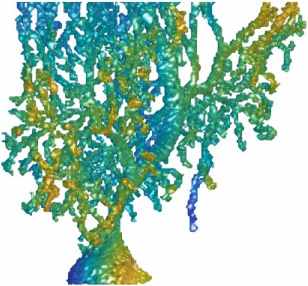
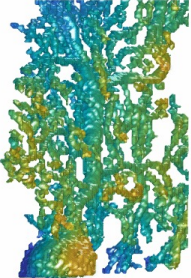


Strahler, A. N. (1957), "Quantitative analysis of watershed geomorphology", Transactions of the American Geophysical Union, 38 (6): 913–920

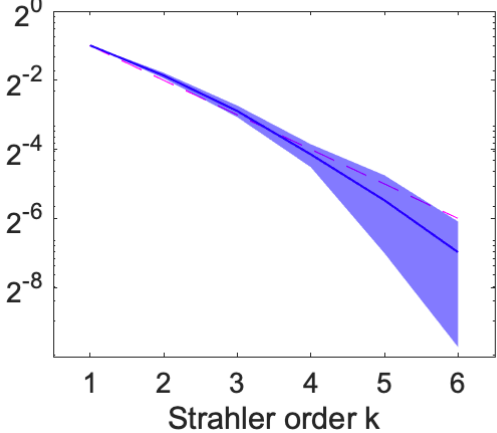
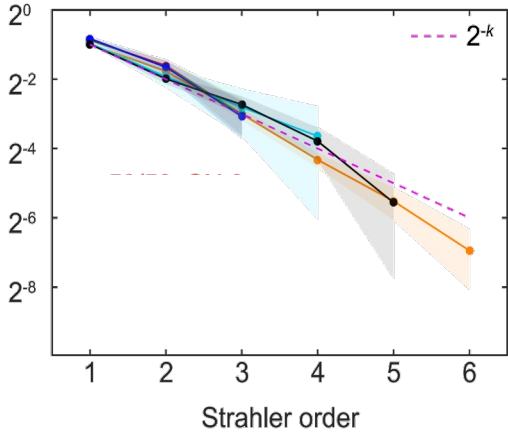
[https://commons.wikimedia.org/wiki/File:Flussordnung\\_\(Strahler\).svg](https://commons.wikimedia.org/wiki/File:Flussordnung_(Strahler).svg)  
Kilom691, CC BY-SA 3.0 <<https://creativecommons.org/licenses/by-sa/3.0/>>, via Wikimedia Commons

# A (possible) Validation: Strahler ordering

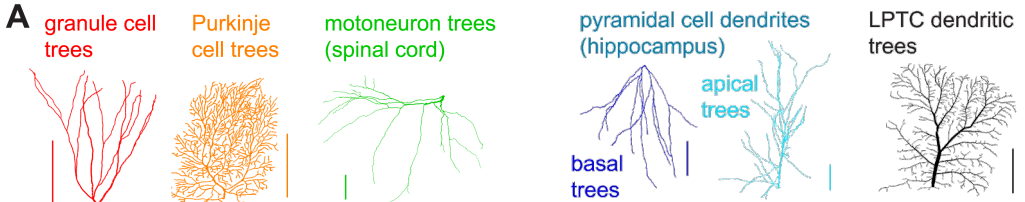
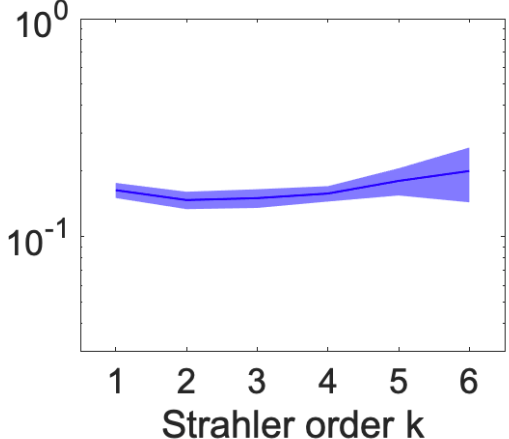
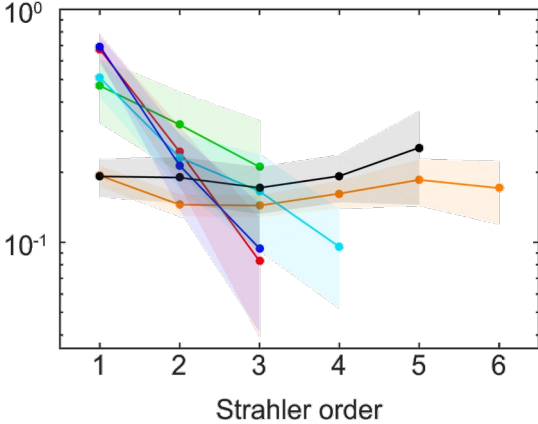
statistical based approach



Segment number



Segment length



Strahler ordering-based statistics for different neural cell types, Vormberg et al., 2017



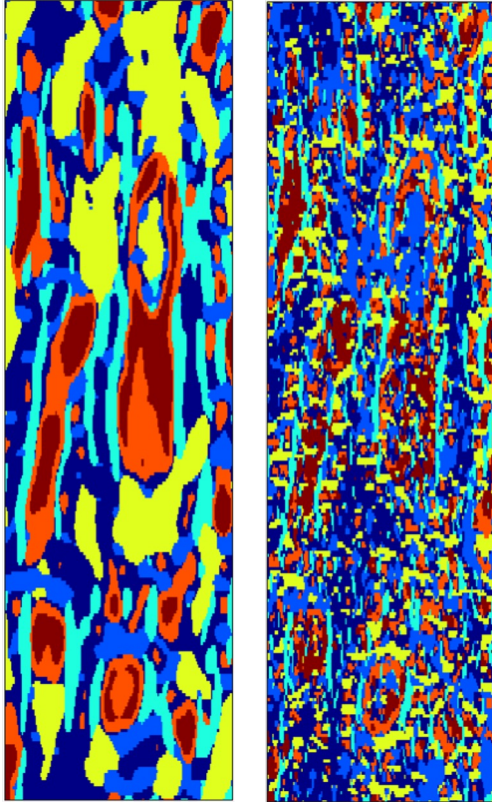
# SENPAI: a topological informed data driven approach to neuronal reconstruction

Spine segmentation

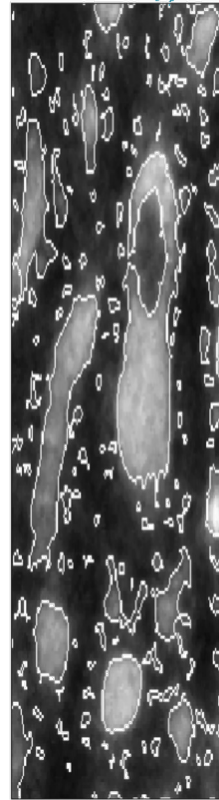
Original Image (93x)



Topology-informed k-means  
At different levels of smoothing



Final result  
(segmentation in white edges)



→  
Class selection  
+  
Level Merging

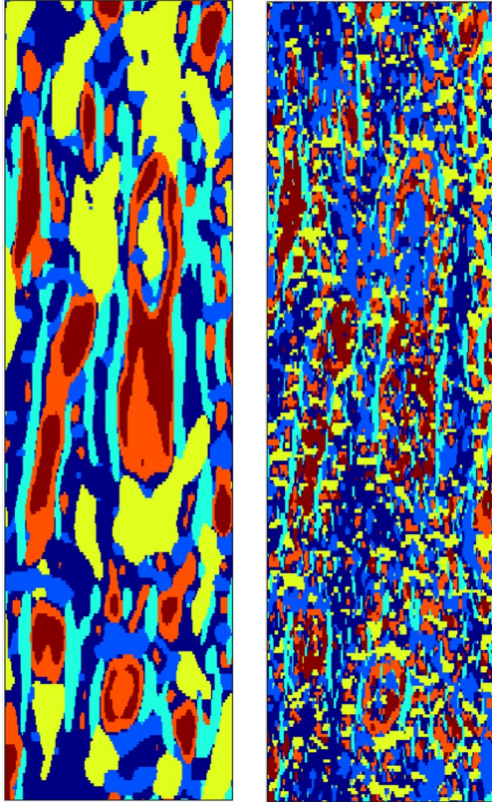
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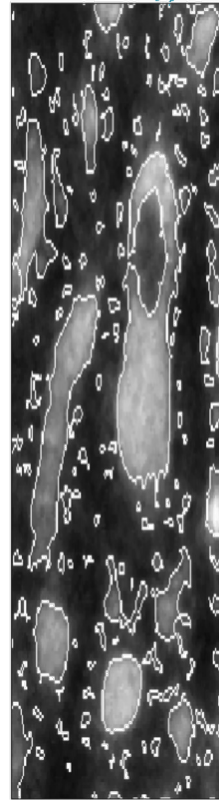
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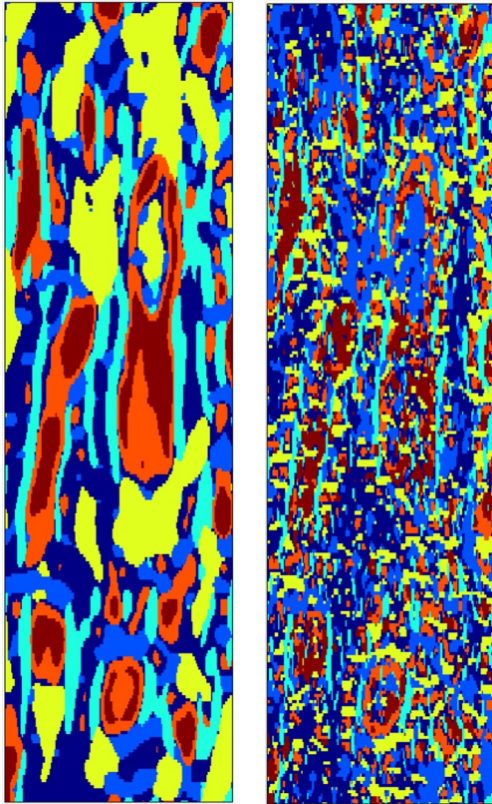
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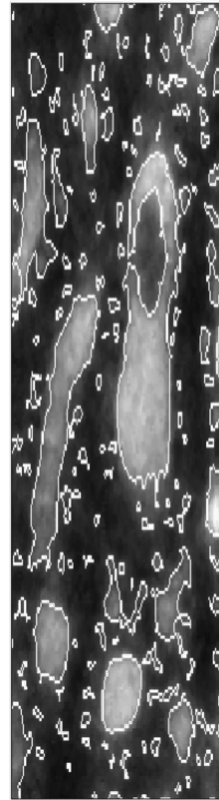
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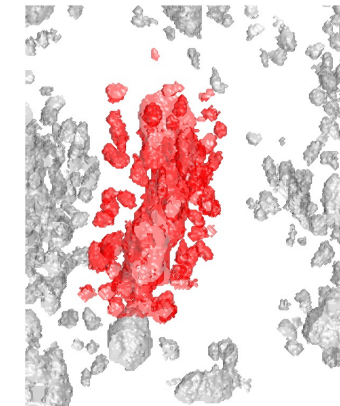
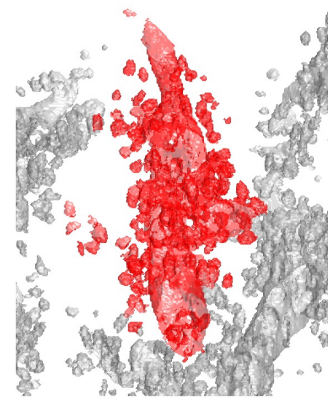


Final result  
(segmentation  
in white edges)



→  
Class selection  
+  
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SENPAI



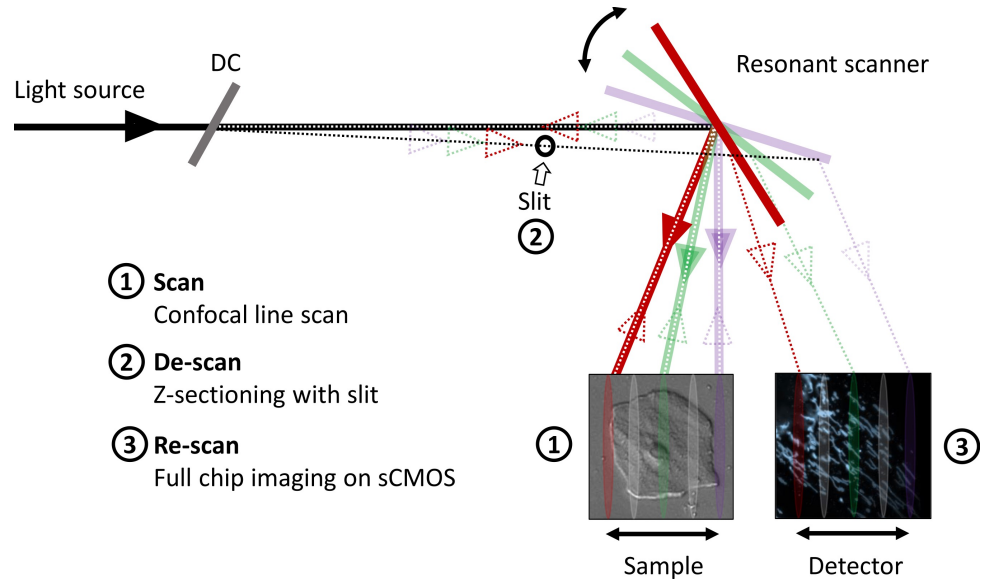
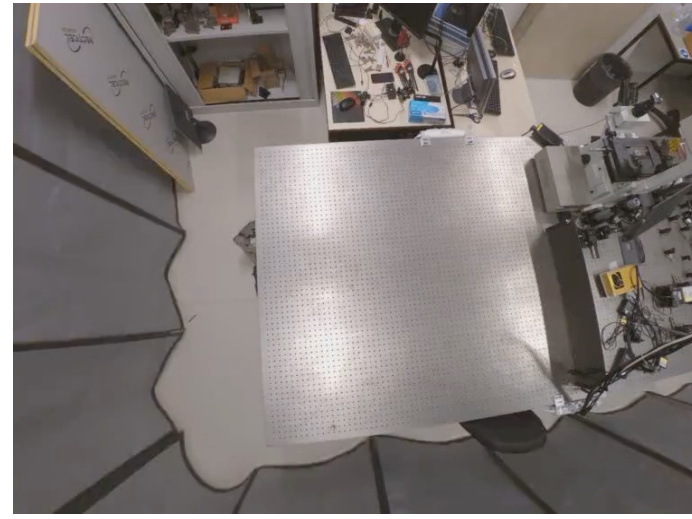
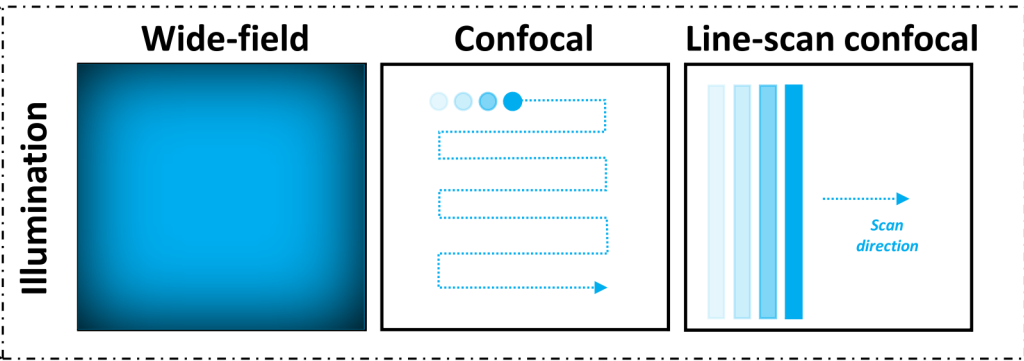
Manual





# Triscan

Fast tissue imaging / SMLM using a line-scan confocal



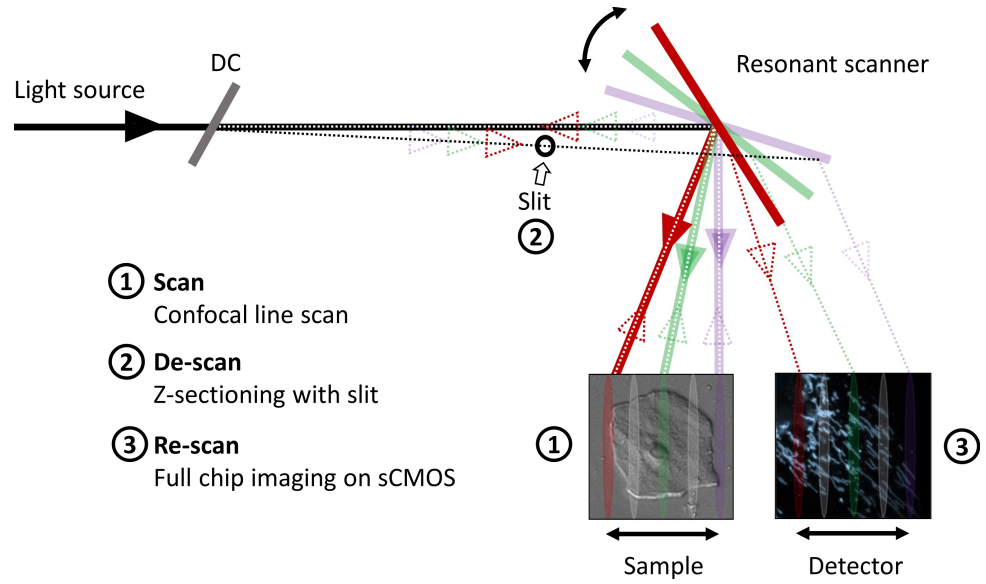
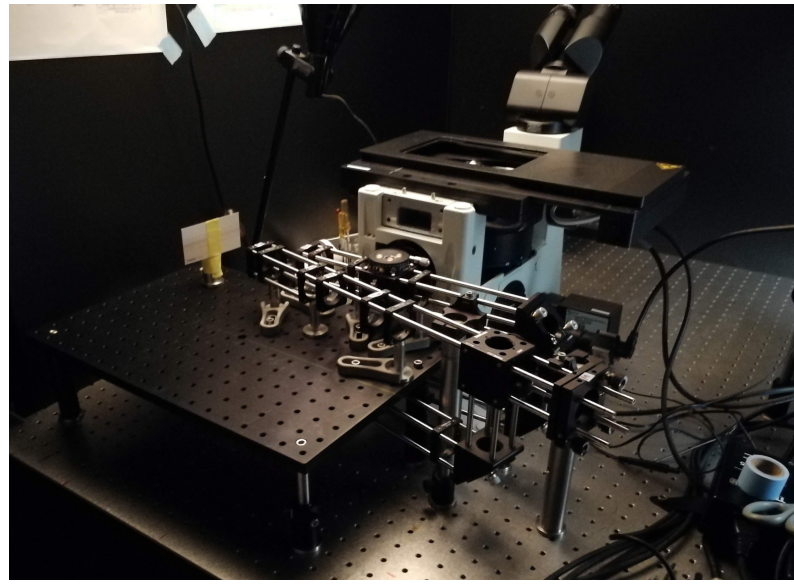
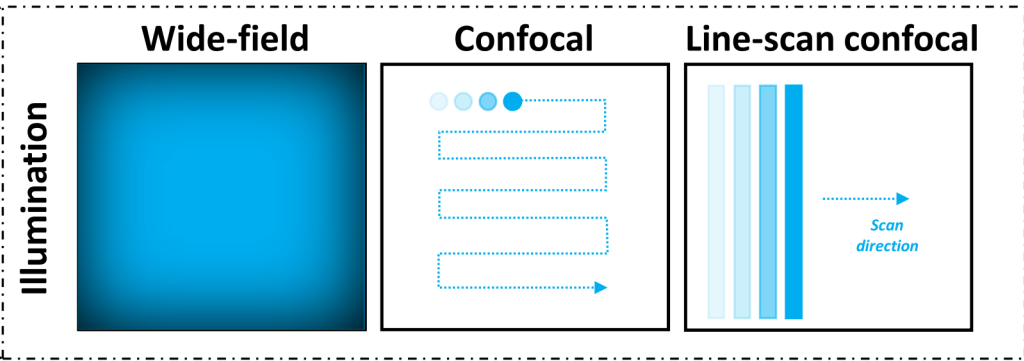
Prototype has been realized  
Optimization ongoing

Expected performance:

- Faster than classical confocal
- Similar resolution in xy
- Slightly reduced sectioning in z
- Single-molecule sensitive

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Fast tissue imaging / SMLM using a line-scan confocal



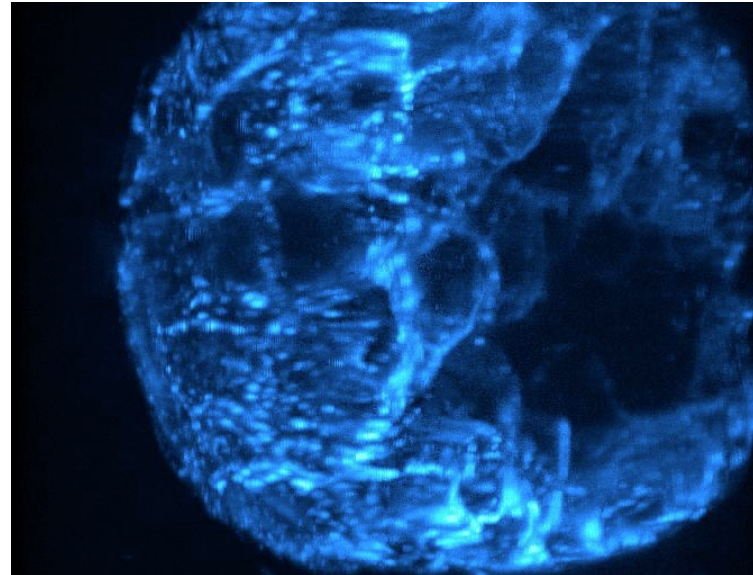
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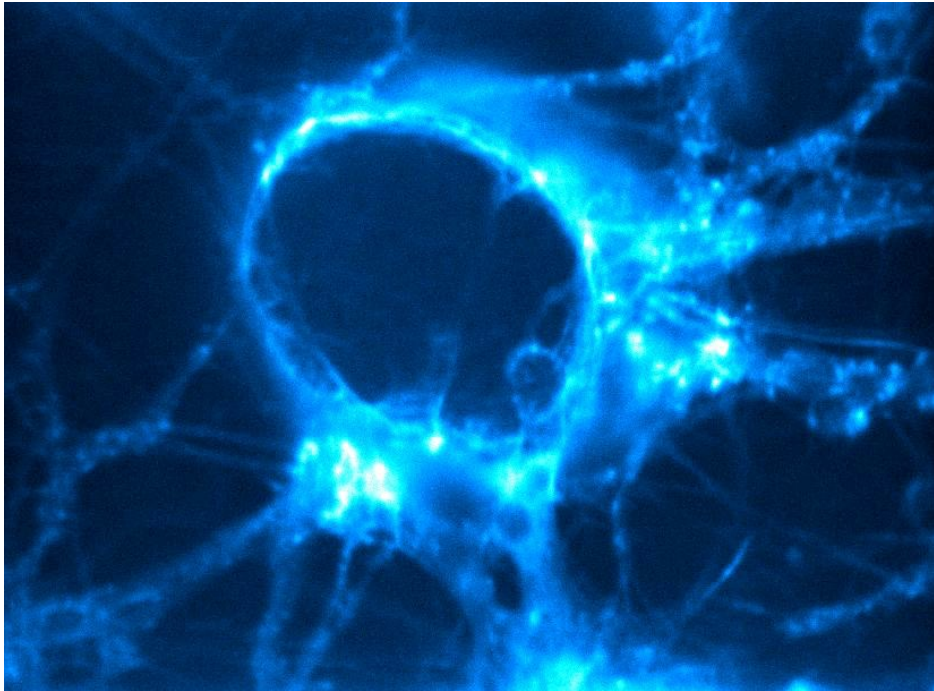
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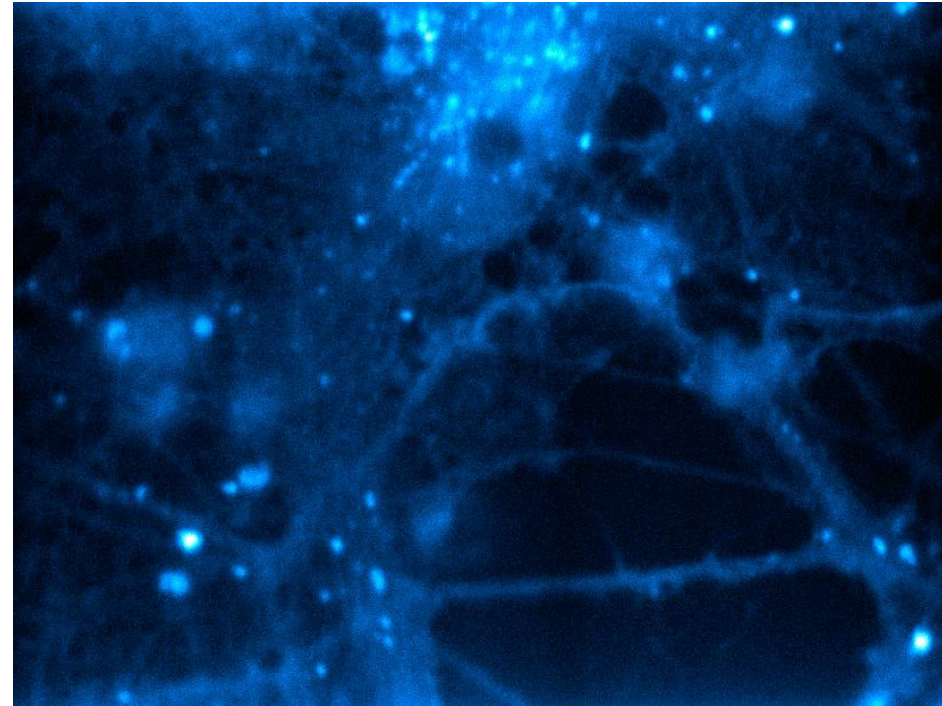


# Triscan

Fast tissue imaging / SMLM using a line-scan confocal



Without Triscan



Triscan

Hippocampal neurons: actin labelled with phalloidin Alexa488



# SENSEI & EBRAINS

- At this time we are not using any EBRAINS tool
- A lot of time was spent to reach SENSEI objectives
  - Limited time/people for learning the use of other tools/services
- SENSEI was not aware about the possible added value of using EBRAINS services/tools, taking into account the corresponding learning curve

# SENSEI outcomes

- Images obtained with different imaging modalities
  -
- Algorithms for neuron segmentation
  - 
  -
- Triscan
  -

# SENSEI outcomes

- Images obtained with different imaging modalities
  - Some images were shared via [Zenodo](#)
- Algorithms for neuron segmentation
  - SmRG (model based segmentation, available on [GitHub](#))
  - SENPAI (soon available)
- Triscan
  - prototype level

Callara A.L., Magliaro C., Ahluwalia A., Vanello N. A Smart Region-Growing Algorithm for Single-Neuron Segmentation From Confocal and 2-Photon Datasets (2020) *Frontiers in Neuroinformatics*, 14, art. no. 9

# SENSEI next steps

- Acquire human samples
  - Surgery resection from S. Anne Hospital, Paris
- Apply the segmentation algorithm to spine morphology/density estimation
- Develop multiscale data integration (e.g. 40x and 63x)
- Share the algorithms and test with other datasets
  - use data in the EBRAINS repository
  - candidate our algorithms for becoming “EBRAINS tools” (i.e. used by EBRAINS users and more...)
- Strengthen the collaborations and creating new ones
  - exploit EBRAINS collaboratory environment
  - use EBRAINS tools for setting up a protocol for registering the images/reconstruction on Atlases





# Thank you



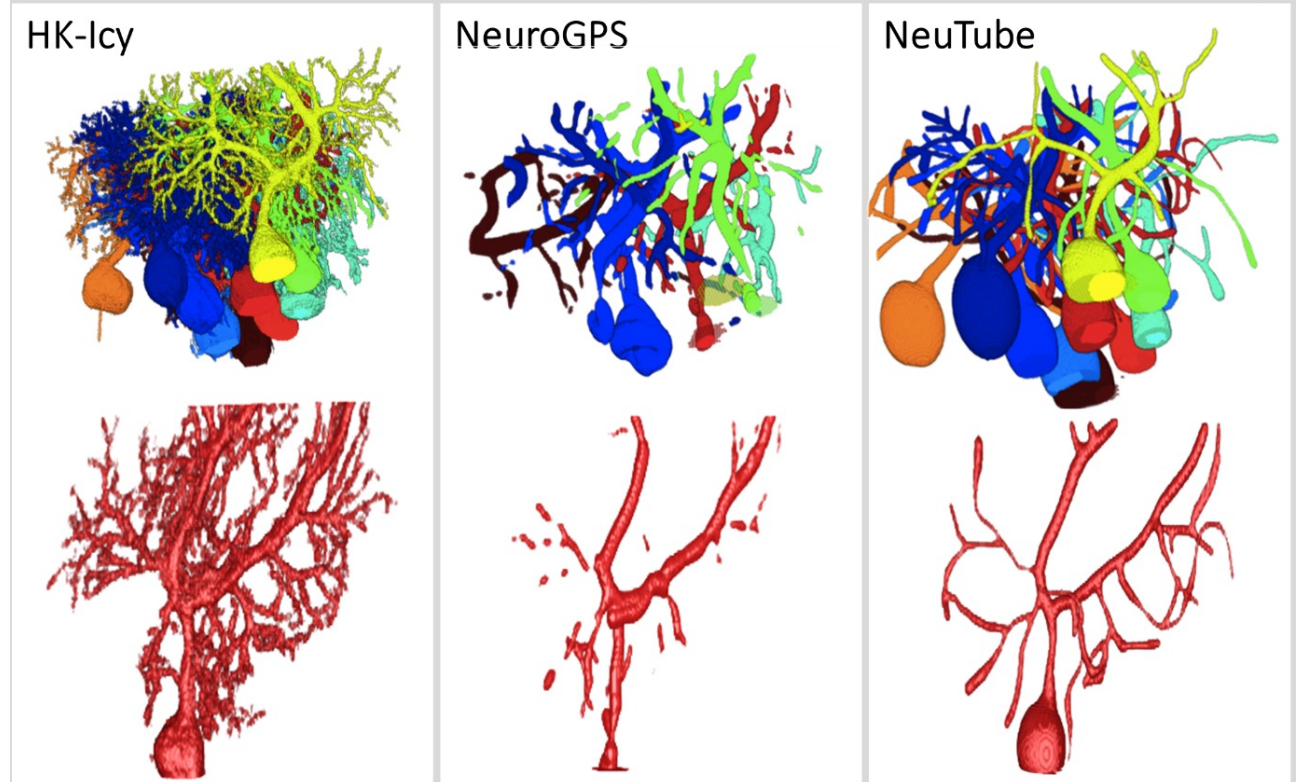
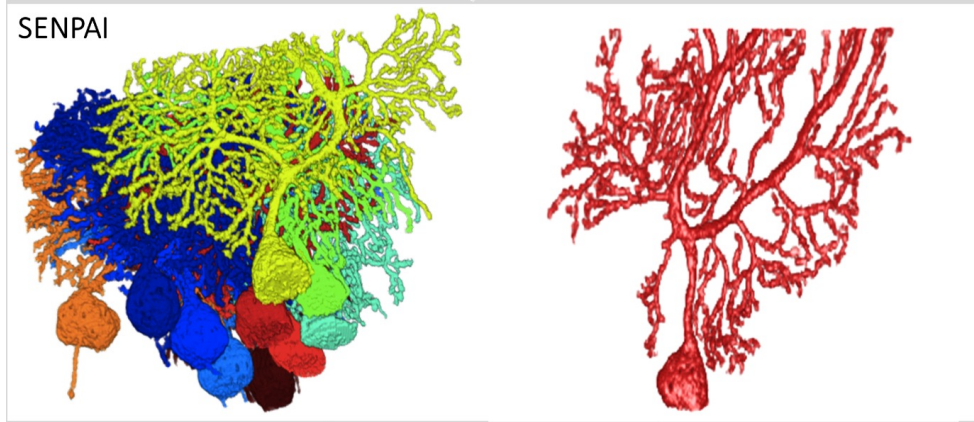
[www.humanbrainproject.eu](http://www.humanbrainproject.eu)

[www.ebrains.eu](http://www.ebrains.eu)



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# Algorithm Comparison



# SENPAI: a topological informed data driven approach to neuronal reconstruction

## Algorithm Comparison

