



Exploiting napari for big data exploration

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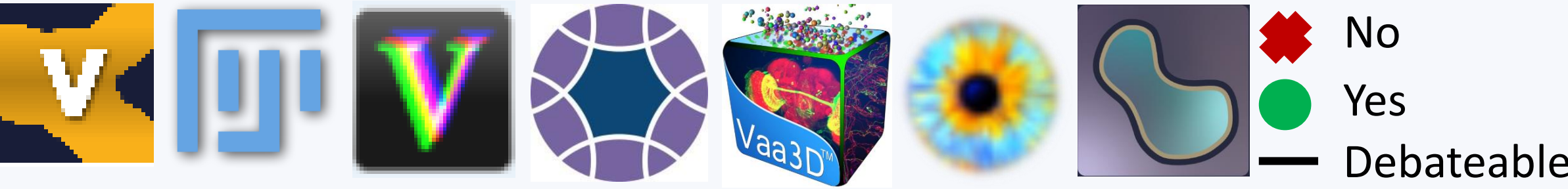


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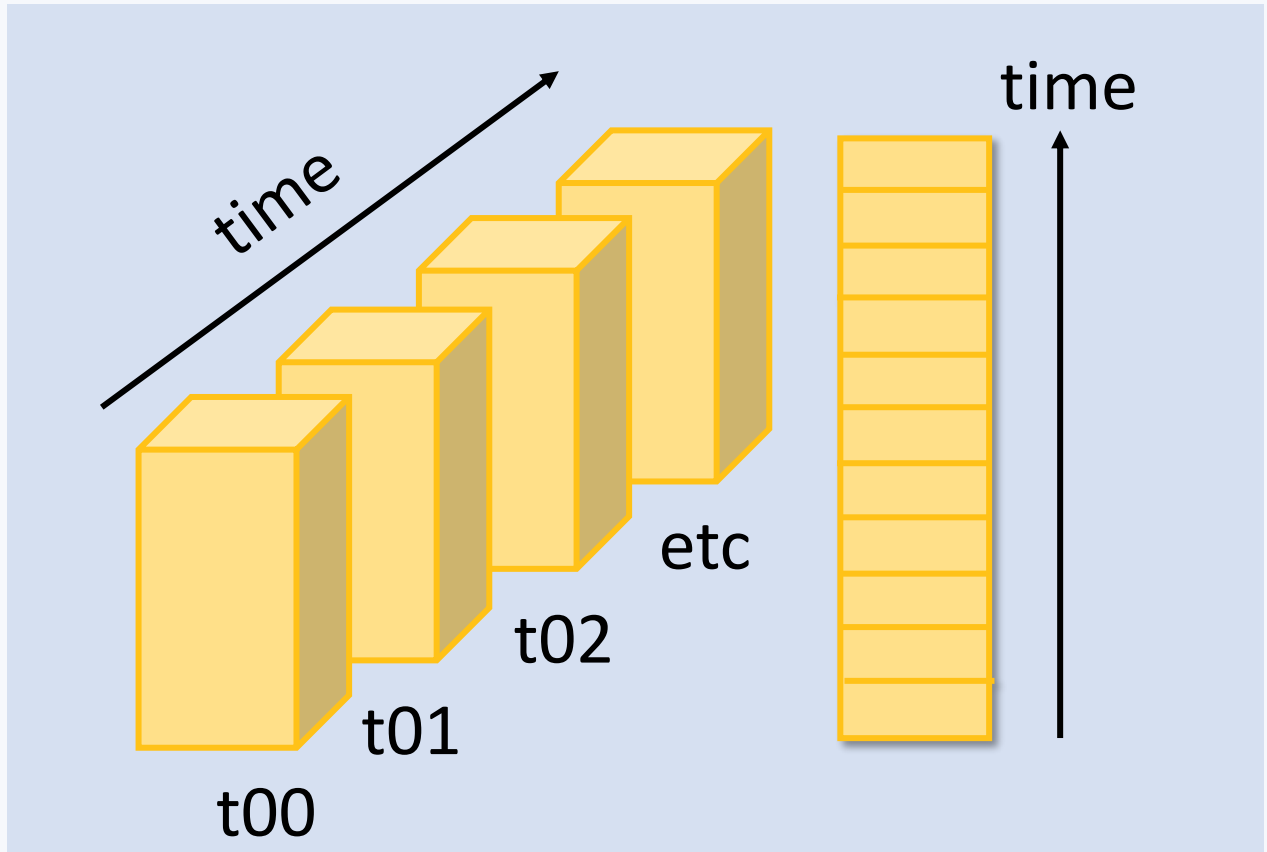
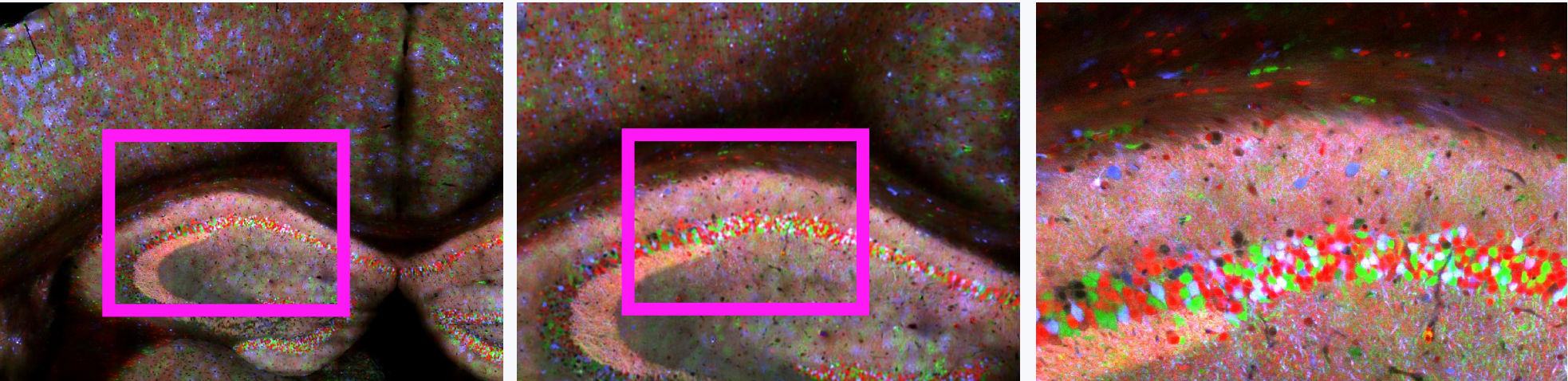
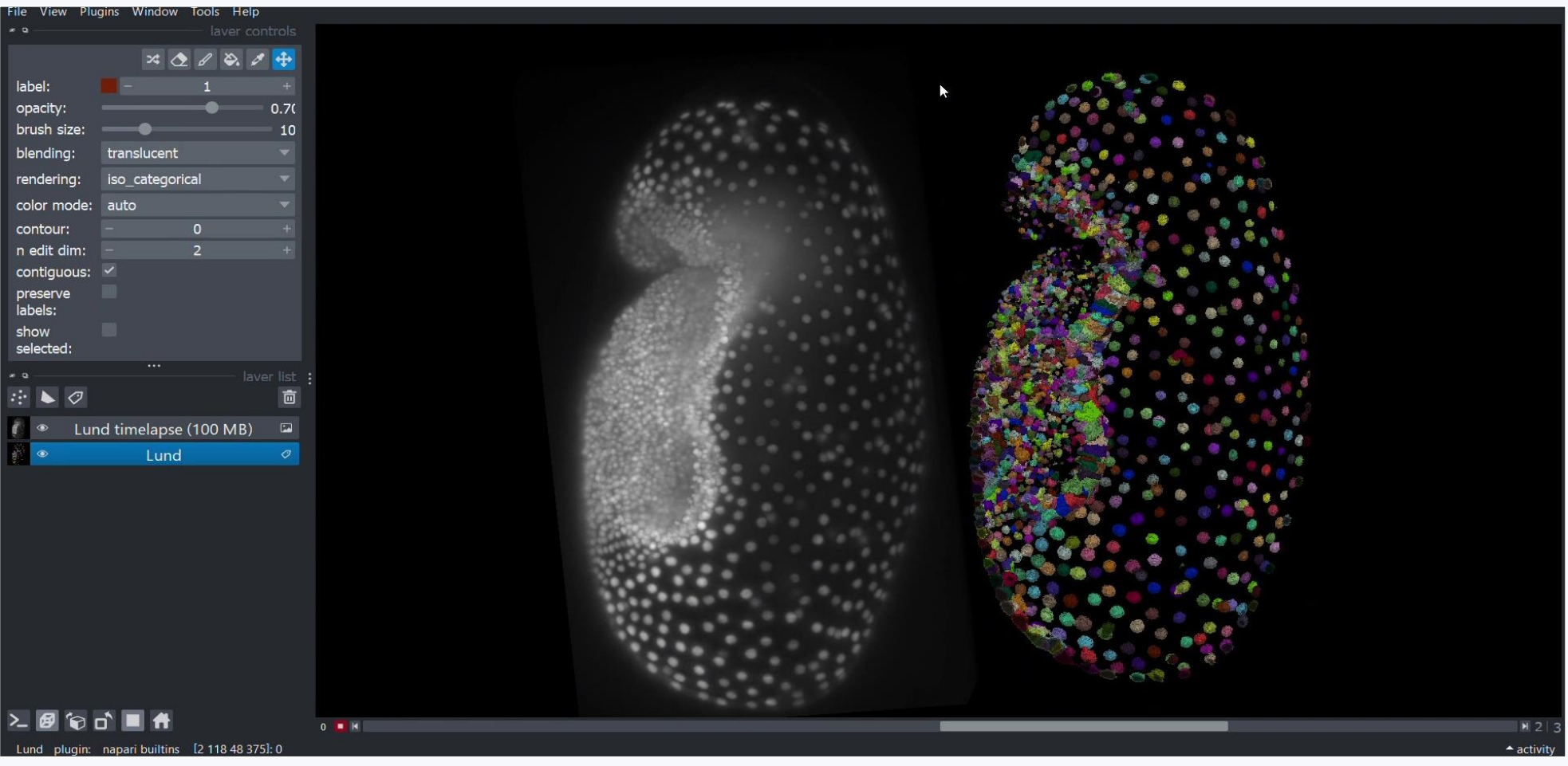
One of the most pressing issues facing the bioimaging community is the rise in data size, with biologists routinely acquiring large image volumes, while lacking an easy-to-use, open-source, freely-available multi-dimensional viewer for big data. Finding the best software and file format will aid in bridging the gap between biologists and image analysts, streamlining workflows with efficient reading of big data and user-friendly interaction with the data. Commercial software such as Imaris and Arivis address the issue, but are unable to support multiple OS platforms and are not freely available to the microscopy community. Open-source solutions, despite being more accessible, typically have inferior features such as the lack of 3D visualisation in BigDataViewer and limited support for next-generation file formats such as OME-Zarr in napari. We are developing a napari-based workflow prototype to optimise the process of big image visualisation with an emphasis on accessibility. By taking advantage of standard python libraries, open-source plugins and next generation file formats we are exploring and comparing different approaches to user-friendly big data interaction. This will ultimately support coordination between data acquisition and processing communities by providing a smooth and responsive means of viewing and interacting with big microscopy data in napari.

Open source viewers

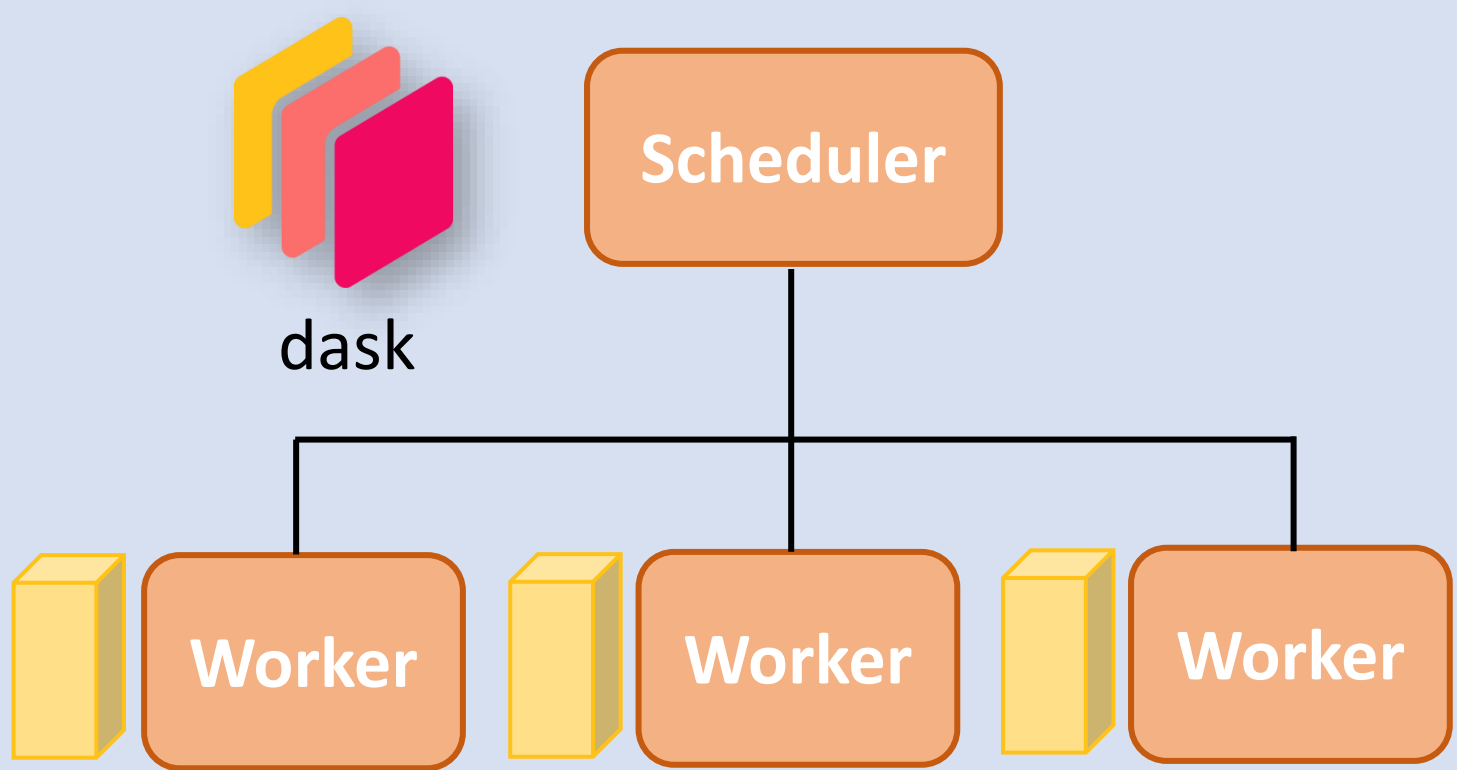


Viewer	Multi - dimensional	NGFF Support	Orthogonal View	Labels Overlay
ImarisViewer	No	Yes	Yes	No
BigVolumeViewer	No	No	No	No
3DViewer	No	No	Yes	No
VVDViewer	Yes	No	No	Yes
Agave	Debateable	Yes	Yes	No
Vaa3D	Yes	No	Yes	Yes
Drishti	Debateable	No	No	No
Napari	Yes	Yes	Debateable	Yes

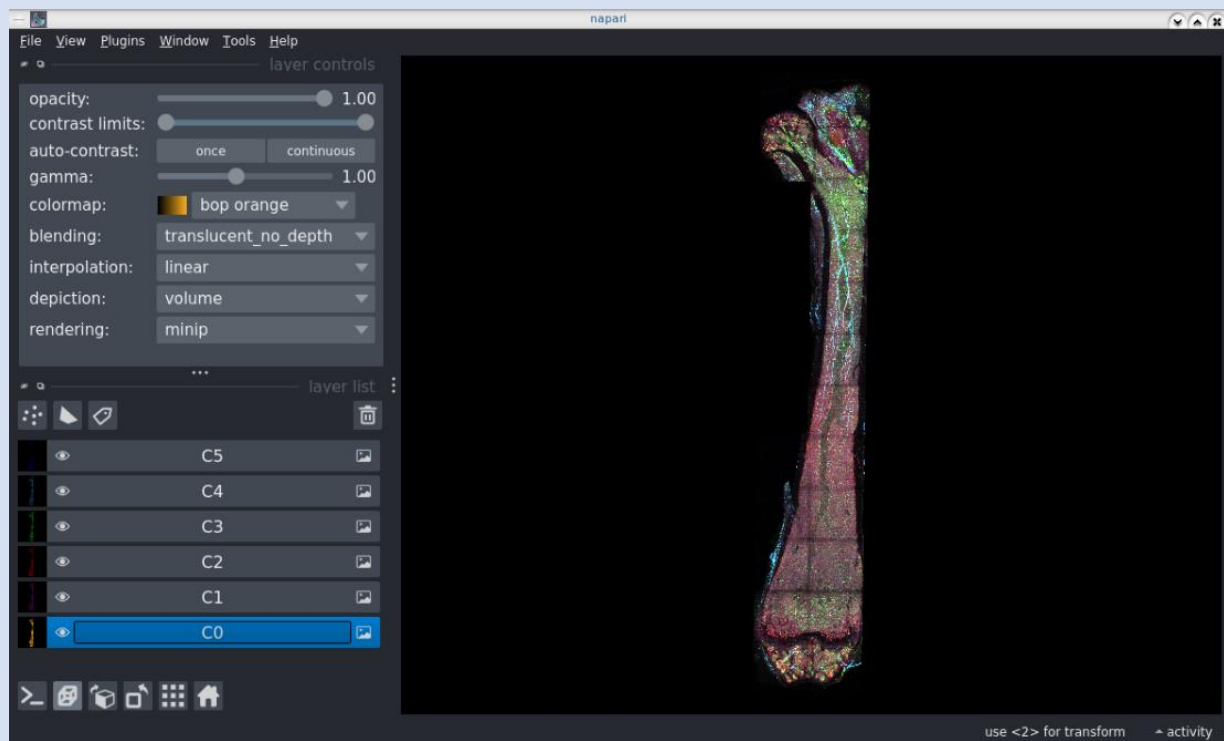
Napari



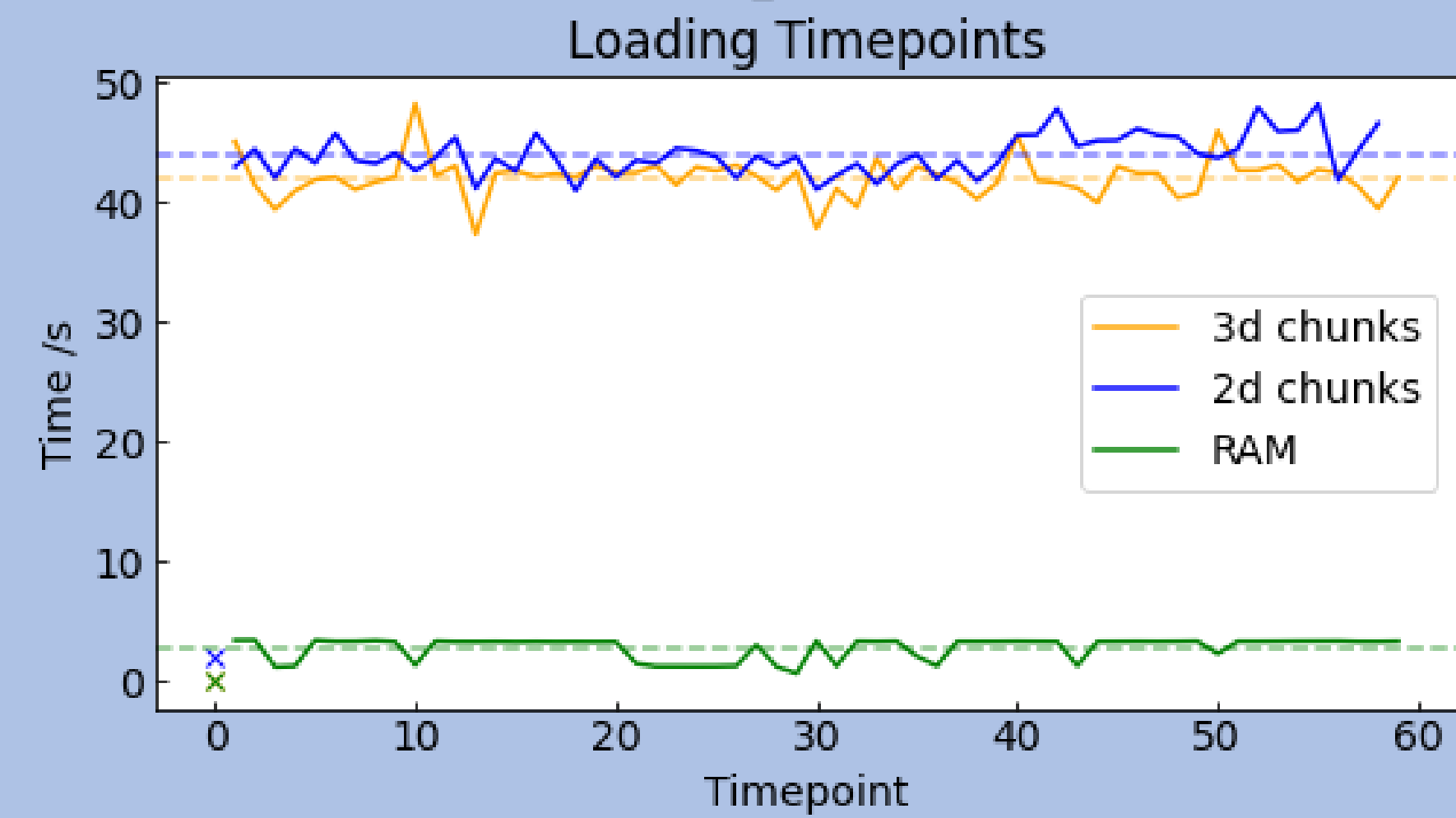
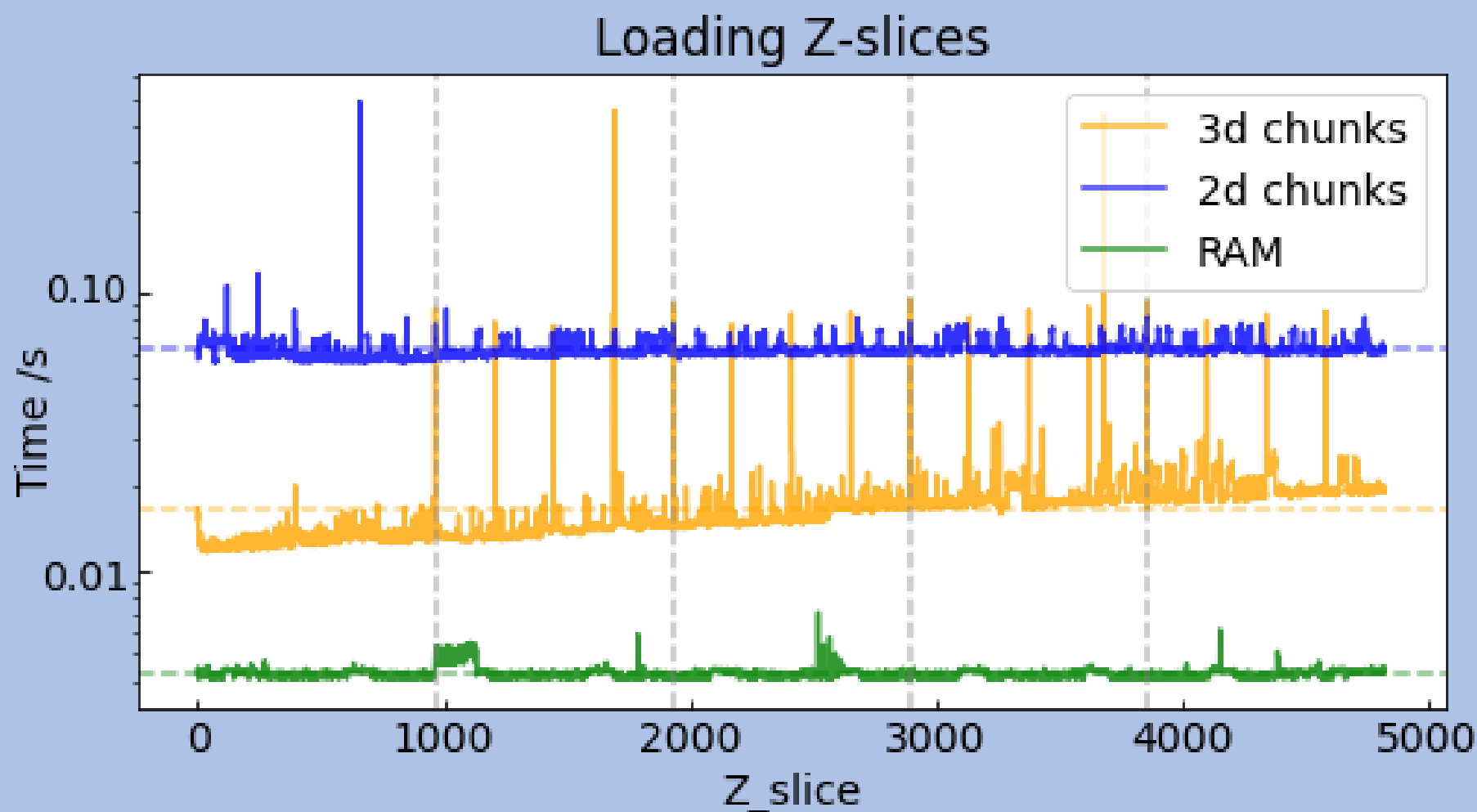
Multidimensional data organised into a dask stack then split into chunks



Dask scheduler manages data transfer between CPU workers and assigns chunks for computation



Data chunks are read into RAM as and when needed by the viewer

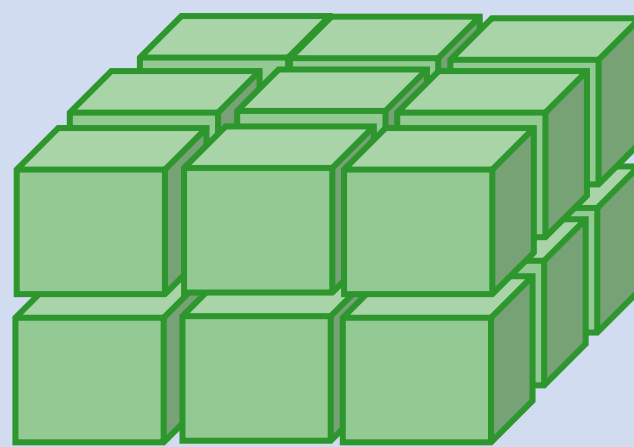
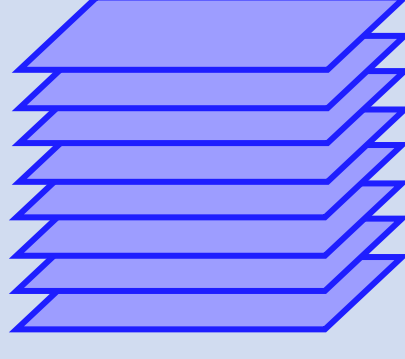
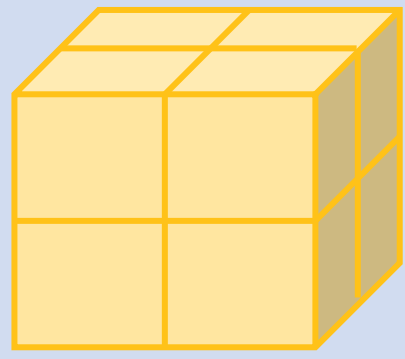


Quantifying interactivity within napari

We compare interactivity performance of dask with different chunk sizes against loading the following dataset into RAM within napari.

Fluo-N3DL-TRIF (60, 991, 1875, 965) TZYX ~ 180 GB
celltrackingchallenge.net/3d-datasets/

Fig. 1



3D chunks + dask
(1, 247, 469, 241)
~ 165 seconds

2D chunks + dask
(1, 1, 1875, 965)
~ 160 seconds

Whole + RAM
(60, 991, 1875, 965)
~ 40 minutes

The graphs display the time taken for napari to load the field of view upon changing z-slice and timepoint, respectively. The legend corresponds to Figure 1.

GitHub Repo



Survey

