Pose Estimation in NWB

ndx-pose





Pose Estimation in NWB

NWB Currently supports writing from two popular pose estimation frameworks

- SLEAP
- DeepLabCut (DLC)

We do not have time to fully explain the details of how these packages work...

...so we will simply explain the input-output (I/O) workflow

- The NWB data type for videos called the ImageSeries
- It has two modes
 - Internal
 - each frame of the video is extracted and written as a Dataset
 - > same as any other array-valued series you might find in an NWB file

```
nwbfile.acquisition

{'Video: Rat10-20140708-01-prerun': Video: Rat10-20140708-01-prerun pynwb.image.ImageSeries
Fields:
   data: <HDF5 dataset "data": shape (5773, 360, 640, 3), type "|u1">
   description: Video recorded by camera.
   timestamps: <HDF5 dataset "timestamps": shape (5773,), type "<f8">
   timestamps_unit: seconds
   unit: Frames
```

github.com/NeurodataWithoutBorders/nwb_hackathons/tree/main/Cosyne_2023

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 - data type instead contains a path to the video file(s) outside the NWB file

```
'OriginalVideoBodyCamera': OriginalVideoBodyCamera pynwb.image.ImageSeries

Fields:
    data: <HDF5 dataset "data": shape (0, 0, 0), type "|u1">
    description: The original video each pose was estimated from.
    external_file: <StrDataset for HDF5 dataset "external_file": shape (1,), type "|O">
    format: external
```

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nwbfile.acquisition["OriginalVideoBodyCamera"].external file[:]

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Best Practices

- Internal mode
 - videos of physiological recordings (such as optical imaging)
 - stimulus presentations
- External mode
 - videos of natural behavior

Note: Reasoning for Best Practices

- Reasoning for these Best Practices is a bit nuanced...
- For some history of decision, see...
 - NWB:#1647, DANDI:#769 and DANDI-helpdesk:#30
- DANDI
 - strongly recommends these practices for storage
 - will automatically remap all paths and organize folder storage at time of upload

Step 1: Install SLEAP or DLC

- SLEAP Talmo Lab <u>instructions</u> <u>GitHub repo</u>
- DLC Mathis Lab instructions GitHub repo

■ This step can take a while, but just be patient and persistent

- If anything goes differently from instructions, when in doubt ask
 - Google
 - ChatGPT
 - or raise an issue on their repository or helpdesk

Step 2: Launch GUI

These can be run headless, but for your first time I recommend using the Graphical User Interface (GUI) to help walk you through it

■ Both are able to load in the original videos (.mpg, .mp4, .avi, etc...)

■ Due to the previous discussion regarding storing ImageSeries of natural behavior, it is recommended to import directly from the raw video files when using either software package

Steps 3-7

■ Step 3: Load videos

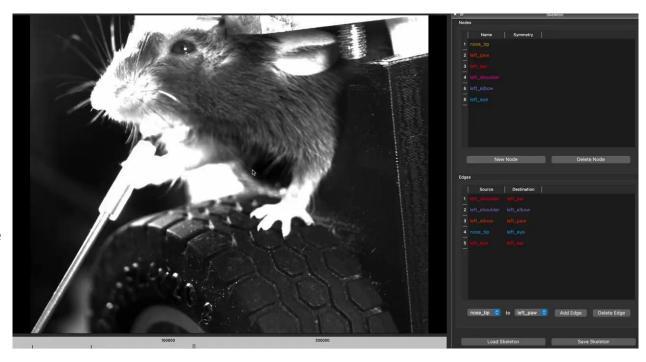


Steps 3-7

■ Step 3: Load videos

Step 4: Extract frames for training

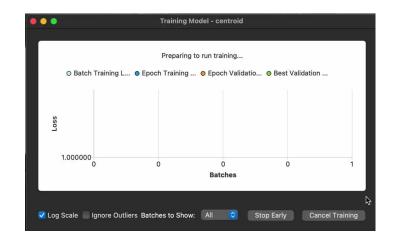
Step 5: Label points on the training frames



Steps 3-7

- Step 3: Load videos
- Step 4: Extract frames for training
- Step 5: Label points on the training frames
- Step 6: Train the model

■ Step 7: Run the prediction





. . .

Output: Saving results

■ The output of the prediction over the entire video can then be saved in either the source .h5 (DLC) or .slp (SLEAP) formats

- These can in turn be converted to NWB format via two utility packages
 - dlc2nwb $(.h5 \rightarrow .nwb)$
 - nwb sub-module of sleap-io (.slp \rightarrow .nwb)

■ BONUS: SLEAP can export directly to NWB through the GUI

Understanding PoseEstimation

■ In either case (or for any future integrations with keypoint tracking tools) these are modelled using a common structure in NWB called the PoseEstimation

```
nwbfile.processing["behavior"]["PoseEstimationLeftCamera"]
PoseEstimationLeftCamera abc.PoseEstimation at 0x140450551033632
Fields:
 description: Estimated positions of body parts using DeepLabCut.
  nodes: <StrDataset for HDF5 dataset "nodes": shape (11,), type "|0">
  pose estimation series: {
   nose tip <class 'abc.PoseEstimationSeries'>,
    paw_1 <class 'abc.PoseEstimationSeries'>,
    paw r <class 'abc.PoseEstimationSeries'>,
    pupil_bottom_r <class 'abc.PoseEstimationSeries'>,
    pupil left r <class 'abc.PoseEstimationSeries'>,
    pupil right r <class 'abc.PoseEstimationSeries'>,
    pupil top r <class 'abc.PoseEstimationSeries'>,
    tongue_end_1 <class 'abc.PoseEstimationSeries'>,
    tongue end r <class 'abc.PoseEstimationSeries'>,
    tube bottom <class 'abc.PoseEstimationSeries'>,
    tube top <class 'abc.PoseEstimationSeries'>
  source software: DeepLabCut
```

Understanding PoseEstimation

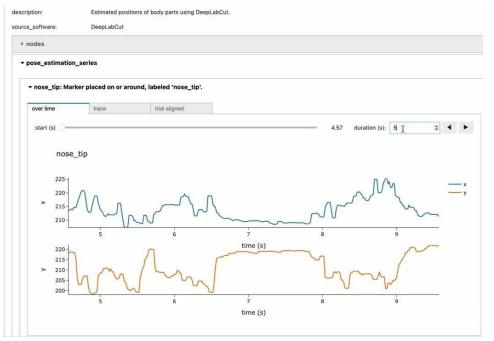
■ A single PoseEstimation is a container of multiple PoseEstimationSeries, one for each node (usually a body part) being tracked

```
nwbfile.processing["behavior"]["PoseEstimationLeftCamera"]
PoseEstimationLeftCamera abc.PoseEstimation at 0x140450551033632
Fields:
  description: Estimated positions of body parts using DeepLabCut.
  nodes: <StrDataset for HDF5 dataset "nodes": shape (11,), type "|0">
  pose estimation series: {
    nose_tip <class 'abc.PoseEstimationSeries'>, -
    paw 1 <class 'abc.PoseEstimationSeries'>,
    paw r <class 'abc.PoseEstimationSeries'>,
                                                                     nwbfile.processing["behavior"]["PoseEstimationLeftCamera"]["nose tip"]
    pupil bottom r <class 'abc.PoseEstimationSeries'>,
    pupil_left_r <class 'abc.PoseEstimationSeries'>,
                                                                     nose tip abc.PoseEstimationSeries at 0x140450549420416
    pupil right r <class 'abc.PoseEstimationSeries'>,
                                                                     Fields:
    pupil top r <class 'abc.PoseEstimationSeries'>,
                                                                       confidence: <HDF5 dataset "confidence": shape (242446,), type "<f8">
    tongue end 1 <class 'abc.PoseEstimationSeries'>,
                                                                       conversion: 1.0
                                                                       data: <HDF5 dataset "data": shape (242446, 2), type "<f8">
    tongue end r <class 'abc.PoseEstimationSeries'>,
                                                                       reference frame: (0,0) corresponds to the upper left corner when using width by height convention.
    tube bottom <class 'abc.PoseEstimationSeries'>,
                                                                       timestamps: <HDF5 dataset "timestamps": shape (242446,), type "<f8">
    tube top <class 'abc.PoseEstimationSeries'>
                                                                       timestamps unit: seconds
                                                                       unit: px
  source software: DeepLabCut
```

github.com/NeurodataWithoutBorders/nwb_hackathons/tree/main/Cosyne_2023

Understanding PoseEstimation

■ The data field of each PoseEstimationSeries represents the x/y/z positions being tracked over time



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