A tutorial on how to use the Brain Observatory Toolbox

This tutorial will show you how to access the Brain Observatory data using MATLAB. In particular, this tutorial will show you how to:

- 1) Instantiate a sessionfilter object to:
- A) Get general information on the brain observatory data
- B) Filter sessions by specified criteria such as brain areas, imaging depth and stimuli type
- 2) Instantiate a session object to:

Retrieve and examine kinds of experiment data as listed below of an interessed session from its Neurodata Without Borders (NWB) file:

- a) ROI masks
- b) fluorescence traces
- c) running speed
- d) motion correction
- e) eye tracking
- 3) Call function convert_fluorescence_trace_into_raster_format to:

Convert fluorescence trace into data that is in "raster format" for further analysis

Organization of the Brain Observatory Data

Before we start, we first want to describe the organization of the Brain Observatory Data.

An experiment container contains three sessions (also called experiments or ophys_experiments) where recordings were made on a singe mouse, in a single brain region and at a particular imaging depth. Each of these three sessions consists of a series of "subexperiments" where a particular stimulus set was shown. Neurodata Without Borders (NWB) files downloaded from the Allen Institute API each consist of data from a single session. As described below, all data that we extract into "raster format" consists of data from a single "subexperiment" where a particular stimulus set was shown.

Note: within an experiment container the same stimuli might be repeated in different sessions (i.e., different sessions can have the same "subexperiment type"). For example, Natural Movie 1 is shown in all three sessions in an experiment container. For more information see: http://alleninstitute.github.io/AllenSDK/brain_observatory.html

sessionfilter is a class that resembels BrainObservatoryCache in allensdk (see Python code at:https://github.com/AllenInstitute/AllenSDK/blob/master/allensdk/core/brain_observatory_cache.py).

The MATLAB sessionfilter enables you to get information and access data using three types of methods:

- A) Get general information on the brain observatory data using methods that start with "get_"
- B) Filter sessions by different criteria using methods that start with "filter_"

To beign, build a brain_observatory_cache object

Note: BOT by default excluds all failed experiment containers and their sessions. If you want to have them included, you can pass 1 to the constructor:

1A) Get general information on the brain observatory data

There are several methods start with "get_" which summairze in the brain observatory data based on particular criteria.

To get the total number of experiment containers we can use:

```
bosf1.get_total_num_of_containers()
ans = 181
```

Note: Total number of sessions is the number of containers times 3 since there are three sessions per container.

To get all the cortical depths (um) that were ever recorded in any experiment container we can use:

```
bosf1.get_all_imaging_depths()

ans =
    175
    265
    275
    300
    320
    325
    335
    350
    365
    375
```

Note: Recordings didn't take place equally among these depths, only the most common depths are shown on homepage of Brain Observatory: http://observatory.brain-map.org/visualcoding

To get all the get all type of cre driver lines from all mice we can use:

```
bosf1.get_all_cre_lines()

ans = 6×1 cell array
    'Cux2-CreERT2'
    'Emx1-IRES-Cre'
    'Nr5a1-Cre'
    'Rbp4-Cre_KL100'
    'Rorb-IRES2-Cre'
    'Scnn1a-Tg3-Cre'
```

Note: All mice had the same reporter line: Ai93 and tTA driver line: Camk2a-tTA

For more information about transgenic lines: http://observatory.brain-map.org/visualcoding/transgenic

To get all the all brain regions that were recorded in any experiment container we can use:

bosf1.get all targeted structures()

```
ans = 6×1 cell array
'VISal'
'VISam'
'VISl'
'VISp'
'VISpm'
'VISrl'
```

VISam

VISl

25

38

For more information on these locations, see homepage of Brain Observatoy: http://observatory.brain-map.org/visualcoding

To get all the all types of sessions that appear in any experiment container we can use:

```
bosf1.get_all_session_types()

ans = 4×1 cell array
    'three_session_A'
    'three_session_B'
    'three_session_C'
    'three session_C2'
```

Note: There are always three sessions in each container: session A, session B, and session C or session C2

To get all the all type of stimuli that were used in any of the four types of sessions we can use:

```
bosf1.get_all_stimuli()

ans = 9×1 cell array
    'drifting_gratings'
    'locally_sparse_noise_4deg'
    'locally_sparse_noise_8deg'
    'natural_movie_one'
    'natural_movie_three'
    'natural_movie_two'
    'natural_scenes'
    'spontaneous'
    'static_gratings'
```

Note: For mapping between session type and stimulus type see http://alleninstitute.github.io/AllenSDK/brain_observatory.html

To get the number of experiment containers recorded in each brain region we can use:

```
bosf1.get_summary_of_containers_along_targeted_structures()

VISal 35
```

| VISp | 66 |
|-------|----|
| VISpm | 36 |
| VISrl | 16 |

To get the number of experiment containers recorded at each cortical depth we can use:

```
bosfl.get_summary_of_containers_along_imaging_depths()

175     53
250     1
265     1
275     75
300     4
320     1
```

To get the number of experiment containers recorded at each cortical depth in each brain region we can use:

bosfl.get summary of containers along depths and structures()

| ans = 11×7 table | | | | | | | | |
|------------------|-------|-------|------|------|-------|-------|-------|--|
| | VISal | VISam | VISl | VISp | VISpm | VISrl | total | |
| | | | | | | | | |
| 175 | 8 | 5 | 10 | 11 | 11 | 5 | 50 | |
| 265 | 1 | 0 | 0 | 0 | 0 | Θ | 1 | |
| 275 | 11 | 9 | 14 | 17 | 9 | 5 | 65 | |
| 300 | 1 | 0 | 1 | 1 | 1 | Θ | 4 | |
| 320 | 0 | 1 | 0 | 0 | 0 | 0 | 1 | |
| 325 | 0 | 1 | 1 | 0 | 1 | 0 | 3 | |
| 335 | 0 | 0 | 0 | 2 | 1 | Θ | 3 | |
| 350 | 2 | 3 | 3 | 10 | 4 | 3 | 25 | |
| 365 | 1 | 0 | 0 | 0 | 0 | 0 | 1 | |
| 375 | 6 | 4 | 5 | 5 | 6 | 2 | 28 | |
| total | 30 | 23 | 34 | 46 | 33 | 15 | 181 | |

1B) Filter sessions by specified criteria such as brain areas, imaging depth and stimuli type

sessionfilter.filtered_session_table is an m-by-n table listing the n fields of meta data of m succesful sessions. As different filter methods get called on sessiionfilter, rows/sessions in filtered_session_table are liminated to meet specified criteria, and other properies (e.g., stimuli, targeted_structure, etc) of sessionfilter, that are essentially important fields of meta data in filtered_session_table, are also updated.

Here we show an example of searching for sessions that Rorb-IRES2-Cre mice had their primary visual cortex recorded at 275 μ m deep, during where drifting gratings were shown and eye tracking went through.

To ensure an "inviolate" sessionfilter:

```
bosf1.clear_filters
```

Eliminate sessions to the ones where drifting gratings were shown:

Note that all properties of bosf have changed except the first top three properties.

Eliminate sessions to the ones that have posterior Primary Visual Cortex recordings:

Eliminate sessions to the ones that were recoreded at 275 um deep

```
bosf1.filter_sessions_by_imaging_depth(275)

ans =
   sessionfilter with properties:
    valid_session_table: [543×15 table]
```

Eliminate sessions to the ones that were operated on mice of Rorb-IRES2-Cre

Eliminate sessions to the ones that eye tracking completely went through

As we can see now, there are 2 sessions that meet all of the criteria given.

Every experiment container in the Brain Obervatory has an unique experiment container ID that was created by the Allen Institute. These container IDs are stored in the <code>experiment_container_id</code> field in <code>filtered_session_table</code>. For example, to see the experiment container IDs that met our filtering criteria we can run:

bosf1.filtered session table.experiment container id

```
ans =
512124562
511510989
```

If we want to find out the three sessions in the second experiment contianer, we can call:

```
bosf2 = bot.sessionfilter
```

```
bosf2 =
  sessionfilter with properties:

    valid_session_table: [543×15 table]
    filtered_session_table: [543×15 table]
        stimulus: {9×1 cell}
        targeted_structure: {6×1 cell}
        imaging_depth: [10×1 double]
        container_id: [181×1 double]
        session_id: [543×1 double]
        session_type: [543×1 categorical]
        cre_line: {6×1 cell}
        eye_tracking_avail: [2×1 logical]
        failed: 0
```

bosf2.filter sessions by container id(511510989)

bosf2.session_id

```
ans = 501788003
501567237
501729039
```

Now let's dig into the real data.

2) Instantiate a session object to:

Retrieve and examine kinds of experiment data as listed below of an interessed session from its Neurodata Without Borders (NWB) file:

To illustrate this, let's instantiate a session object for the first session in bosf2.

When you run this, BOT automatiacally searchs for the corresponding NWB file and downloads it if it is not there yet.

Dowanloading an NWB file from their server can take about 20 min.

All the meta data stored in bosf2.filtered_session_table for this session are carried over to bos1.sSessionInfo

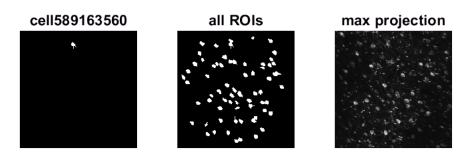
bos1.sSessionInfo

2a) ROI Masks

If you want to take a look at specific cells or all cells visually, you can extract and plot thier pixel masks. You can also pull out the maximum intensity projection of the movie for context.

```
% get all the cell ids cell_ids = bos1.get_cell_specimen_ids()
```

```
% plot the ROI mask for the first cell in the list
subplot(1,3,1)
imshow(bos1.get_roi_mask_array(cell_ids(1)))
title(['cell' num2str(cell ids(1))])
% plot the ROI masks for all celss
all mask = logical(zeros(512,512));
for iCell = 1: length(cell ids)
    all mask = or(all mask, bos1.get roi mask array(cell ids(iCell)));
end
subplot(1,3,2)
imshow(all_mask)
title('all ROIs')
% plot the maximum intensity projection
subplot(1,3,3)
imshow(double(bos1.get max projection)/max(max(double(bos1.get max projection))))
% colormap(gray)
title('max projection')
```



2d) Fluorescence Traces

iCell = 1

You can get the fluorescence traces of a specific cell by its index. Let's take cell 589163560 as an example.

```
[time, raw] = bos1.get_fluorescence_traces;
[time, demixed] = bos1.get_demixed_traces;
[time, neuropil] = bos1.get_neuropil_traces;

neuropil_r = bos1.get_neuropil_r;

[time, corrected] = bos1.get_corrected_fluorescence_traces;

% compute it yourself
% corrected = demixed - neuropil * (neuropil_r .* eye(size(neuropil_r,1)));

[time, dff] = bos1.get_dff_traces;

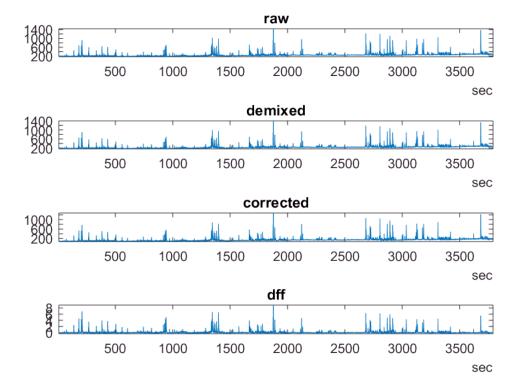
iCell = bos1.get_cell_specimen_indices(589163560)
```

```
subplot(4,1,1)
plot(time, raw (:,iCell))
axis tight
title ('raw')
```

```
subplot(4,1,2)
plot(time, demixed (:,iCell))
axis tight
title ('demixed')

subplot(4,1,3)
plot(time, corrected(:,iCell))
axis tight
title ('corrected')

subplot(4,1,4)
plot(time, dff(:,iCell))
axis tight
title ('dff')
```



You can also examine the fluorescence traces of all cells recorded in a session one cell at a time in sequence. The traces will be plotted in an external figure, hit any key to proceed to the next cell.

```
figure('Position', get(0, 'Screensize'))
set(gcf, 'Visible','on')

for iCell = 1:length(cell_ids)
clf

ft_to_be_plotted = {'raw', 'demixed', 'corrected', 'dff'};
```

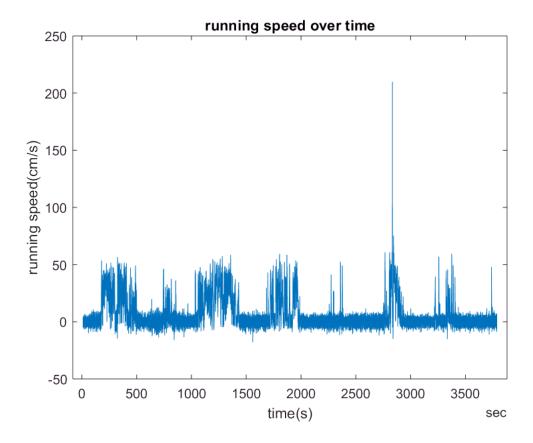
```
for ift = 1: length(ft to be plotted)
subplot(length(ft to be plotted),1,ift)
cur ft = eval(ft to be plotted{ift});
plot(time, cur ft(:,iCell))
% axis tight
ylim([0,max(max(cur ft))])
title(ft to be plotted{ift})
end
% figure('KeyPressFcn',@keyDownListener)
%
      function keyDownListener(
fprintf('Program paused. Press any key to continue to the next cell.\n');
pause;
end
Program paused. Press any key to continue to the next cell.
Program paused. Press any key to continue to the next cell.
Program paused. Press any key to continue to the next cell.
```

close all

2c) Running Speed

All experiments contain running speed. Extreme outliers from the tracking have been removed and replaced with NaN, which will appear as gaps in the plotted data.

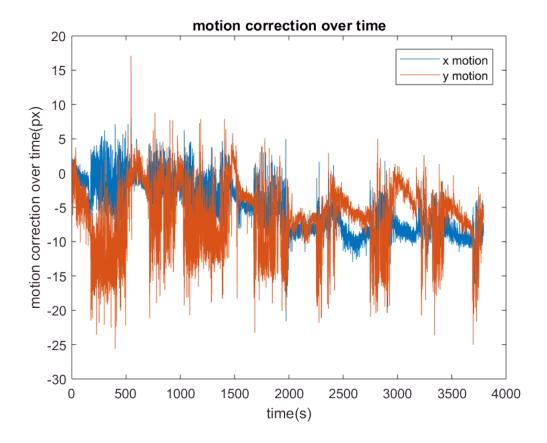
```
[time, runningSpeed] = bos1.get_running_speed;
plot(time, runningSpeed)
title('running speed over time')
xlabel('time(s)')
ylabel('running speed(cm/s)')
```



2b) Motion Correction

X and Y translation values in pixels required to correct for motion artifacts during the experiment are available as well.

```
close all
mc = bos1.get_motion_correction;
plot(table2array(mc(:,'timestamp')), table2array(mc(:,'x_motion')),table2array(mc(:,'timestamp'))
xlabel('motion correction over time')
xlabel('time(s)')
ylabel('motion correction over time(px)')
legend('x motion','y motion')
xlabel('time(s)')
```



Eye Tracking

Many experiments contain pupil position and pupil size from eye tracking. Extreme outliers from the tracking have been removed and replaced with NaN, which will appear as gaps in the plotted data. If an experiment does not have eye tracking data, a NoEyeTrackingException will be raised.

As shown in bos1.sSessionInfo, the session in bos1 doesn't have eye tracking data, so let's filer out all sessions with eye tracking data and takes the first session in the list to demonstrate this functionality.

```
bosf3 = bot.sessionfilter
```

bosf3.filter session by eye tracking(1)

```
ans =
  sessionfilter with properties:
```

```
filtered_session_table: [308×15 table]
    stimulus: {9×1 cell}
    targeted_structure: {6×1 cell}
    imaging_depth: [10×1 double]
        container_id: [137×1 double]
        session_id: [308×1 double]
        session_type: [308×1 categorical]
            cre_line: {6×1 cell}
        eye_tracking_avail: 1
            failed: 0
```

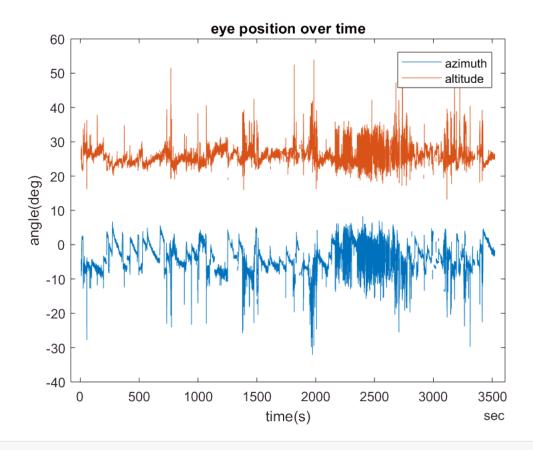
bos2 = bot.session(bosf3.session id(1))

```
bos2 =
  session with properties:
```

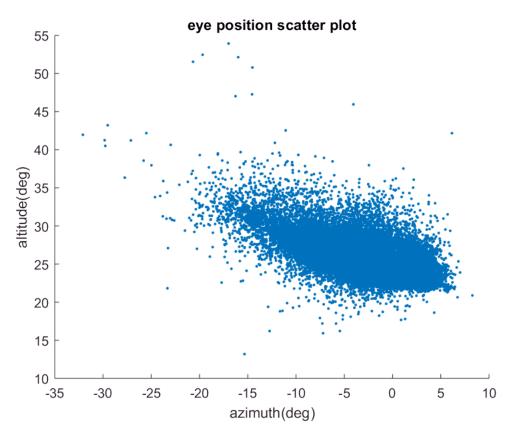
sSessionInfo: [1×1 struct]

 $strLocal NWBFile Location: \ 'C: \ Users \ 14868 \setminus Brain-Observatory-Toolbox \ +bot \ Cache \setminus extends \ Archive the location \ 'C: \ Users \ 14868 \setminus Brain-Observatory-Toolbox \ +bot \ Cache \setminus extends \ Archive the location \ 'C: \ Users \ 14868 \setminus Brain-Observatory-Toolbox \ +bot \ Cache \setminus extends \ Archive the location \ 'C: \ Users \ 14868 \setminus Brain-Observatory-Toolbox \ +bot \ Cache \setminus extends \ Archive the location \ 'C: \ Users \ 14868 \setminus Brain-Observatory-Toolbox \ +bot \ Cache \setminus extends \ Archive the location \ 'C: \ Users \ 14868 \setminus Brain-Observatory-Toolbox \ +bot \ Cache \setminus extends \ Archive the location \ 'C: \ Users \ 14868 \setminus Brain-Observatory-Toolbox \ +bot \ Cache \setminus extends \ Archive the location \ Archive the locati$

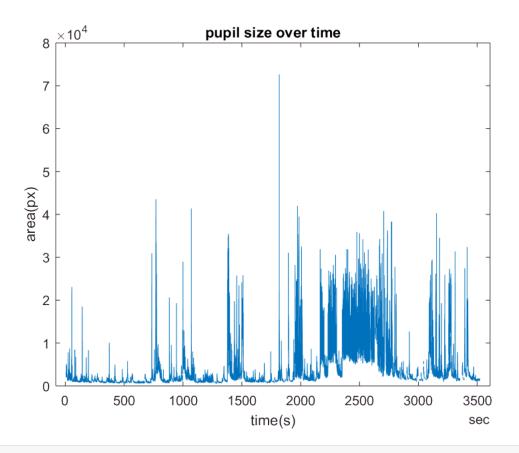
```
[time, pupilLocation] = bos2.get_pupil_location;
azimuth = pupilLocation(:,1);
altitude = pupilLocation(:,2);
plot(time, azimuth, time, altitude)
title('eye position over time')
legend ('azimuth', 'altitude')
xlabel('time(s)')
ylabel('angle(deg)')
```



```
scatter(azimuth, altitude, '.')
title('eye position scatter plot')
xlabel('azimuth(deg)')
ylabel('altitude(deg)')
```



```
[time, pupilArea] = bos2.get_pupil_size;
plot(time, pupilArea)
title('pupil size over time')
xlabel('time(s)')
ylabel('area(px)')
```



3) Call function convert_fluorescence_trace_into_raster_format to:

Convert fluorescence trace into data that is in "raster format" for further analysis

Let's save the 'raster format' DfOVerF data responding to static gratings during the session of bos1 in dir ~/raster/static_gratings/static_gratings_511458874/

convert_fluorescence_trace_into_raster_format('Df0verF', bos1.sSessionInfo.id,'static_grating

All raster files already exist in directory [raster\static_gratings\static_gratings_511458874].