

# WS: BRAIN-COMPUTER INTERFACE USING OPENVIBE, AN OPEN-SOURCE SOFTWARE PLATFORM – PART 2



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# CHAPTER 1

PREREQUISITES BEFORE PERFORMING A MI-BCI EXPERIMENT

# DIFFERENT TYPES OF BCI

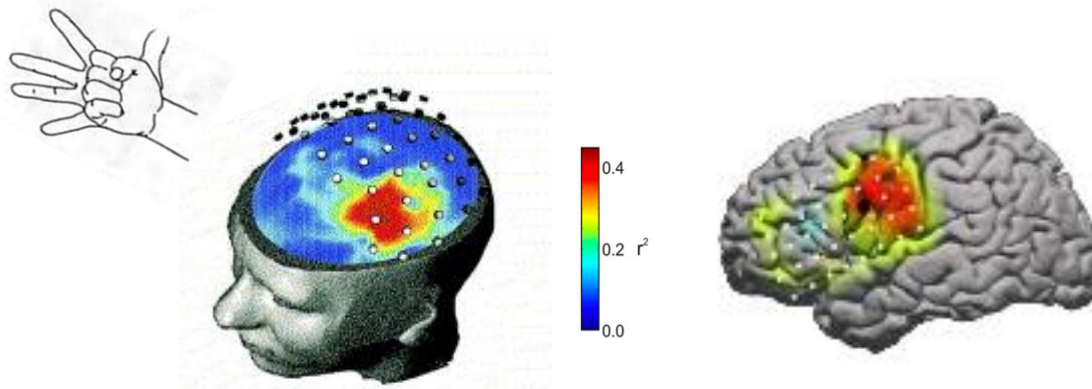
## Underlying idea

Taking advantage of a neurophysiological phenomenon to establish a communication between the brain and the computer

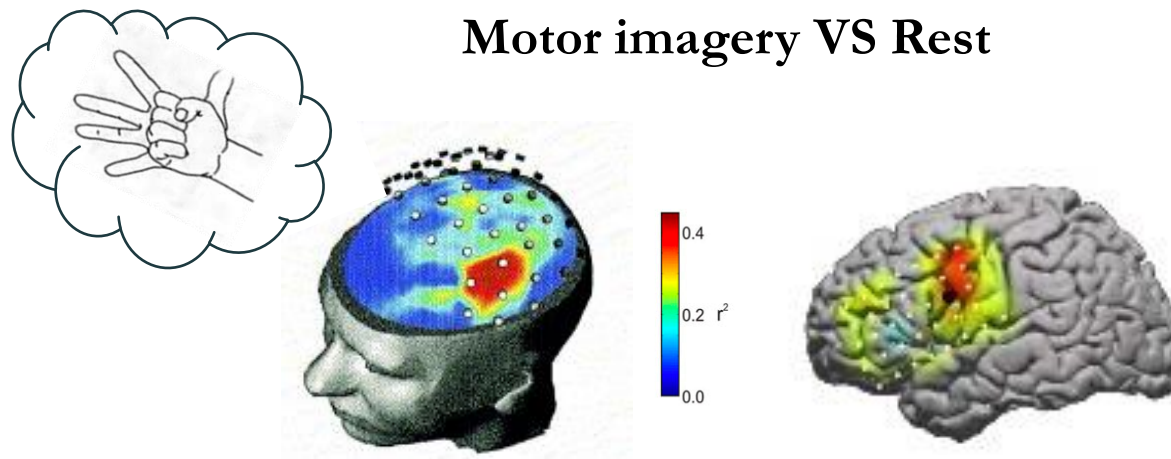
Illustration with Motor imagery-based BCI

# MOTOR IMAGERY – OBSERVATIONS

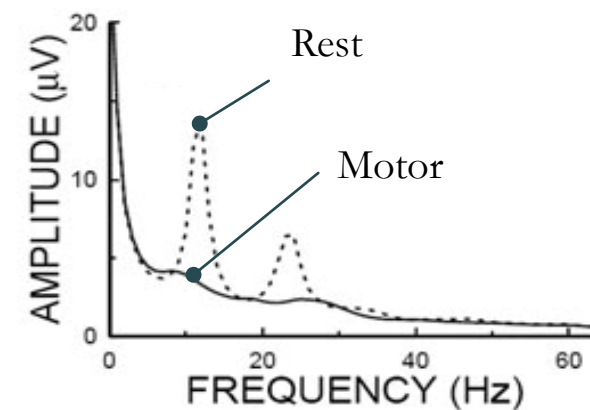
## Motor execution VS Rest



## Motor imagery VS Rest



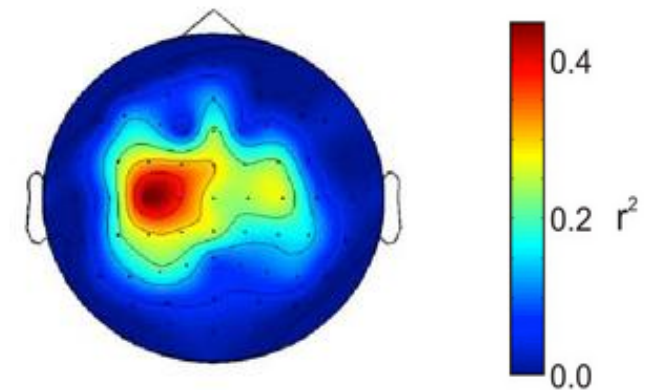
## Power decrease



Desynchronization effect  
(Pfurtscheller et al, 1999)

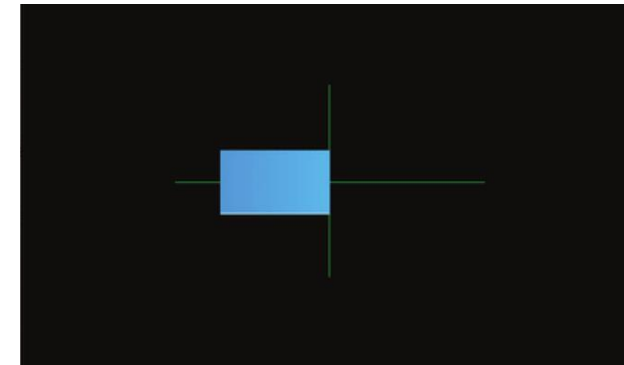
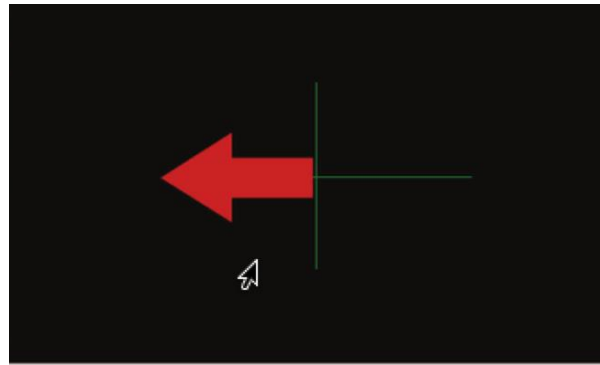
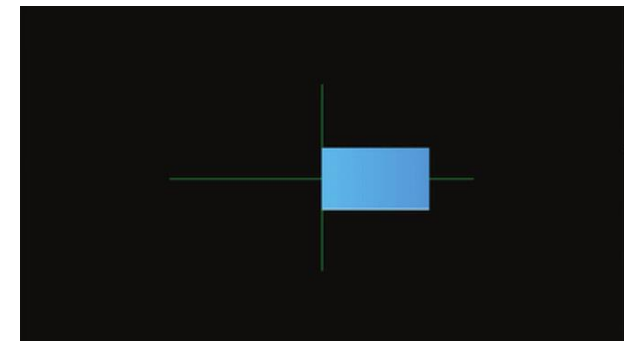
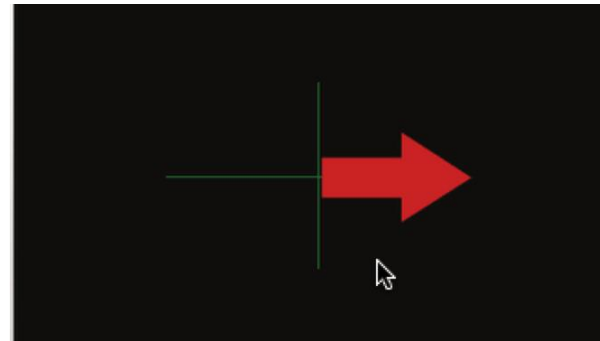
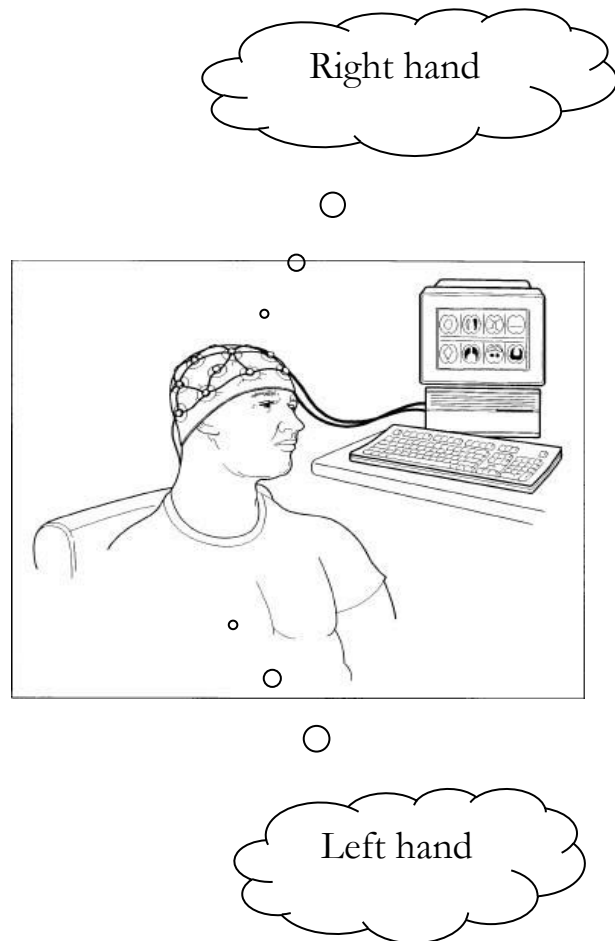
# MOTOR IMAGERY – MU-BETA RHYTHM

- Behavioral properties
  - Movement / preparation for movement : Event-related desynchronization (ERD) (Pfurtscheller, G, Lopes da Silva, FH, 1999)
  - With relaxation/post-movement period : ERS
- Why using it in BCI ?
  - Mu/Beta activity modulation by motor-imagery, a way to communicate
  - Use of power spectra
  - To establish this communication :
    - Spatial selection
    - Frequency selection

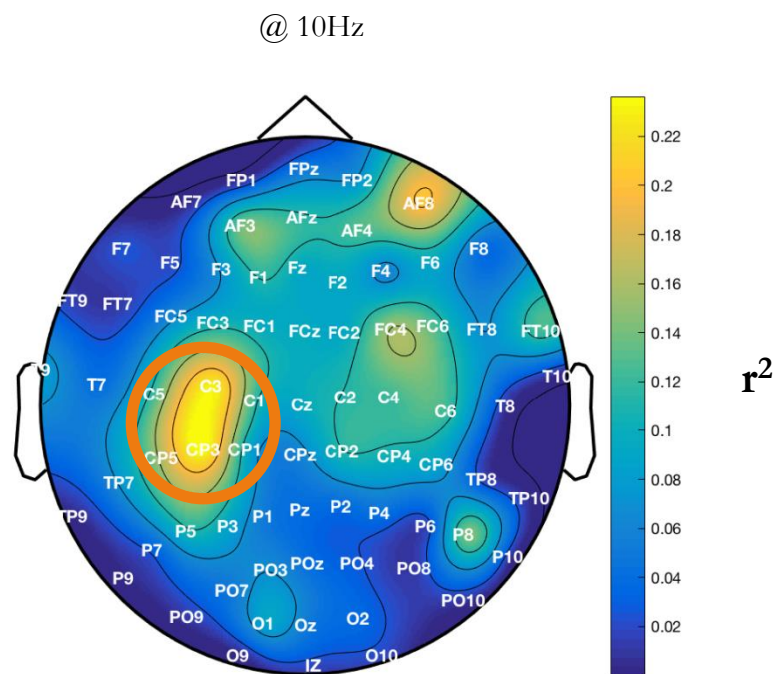


Illustrations from BCI2000 website

# MOTOR IMAGERY – IN PRACTICE

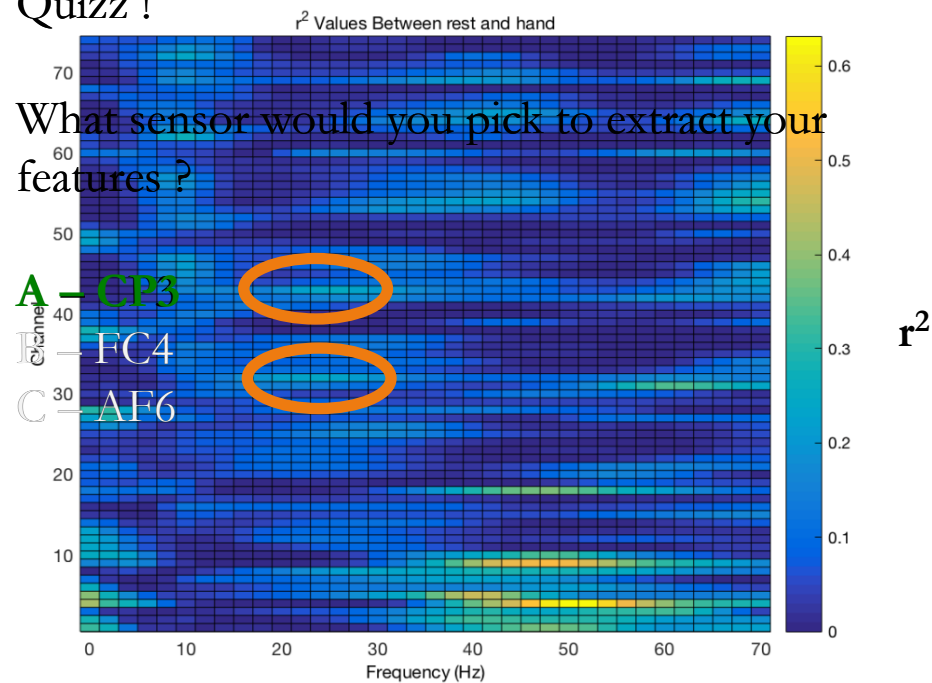


# MOTOR IMAGERY – IN PRACTICE

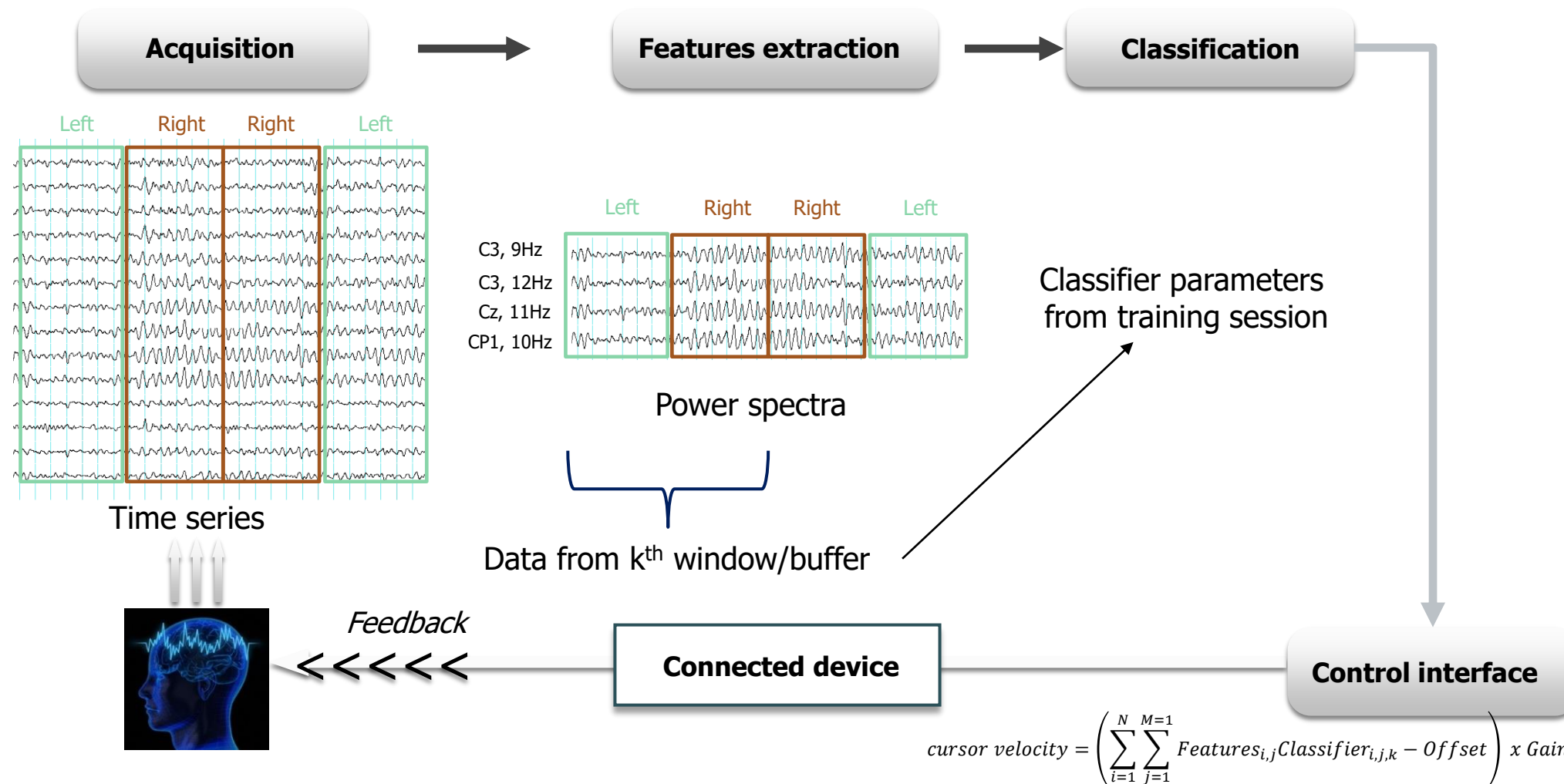


Quizz !

What sensor would you pick to extract your features ?

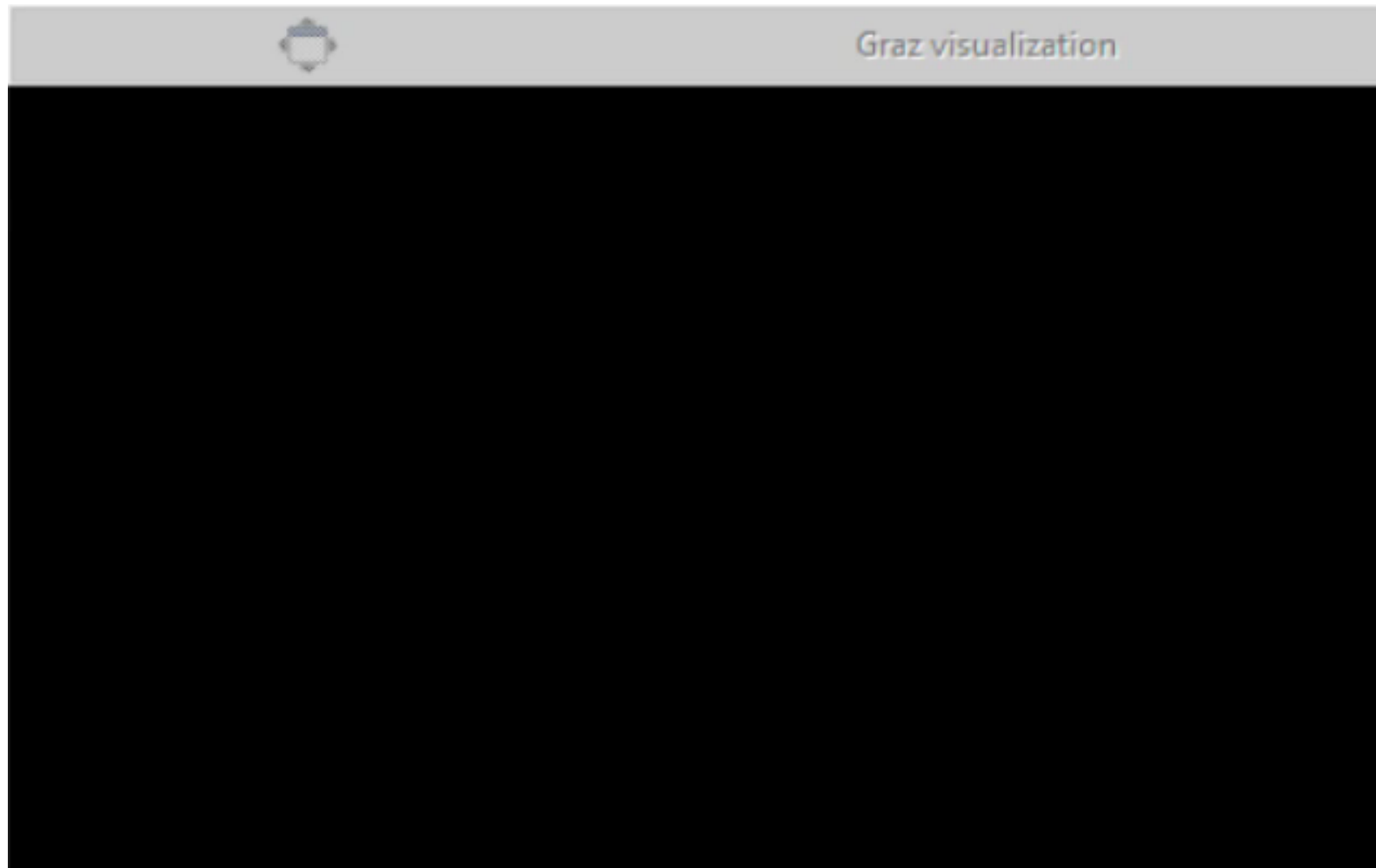


# MOTOR IMAGERY – IN PRACTICE





# MOTOR IMAGERY – IN PRACTICE





# CHAPTER 2

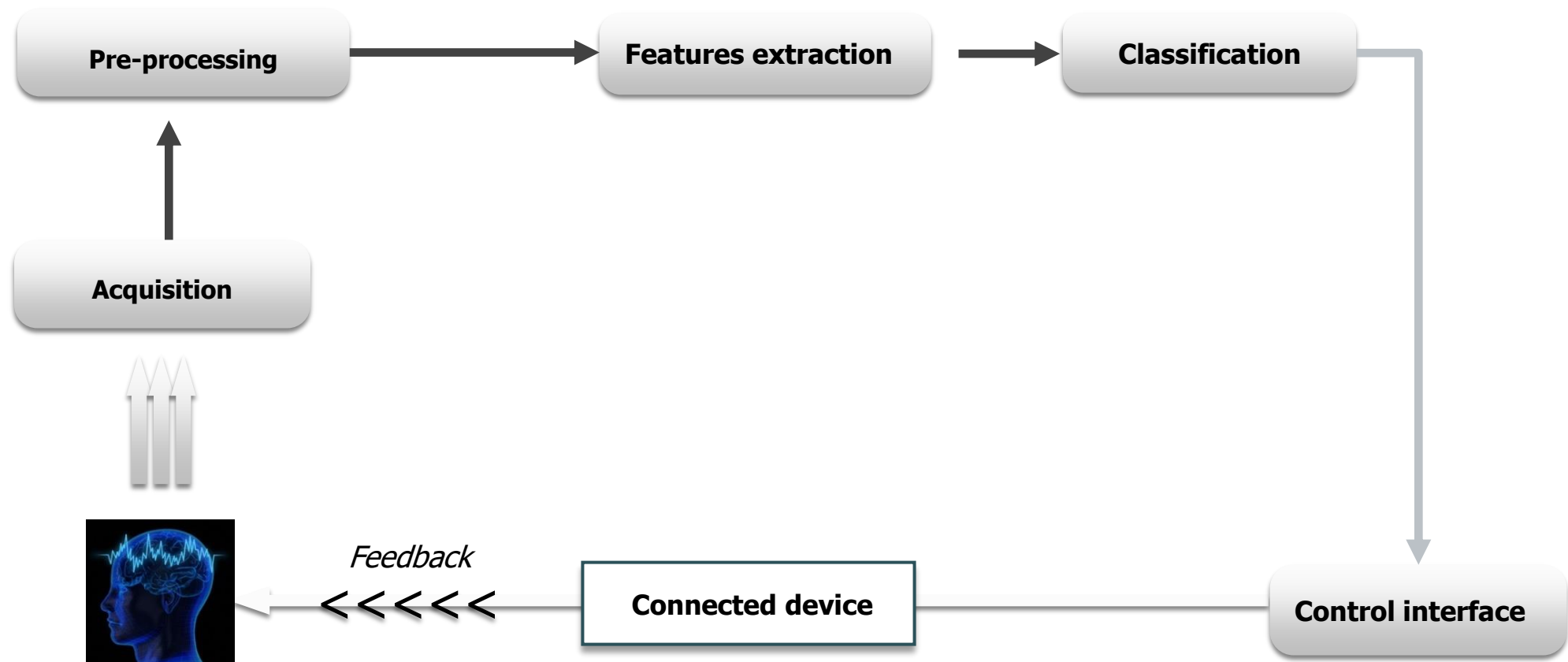
## BCI-CUTTINGEEG PROTOCOL SET-UP

## RESOURCES

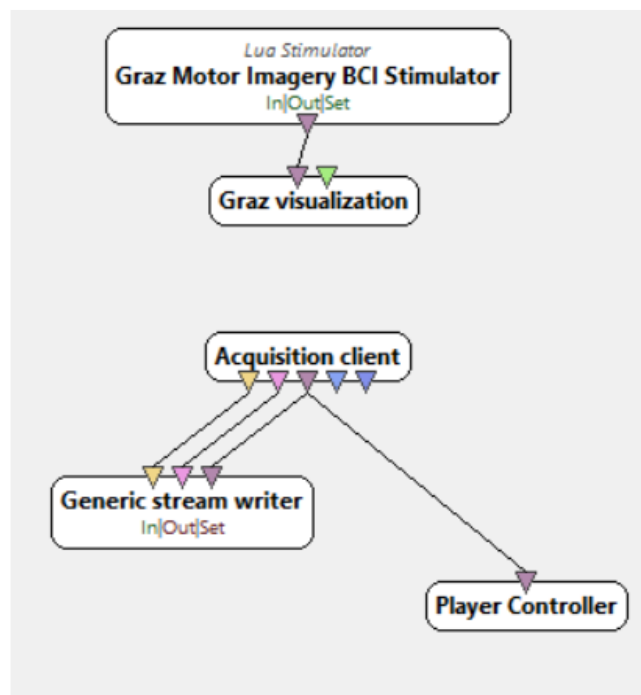
- GitHub repo:

<https://github.com/BCI-NET/BCI-OpenViBE-CuttingEEG2021/>

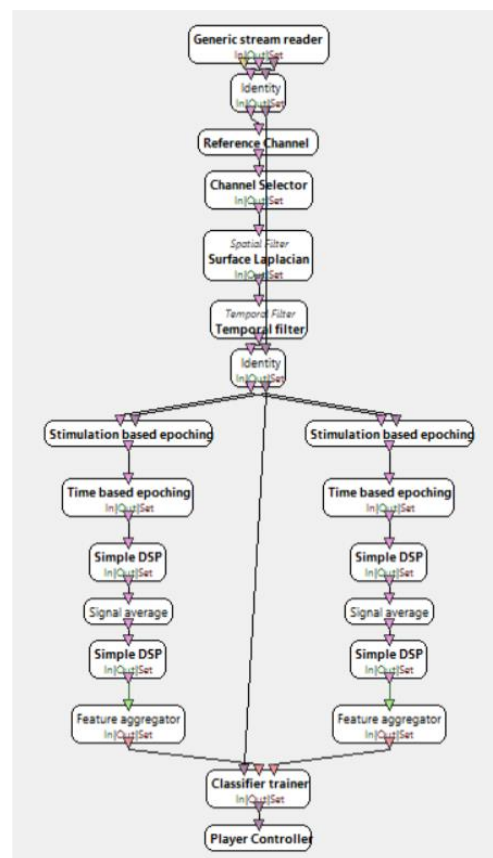
# AIM OF THE PART



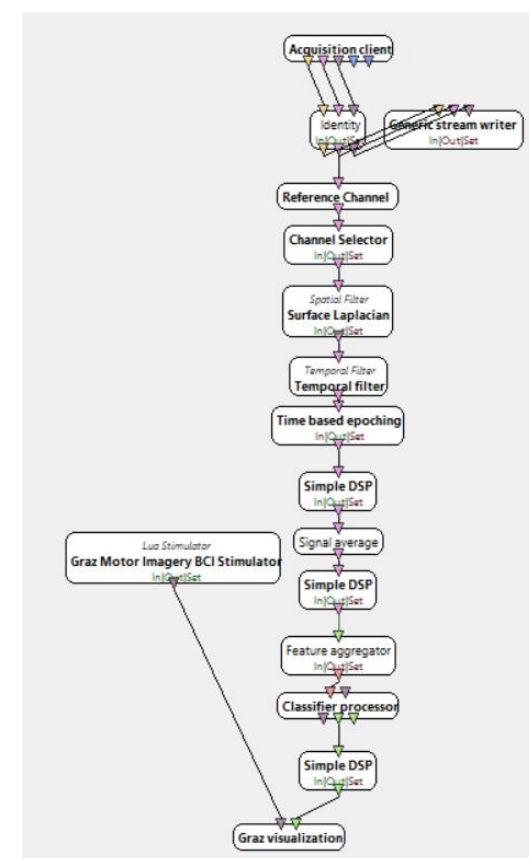
# AIM OF THE PART



1- Data acquisition

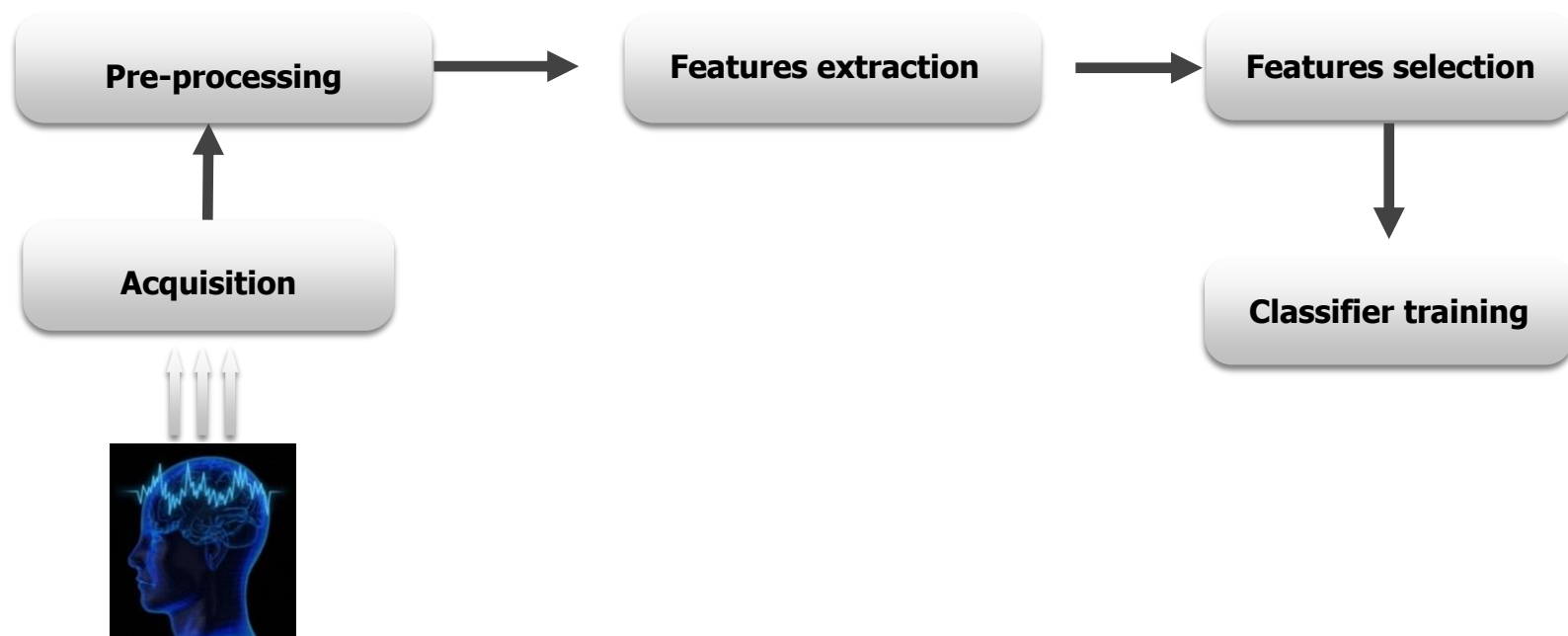


2- Features extraction  
& Classification



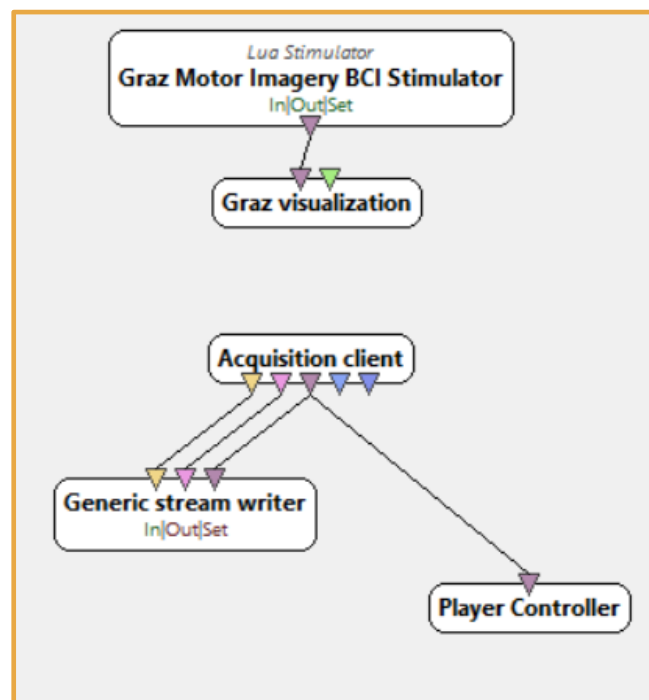
3- Online feedback

# AIM 1 – BUILDING THE CLASSIFIER...

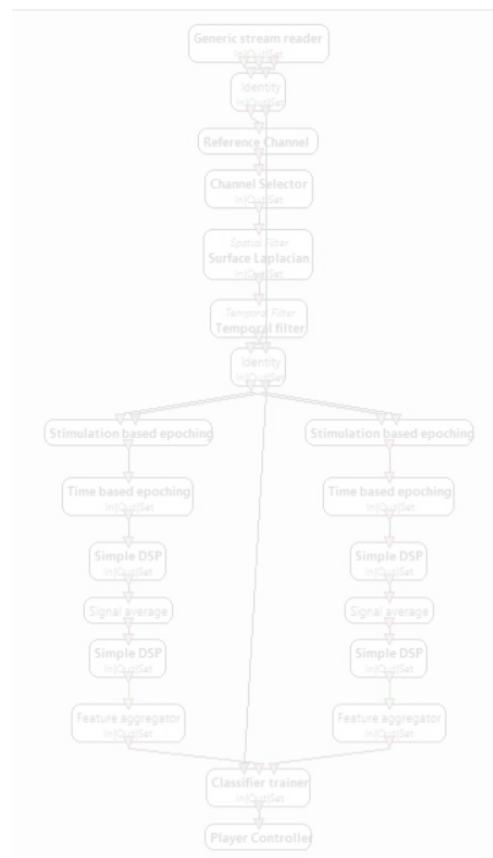


*Adapted from (X. Navarro & F. Grosselin)*

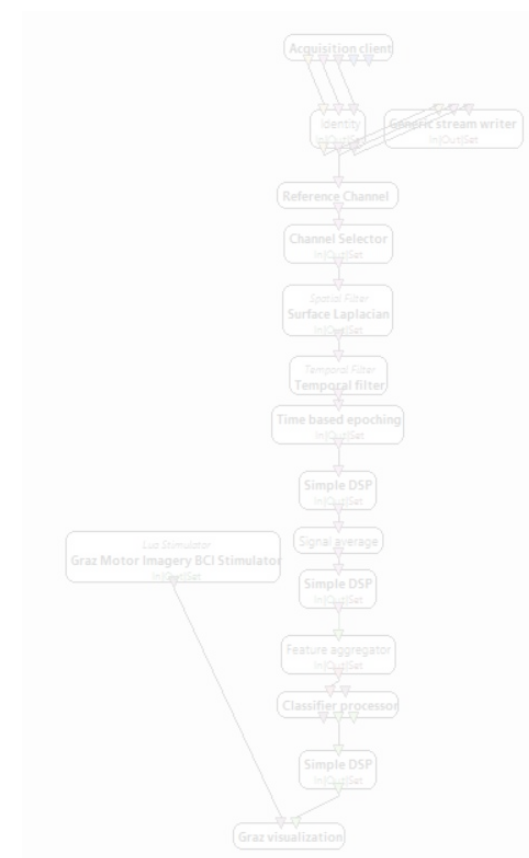
# SC 1 – DATA ACQUISITION



1- Data acquisition

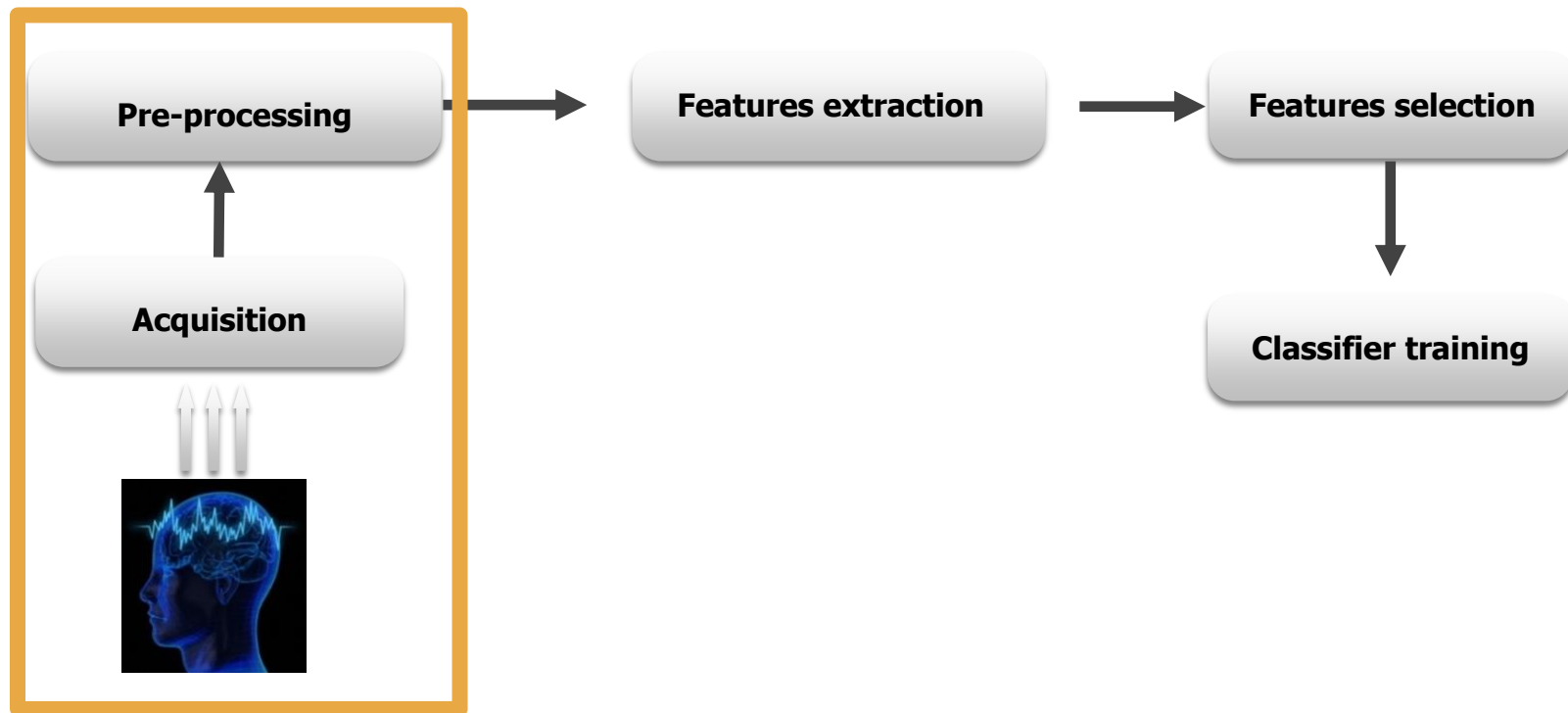


2- Features extraction  
& Classification



3- Online feedback

# SC 1 – DATA ACQUISITION



*Adapted from (X. Navarro & F. Grosselin)*



# SC1 – DATA ACQUISITION

- **Sub-chapters:**
  - Warm-up, first steps: simple I/O
  - Warm-up 2: acquisition server
  - BCI acquisition protocol, Stimulations

# SC1 - STEP 1 – WARMUP ! SIMPLE I/O AND DISPLAY

- **Sub-chapters:**
  - Warm-up, first steps: simple I/O
  - Warm-up 2: acquisition server
  - BCI acquisition protocol, Stimulations

# SC1 - STEP 1 – WARMUP ! SIMPLE I/O AND DISPLAY

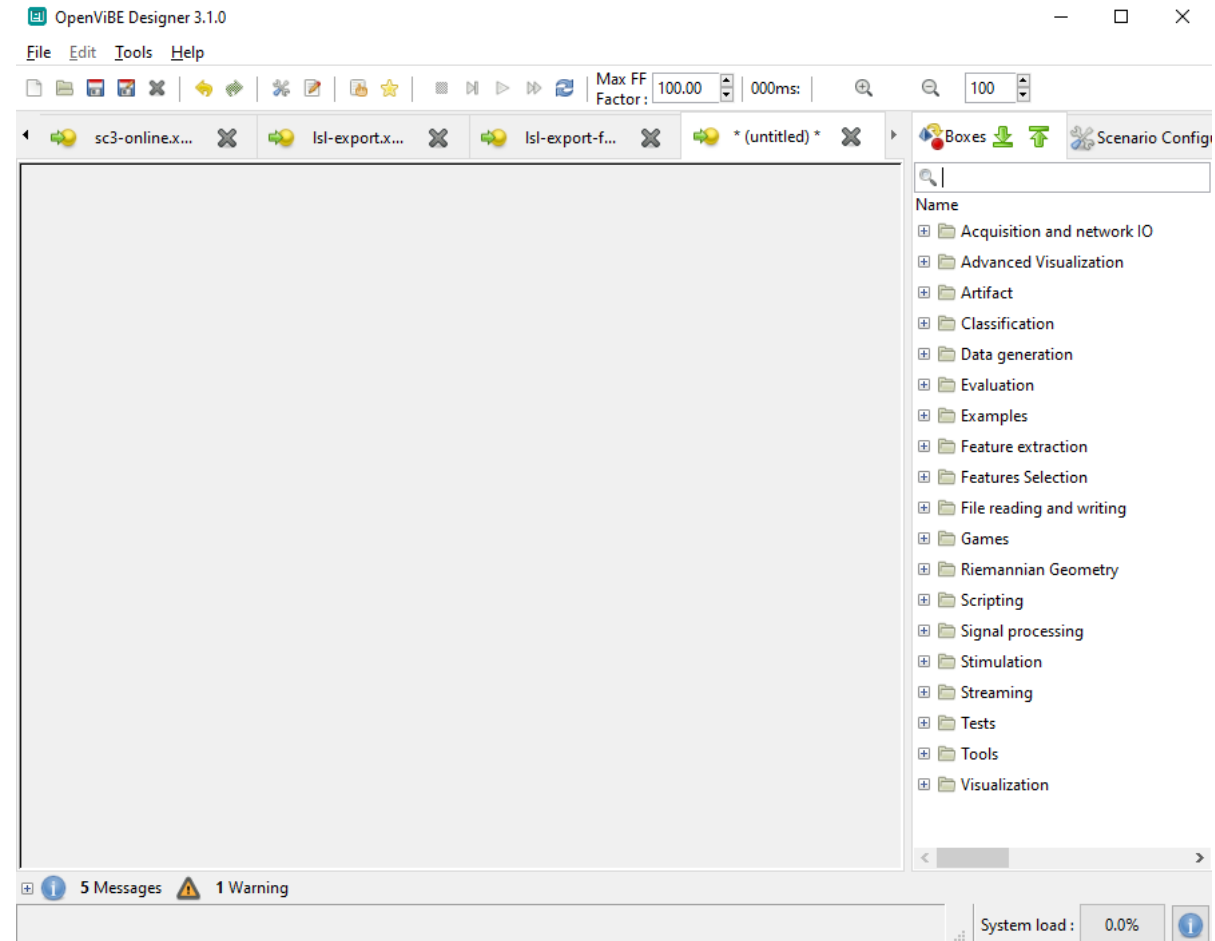
- Goal : load & display file :

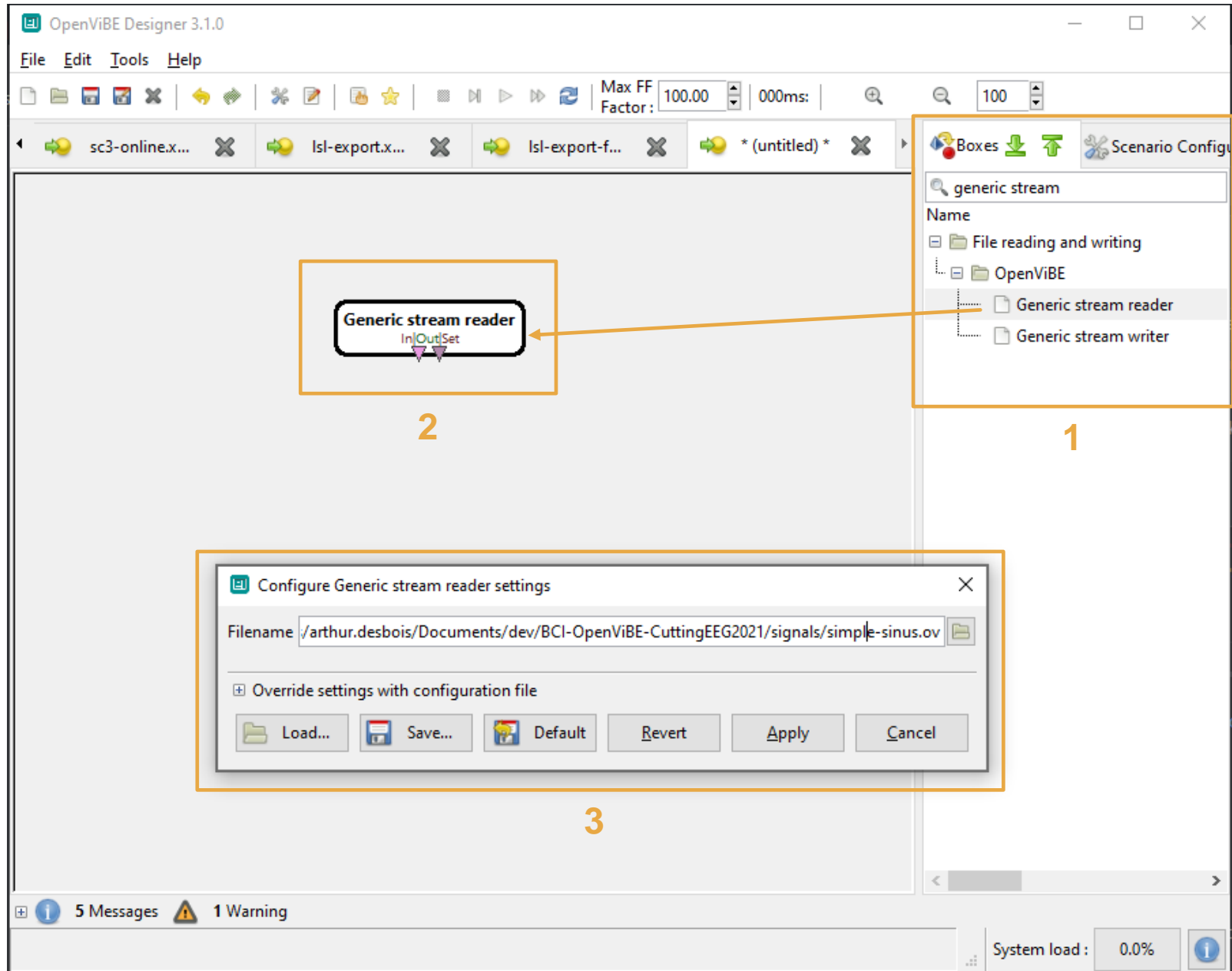
signals/[simple-sinus.ov](https://github.com/owensml/ovibe-simple-sinus)

(in Github repo)

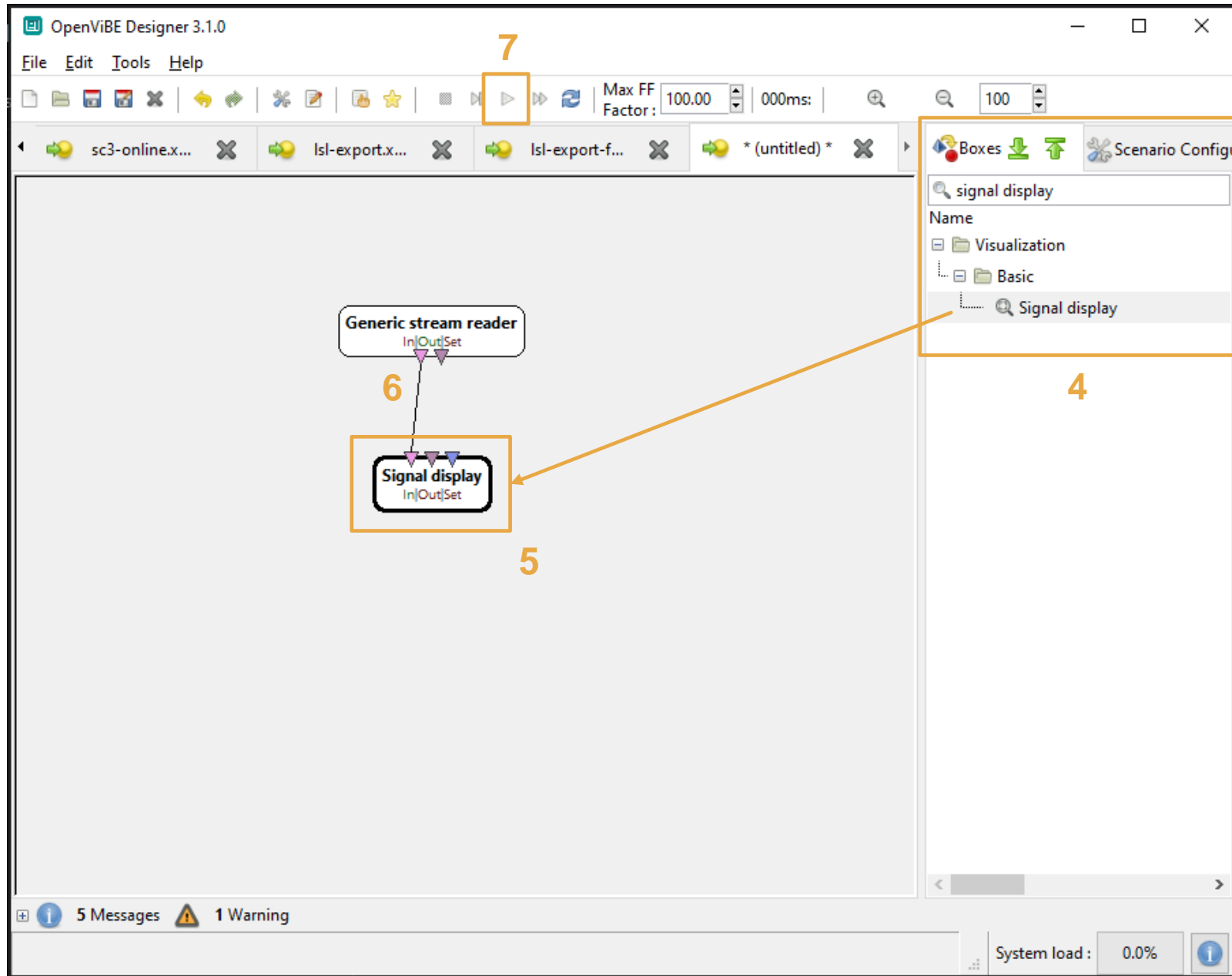
- 0- Launch OpenViBE Designer :

`openvibe-designer.cmd`





- 1- Search for “generic stream reader” in boxes list
- 2- Drag & drop to designer window
- 3- Set filename with browser



- 4- Search for “signal display” in boxes list
- 5- Drag & drop to designer window
- 6- Drag & drop connection between boxes, using the “signal” stream type
- 7- Press Play!

# SC1 - STEP 2 – WARMUP ! ACQ SERVER

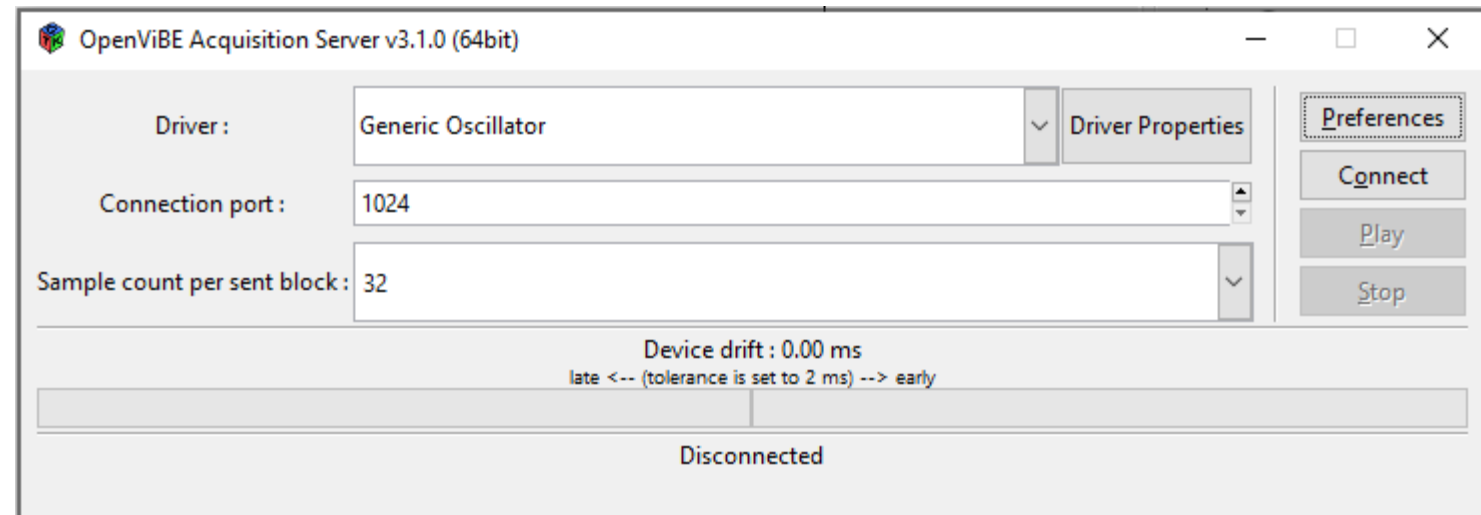
- **Sub-chapters:**

- Warm-up, first steps: simple I/O
- Warm-up 2: acquisition server
- BCI acquisition protocol, Stimulations

## SC1 - STEP 2 – WARMUP ! ACQ SERVER

- **Goal :** use acquisition server/client with a simple sine generator
- **0- Launch OpenViBE Acquisition server :**

`openvibe-acquisition-server.cmd`

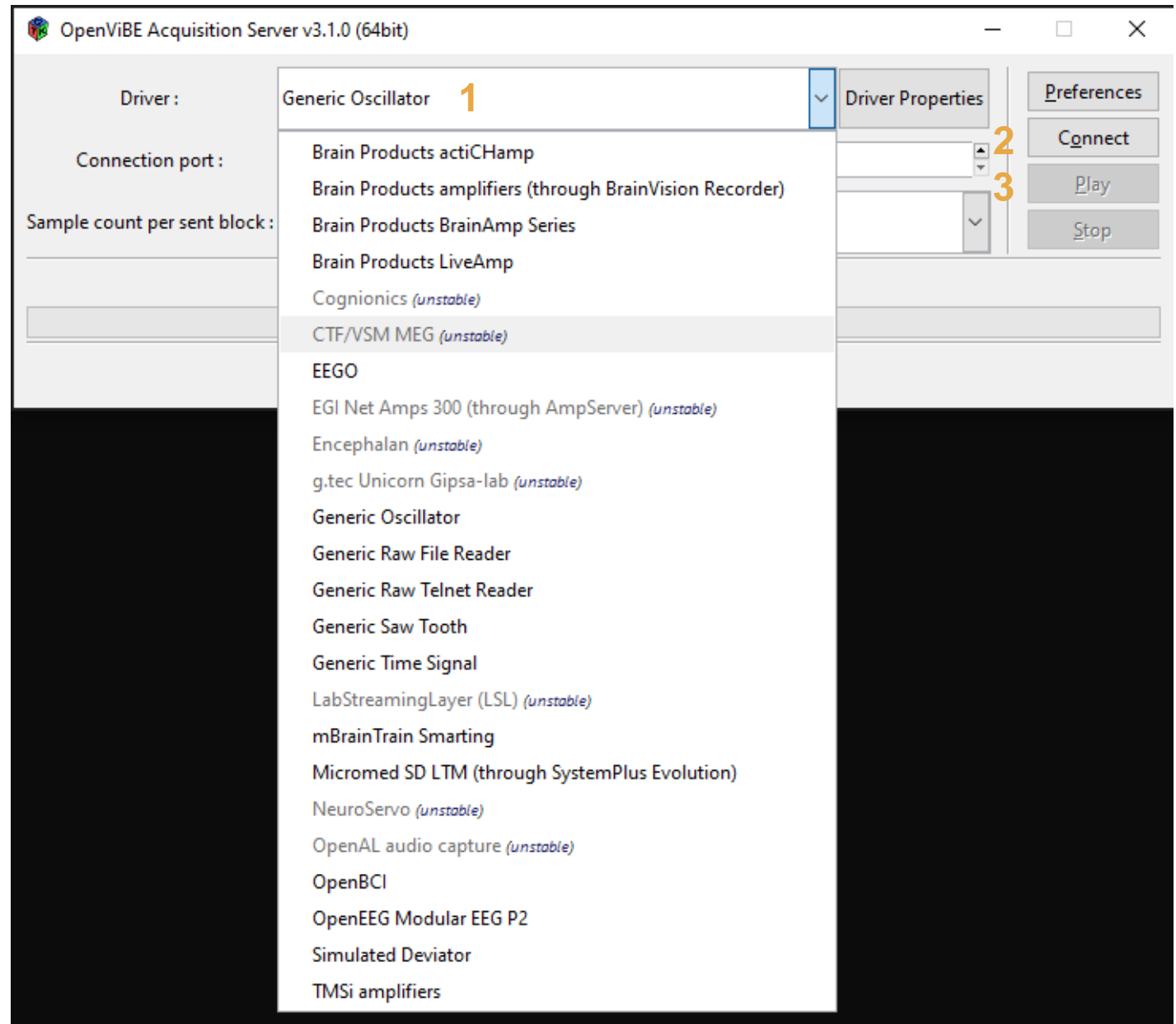


- 1- In the “Driver” list, select “Generic Oscillator”

In this list, you’ll find all the drivers for OpenViBE’s supported EEG hardware

- 2- Click “Connect”

- 3- Click “Play”







## SC1 - STEP 3 – PROTOCOL MANAGEMENT & STIMULATIONS...

- **Sub-chapters:**
  - Warm-up, first steps: simple I/O
  - Warm-up 2: acquisition server
  - BCI acquisition protocol, Stimulations

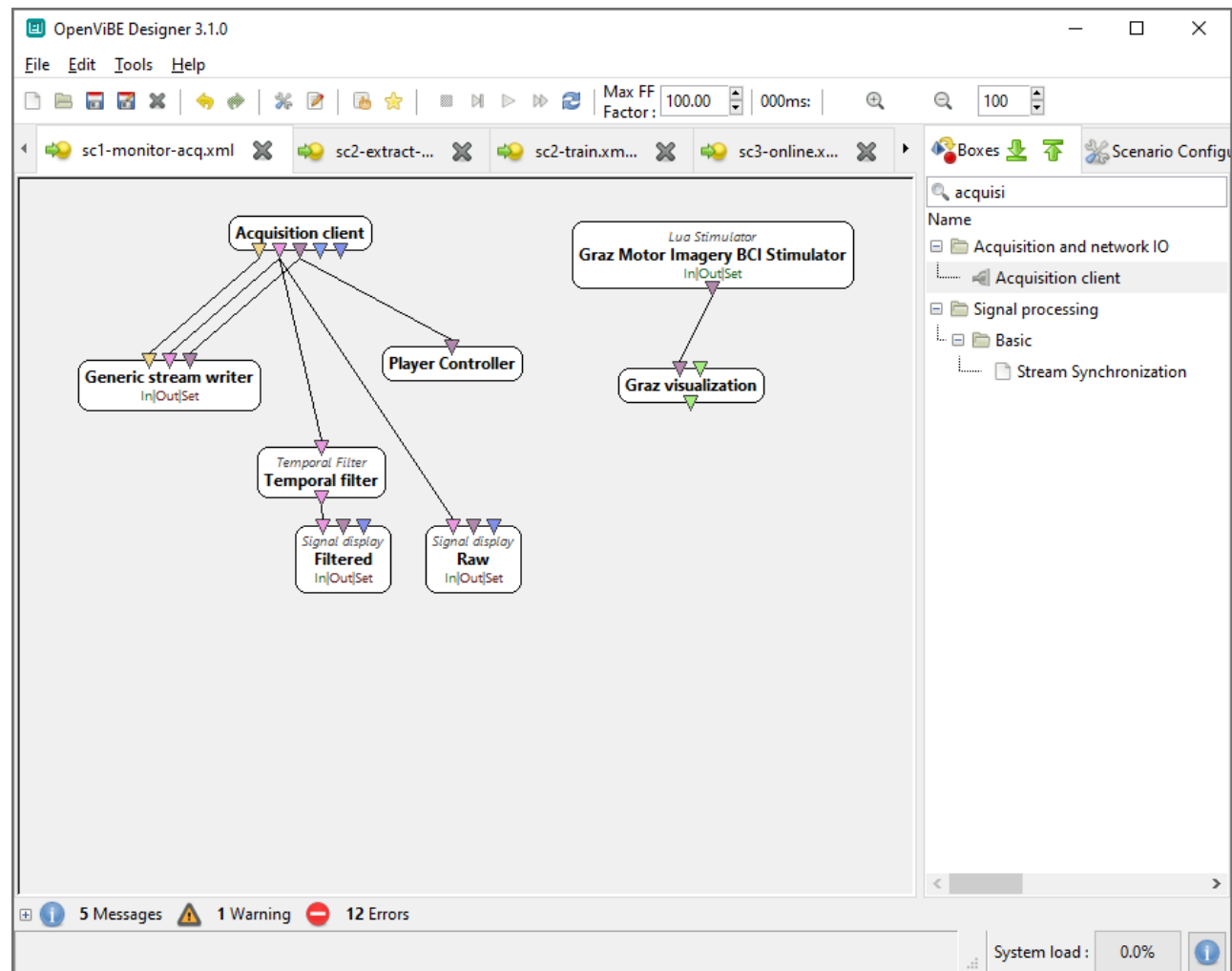
# SC1 - STEP 3 – PROTOCOL MANAGEMENT & STIMULATIONS...

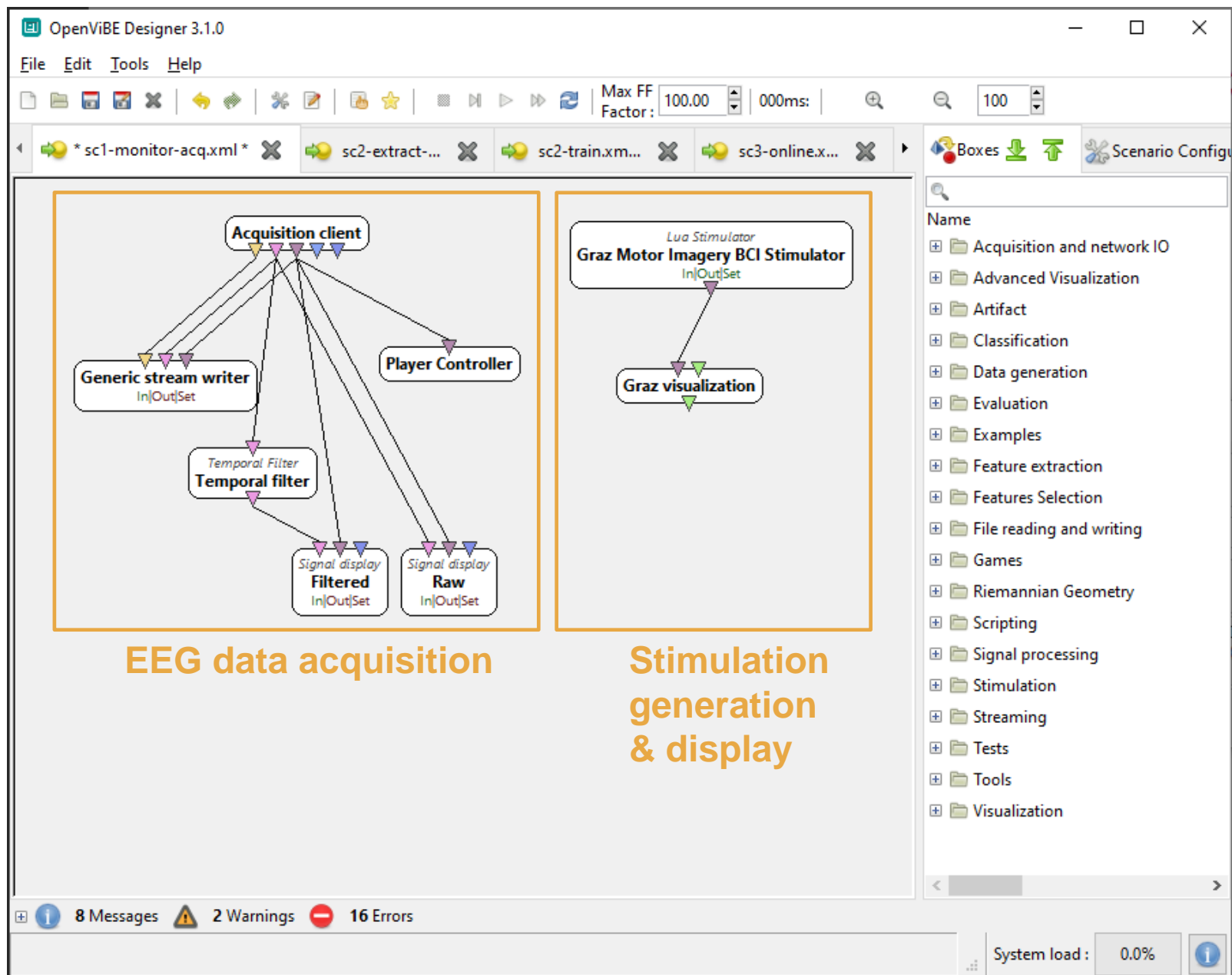
- Goal : understand data acquisition & synchro. with stimulations

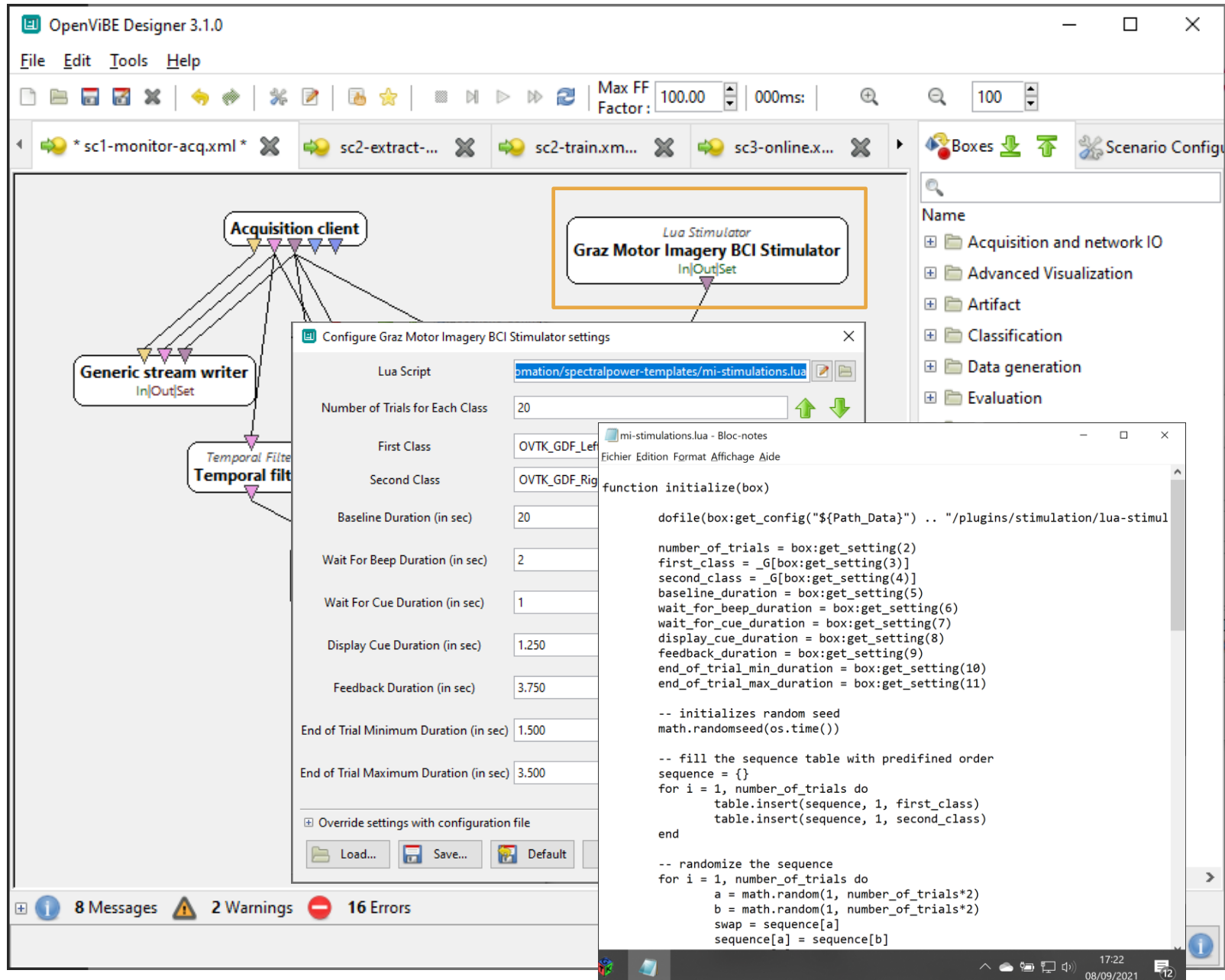
- Load scenario:

**BCI-OpenViBE-CuttingEEG2021/  
scenarios/sc1-monitor-acq.xml**

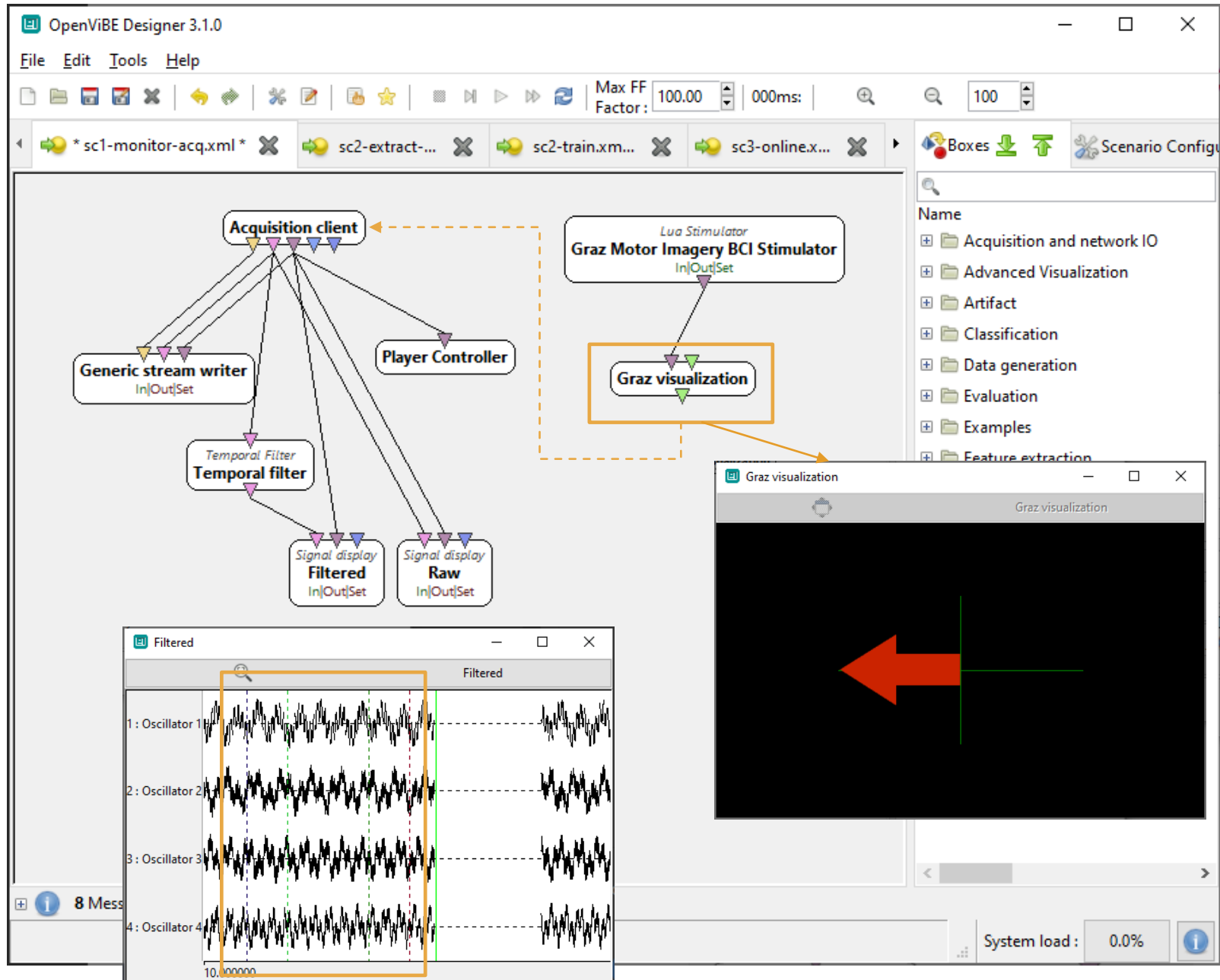
(github)



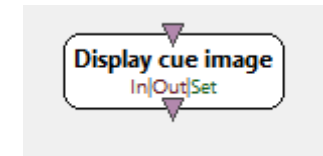


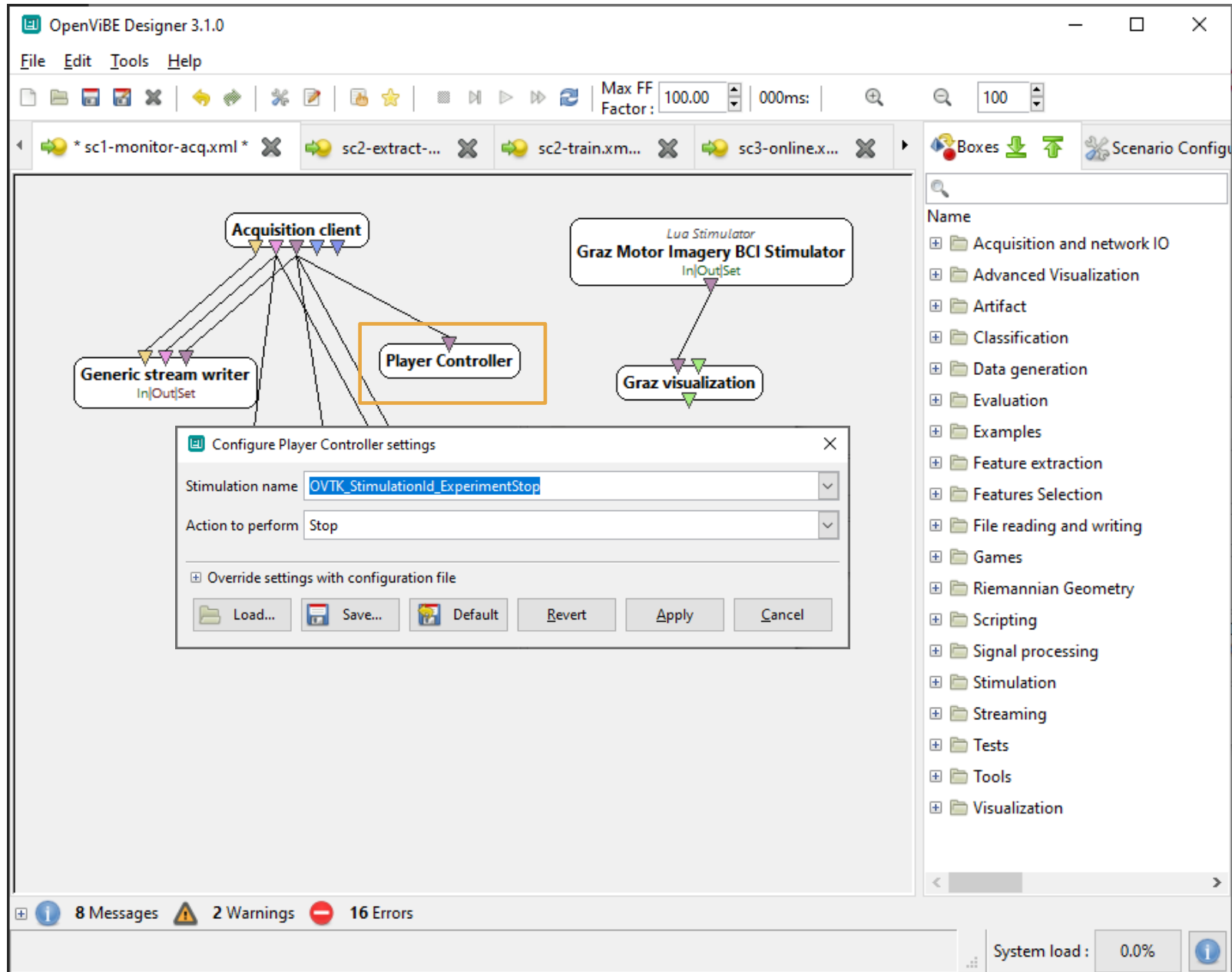


- **LUA Stimulator Box**  
(LUA = scripting language)
- **Experiment example:**  
Different stimulation/event codes at different times.  
Useful for signal segmentation (“epoching”)
- **Stimulation label examples:**
  - Experiment Start/Stop
  - Trial Start/Stop
  - LEFT / RIGHT
  - Button pressed
  - etc



- **“Graz Visualization”**  
(specific for “Graz Protocol”, based on Box “Display Cue Image”)
- Displays specified image upon receiving specified stimulation code
- Transmits the **stimulation code & time** to the Acquisition Server
- The stimulations are received by the ACQ Client, synchronized with the signal





- **“Player Controller”**  
Orchestrates the experiment course, by applying an action upon receiving a stimulation

OpenViBE Designer 3.1.0

File Edit Tools Help

Max FF Factor: 100.00 | 000ms: | 100

\*sc1-monitor-acq.xml\* sc2-extract-... sc2-train.xm... sc3-online.x...

Boxes Scenario Config

Acquisition client

Generic stream writer In|Out|Set

Player Controller

Graz Motor Imagery BCI Stimulator Lua Stimulator In|Out|Set

Graz visualization

Configure Player Controller settings

Stimulation name: OVTk\_StimulationId\_ExperimentStop

Action to perform: Stop

Override settings with configuration file

Load... Save... Default Revert Apply Cancel

8 Messages 2 Warnings 16 Errors

System load: 0.0%

- Check the embedded log console if anything goes wrong!



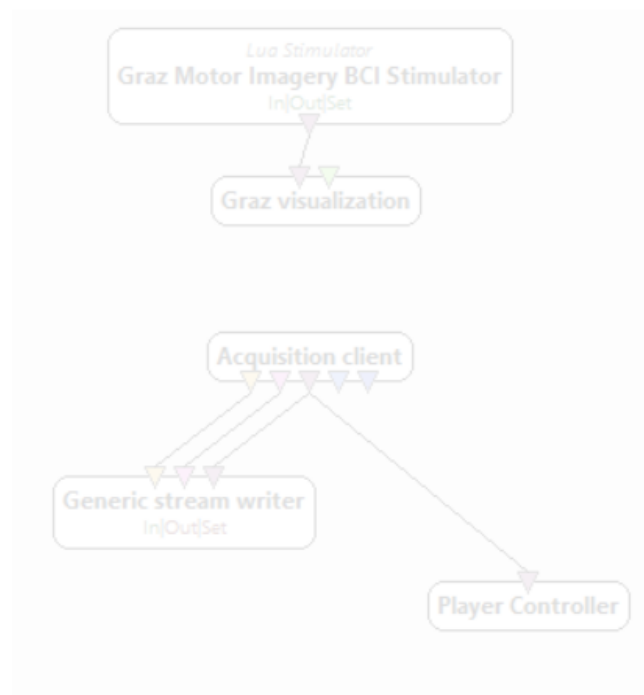
## SC1 – FINAL WORDS / Q&A

### ■ **Recap:**

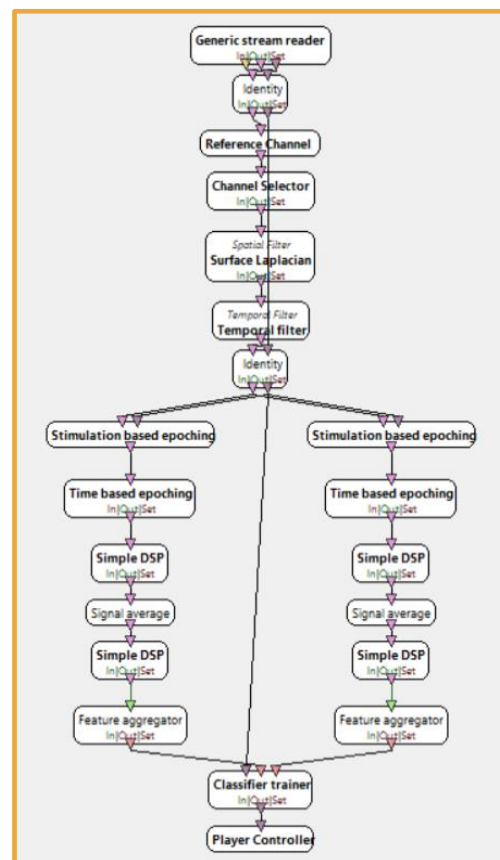
- Simple I/O operations (signal loading, writing)
- Signal Display
- Acquisition Server / Client concept
- Stimulations
- Typical BCI protocol acquisition scenario

- **Note: from now on, we will only be using pre-generated/recorded signals**

# SC2 – CLASSIFIER TRAINING



1- Data acquisition

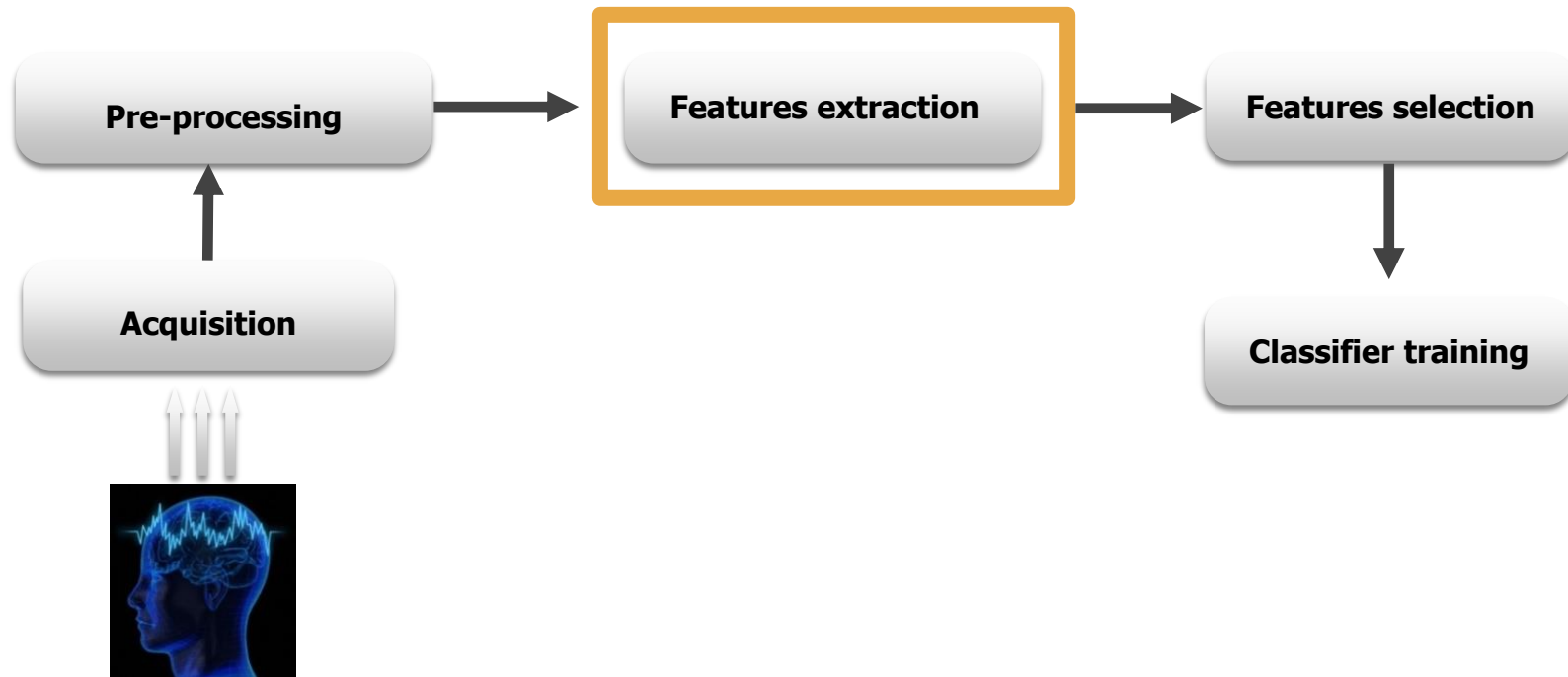


2- Features extraction  
& Classification



3- Online feedback

## SC2 – CLASSIFIER TRAINING

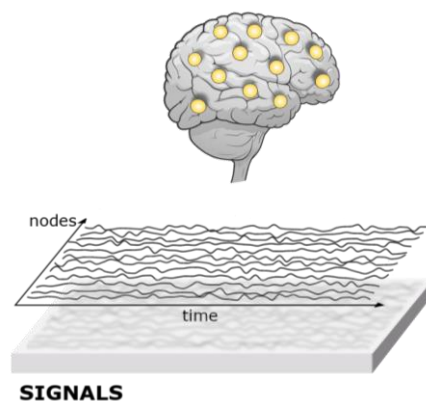


*Adapted from (X. Navarro & F. Grosselin)*

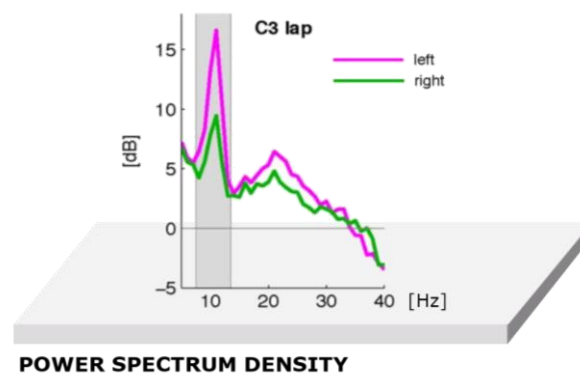
## STEP 2 – SIGNAL PROCESSING & VISUALIZATION

### Features to extract (recap)

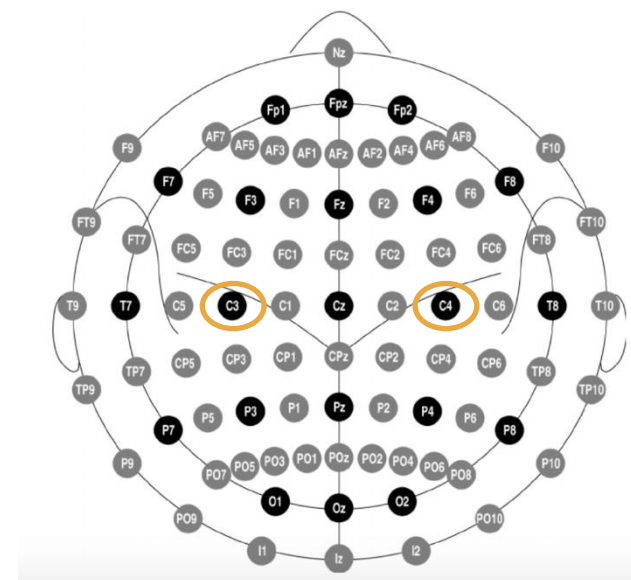
- Power spectra
- Sensorimotor area
- Mu: 8-12Hz &/OR Beta: 14-29Hz



(Gonzalez-Astudillo et al, 2020)



(Maeder et al., 2012)



## SC2 – CLASSIFIER TRAINING

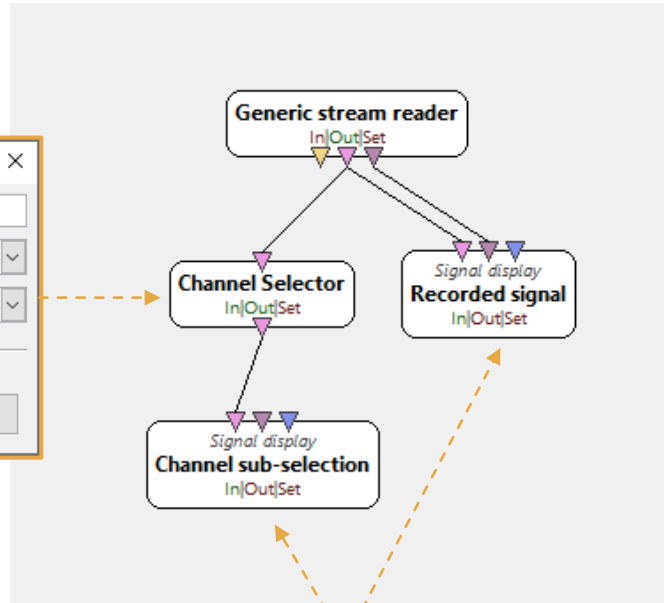
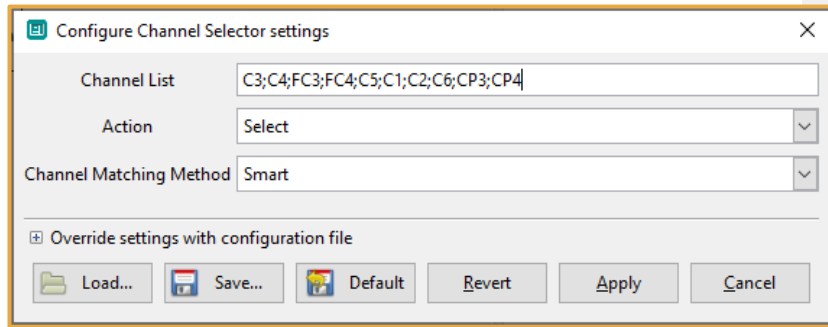
- **Sub-chapters:**
  - Channel Selection
  - Basic Signal Processing / Spectral analysis
  - Topography visualization
  - Feature Selection + Spatial Filtering
  - Classifier training

## SC2 – STEP 1 - CHANNEL SELECTION

- **Sub-chapters:**
  - Channel Selection
  - Basic Signal Processing / Spectral analysis
  - Topography visualization
  - Feature Selection + Spatial Filtering
  - Classifier training

## SC2 – STEP 1 - CHANNEL SELECTION

- **OpenViBE** can manage multiple signal streams in parallel
- In our example, **1 channel = 1 EEG electrode**
- We will load a pre-recorded signal file, that used 11 labeled electrodes and consider only a sub-set of those electrodes.



Note:  
Right click + Rename  
can be useful when  
designing a scenario...

Display windows will use  
« renamed » label

- 1- Import a “Generic stream reader” box, and set the filename to:

```
<opencv-3.1.0-64bit>\
share\opencv\scenarios\signals\
bci-motor-imagery.ov
```

- (optional – have a look at the recorded signals with “**Signal Display**”)
- 2- Import a “**Channel Selector**” & link the boxes
- 3- Set the selection to :  
  
C3;C4;FC3;FC4;C5;C1;  
C2;C6;CP3;CP4
- 4- Display the output



## SC2 – STEP 2 – SIGNAL PROCESSING & SPECTRAL ANALYSIS

- **Sub-chapters:**
  - Channel Selection
  - Basic Signal Processing / Spectral analysis
  - Topography visualization
  - Feature Selection + Spatial Filtering
  - Classifier training

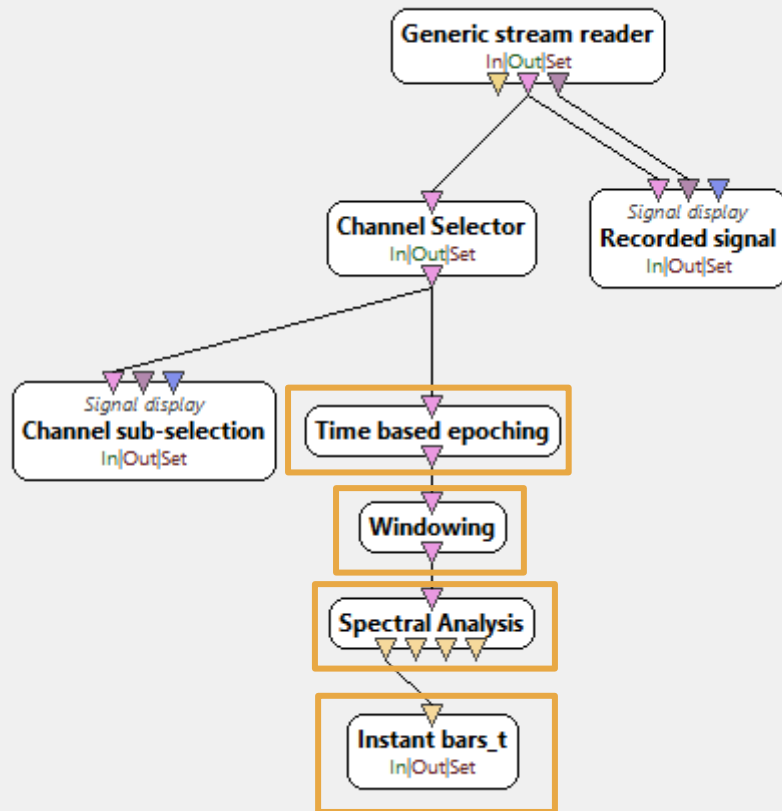
## SC2 – STEP 2 – SIGNAL PROCESSING & SPECTRAL ANALYSIS

### ■ Goals:

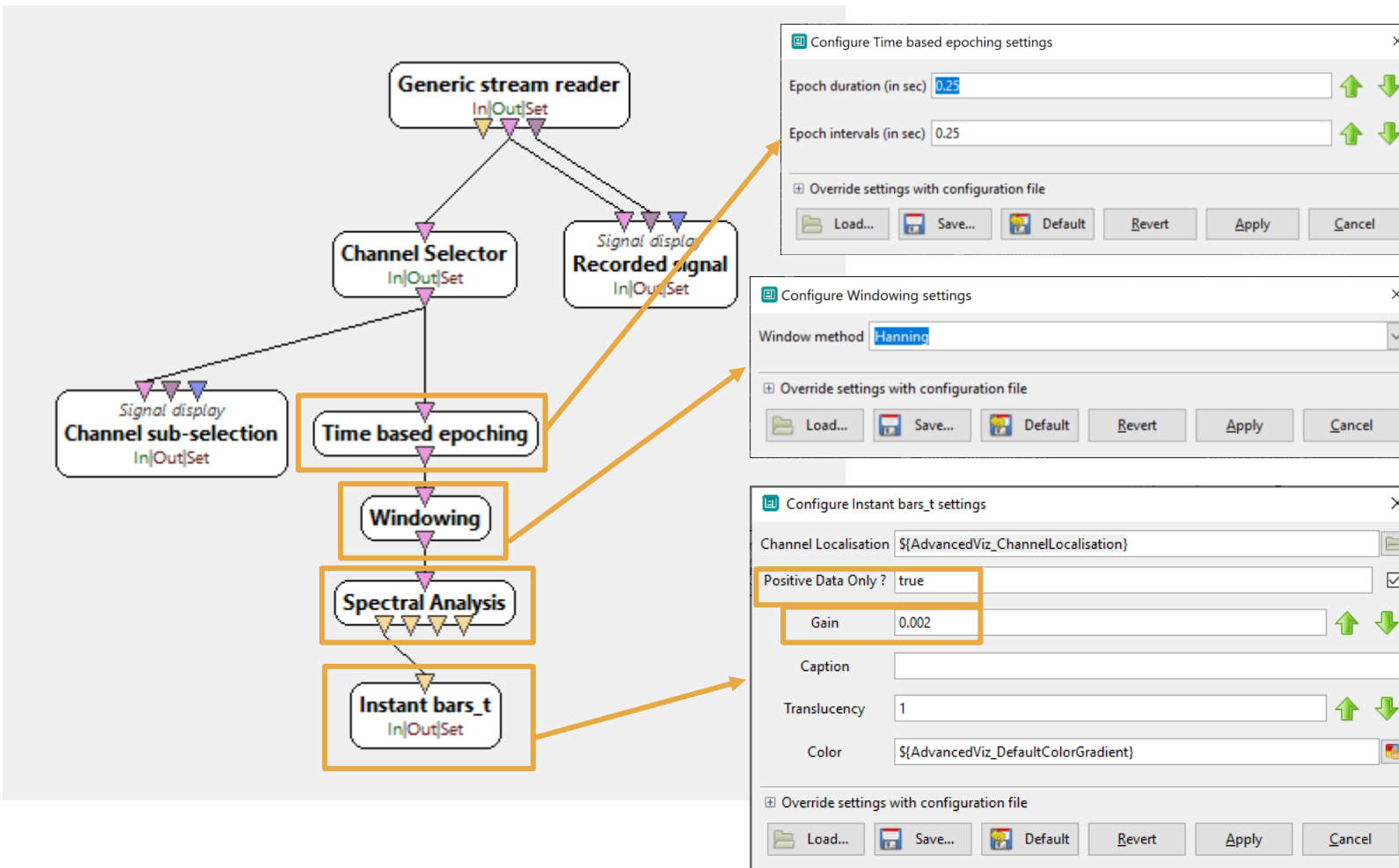
Discover & use filtering, epoching, windowing tools

Compute & display spectra for each channel/electrodes:

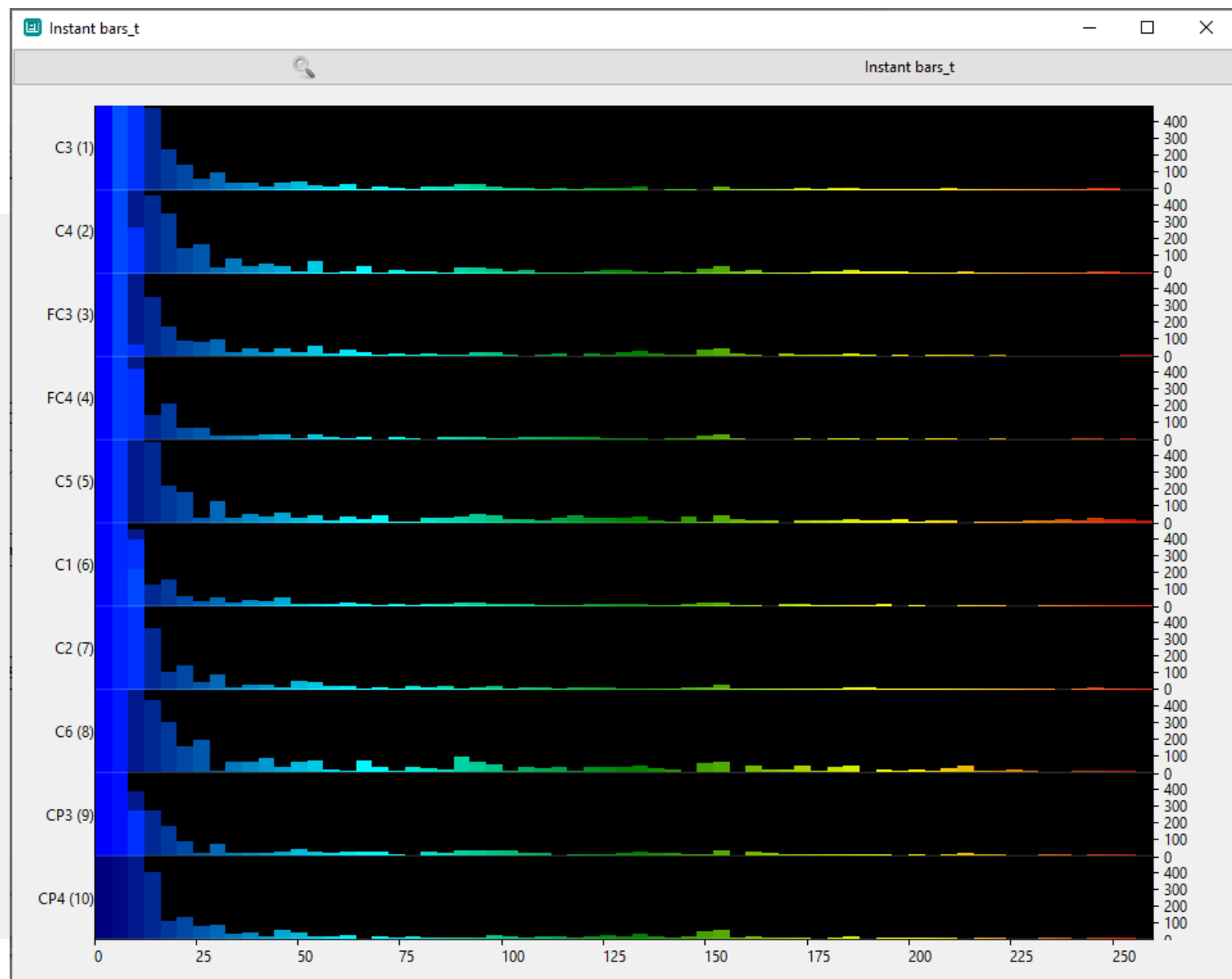
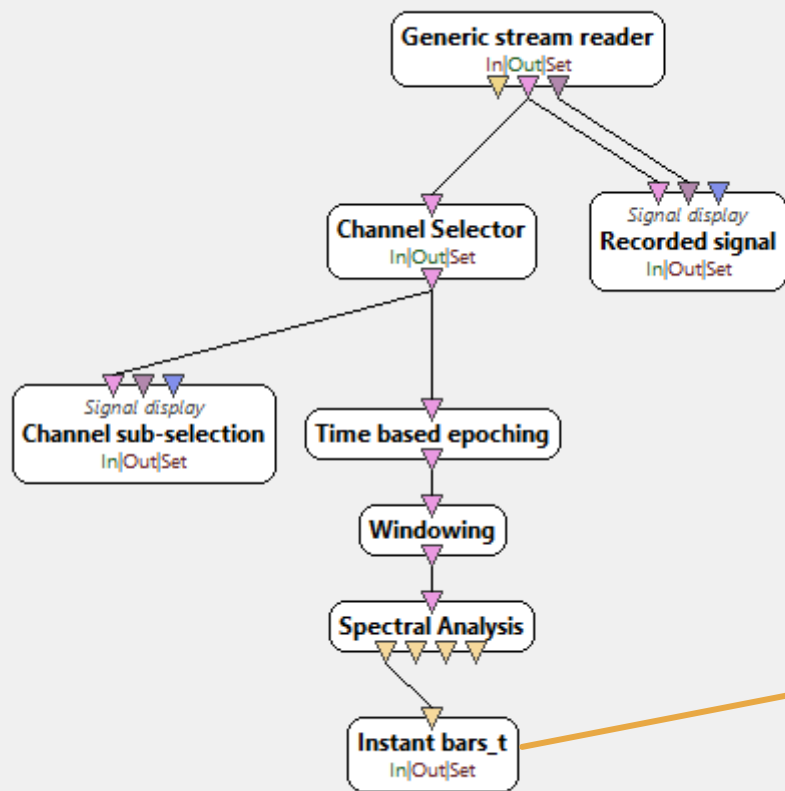
- for the whole band
- for different frequency bands (alpha, beta, etc)



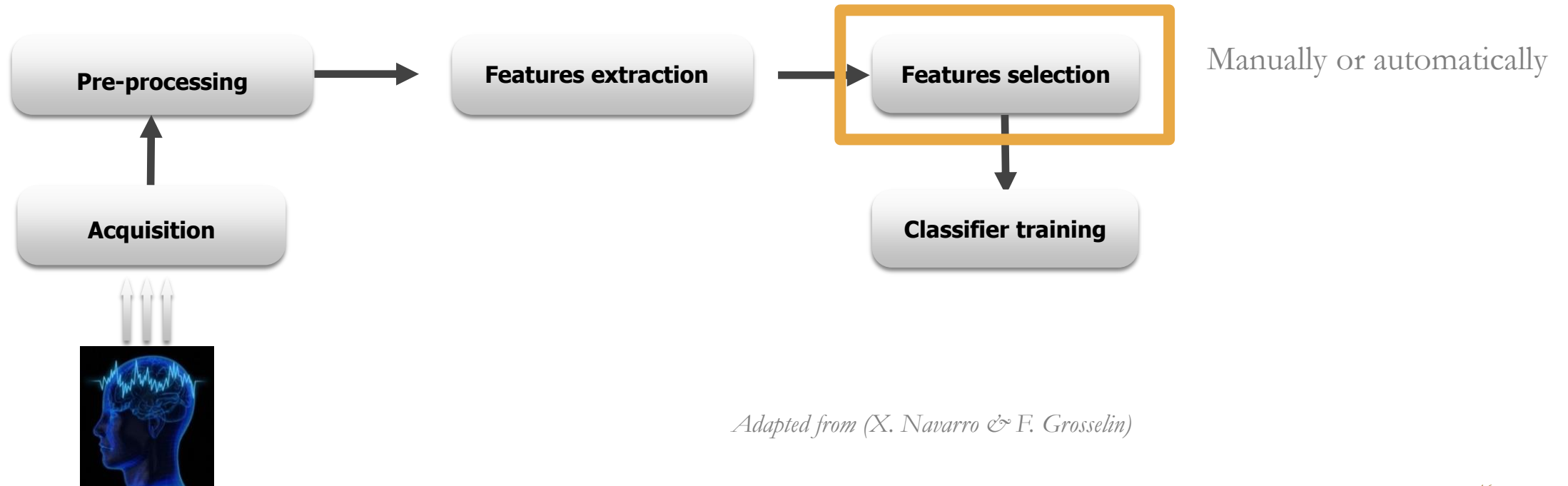
- 1- Import:  
“Time Based Epoching”,  
“Windowing”,  
“Spectral Analysis”,  
“Instant bars”

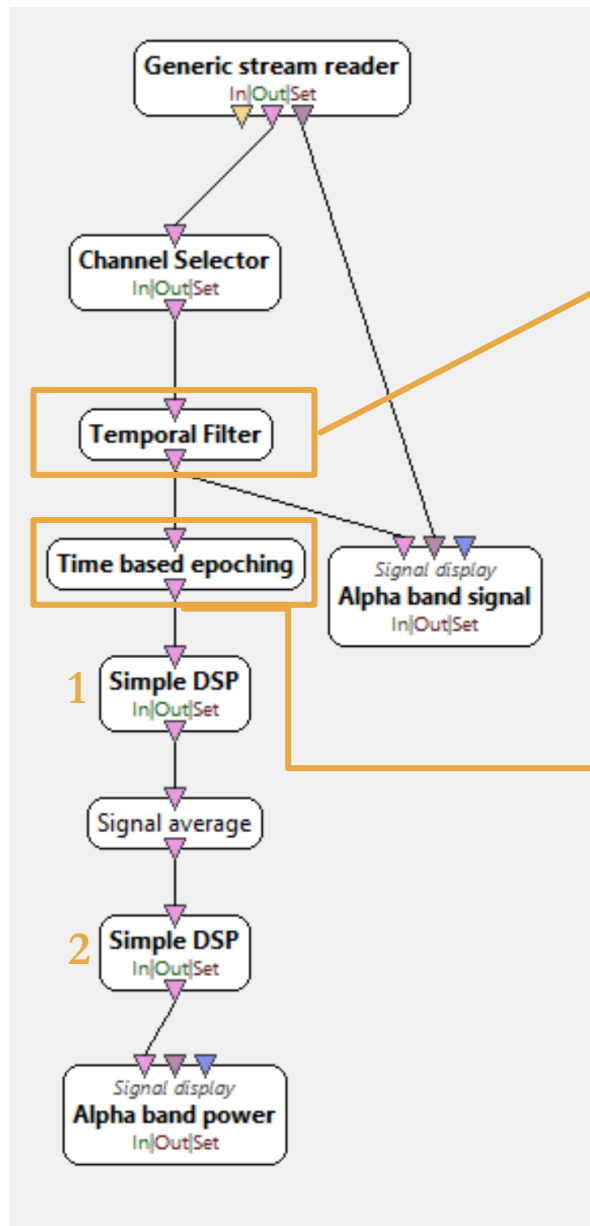


- 1- Import:  
“Time Based Epoching”,  
“Windowing”,  
“Spectral Analysis”,  
“Instant bars”
- 2- Set preferred settings for Epoching and windowing (ex: 0,25s, no overlap, Hann Window)
- Note : Spectral Analysis has 4 outputs (see parameters for details).
- 3- Set gain for “instant bars” ! No automatic scaling...



- We now have access to a view of the full band spectrum
- However, the feature we need for training the classifier is the **spectral power in the alpha/beta band**





Configure Temporal Filter settings

Filter Method:

Filter Type:

Filter Order:

Low Cut-off Frequency (Hz):

High Cut-off Frequency (Hz):

Band Pass Ripple (dB):

☒ Override settings with configuration file

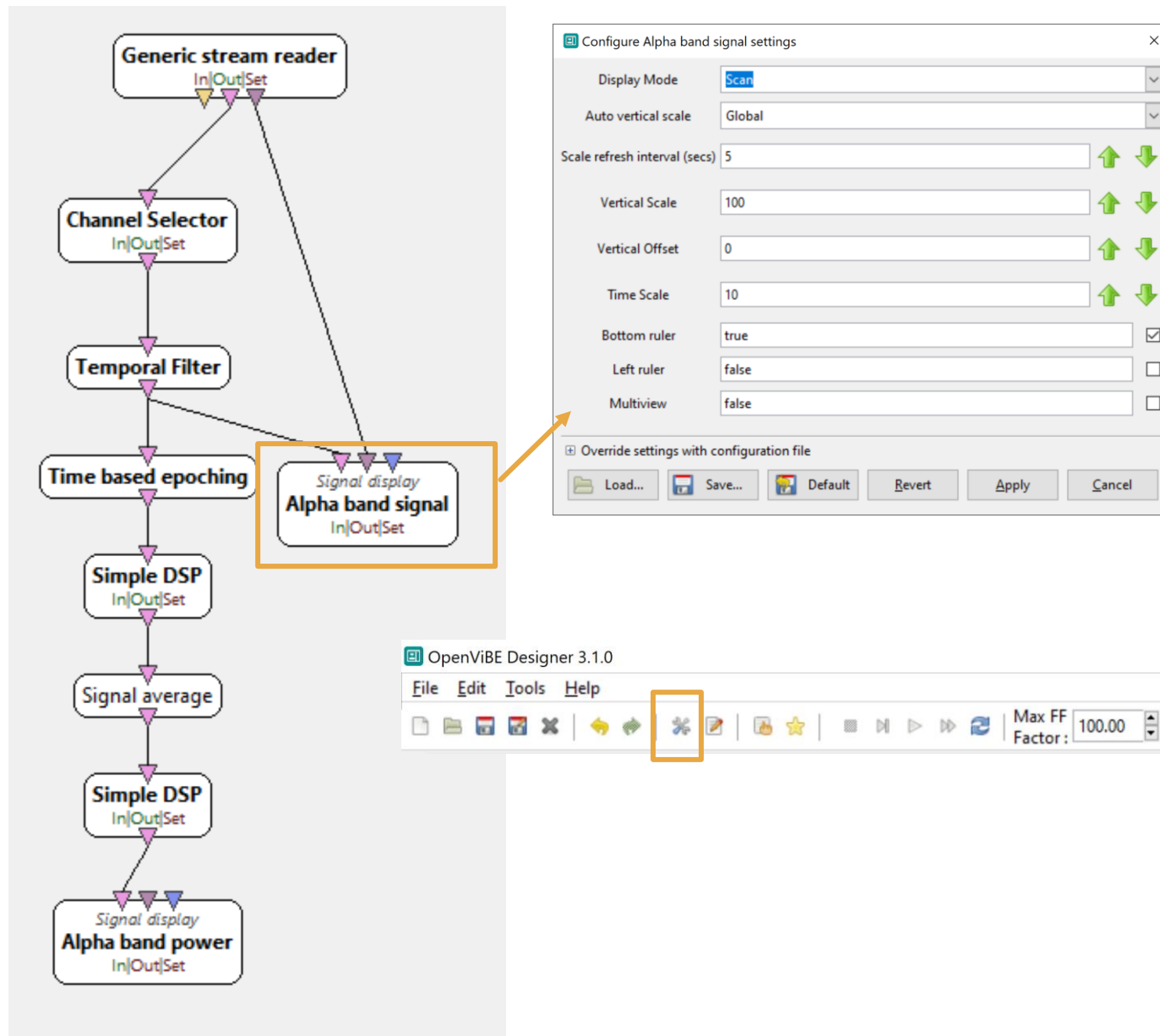
Configure Time based epoching settings

Epoch duration (in sec):

Epoch intervals (in sec):

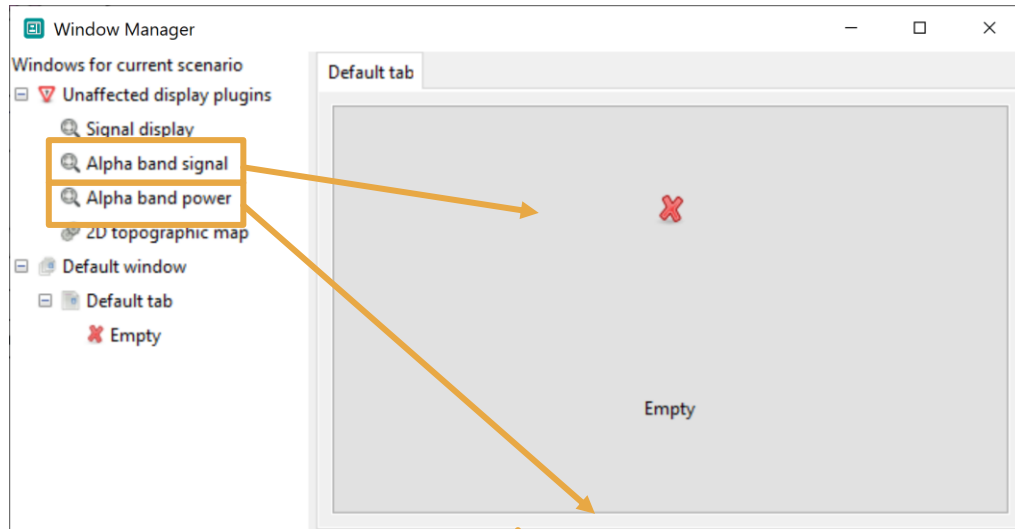
☒ Override settings with configuration file

- “Temporal Filter”, set to [8;12] Hz band
- Signal processing chain:
  - “Time based epoching” of 1s, every 0.2s (overlapping windows of signal for power computation)
  - DSP 1 formula:  $x*x$
  - Average (= of  $x^2$  across a window of 1s)
  - DSP 2 formula:  $\text{Log}_{10}(1+x)$
  - $\Rightarrow \log_{10}(1+\text{avg}(x^2))$
- Add **Displays** and rename them
  - Set time scales & vertical scaling

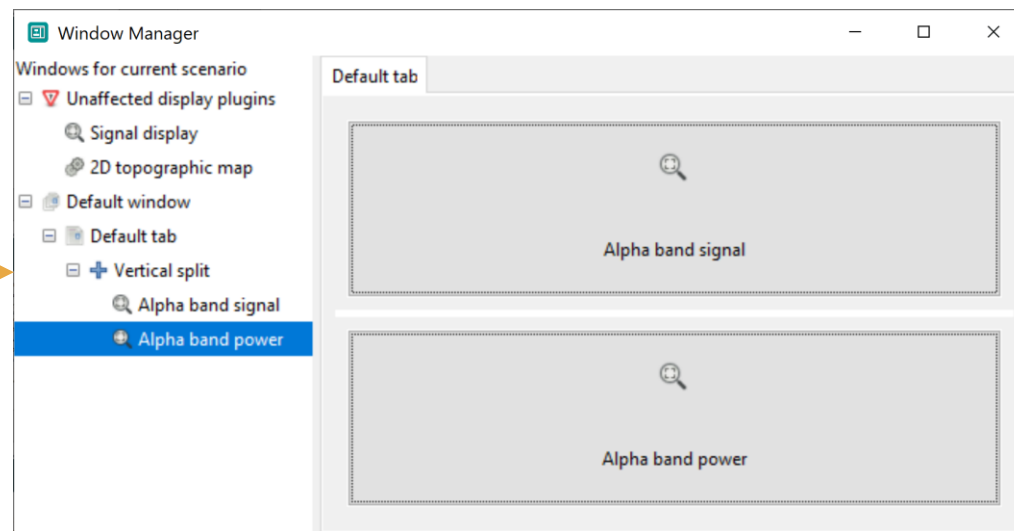


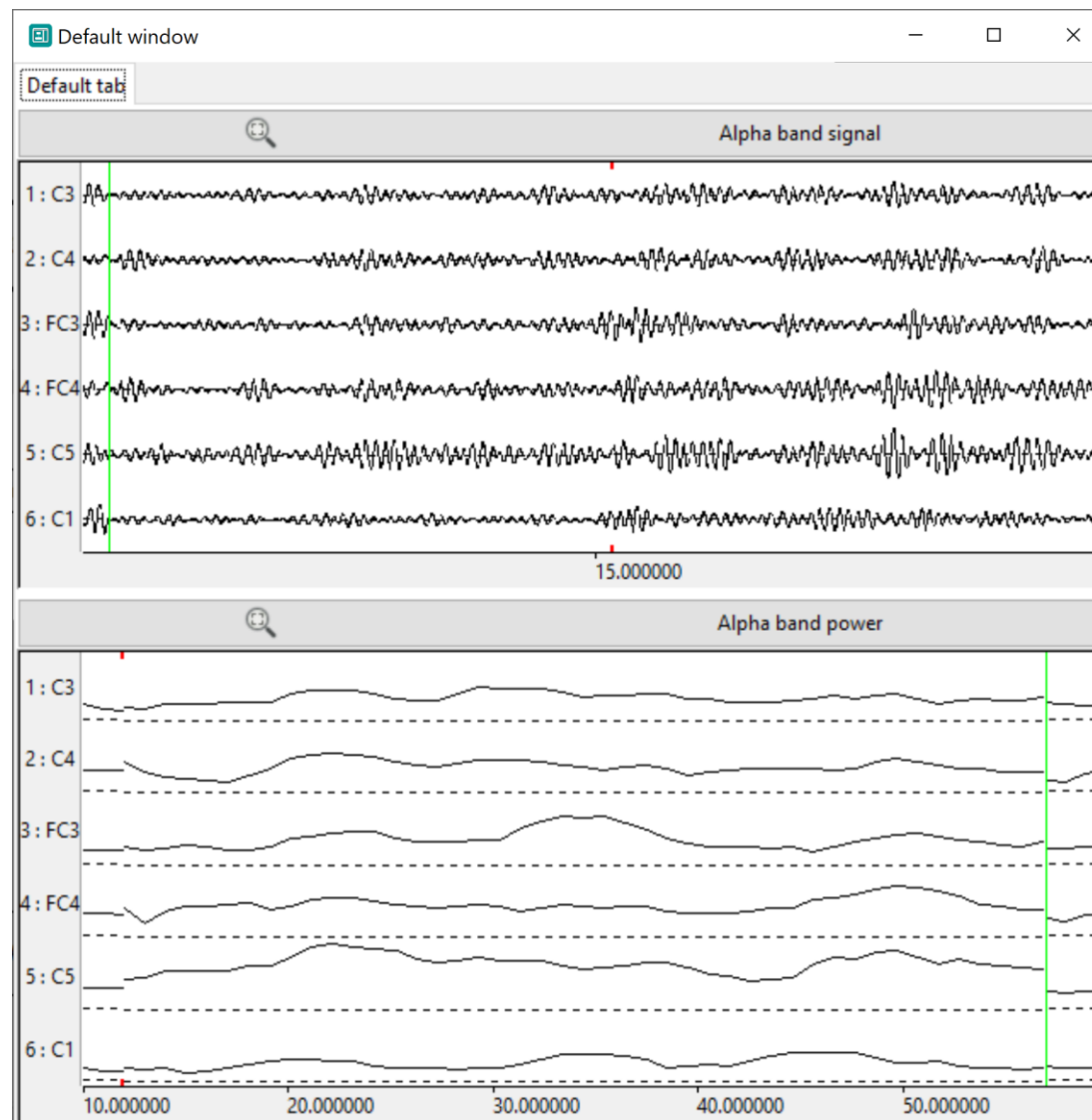
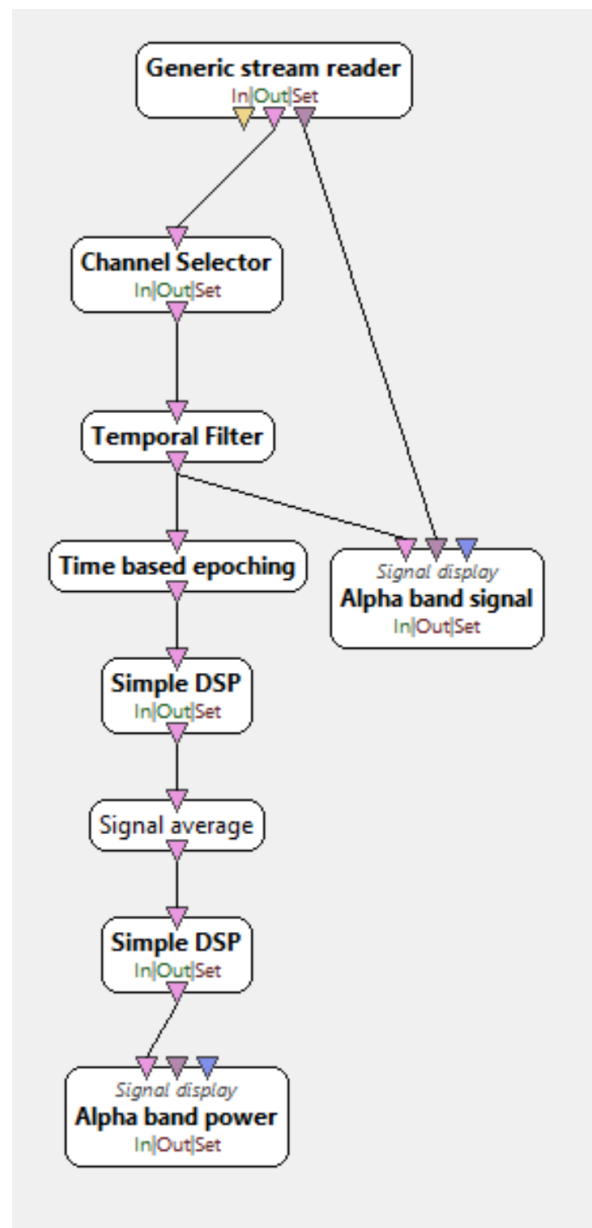
- Add **Displays** and rename them
  - Set time scales : 10 (seconds) for alpha band signal display, 50 for alpha band power (since we have 5 times more data to display)
  - Set vertical scaling to “global”
- Use widget reordering tool for ease of use!

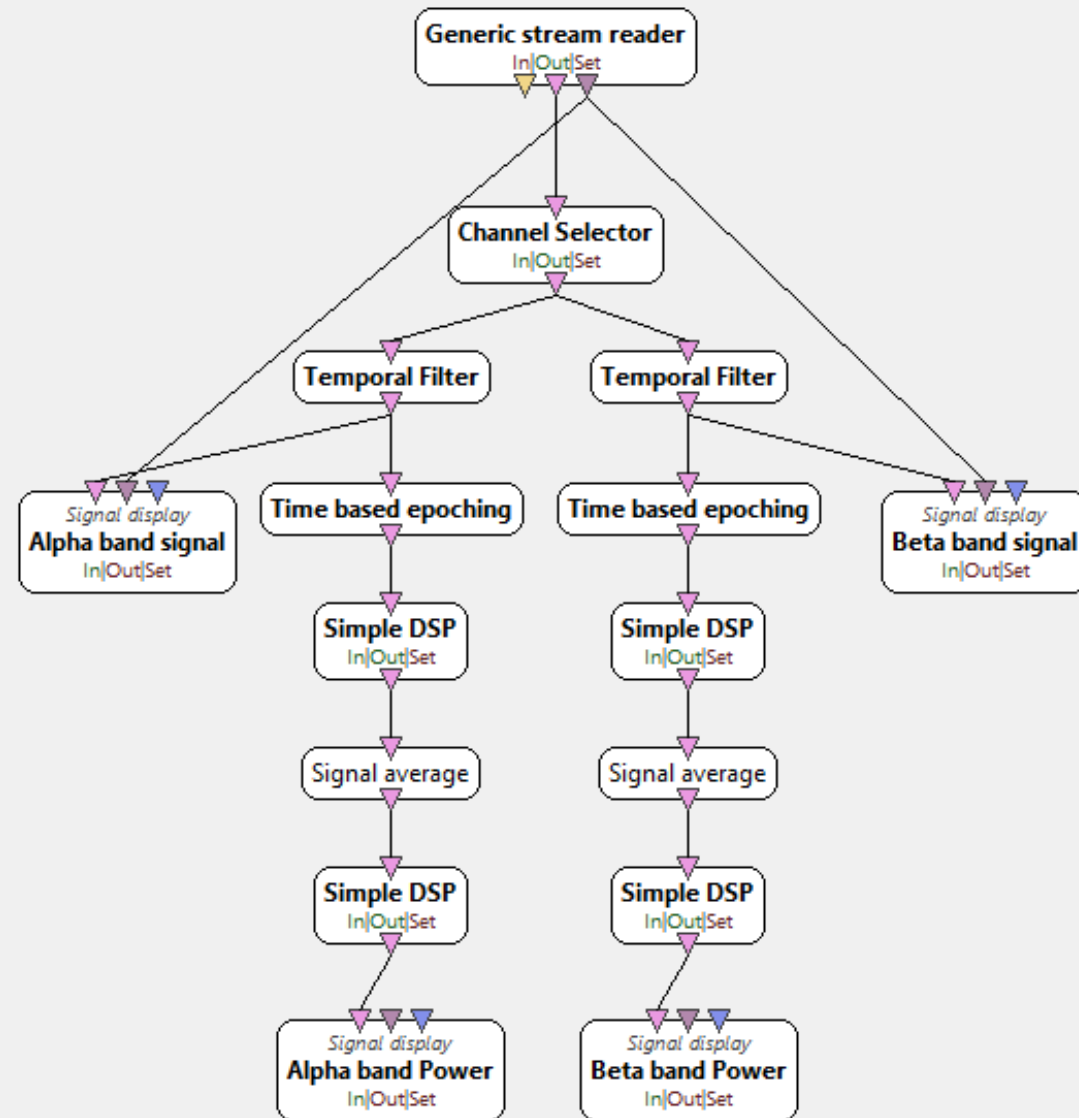




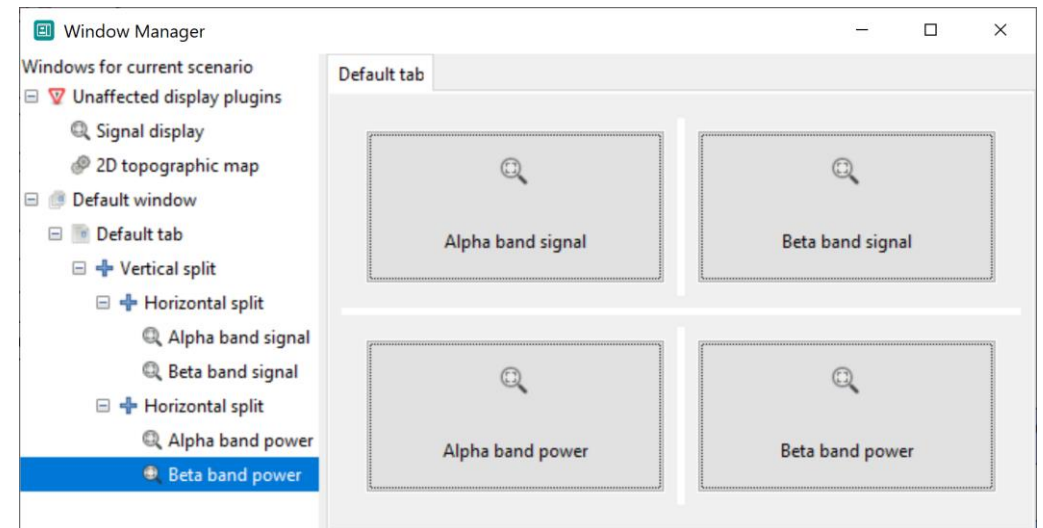
- Use widget reordering tool for ease of use!

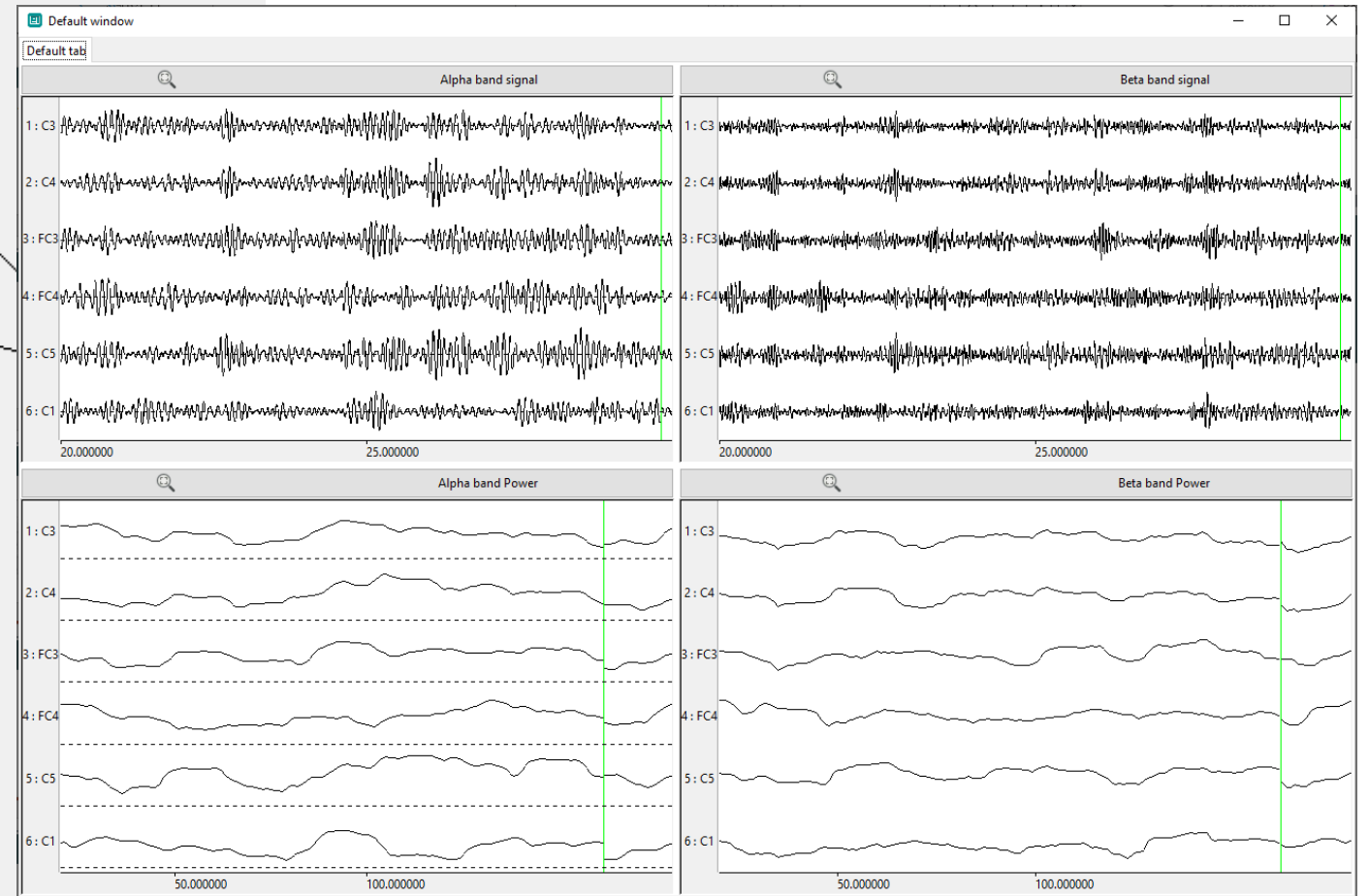
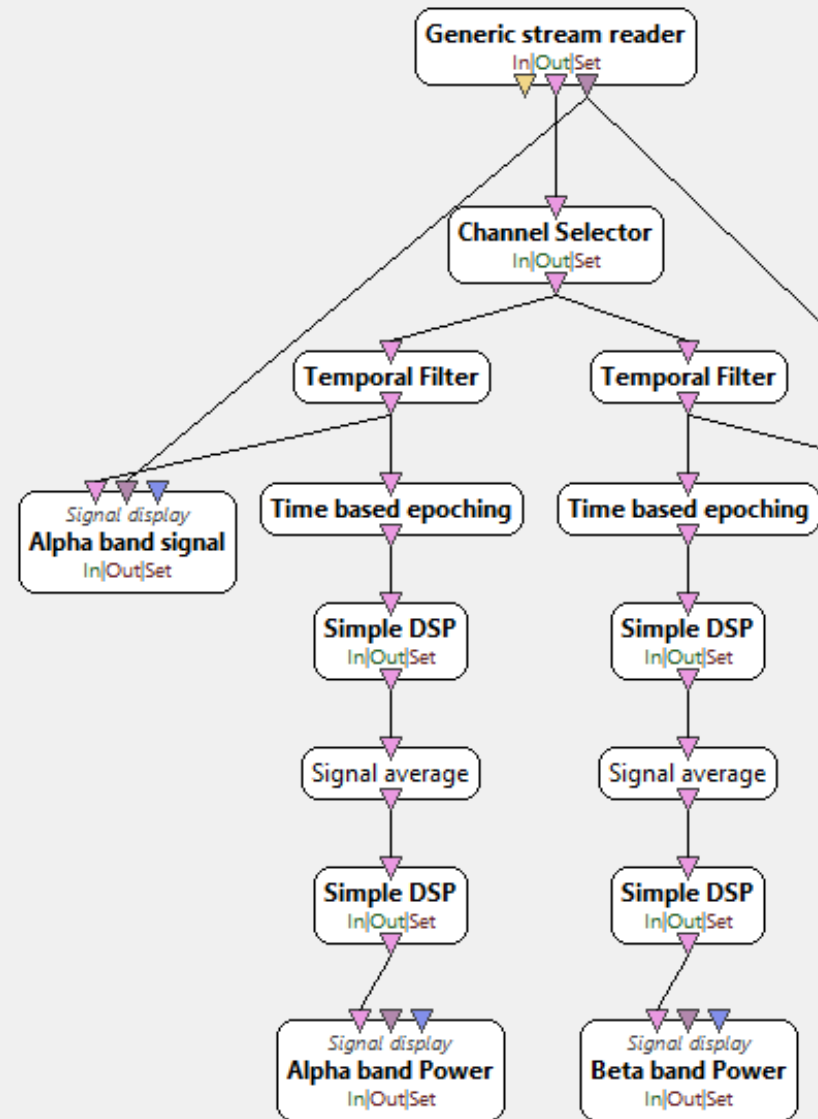


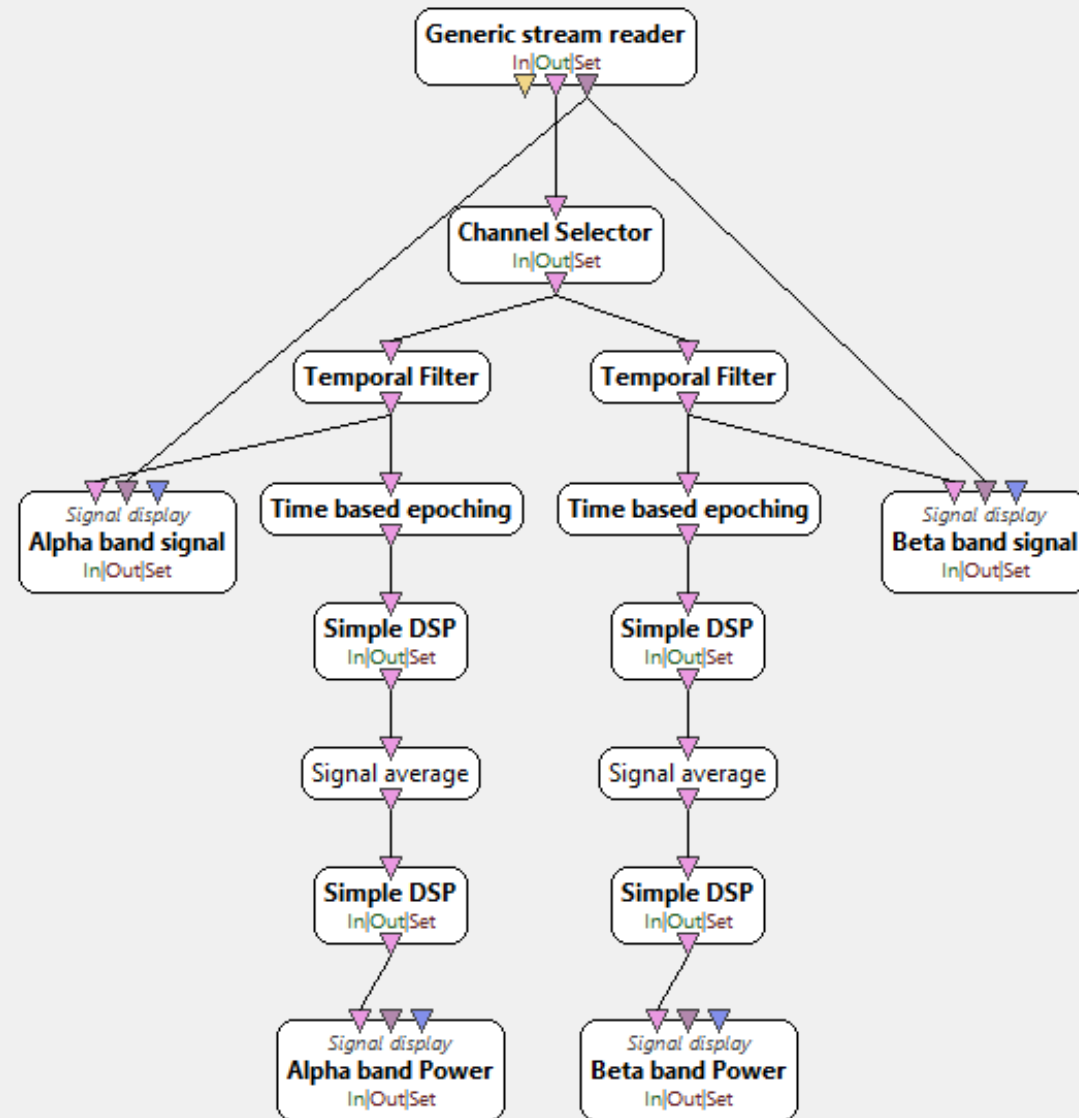




- Repeat for **Beta Band** [12;24]Hz
- Don't hesitate to copy/paste boxes...!







- Scenario is available on the github repo:

BCI-OpenViBE-CuttingEEG2021/  
scenarios/sc2-spectralAnalysis.xml

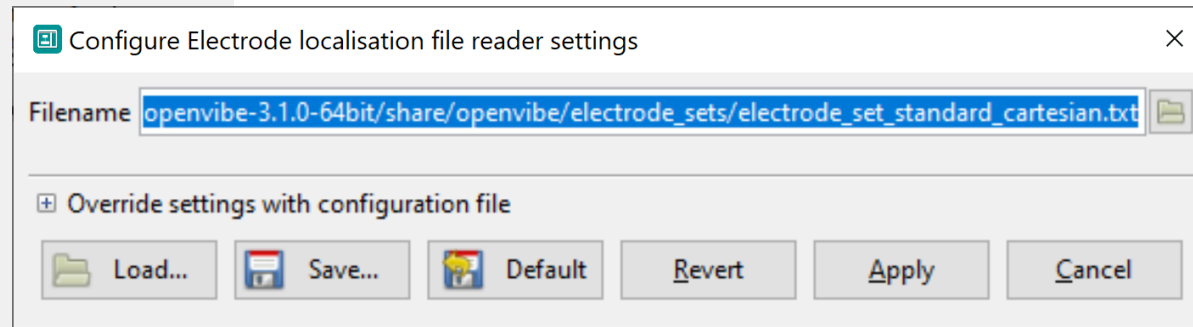
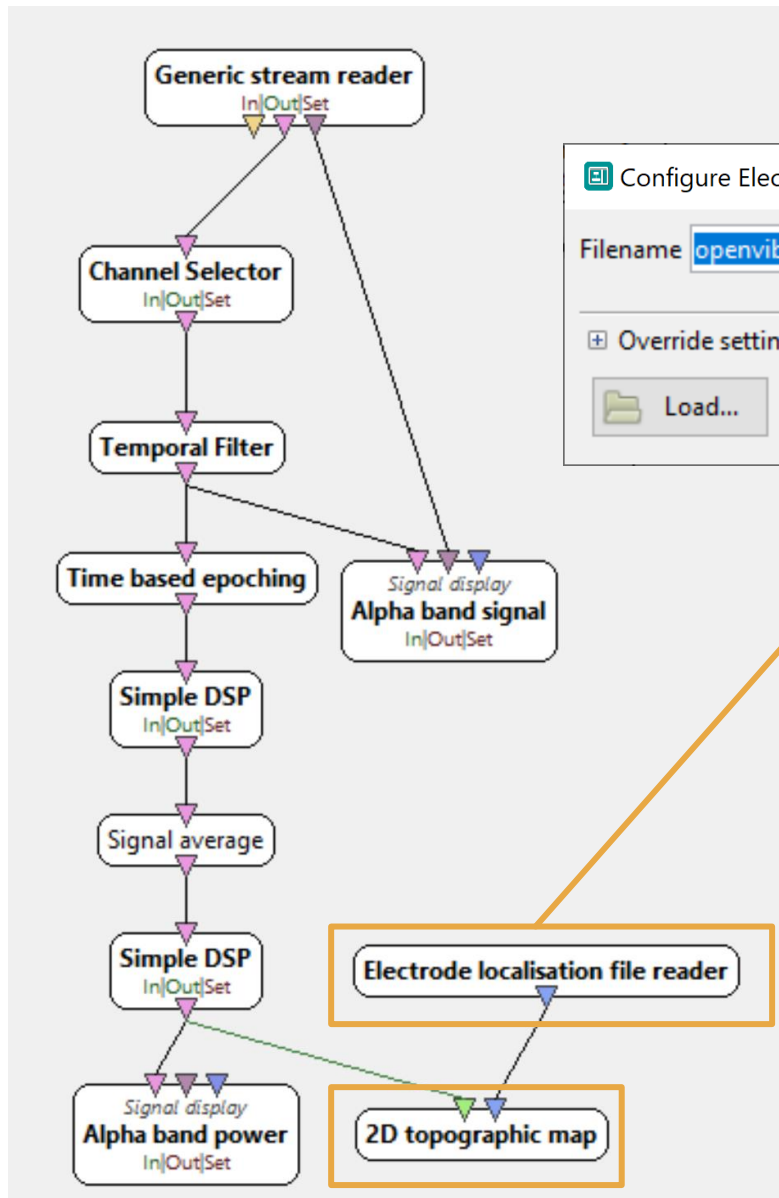
## SC2 – STEP 3 – VISUALIZING BRAIN TOPOGRAPHY

- **Sub-chapters:**

- Channel Selection
- Basic Signal Processing / Spectral analysis
- Topography visualization
- Feature Selection + Spatial Filtering
- Classifier training

## SC2 – STEP 3 – VISUALIZING BRAIN TOPOGRAPHY

- **Goals:**
- Load electrode localization data
- Topography viewer for a specific band

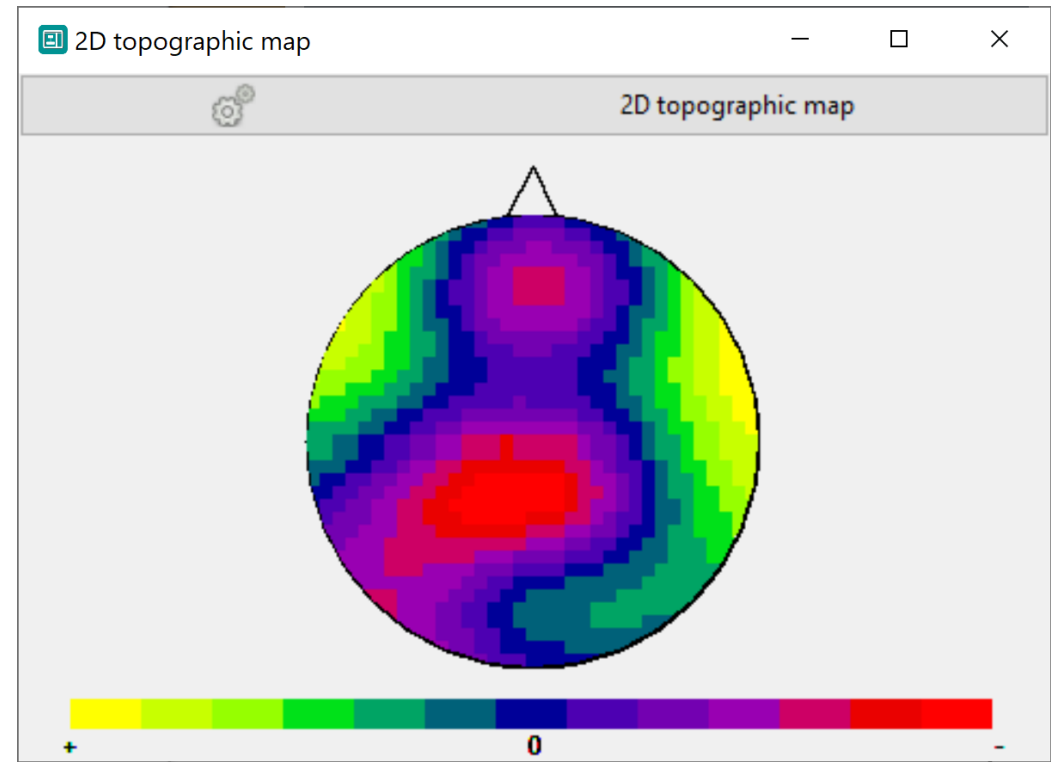
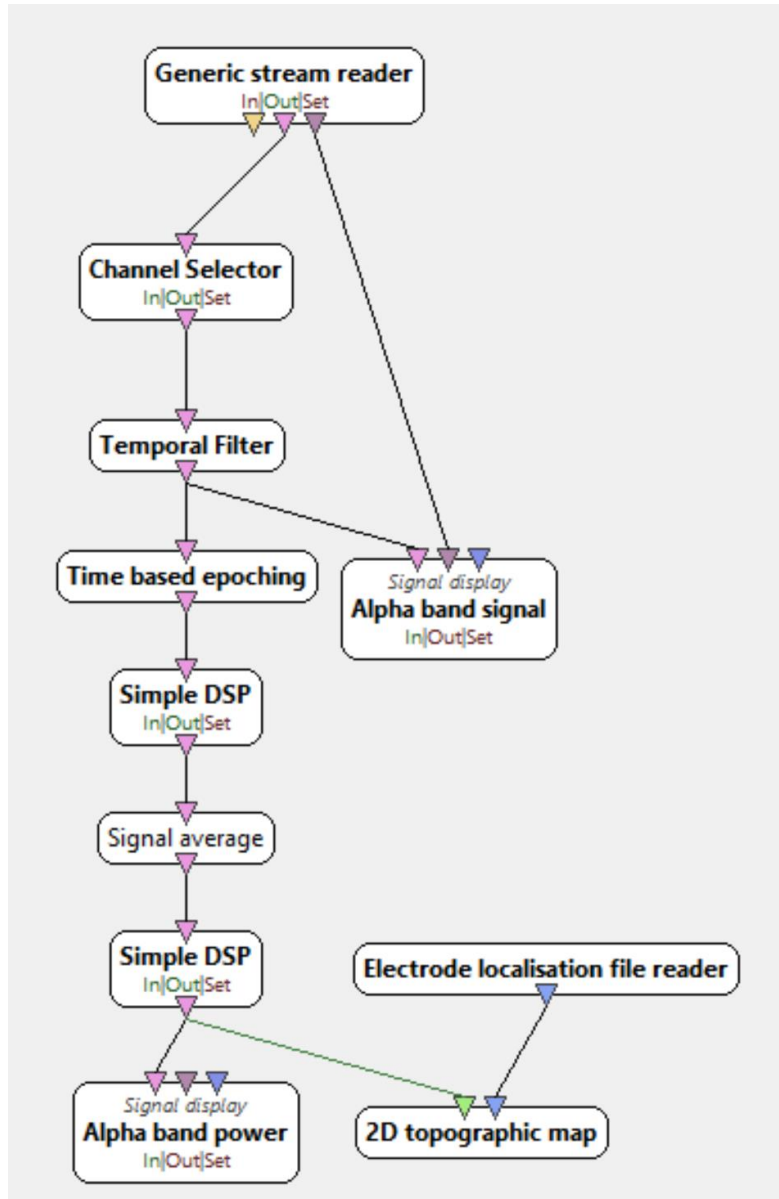


- Using the previous scenario computing the spectral power (in band alpha only):
- Add “**Electrode localization file reader**” and check the file location

`<openvibe-3,1,0-64bits>/  
share/openvibe/electrode_sets/electrode_set_standard_cartesian.txt`

- Add “**2D topographic map**”
- Connect boxes and play the scenario





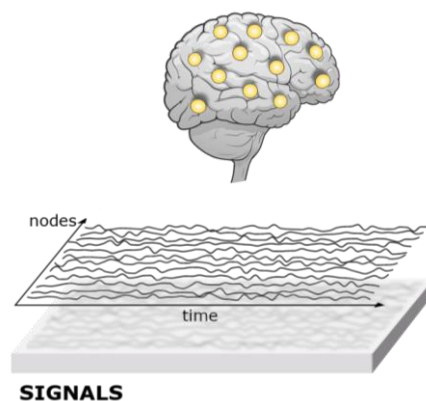
## SC2 – STEP 4 – FEATURE SELECTION

- **Sub-chapters:**
  - Channel Selection
  - Basic Signal Processing / Spectral analysis
  - Topography visualization
  - Feature Selection + Spatial Filtering
  - Classifier training

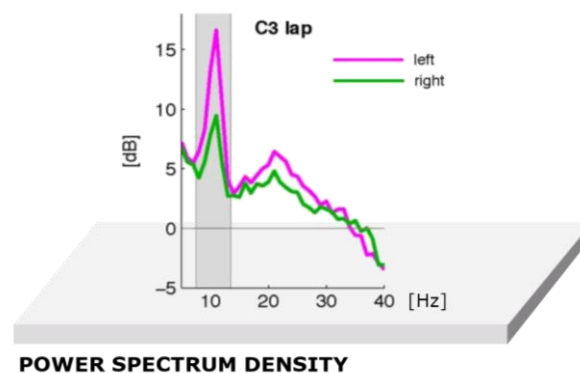
## SC2 – STEP 4 – FEATURE SELECTION

## Features to extract (recap)

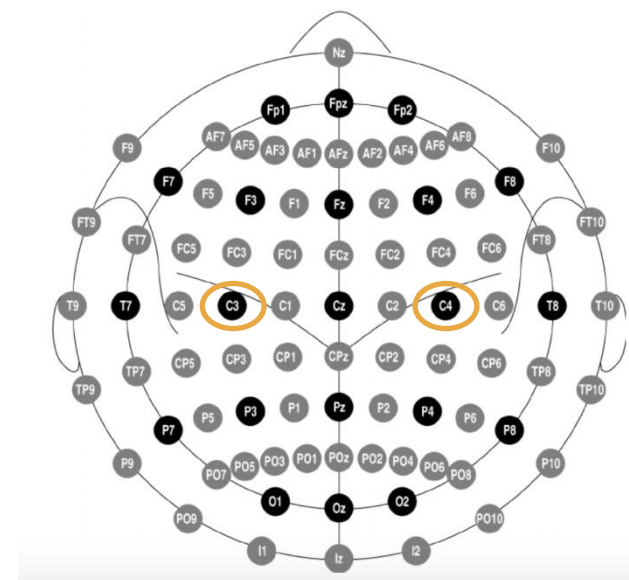
- Power spectra
- Sensorimotor area
- Mu: 8-12Hz &/OR Beta: 14-29Hz



(Gonzalez-Astudillo et al, 2020)



(Maeder et al., 2012)



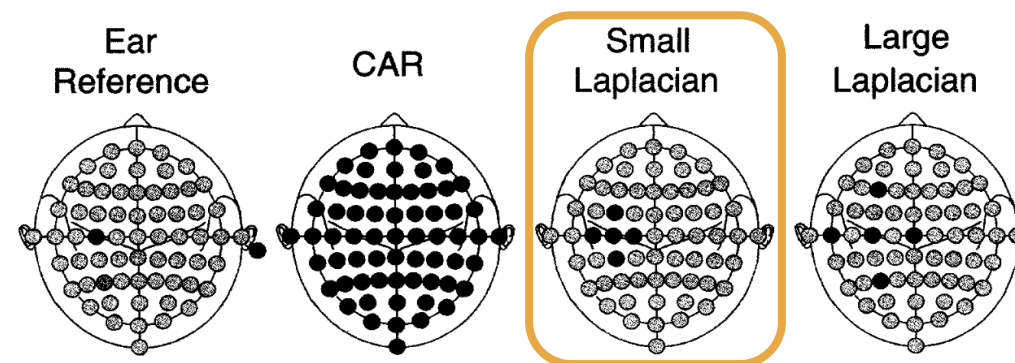
## SC2 – STEP 4 – FEATURE SELECTION – SPATIAL FILTERING

### ■ Spatial Filtering

- Common Average Reference (CAR)
- Surface Laplacian – used here

$$C3' = 4 * C3 - FC3 - C5 - C1 - CP3$$

$$C4' = 4 * C4 - FC4 - C2 - C6 - CP4$$



Spatial filtering  
(McFarland et al, 1997)

- OV box: **Spatial Filter**

- 10 inputs => 2 outputs

$$C3' = 4 * C3 - FC3 - C5 - C1 - CP3$$

$$C4' = 4 * C4 - FC4 - C2 - C6 - CP4$$

- Our list of channels:

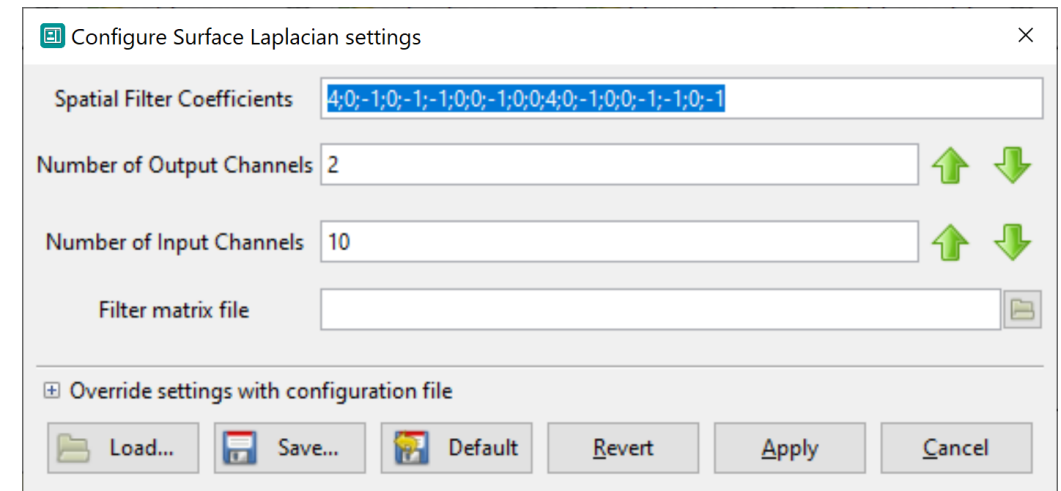
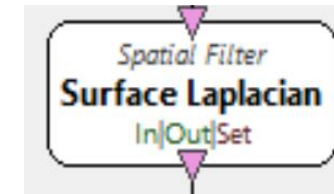
C3; C4; FC3; FC4; C5; C1; C2; C6; CP3; CP4

$$C3' = 4; 0; -1; 0; -1; -1; 0; 0; -1; 0$$

$$C4' = 0; 4; 0; -1; 0; 0; -1; -1; 0; -1$$

- Serialize formulae in box parameters

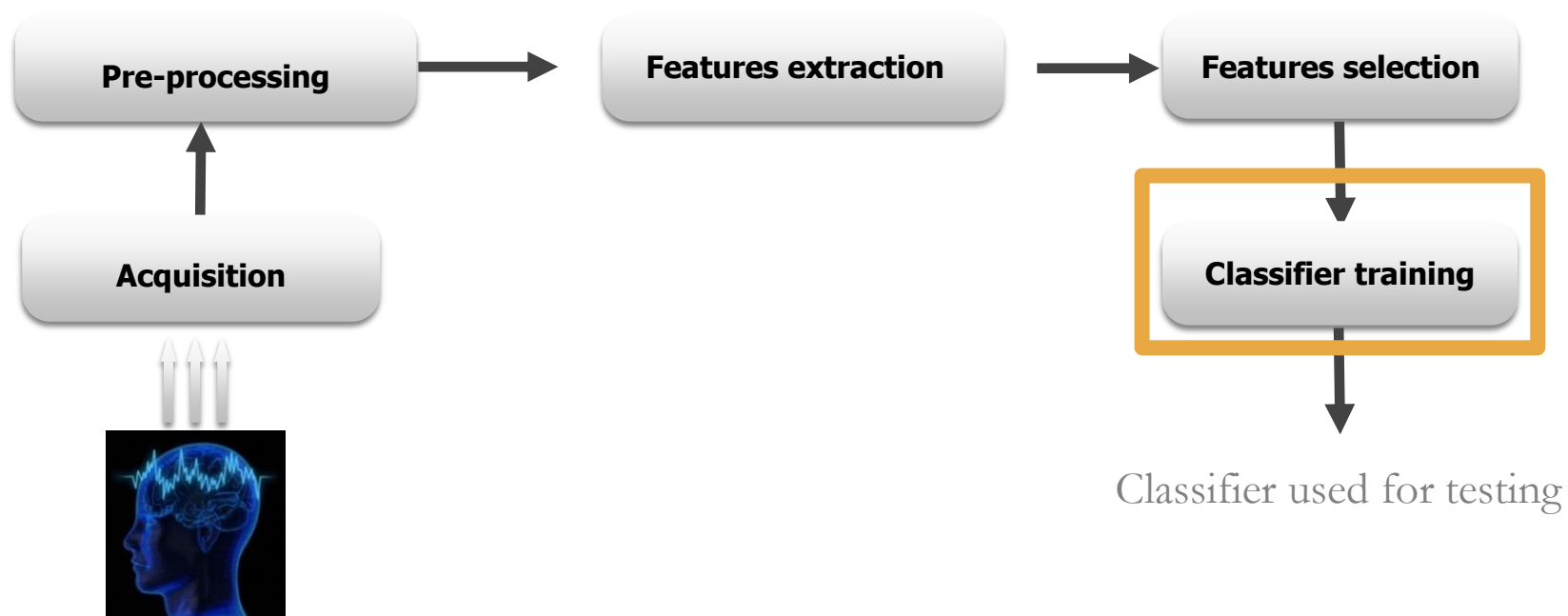
4;0;-1;0;-1;-1;0;0;-1;0;0;4;0;-1;0;0;-1;-1;0;-1



## SC2 – STEP 5 – CLASSIFIER TRAINING

- **Sub-chapters:**
  - Channel Selection
  - Basic Signal Processing / Spectral analysis
  - Topography visualization
  - Feature Selection + Spatial Filtering
  - Classifier training

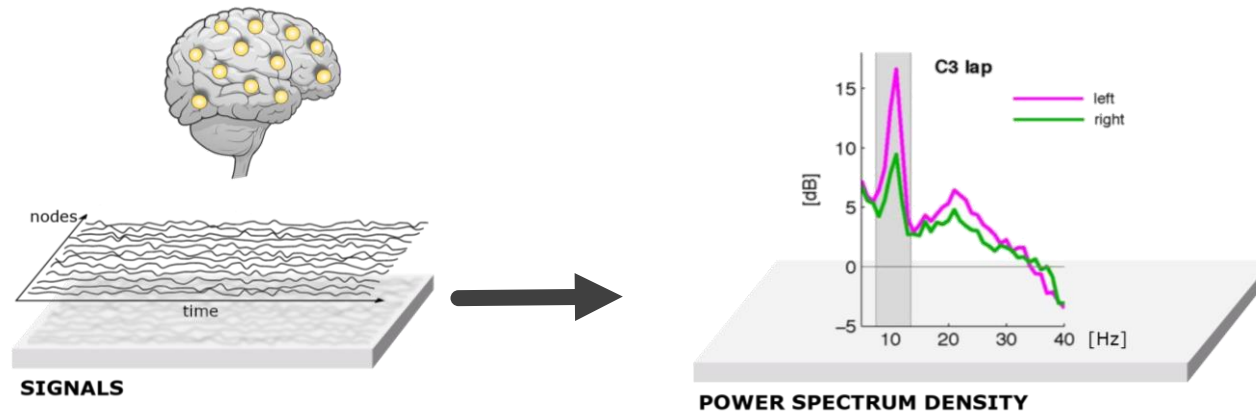
## SC2 – STEP 5 – CLASSIFIER TRAINING



*Adapted from (X. Navarro & F. Grosselin)*

## SC2 – STEP 5 – CLASSIFIER TRAINING

- **Goal reminder:**
- We want to train a classifier to distinguish between mental states, based on instantaneous spectral power in bands alpha and/or beta



(Gonzalez-Astudillo et al, 2020)

(Maeder et al., 2012)



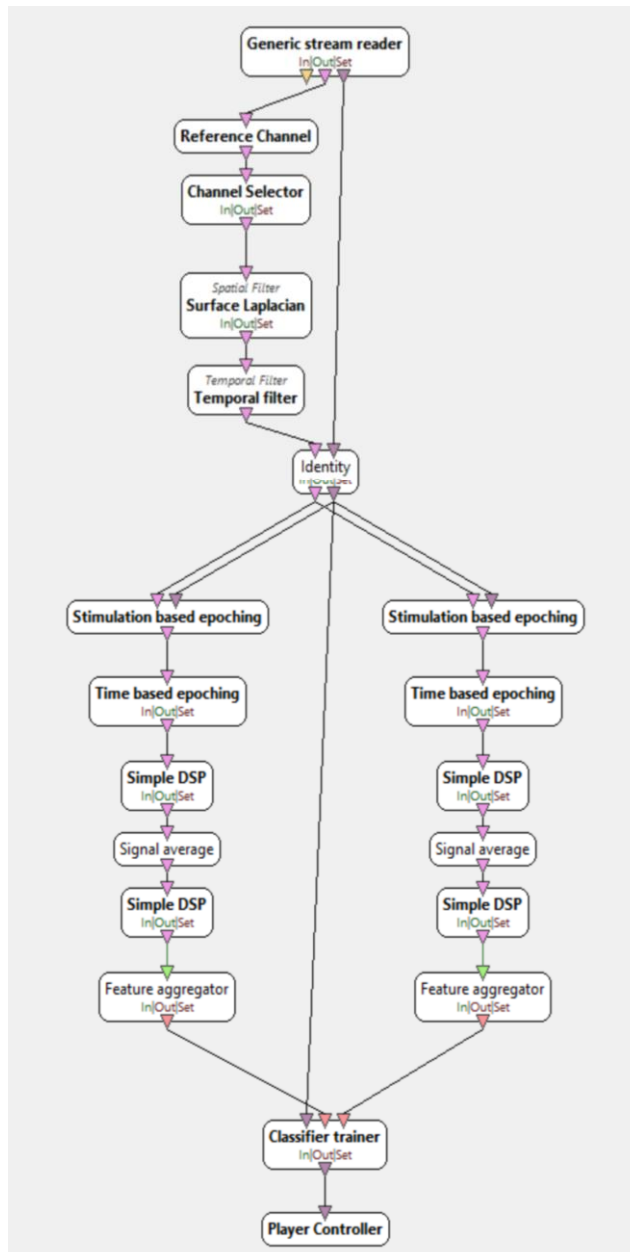
## SC2 – STEP 5 – CLASSIFIER TRAINING

- Load the full scenario so we can analyze it step by step...

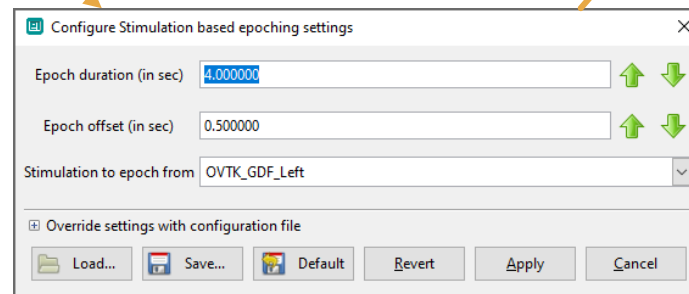
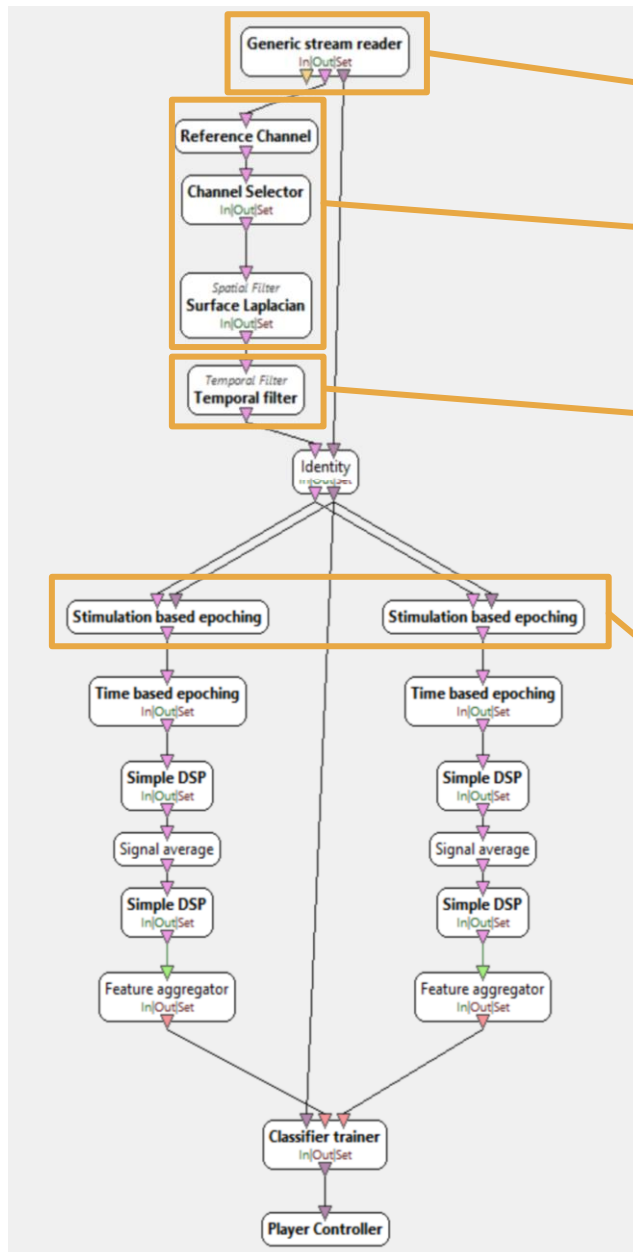
```
<OpenViBE_install_path>\  
share\openvibe\scenarios\bci-examples\motor-imagery\  
motor-imagery-bci-2-classifier-trainer.xml
```

## SC2 – STEP 5 – CLASSIFIER TRAINING

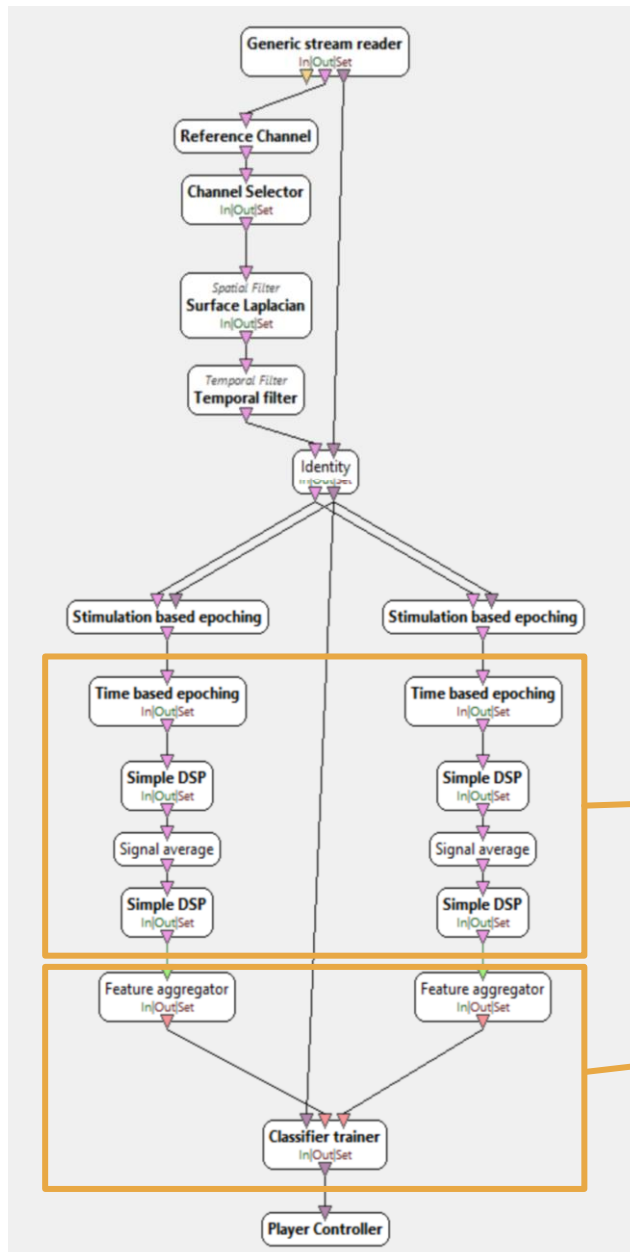
- **Goal: putting it all together in a BCI scenario...**
  - **Load** pre-recorded signal file
  - Select **sub-set of electrodes**
  - Surface Laplacian
  - Filter to the **frequency band** of interest
  - Stimulation-based Epoching: **split between conditions** (LEFT vs RIGHT...)
  - Epoching & Signal processing: compute a “**classifier feature**” based on **spectral power**
  - **Train the classifier!**



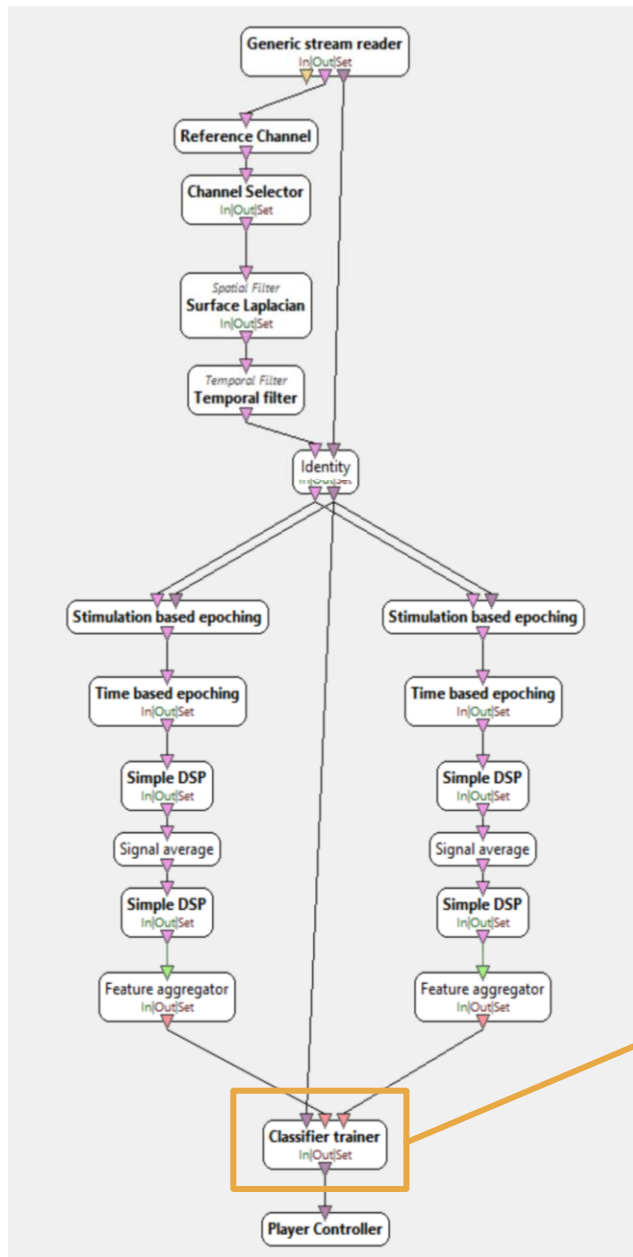
- **Load** pre-recorded signal file
- Select **sub-set of electrodes**, set **Reference Channel** + apply **spatial filtering**
- Filter to the **frequency band** of interest
- Stimulation-based Epoching: **split between conditions** (LEFT vs RIGHT...)
- Epoching & Signal processing: compute a “**classifier feature**” based on **spectral power**
- **Train the classifier!**



- Load pre-recorded signal file
- Select sub-set of electrodes, set Reference Channel + apply spatial filtering
- Filter to the frequency band of interest
- Stimulation-based Epoching: split between conditions (LEFT vs RIGHT...)
- Epoching & Signal processing: compute a “classifier feature” based on spectral power
- Train the classifier!



- Load pre-recorded signal file
- Select sub-set of electrodes, set Reference Channel + apply spatial filtering
- Filter to the frequency band of interest
- Stimulation-based Epoching: split between conditions (LEFT vs RIGHT...)
- Epoching & Signal processing: compute a “classifier feature” based on spectral power
- Train the classifier!



**Configure Classifier trainer settings**

Train trigger	OVTk_StimulationId_Train
Filename to save configuration to	\${Player_ScenarioDirectory}/motor-imagery-b...
Multiclass strategy to apply	Native
Class 1 label	OVTk_GDF_Left
Class 2 label	OVTk_GDF_Right
Algorithm to use	Linear Discriminant Analysis (LDA)
Use shrinkage	false
Shrinkage coefficient (-1 == auto)	-1.000000
Shrinkage: Force diagonal cov (DDA)	false
Number of partitions for k-fold cross-validation test	7
Balance classes	false

☐ Override settings with configuration file

■ Train the classifier!

Configure Classifier trainer settings

Train trigger	OVTk_StimulationId_Train	
Filename to save configuration to	\${Player_ScenarioDirectory}/motor-imagery-b	
Multiclass strategy to apply	Native	
Class 1 label	OVTk_GDF_Left	
Class 2 label	OVTk_GDF_Right	
Algorithm to use	Linear Discriminant Analysis (LDA)	
Use shrinkage	false	<input type="checkbox"/>
Shrinkage coefficient (-1 == auto)	-1.000000	↑ ↓
Shrinkage: Force diagonal cov (DDA)	false	<input type="checkbox"/>
Number of partitions for k-fold cross-validation test	7	↑ ↓
Balance classes	false	<input type="checkbox"/>

☐ Override settings with configuration file

- **“Train trigger”** stimulation should be set in the acquisition LUA script, at the end of the experiment
- Set the **path/filename** for the training “weights” (results)
- Check that the classifier classes are correctly labeled
- Select the classifying algorithm (LDA, SVM...) and set the parameters

```

[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Received train stimulation. Data dim is [1764x2]
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> For information, we have 931 feature vector(s) for input 1
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> For information, we have 833 feature vector(s) for input 2
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> k-fold test could take quite a long time, be patient
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Finished with partition 1 / 7 (performance : 75.000000%)
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Finished with partition 2 / 7 (performance : 74.206349%)
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Finished with partition 3 / 7 (performance : 59.126984%)
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Finished with partition 4 / 7 (performance : 88.492063%)
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Finished with partition 5 / 7 (performance : 75.000000%)
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Finished with partition 6 / 7 (performance : 88.095238%)
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Finished with partition 7 / 7 (performance : 82.539683%)
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Cross-validation test accuracy is 77.494331% (sigma = 9.406674%)
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer>
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Cls vs cls      1      2
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Target 1:    82.7  17.3 %, 931 examples
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Target 2:    28.3  71.7 %, 833 examples
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Training set accuracy is 80.045351% (optimistic)
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer>
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Cls vs cls      1      2
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Target 1:    85.2  14.8 %, 931 examples
[ INF ] At time 462.008 sec <Box algorithm::(0x02e67945, 0x5ea8d309) aka Classifier trainer> Target 2:    25.7  74.3 %, 833 examples

```

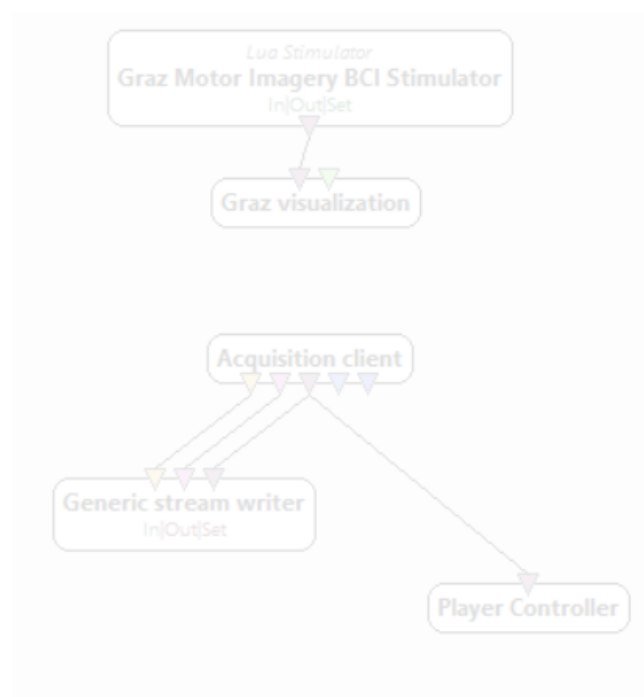


## SC2 – FINAL WORDS / Q&A

### ■ **Recap:**

- Signal processing / Channel selection / Filtering / Epoching
- Spectral Analysis, spectrum display
- Brain topography
- Feature selection
- Classifier training

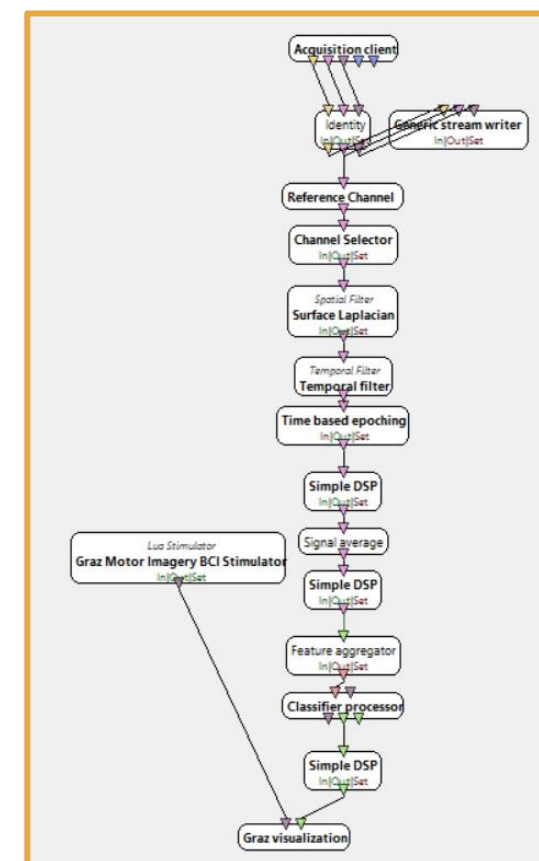
# SC3 – ONLINE CLASSIFICATION & FEEDBACK



1- Training data acquisition

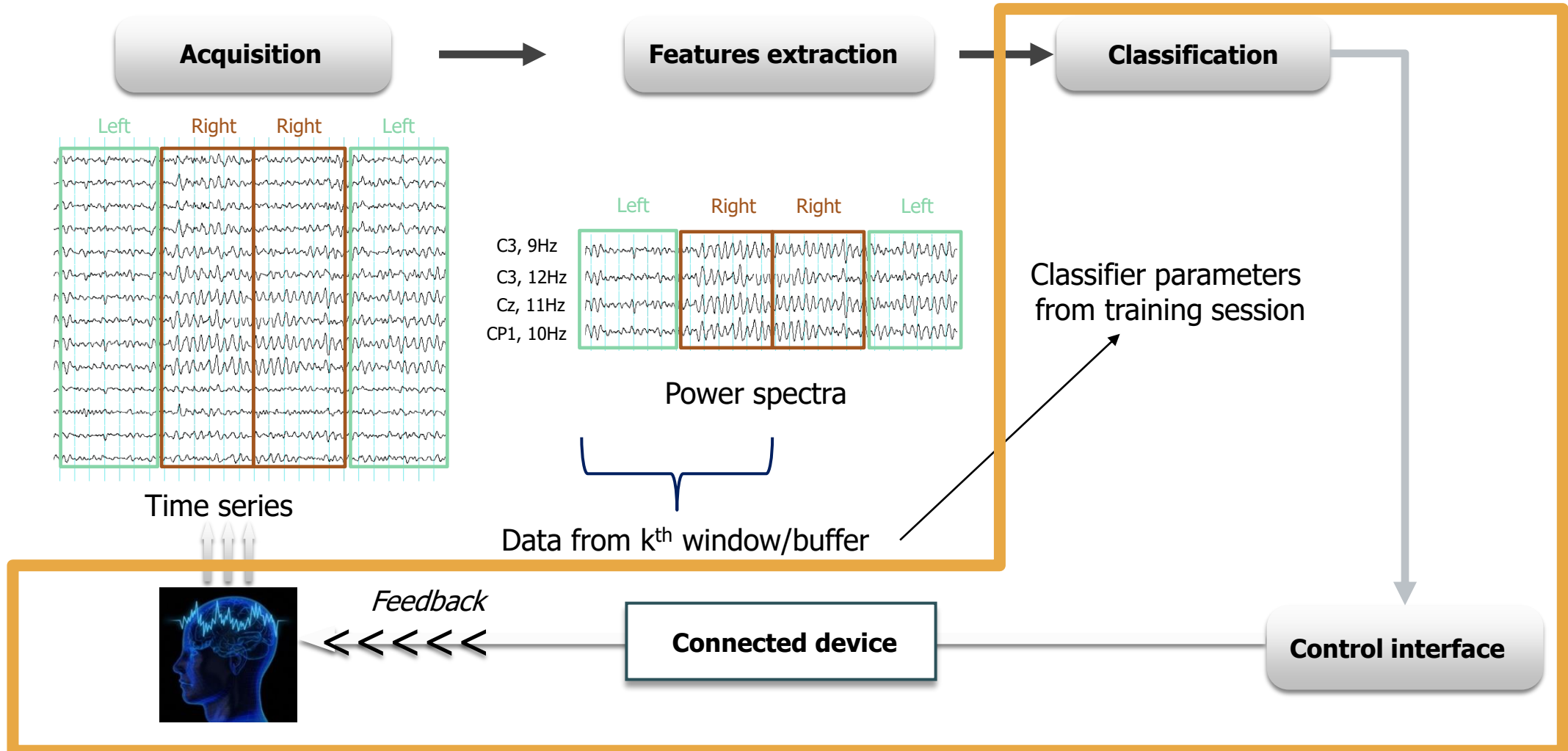


2- Features extraction  
& Classification



3- Online feedback

# SC3 – ONLINE CLASSIFICATION



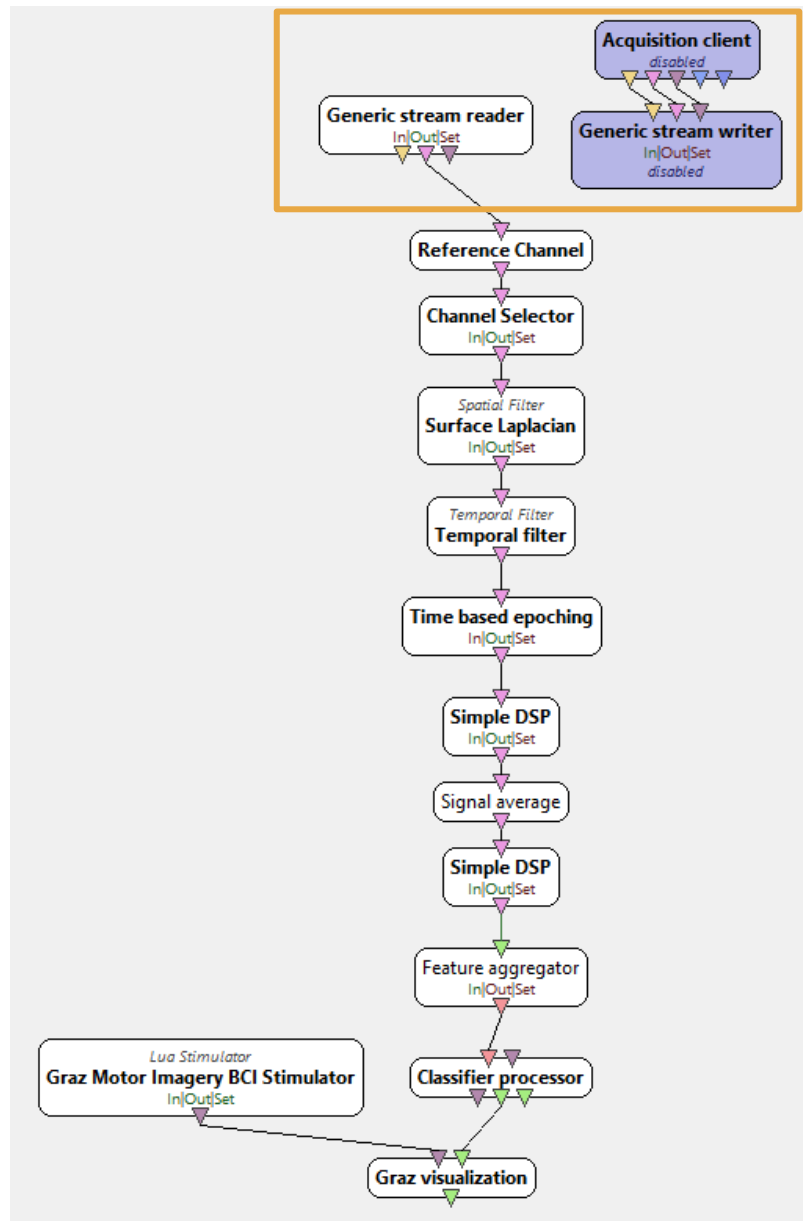
## SC3 – ONLINE CLASSIFICATION

- **Goals:**

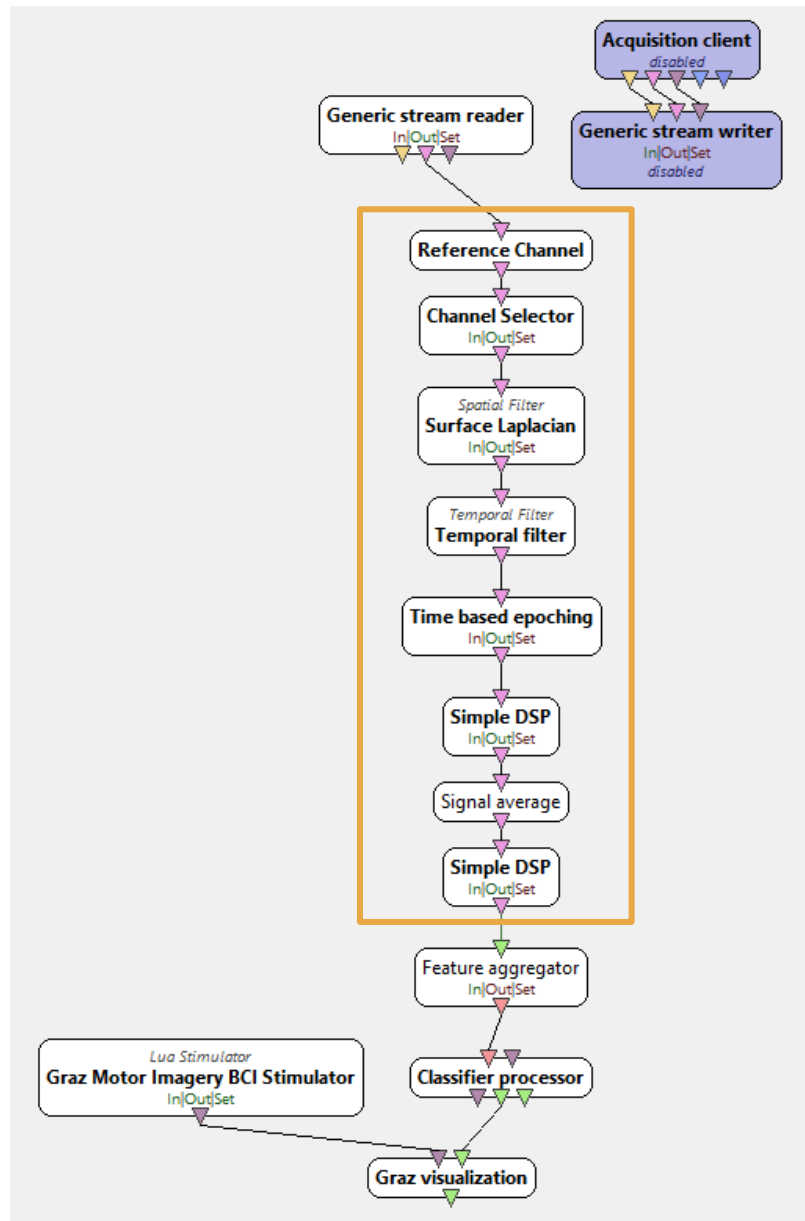
- Acquire EEG data in real-time, while providing “tasks” to the subject
- Classify his/her mental states (also in real-time) using the classifier trained in SC2
- (almost) everything we need is already there!

- **Let's load the scenario and analyze it...**

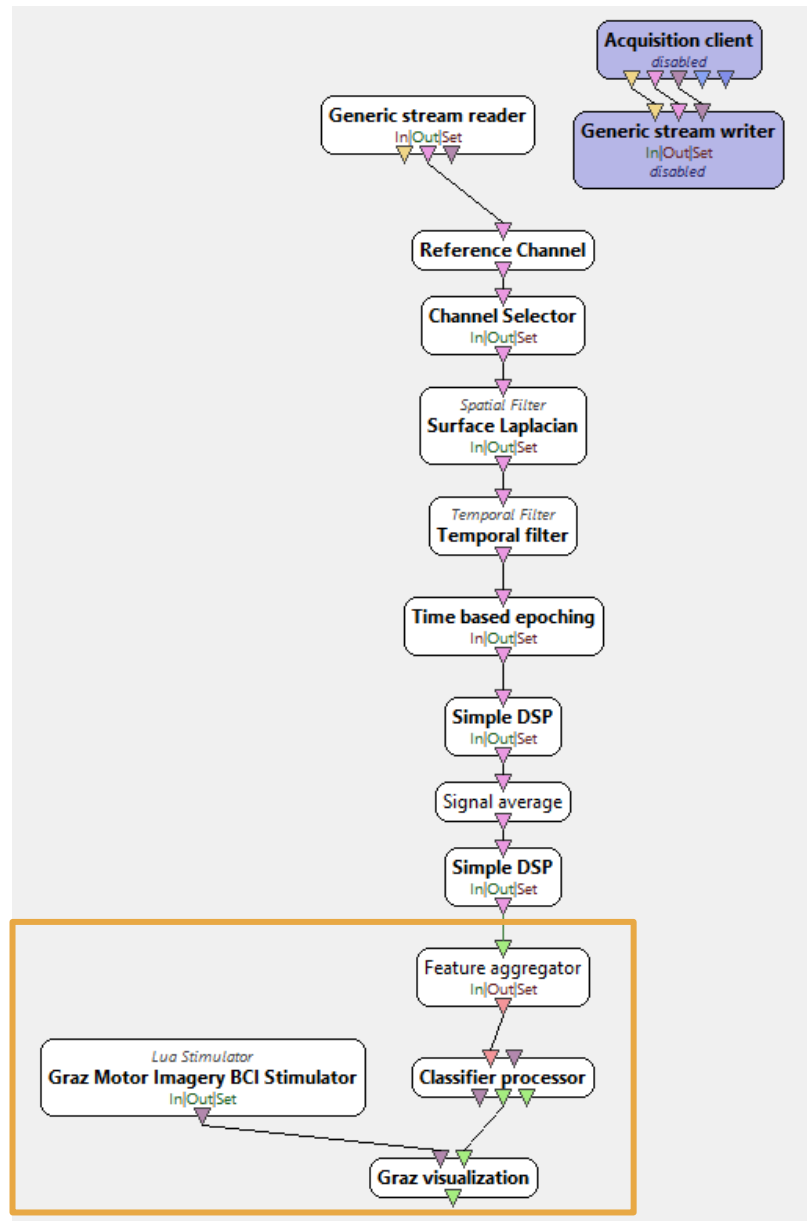
```
<OpenViBE_install_path>\  
share\openvibe\scenarios\bci-examples\motor-imagery\  
motor-imagery-bci-3-online.xml
```



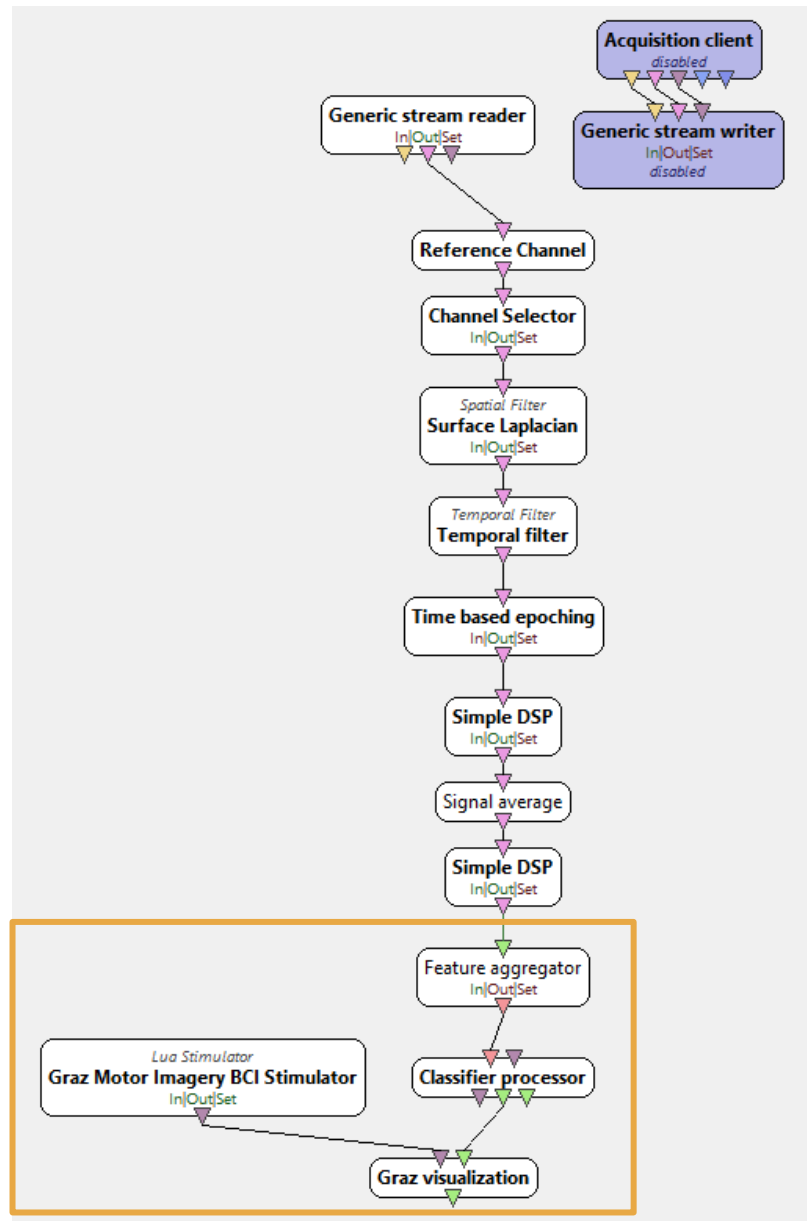
- In a real BCI experiment, we should of course use the Acquisition Client (and record the incoming signals for future replays and studies)
- Here, for the sake of an already long workshop, we'll use the same signal file as before...



- We want the classifier to have the same type of data used for training
- (It's actually the other way around: train your classifier with the same data type you use in the online step)
- The whole selection, filtering, epoching & Signal Processing chain is the same as in SC2 !

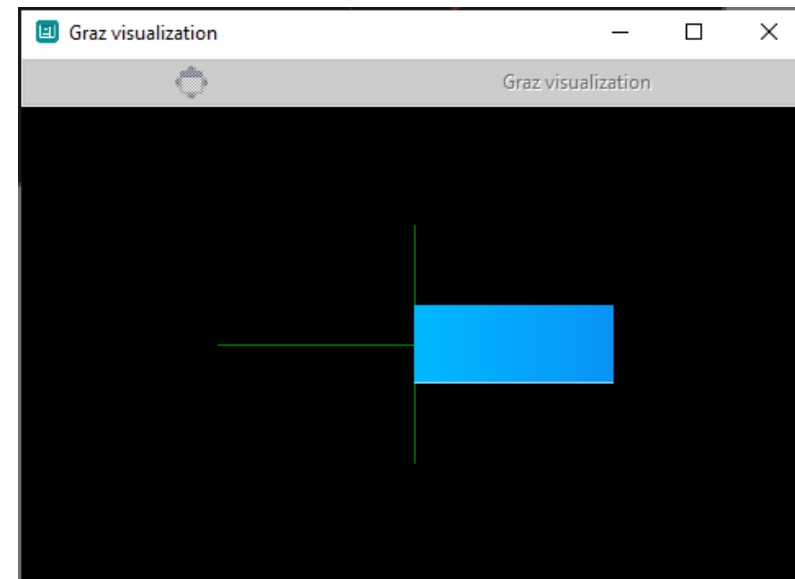
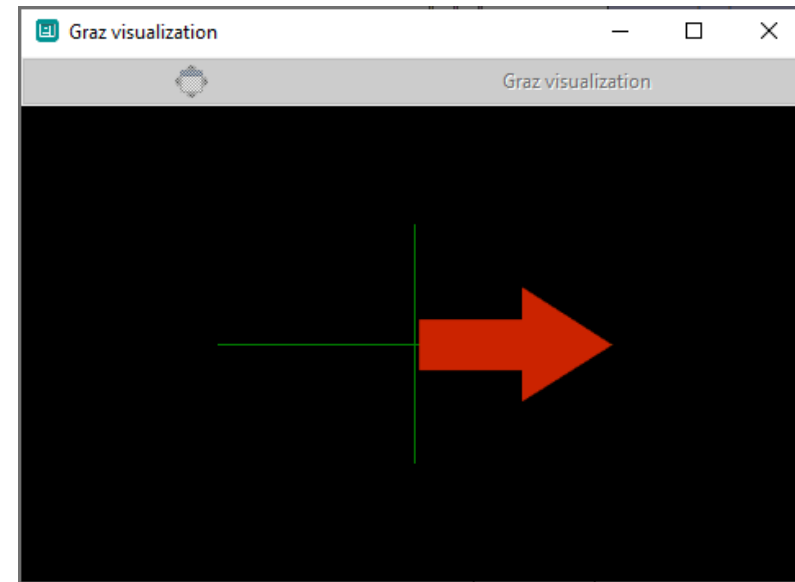
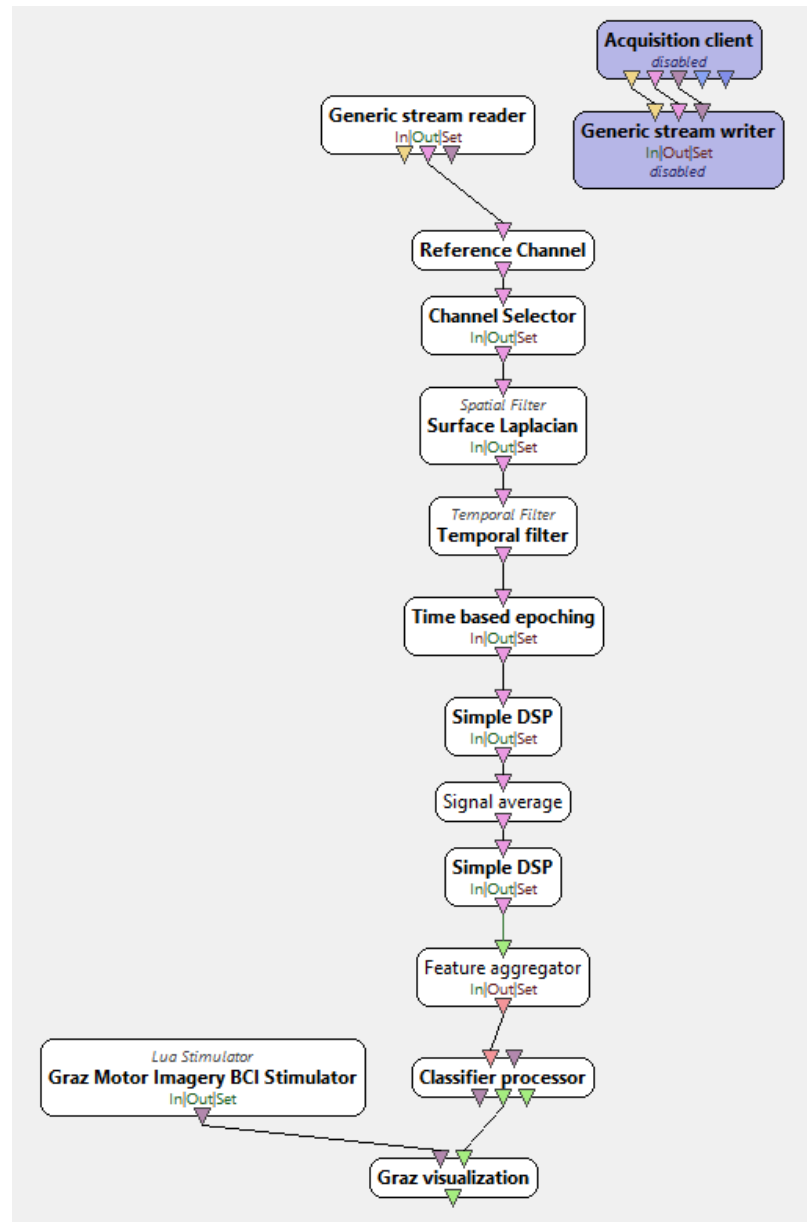


- Use the “**Classifier Processor**” box, loading the correct configuration file (= the training step output)
- Use the same LUA script & Graz Visualization boxes as for the SC1 (acquisition) to generate stimulations & “tasks”
- The “**streamed matrix**” stream from **Classifier processor** to **Graz Visualization** corresponds to the “classification **accuracy**”



- **Notes** regarding the difference between this simulation and an actual BCI/MI protocol
- In a real BCI experiment, the user would receive as visual inputs:
  - first, an arrow towards the side on which to perform Motor Imagery task,
  - then a continuous feedback of the classifier's performance (in separating MI from rest)
- As in SC1, the Graz visualization box is used both to update the display, and to propagate the stimulations to the ACQ server for synchronization with the signal stream.
- Here, we use pre-recorded signals, so this stimulation sync. is not done...





## SC3 – FINAL WORDS / Q&A

- **Recap:**

- Online classification pipeline
- Example of real-time visual feedback

## CHAPTER 2 – Q&A



BCI Motor Imagery with OpenViBE in X-Men: First Class