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GRA	PHICAL USER INTERFACE	
	% % 1 DIRPOSE:	
	% % 1. PURPOSE: % For visualizing single-channel LFD signals, their time-frequency	
	% For visualizing single-channel LFP signals, their time-frequency	
	<pre>% For visualizing single-channel LFP signals, their time-frequency % characteristics; and detect epileptic seizures in zebrafish larvae</pre>	
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	<pre>% For visualizing single-channel LFP signals, their time-frequency % characteristics; and detect epileptic seizures in zebrafish larvae % Note that the algorithm (feature extraction and classifier) % was optimized for detecting seizures in a specific genetic and</pre>	
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	<pre>% For visualizing single-channel LFP signals, their time-frequency % characteristics; and detect epileptic seizures in zebrafish larvae % Note that the algorithm (feature extraction and classifier) % was optimized for detecting seizures in a specific genetic and % chemical model of epilepsy as described in the reference paper. % For optimal performance on different zebrafish models and signals % recorded under different conditions, the classifier needs to be % retrained on data from a similar setting. % 2. LANGUAGE: % Matlab 2014b (GUI) and 2016 (Classifier training) % 3. REFERENCE: % Borbála Hunyadi, Aleksandra Siekierska, Jo Sourbron, Daniëlle % Copmans, Peter A.M. de Witte, % Automated analysis of brain activity for seizure</pre>	
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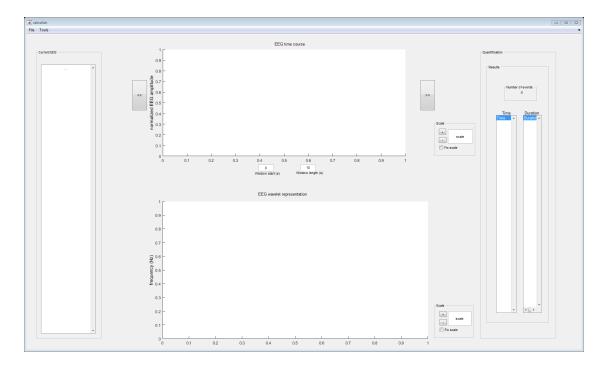
```
% Journal of Neuroscience Methods, Volume 287, 1 August 2017,
% Pages 13-24, ISSN 0165-0270,
% https://doi.org/10.1016/j.jneumeth.2017.05.024.
% 4. OWNER AND DEVELOPER
% Owner: KU Leuven
% Developer: Borbala Hunyadi, Stadius, ESAT, KU Leuven
% Contact person: Borbala Hunyadi, Stadius, ESAT, KU Leuven
% Description of code:
% The functions below initialize the GUI layout and default function
% as created by GUIDE. It also specifies some global parameters
% specific for the application.
% To run the GUI, simply type zebrafish in the Matlab command line and
% press Enter. The GUI will initialize.
function varargout = zebrafish(varargin)
% Last Modified by by Borbála Hunyadi 28/06/2017
% STADIUS, Department of Electrical Engineering, KU Leuven
qui Singleton = 1;
gui_State = struct('gui_Name',
                                    mfilename, ...
    'gui_Singleton', gui_Singleton, ...
    'gui_OpeningFcn', @zebrafish_OpeningFcn, ...
    'gui_OutputFcn', @zebrafish_OutputFcn, ...
    'gui_LayoutFcn', [], ...
    'gui_Callback',
                     []);
if nargin && ischar(varargin{1})
   gui_State.gui_Callback = str2func(varargin{1});
end
if nargout
    [varargout{1:nargout}] = gui_mainfcn(gui_State, varargin{:});
   gui_mainfcn(gui_State, varargin{:});
% End initialization code - DO NOT EDIT
% --- Executes just before zebrafish is made visible.
function zebrafish OpeningFcn(hObject, eventdata, handles, varargin)
% This function has no output args, see OutputFcn.
% hObject handle to figure
% eventdata reserved - to be defined in a future version of MATLAB
            structure with handles and user data (see GUIDATA)
% handles
% varargin command line arguments to zebrafish (see VARARGIN)
%set(handles.AnalyzeBatchMenuItem, 'Enable', 'on');
```

2

```
%Initial setting Zebrafish
handles.fs=1000;
handles.window=10*handles.fs; % 10 s long window
handles.start=1;
handles.step=10*handles.fs; %initial length of EEG window is
 10s=10000ms
handles.ylimits1=[];
handles.fixaxes1=0;
handles.fixaxes2=0;
handles.climit=[1 64];
% Choose default command line output for zebrafish
handles.output = hObject;
% Update handles structure
guidata(hObject, handles);
% This sets up the initial plot - only do when we are invisible
% so window can get raised using zebrafish.
if strcmp(get(hObject,'Visible'),'off')
    axes(handles.axes1);
    ylabel('normalized EEG amplitude')
    axes(handles.axes2);
    ylabel('frequency (Hz)')
end
% UIWAIT makes zebrafish wait for user response (see UIRESUME)
% uiwait(handles.figure1);
% --- Outputs from this function are returned to the command line.
function axes1_CreateFcn(hObject, eventdata, handles)
% hObject
           handle to axes1 (see GCBO)
% eventdata reserved - to be defined in a future version of MATLAB
% handles
            empty - handles not created until after all CreateFcns
 called
% Hint: place code in OpeningFcn to populate axes1
function varargout = zebrafish_OutputFcn(hObject, eventdata, handles)
% varargout cell array for returning output args (see VARARGOUT);
           handle to figure
% hObject
% eventdata reserved - to be defined in a future version of MATLAB
% handles structure with handles and user data (see GUIDATA)
% Get default command line output from handles structure
varargout{1} = handles.output;
% --- Executes on button press in pushbutton1.
function pushbutton1 Callback(hObject, eventdata, handles)
           handle to pushbutton1 (see GCBO)
% hObject
% eventdata reserved - to be defined in a future version of MATLAB
```

```
% handles
                             structure with handles and user data (see GUIDATA)
% --- Executes when figure1 is resized.
function figure1_ResizeFcn(hObject, eventdata, handles)
% hObject
                            handle to figure1 (see GCBO)
% eventdata reserved - to be defined in a future version of MATLAB
                             structure with handles and user data (see GUIDATA)
% handles
% --- Executes on selection change in popupmenul.
function popupmenul Callback(hObject, eventdata, handles)
function popupmenul_CreateFcn(hObject, eventdata, handles)
if ispc && isequal(get(hObject, 'BackgroundColor'),
 get(0,'defaultUicontrolBackgroundColor'))
        set(hObject, 'BackgroundColor', 'white');
end
set(hObject, 'String',
  \{ 'plot(rand(5))', 'plot(sin(1:0.01:25))', 'bar(1:.5:10)', 'plot(membrane)', 'surf (1:.5:10)', 'surf
function popupmenu2_Callback(hObject, eventdata, handles)
function popupmenu2_CreateFcn(hObject, eventdata, handles)
if ispc && isequal(get(hObject, 'BackgroundColor'),
 get(0,'defaultUicontrolBackgroundColor'))
         set(hObject, 'BackgroundColor', 'white');
end
function listbox3 Callback(hObject, eventdata, handles)
function listbox3 CreateFcn(hObject, eventdata, handles)
if ispc && isequal(get(hObject, 'BackgroundColor'),
 get(0,'defaultUicontrolBackgroundColor'))
        set(hObject,'BackgroundColor','white');
end
function listbox4_Callback(hObject, eventdata, handles)
function listbox4 CreateFcn(hObject, eventdata, handles)
if ispc && isequal(get(hObject, 'BackgroundColor'),
 get(0,'defaultUicontrolBackgroundColor'))
         set(hObject, 'BackgroundColor', 'white');
end
function text3_CreateFcn(hObject, eventdata, handles)
```

```
function text7_Callback(hObject, eventdata, handles)
function text7_CreateFcn(hObject, eventdata, handles)
if ispc && isequal(get(hObject, 'BackgroundColor'),
   get(0, 'defaultUicontrolBackgroundColor'))
   set(hObject, 'BackgroundColor', 'white');
end
```



I. FILE MENU

function FileMenu_Callback(hObject, eventdata, handles)

1. Open an EEG file

The expected format is a .txt file with a single column of values, correspoding to each LFP sample. See example.txt for an example EEG/LFP file. WARNINING: The software assumes a sampling rate of 1000Hz.

```
function open_file_Callback(hObject, eventdata, handles)
% hObject    handle to open_file (see GCBO)
% eventdata    reserved - to be defined in a future version of MATLAB
% handles    structure with handles and user data (see GUIDATA)

[FileName,PathName] = uigetfile('*.txt','Select an EEG file','MultiSelect','off');
```

```
if isequal(FileName,0)
    return
end
if isequal(FileName, 0)
    return;
end
handles.eegFile=FileName;
eeg=csvread([PathName,FileName]);
handles.eeq=eeq(:,1);
handles.FileName=FileName;
set(handles.text7, 'String', (FileName));
disp('EEG data loaded...')
handles.start=1;
handles.stop=handles.start+handles.window-1;
msg=msgbox('Preprocessing, please wait...');
[handles.eegfilt, handles.wc, handles.eegplot]=processEEG(eeg, 10);
handles.wc(isnan(handles.wc))=0;
handles.eegfilt=handles.eegfilt';
plotEEG( handles)
if ishandle(msq)
    delete(msg);
guidata(hObject,handles);
%plotting whole signal
%plot(handles.eeg);
```

2. Create a batch file of multiple LFP recordings

A batch file is a list of EEG/LFP files. It allows the user to run a certian analysis on multiple files together. The batch file is saved in .mat format and contains the name of the LFP files

```
function create_batch_Callback(hObject, eventdata, handles)

% hObject handle to create_batch (see GCBO)
% eventdata reserved - to be defined in a future version of MATLAB
% handles structure with handles and user data (see GUIDATA)
[eegFiles,eegPath] = uigetfile('*.txt','Select EEG files for the batch','MultiSelect','on');
if isequal(eegFiles,0)
    return
end
BatchName = uiputfile('.mat','Specify a name for the batch file');
save(BatchName,'eegFiles','eegPath');
```

3. Open a batch file of LFP files, created using this GUI.

```
function open_batch_Callback(hObject, eventdata, handles)

[batchFile,batchPath] = uigetfile('*.mat','Select batch
    file','MultiSelect','off');

if isequal(batchFile,0)
        return
end
load([batchPath, batchFile]);
handles.eegFiles=eegFiles;
handles.eegPath=eegPath;
set(handles.text7, 'String', (eegFiles));

axes(handles.axes1);
axes(handles.axes2);

guidata(hObject,handles);
```

4. Save analysis results in a .csv file

```
function save_analysis_Callback(hObject, eventdata, handles)
headers = {'spike time', 'spike duration'};
cdir = getappdata(0, 'currentdir');
[fn p]=uiputfile([cdir,'*.csv'], 'Save to');
if isequal(fn,0) | isequal(p,0)
    return
end

data=[handles.spiketime',handles.duration'];
cd(pwd);
savef = fullfile(p, fn);

fid = fopen(savef, 'w');
fprintf(fid, '%s;', headers{1});fprintf(fid, '%s\n', headers{2});
fclose(fid);
dlmwrite(savef,data,'-
append','roffset',1,'roffset',0,'delimiter',';');
```

5. Close GUI

```
function CloseMenuItem_Callback(hObject, eventdata, handles)

selection = questdlg(['Close ' get(handles.figure1,'Name') '?'],...
    ['Close ' get(handles.figure1,'Name') '...'],...
    'Yes','No','Yes');
```

```
if strcmp(selection,'No')
    return;
end

delete(handles.figure1)
```

II. TOOLS MENU

```
% -----
function ToolsMenu_Callback(hObject, eventdata, handles)
```

1. Seizure detection in single file

This option will run the seizure detection on the LFP file which has been loaded previously using the File menu > Open EEG option. The software prompts for 2 values, an energy threshold and a minimum seizure length value. Recommended values are filled in by default, or the user can specify an appropriate value. Energy threshold: the larger it is, the less candidate seizure segments will be selected. Extreme values are between 5-100, a reasonable values are 10-20. Minimum seizure length (in ms): the larger it is, the less candidate seizure segments will be selected. Typical values vary depending on the specific epilepsy model. The software also prompts the user to choose a classifier file (class_svm_genetic or class_svm_chemical): see the reference paper for details.

```
function AnalyzeFileMenuItem_Callback(hObject, eventdata, handles)
prompt = {'Enter energy threshold:','Enter mininum seizure length:'};
dlg_title = 'Input';
num lines = 1;
def = \{ '20', '60', '0.5' \};
answer = inputdlg(prompt,dlg_title,num_lines,def);
if isequal(answer,{})
    return
end
nleo thr=str2double(answer{1});
len_thr=str2double(answer{2});
%prior=str2double(answer{3}); prior=[prior 1-prior];
classifier = uigetfile('*.mat','Select a classifier
 file','MultiSelect','off');
%classifier='class_svm_chemical.mat';
 if isequal(classifier,0)
     return
 end
[handles.spiketime, handles.duration] = detect_spikes_final(handles.eeg, handles.eegFi
no_spikes=size(handles.spiketime,2);
set(handles.text3, 'String', num2str(no_spikes));
spiketime = sprintf('%.2f|', handles.spiketime);
duration = sprintf('%.2f|', handles.duration);
set(handles.listbox3,'String',num2str(spiketime));
set(handles.listbox4, 'String', num2str(duration));
```

2. Seizure detection in a batch of files

This option will run the seizure detection on all the LFP files specified in the batch file perviously loaded using the File > Open batch file option. The batch file can be created using File > Create batch file option. The software prompts for 2 values, an energy threshold and a minimum seizure length value. Recommended values are filled in by default, or the user can specify an appropriate value. Energy threshold: the larger it is, the less candidate seizure segments will be selected. Extreme values are between 5-100, a reasonable values are 10-20. Minimum seizure length (in ms): the larger it is, the less candidate seizure segments will be selected. Typical values vary depending on the specific epilepsy model. The software also prompts the user to choose a classifier file (class_svm_genetic or class_svm_chemical): see the reference paper for details.

```
function AnalyzeBatchMenuItem_Callback(hObject, eventdata, handles)
prompt = {'Enter energy threshold:','Enter mininum seizure length:'};
dlg_title = 'Input';
num lines = 1;
def = \{ '20', '60' \};
answer = inputdlg(prompt,dlg_title,num_lines,def);
if isequal(answer,{})
    return
end
nleo thr=str2double(answer{1});
len_thr=str2double(answer{2});
%prior=str2double(answer{3}); prior=[prior 1-prior];
classifier = uigetfile('*.mat','Select a classifier
 file','MultiSelect','off');
%classifier='class_svm_chemical.mat';
 if isequal(classifier,0)
     return
 end
%classifier=load('class svm.mat');
for i=1:numel(handles.eegFiles)
    msg=msgbox(['Processing file ' num2str(i) 'out of '
 num2str(numel(handles.eegFiles)) '...']);
    eeq=csvread([handles.eeqPath,handles.eeqFiles{i}]);
    handles.eeg=eeg(:,1);
 [spiketime{i},duration{i}]=detect_spikes_final(handles.eeg,handles.eegFiles{i},nl
    no spikes(i)=size(spiketime,2);
    if ishandle(msq)
        delete(msg);
    end
end
```

```
waitfor(msgbox('Batch analysis is done. Press OK to save
 results...'));
%ResultsName = uiputfile('.mat','Specify a filename to save analysis
 results...');
headers = {'seizure start time','duration'};
cdir = getappdata(0,'currentdir');
[fn p]=uiputfile([cdir,'*.csv'], 'Specify a filename to save batch
 analysis results');
if isequal(fn,0) | isequal(p,0)
    return
end
data=[];
for i=1:numel(handles.eegFiles)
    data=[data; [{'filename'}, handles.eegFiles{i}]];
    data=[data; [{'seizure start time'},{'duration'}]];
    data=[data; num2cell([spiketime{i}',duration{i}'])];
    data=[data; [{''},{''}]];
end
cd(pwd);
savef = fullfile(p, fn);
fid = fopen(savef, 'w');
cell2csv(savef,data,';')
fclose(fid);
```

3. Time-frequency analysis: visualization

```
This option visualizes the spectrogram of the currently displayed LFP segment.

------

function tf_anal_Callback(hObject, eventdata, handles)

[S] = spectrogram(handles.eegfilt(handles.start:handles.stop),100,99,1:1:400,handles.fs figure; imagesc(abs(S)); 
xlabel('time (ms)'); 
ylabel('frequency (Hz)');
```

4. Time-frequency quantification

This option computes the power spectral density estaimte, using Welch's method, of the full LFP signal (all LFP files, in case a batch was loaded) in a user-defined frequency band. The software will prompt the user to define the lower and higher limits of the frequency band (in Hz). It is possible to define multiple frequency bands. In this case, the lower limits and the higher limits of the frequency bands have to be specified as Matlab arrays. For example, to define the frequency bands 1-20Hz, 21-40Hz, 41-60Hz, the following parameters should be specified: lower frequency of the band: 1:20:41. higher frequency of the band: 20:20:60. Note that the signals are analysed in short windows, therefore, the estimation will not be precise for low frequencies (below 10Hz). The results will be save in a csv file. The user will be prompted to specify a file name.

```
function tf_quant_Callback(hObject, eventdata, handles)
prompt = {'Enter lower frequency of the band:','Enter higher frequency
of the band: '};
dlg_title = 'Input';
num_lines = 1;
def = \{ '20', '40' \};
answer = inputdlg(prompt,dlg_title,num_lines,def);
if isequal(answer,{})
    return
end
f1=eval(answer{1});
f2=eval(answer{2});
try
    singlemode=0;
    handles.eegFiles
catch
    singlemode=1;
handles.eegFiles{1}=handles.eegFile;
end
for i=1:numel(handles.eegFiles)
    msg=msgbox(['Processing file ' num2str(i) 'out of '
 num2str(numel(handles.eegFiles)) '...']);
    if singlemode
        % do nothing
    else
    eeg=csvread([handles.eegPath,handles.eegFiles{i}]);
    handles.eeg=eeg(:,1);
    end
    % freq. anal
    [pW(i,:)] = pwelch(handles.eeg,100,80,512); %% 100 sample windows
 with 80% overlap
    if ishandle(msq)
        delete(msg);
    end
end
f=0:(handles.fs/2)/(size(pW,2)-1):(handles.fs/2);
for j=1:numel(f1)
f1x(j)=find(f>f1(j),1);
f2x(j)=find(f>f2(j),1);
f50=find(f<50,1,'last');
figure
for i=1:numel(handles.eegFiles)
semilogy(pW(i,:));
```

```
hold all;
end
legend(handles.eegFiles)
title('Power Spectral Density')
xlabel('Frequency (Hz)')
ylabel('log(Y(f).^2)')
set(gca,'XTick',f50:f50:max(f))
set(gca,'XTickLabel',50:50:500)
xlim([1 numel(f)-1]);
waitfor(msqbox('Batch analysis is done. Press OK to save
 results...'));
cdir = getappdata(0,'currentdir');
[fn p]=uiputfile([cdir,'*.csv'], 'Specify a filename to save batch
 analysis results');
if isequal(fn,0) | isequal(p,0)
    return
end
data=[];
hdrs=[];
for j=1:numel(f1)
hdrs=[hdrs {[num2str(f1(j)) ' - ' num2str(f2(j)) ' Hz band']}]
data=[data; [{'filename'},hdrs]];
for i=1:numel(handles.eegFiles)
    bandpow=[];
    for j=1:numel(f1)
        f1x(j)
        f2x(j)
    bandpow=[bandpow num2cell(sum(pW(i,f1x(j):f2x(j))))];
    data=[data; [handles.eegFiles{i} bandpow]];
end
cd(pwd);
savef = fullfile(p, fn);
fid = fopen(savef, 'w');
cell2csv(savef,data,';')
fclose(fid);
```

III. DISPLAY OPTIONS

1. Fix axes scale for time course

```
function checkbox1_Callback(hObject, eventdata, handles)
handles.fixaxes1 = get(hObject, 'Value');
axes(handles.axes1);
```

```
y=get(gca,'ylim');
handles.ylimits1=y;
quidata(hObject,handles);
```

2. Fix color scale for wavelet transformed signal

```
function checkbox2_Callback(hObject, eventdata, handles)
handles.fixaxes2 = get(hObject, 'Value');
axes(handles.axes2);
y=get(gca,'clim');
handles.climit=y;
guidata(hObject, handles);
```

3. Manual adjust window length

```
function edit2_CreateFcn(hObject, eventdata, handles) % window length

if ispc && isequal(get(hObject, 'BackgroundColor'),
    get(0, 'defaultUicontrolBackgroundColor'))
        set(hObject, 'BackgroundColor', 'white');
end

function edit2_Callback(hObject, eventdata, handles) % window length

handles.window=str2double(get(hObject, 'String'))*handles.fs;
handles.stop=handles.start+handles.window;
if handles.stop<numel(handles.eegfilt)

plotEEG(handles)

end
handles.step=handles.window;
guidata(hObject, handles);</pre>
```

4. Semi-auto adjust timecourse scale limits: increase or decrease

```
function pushbutton6_Callback(hObject, eventdata, handles) %% adjust
    scale timecourse

if ~handles.fixaxes1
    axes(handles.axes1);
    y=get(gca,'ylim');
    handles.ylimits1=[-max(abs(y))*2 max(abs(y))*2];
    set(handles.axes1,'ylim',handles.ylimits1);
    set(handles.edit3,'String',['['
    num2str(handles.ylimits1(1),'%.1f') ' '
    num2str(handles.ylimits1(2),'%.1f') ' ']' ]);
```

```
guidata(hObject,handles);
end

function pushbutton7_Callback(hObject, eventdata, handles) %% adjust
    scale timecourse

if ~handles.fixaxes1
    axes(handles.axes1);
    y=get(gca,'ylim');
    handles.ylimits1=[-max(abs(y))/2 max(abs(y))/2];
    set(handles.axes1,'ylim',handles.ylimits1);
    set(handles.edit3,'String',['['
    num2str(handles.ylimits1(1),'%.1f') ' '
    num2str(handles.ylimits1(2),'%.1f') ' ']' ]);
    guidata(hObject,handles);
end
```

5. Semi-auto adjust wavelet scale

```
function pushbutton8_Callback(hObject, eventdata, handles)
if ~handles.fixaxes2
   axes(handles.axes2);
   y=get(gca,'clim');
   handles.climit=[y(1)/2 y(2)*2];
   set(handles.axes2,'clim',handles.climit);
    set(handles.edit4, 'String', ['['
num2str(handles.climit(1),'%.1f') ' '
num2str(handles.climit(2),'%.1f') ']']);
   guidata(hObject,handles);
end
function pushbutton9_Callback(hObject, eventdata, handles)
if ~handles.fixaxes2
    axes(handles.axes2);
   y=get(gca,'clim');
    try
       handles.climit=[y(1)*2 y(2)/2];
        set(handles.axes2,'clim', handles.climit);
        handles.climit=[y(1)*2 max(y(2)/2, y(1)*2+1)];
        set(handles.axes2,'clim', handles.climit);
    end
   set(handles.edit4,'String',['['
num2str(handles.climit(1),'%.1f') ' '
num2str(handles.climit(2),'%.1f') ']']);
   quidata(hObject,handles);
end
```

6. Manual adjust timecourse scale

```
function edit4_Callback(hObject, eventdata, handles) % manual adjust
 timecourse scale
if ~handles.fixaxes2
   handles.climit = eval(get(hObject,'String'));
    set(handles.axes2, 'clim', handles.climit);
   guidata(hObject,handles);
end
function edit4_CreateFcn(hObject, eventdata, handles) % manual adjust
timecourse scale
% hObject
            handle to edit4 (see GCBO)
% eventdata reserved - to be defined in a future version of MATLAB
             empty - handles not created until after all CreateFcns
% handles
called
% Hint: edit controls usually have a white background on Windows.
       See ISPC and COMPUTER.
if ispc && isequal(get(hObject, 'BackgroundColor'),
get(0, 'defaultUicontrolBackgroundColor'))
    set(hObject, 'BackgroundColor', 'white');
end
if exist('handles.axes1')
   axes(handles.axes1);
   y=get(gca,'ylim');
    set(hObject, 'String',['[' num2str(y) ']' ]);
end
```

7. Manual adjust wavelet scale

```
function edit3_Callback(hObject, eventdata, handles) % manual adjust
wavelet scale

if ~handles.fixaxes1
    handles.ylimit1 = eval(get(hObject, 'String'));
    set(handles.axes1, 'ylim', handles.ylimit1);
    guidata(hObject, handles);
end

% --- Executes during object creation, after setting all properties.

function edit3_CreateFcn(hObject, eventdata, handles)

if ispc && isequal(get(hObject, 'BackgroundColor'),
    get(0, 'defaultUicontrolBackgroundColor'))
    set(hObject, 'BackgroundColor', 'white');
end
```

```
if exist('handles.axes2')
    axes(handles.axes2);
    y=get(gca,'clim');
    set(hObject,'String',['[' num2str(y) ']' ]);
end
```

8. Adjust window start time

```
function edit5_Callback(hObject, eventdata, handles)
handles.start = str2double(get(hObject, 'String'))*handles.fs;
handles.stop = handles.start+handles.window;
guidata(hObject, handles);
plotEEG(handles);

function edit5_CreateFcn(hObject, eventdata, handles) % window start time

if ispc && isequal(get(hObject, 'BackgroundColor'),
    get(0,'defaultUicontrolBackgroundColor'))
    set(hObject,'BackgroundColor','white');
end
```

9. Scroll forward and backward

```
function pushbutton4_Callback(hObject, eventdata, handles)
handles.start=handles.start-handles.step;
handles.stop=handles.start+handles.window;
if handles.stop>=handles.window
    plotEEG( handles)
end
guidata(hObject, handles);

function pushbutton5_Callback(hObject, eventdata, handles) %%
handles.start=(handles.start+handles.step);
handles.stop=handles.start+handles.window;
if handles.stop<numel(handles.eegfilt)
    plotEEG( handles)
end
guidata(hObject, handles);</pre>
```

10. Data Visualization

Visualizes the signle channel LFP signal (top) and its wavelet transform (bottom)

```
function plotEEG( handles)
```

```
axes(handles.axes1);
plot([handles.start:handles.stop]./
handles.fs, handles.eegplot(handles.start:handles.stop));
if handles.fixaxes1
    set(handles.axes1, 'ylim', handles.ylimits1, 'xlim', [[handles.start
handles.stop]./handles.fs])
else
    axis('tight')
end
axes(handles.axes2);
if handles.fixaxes2
    imagesc([handles.start:handles.stop]./handles.fs,
 [1:size(handles.wc,1)], handles.wc(:,[handles.start:handles.stop]),
[ handles.climit]);
else
    maxval=max(max(handles.wc(:,[handles.start:handles.stop])));
    minval=min(min(handles.wc(:,[handles.start:handles.stop])));
    imagesc([handles.start:handles.stop]./handles.fs,
 [1:size(handles.wc,1)], handles.wc(:,[handles.start:handles.stop]),
[minval maxval;]);
end
colorbar('location','EastOutside')
axes(handles.axes1);
y=get(gca,'ylim');
set(handles.edit3, 'String', ['[' num2str(y, '%.1f') ']' ]);
axes(handles.axes2);
y=get(gca,'clim');
set(handles.edit4,'String',['[' num2str(y,'%.1f') ']' ]);
% Preprocess EEG prior to visualization
function [eeg,wc,eegplot]=processEEG(eeg,cutoff)
fs=1000;
% settings
res=10; % 10 windows per second
% 'From which second is the EEG stable? (detection starts from here):
tmp=zeros(1,100);
for i=1:100
tmp(i) = max(abs(eeg((i-1)*fs+1:i*fs)));
if tmp(i) < 1
    break;
end
end
```

```
eeg(1:(i-1)*fs)=eeg((i-1)*fs+1);
% interpolate large artifacts
%eeg(1)=0;eeg
x=1:numel(eeq);
xi=x(find(abs(eeg)<1));</pre>
yi=eeg(find(abs(eeg)<1));</pre>
eegi=interp1(xi,yi,x,'linear');
clear xi yi;
% high pass filter again baseline fluctuation
[b,a] = butter(6,cutoff/(fs/2),'high'); %%select appropriate filter
 curoff
eegfilt=filtfilt(b,a,eegi);
eegfilt=zscore(eegfilt);
[b,a] = butter(6,1/(fs/2),'high'); %%%select appropriate filter curoff
eegplot=filtfilt(b,a,eegi);
eegplot=zscore(eegplot);
eeg=eegfilt;
clear eegi eegfilt;
disp('preprocessing...')
% compute wavelet coefficients
scales = make_scales(2,200,1,'bior1.5',fs);
wc=zeros(length(scales),length(eeg));
wc=cwt(eeq,scales,'bior1.5').^2;
for i=1:length(scales)
    wc(i,:)=zscore(wc(i,:));
end
```

IV. CORE FUNCTIONS

```
% These functions are called by the buttons of the GUI, the user does
not
% need to interact with them directly.
```

1. Seizure detection

```
function [spike_timing,
  duration] = detect_spikes_final(eeg,fname,nleo_thr,len_thr,classifier)
load(classifier);
if size(SVMModel.X,2) == 112
    flag='genetic'
```

```
elseif size(SVMModel.X,2)==101
    flag='chemical';
end
fs=1000;
res=10;
%nleo thr=10;
%len thr=50;
% preprocess before feature extraction
         disp('preprocessing, please wait...')
        % 'From which second is the EEG stable? (detection starts from
 here): ';
        tmp=zeros(1,100);
        for j=1:100
            tmp(j) = max(abs(eeg((j-1)*fs+1:j*fs)));
            if tmp(j) < 1
                break;
            end
        end
        eeg(1:(j-1)*fs)=eeg((j-1)*fs+1);
        % interpolate large artifacts
        %eeq(1)=0;
        x=1:numel(eeg);
        xi=x(find(abs(eeg)<1));</pre>
        yi=eeg(find(abs(eeg)<1));</pre>
        eegi=interp1(xi,yi,x,'linear');
        clear xi yi;
        % high pass filter again baseline fluctuation
        if strcmp(flag, 'chemical')
        [b,a] = butter(6,10/(fs/2),'high');
        elseif strcmp(flag, 'genetic')
        [b,a] = butter(6,1/(fs/2),'high');
        end
        eegfilt=filtfilt(b,a,eegi);
        eegfilt=zscore(eegfilt);
        eeg=eegfilt;
        clear eegi eegfilt;
disp('extracting features, please wait...')
% segmentation: select segment which an NLEO larger than threshold
eegs=zscore(eeg);
```

```
if size(eegs,1)==1; eegs=eegs'; end
sig kaiser = kaiserVarRT(eegs);
cand=find(sig_kaiser>=nleo_thr);
len=zeros(size(cand));
fin=cand;
% cluster the close high energy samples together
for j=length(cand):-1:2
    if cand(j)-cand(j-1)<200 %% if they are not 200ms apart, they
belong to the same event
        fin(j-1)=fin(j);
        len(j-1)=fin(j-1)-cand(j-1);
        len(j)=0;
    end
end
% feature extraction
clear selector segment featvec;
c=find(len>len_thr);
for j=1:numel(c)
   candfin=zeros(size(eegs));
   candfin(cand(c(j)):cand(c(j))+len(c(j)))=1;
    selector{j}=unique(ceil(find(candfin)/(fs/res)));
    segment{j}=cand(c(j)):cand(c(j))+len(c(j));
   featvec(j,:)=extract_features_segment(eeg(segment{j})),flag);
end
if(isempty(c))
    featvec=[];
   spike_timing=[];
   duration=[];
else
   disp('classification...')
   detections = find(predict(SVMModel,featvec));
    if(isempty(detections))
        spike timing=[];
       duration=[];
    else
        for i=1:numel(detections)
            spike timing(i)=segment{detections(i)}(1)/1000;
            duration(i)=numel(segment{detections(i)});
        end
```

```
end
end
disp('done...')
```

2. Feature extaction

```
function featvec=extract_features_segment(eeg,flag)
% flag: specify if feature extraction according to the genetic or
 chemical model should be run (see reference paper for details!)
 fs=1000;
응
if strcmp(flag,'chemical')
% compute wavelet coefficients
 scales = make_scales(10,100,1,'morl',fs);
 freqs=10:1:100;
 f10=find(freqs==10);
 f20=find(freqs==20);
 f50=find(freqs==50);
    wc=cwt(eeg,scales,'morl').^2;
for i=1:length(scales)
    wc(i,:)=wc(i,:)./std(wc(i,:));
end
% mean wavelet coef. in 1-50Hz band
wc20=mean(wc(f10+1:f20,:),1);
wc50=mean(wc(f20+1:f50,:),1);
wc100=mean(wc(f50+1:end,:),1);
    i=1;
% feature 1: time domain
     zamp(i)=max(abs(eeq));
     rms(i)=sqrt(mean(eeg.^2));
% feature 2: the aveage wavelet coef. in 1-50Hz band exceeds the 20x
 standard deviation
   % mwc10(i)=mean(wc10);
     mwc20(i)=mean(wc20);
     mwc50(i)=mean(wc50);
     mwc100(i)=mean(wc100);
     mwc(:,i)=mean(wc,2);
```

```
% criterion 3: zero crossings, mins, maxes
    [zeronum(i),
 maxnum(i),minnum(i),m_zero_intvl(i),std_zero_intvl(i)]=countzerominmax(zscore(eeg
    zeronum=zeronum/length(eeg);
    maxnum=maxnum/length(eeg);
    minnum=minnum/length(eeg);
%GEN
%featvec=[zamp; rms; mwc10; mwc20; mwc50; mwc100; mwc; wcr10; wcr20;
 zeronum; maxnum; minnum; m_zero_intvl; std_zero_intvl]';
featvec=[zamp; rms; mwc; mwc20; mwc50; mwc100; zeronum; maxnum;
minnum; m_zero_intvl; std_zero_intvl]';
else
    % compute wavelet coefficients
 scales = make_scales(2,100,1,'bior1.5',fs);
    wc=cwt(eeg,scales,'bior1.5').^2;
for i=1:length(scales)
    wc(i,:)=zscore(wc(i,:));
end
% mean wavelet coef. in 1-50Hz band
wc10=mean(wc(1:10,:),1);
wc20=mean(wc(1:20,:),1);
wc50=mean(wc(1:50,:),1);
wc100=mean(wc(1:end,:),1);
    i=1;
% feature 1: time domain
     zamp(i)=max(abs(eeg));
     rms(i)=sqrt(mean(eeg.^2));
% feature 2: the aveage wavelet coef. in 1-50Hz band exceeds the 20x
 standard deviation
     mwc10(i) = mean(wc10);
     mwc20(i)=mean(wc20);
     mwc50(i) = mean(wc50);
     mwc100(i)=mean(wc100);
     mwc(:,i)=mean(wc,2);
% criterion 3: high and low frequency ratio
     wcr10(i)=mean(abs(wc100))/mean(abs(wc10));
     wcr20(i)=mean(abs(wc100))/mean(abs(wc20));
% criterion 4: zero crossings, mins, maxes
    [zeronum(i),
 maxnum(i),minnum(i),m_zero_intvl(i),std_zero_intvl(i)]=countzerominmax(zscore(eeg
```

```
zeronum=zeronum/length(eeg);
    maxnum=maxnum/length(eeg);
    minnum=minnum/length(eeg);
featvec=[zamp; rms; mwc10; mwc20; mwc50; mwc100; mwc; wcr10; wcr20;
 zeronum; maxnum; minnum; m_zero_intvl; std_zero_intvl]';
end
function [zeronum,
maxnum,minnum,m_zero_intvl,std_zero_intvl]=countzerominmax(x)
% This function calualates the number zero crossing, minima, maxima,
% the mean and standard deviation of zero-to-zero intervals
zeronum=0;
maxnum=0;
minnum=0;
zero_idx=[];
for i=2:length(x)-1
    if sign(x(i-1)) \sim = sign(x(i))
        zeronum=zeronum+1;
        zero_idx=[zero_idx i];
    end
    zero_intvl=zero_idx(2:end)-zero_idx(1:end-1);
    m zero intvl=mean(zero intvl);
    std_zero_intvl=std(zero_intvl);
    if x(i-1) < x(i) && x(i) > x(i+1)
        maxnum=maxnum+1;
    end
    if x(i-1)>x(i) && x(i)<x(i+1)
        minnum=minnum+1;
    end
end
function sig_kaiser = kaiserVarRT(sig)
This function calculates the energy of the signal with the teager-
kaiser
%energy operator. This energy is related to the square of the
 immediate
%amplitude and frequency. The operator enhances spike activity
 facilitating
%the segmentation.
[ySize, xSize] = size(sig);
```

```
sig_kaiser = zeros(ySize, xSize);
for c=1:xSize
    sig kaiser(4:end,c) = (sig(3:end-1,c).*sig(2:end-2,c))-
(sig(4:end,c).*sig(1:end-3,c));
sig_kaiser=abs(sig_kaiser);
return;
function scales = make_scales(startFreq,endFreq,freqRes,wl,fs)
% scales = make scales(startFreq,endFreq,freqRes,wl)
% Function that determines the scales needed as input for a CWT with a
% certain wavelet. The start frequency (in Hz) is the first input
argument,
% the end frequency (in Hz) the second, the frequency resolution at
which
% the transformation should be calculated, the third. The final
argument is
% the name of the wavelet used for the CWT. A sampling of 250Hz of the
data
% is assumed.
centerFreq = centfrq(wl);
counter=1;
for i=startFreq:freqRes:endFreq
    scales(counter) = ((centerFreq*fs)/(i));
    counter=counter+1;
end
```

x. Others

```
function cell2csv(filename,cellArray,delimiter)
% Writes cell array content into a *.csv file.
% CELL2CSV(filename,cellArray,delimiter)
               = Name of the file to save. [ i.e. 'text.csv' ]
% filename
% cellarray = Name of the Cell Array where the data is in
% delimiter = seperating sign, normally:',' (default)
% by Sylvain Fiedler, KA, 2004
% modified by Rob Kohr, Rutgers, 2005 - changed to english and fixed
delimiter
if nargin<3
    delimiter = ',';
end
datei = fopen(filename, 'w');
for z=1:size(cellArray,1)
    for s=1:size(cellArray,2)
```

```
var = eval(['cellArray{z,s}']);

if size(var,1) == 0
     var = '';
end

if isnumeric(var) == 1
     var = num2str(var);
end

fprintf(datei,var);

if s ~= size(cellArray,2)
     fprintf(datei,[delimiter]);
end
end
fprintf(datei,'\n');
end
fclose(datei);
```

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