

Across-trial phase-lag index: tutorial

The following is a step-by-step tutorial demonstrating how the function 'across_trial_pli.m' is implemented.

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Generate pseudo EEG data.

First, we will create pseudo-random EEG data. This data will be event-related (70 trials, 64 channels, 6000 samples per trial at a sampling rate of 1000Hz). Each channel time-series will be a weighted combination of pink noise (i.e., 1/f) and two sinusoids (10Hz and 25Hz), where the phase of each sinusoid is drawn uniformly at random from $-\pi$ to π .

```
% Define the dimensions of the data.
n_channels = 64;
n_samples = 6000;
n_trials = 70;
samp_rate = 1000;

% Define the frequencies and weights of the sinusoids.
peaks = [];
peaks.hz = [10,25];
peaks.w = [.1,.05];

% Define the amount of noise.
noise = 3;

% Set seed for reproducibility
rng(1)

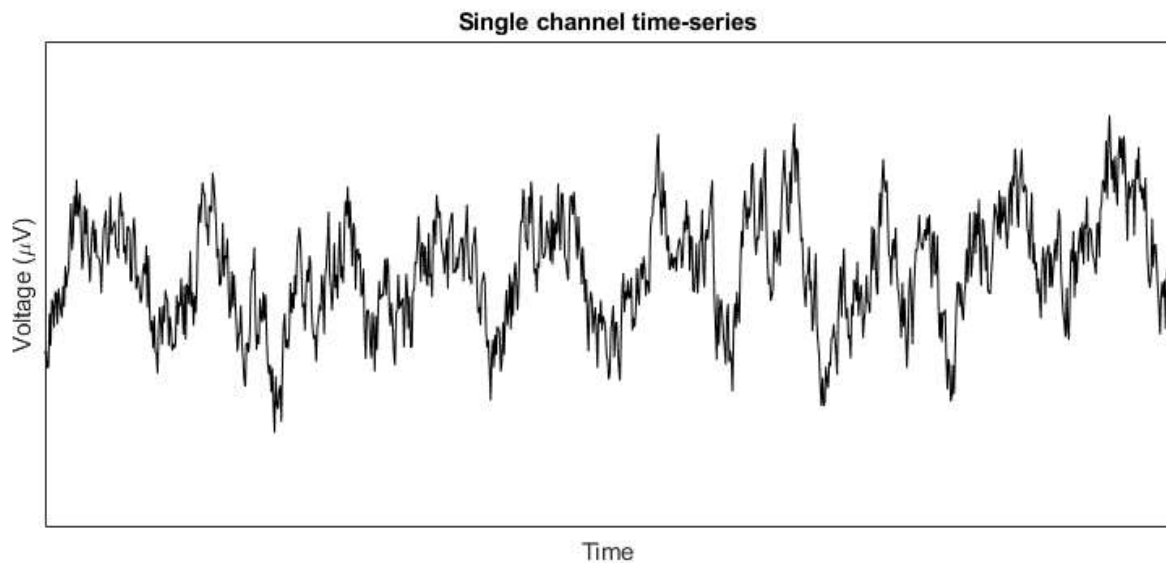
xs = [];
xs.raw = zeros(n_channels,n_samples,n_trials);
time = 0:(1/samp_rate):(n_samples/samp_rate)-(1/samp_rate);

for trial = 1:n_trials
    for channel = 1:n_channels
        sine_1 = sin(2*pi*peaks.hz(1)*time + rand(1)*pi);
        sine_2 = sin(2*pi*peaks.hz(2)*time + rand(1)*pi);
        xs.raw(channel,:,trial) = peaks.w(1)*sine_1 + peaks.w(2)*sine_2;
    end
    xs.raw(:, :, trial) = xs.raw(:, :, trial) + noise*pinknoise(n_samples,n_channels)';
end
```

Let's have a look at the time-series of a single channel to ensure everything worked.

```
% Plot a single channel time-series.

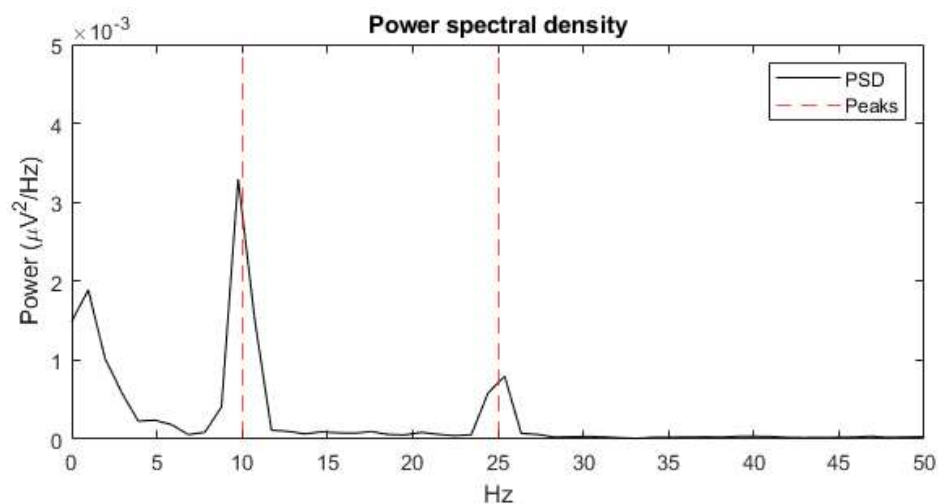
figure('Position',[505 428 895 369]);
plot(xs.raw(1,:,1), 'k');
xlim([650,1650])
ylim([-0.5,0.5])
xlabel('Time')
ylabel('Voltage ( $\mu$ V)')
title('Single channel time-series')
set(gca,'XTick',[], 'YTick', [])
```



Let's inspect the power spectral density of said time-series as well, to ensure that the oscillations we created are present as peaks. With real data, I suggest parameterizing this power spectrum to assess the characteristics of any peaks that exist within the data prior to implementing these analyses. See, for example: <https://foof-tools.github.io/foof/>

```
[~,freq,~,psd] = spectrogram(xs.raw(1,:,1),samp_rate,[],[],samp_rate);
mean_psd = mean(psd,2);

figure('Position',[576 479 678 309]);
plot(freq,mean_psd,'k')
title('Power spectral density')
xlim([0,50])
ylim([0,.005])
xlabel('Hz')
ylabel('Power ({\mu}V^2/Hz)')
xline(peaks.hz,'--','color','red')
legend('PSD','Peaks')
```



Create filter.

Now, let's create a band-pass filter so we can estimate the instantaneous phase of the oscillation at 10Hz. I'll use a window from 5-15Hz for this filter. However, visual inspection, as well as careful consideration of the research question (in terms of time/frequency resolution trade-off) is crucial in selecting this parameter. In an ideal world, the sensitivity of any results to a range of filter widths would be assessed.

```
filt_band = [5 15];

filt_params = [];

filt_params.transition = mean(filt_band) * 0.2;
```

```
filt_params.frequencies = [filt_band(1) - filt_params.transition, filt_band(1), filt_band(2), filt_band(2) + filt_params.transition];
filt_params.order = kaiserord(filt_params.frequencies, [0 1 0], [0.1 0.05 0.1], samp_rate);
filt_params.coefficients = fir1(filt_params.order, filt_band*(2/samp_rate), 'bandpass');
```

The parameters of the filter are stored in the structure 'filt_params', shown here:

```
disp(filt_params)
```

```
transition: 2
frequencies: [3 5 15 17]
order: 630
coefficients: [-4.3094e-05 -4.3982e-05 -4.4236e-05 -4.3866e-05 ... ]
```

Apply filter, estimate instantaneous phase.

Now that we've created the filter, let's apply it and calculate the instantaneous phase of the time-series using the Hilbert transform. First, we will Z-score each channel time-series, then filter, then calculate the instantaneous phase.

```
tic
for trial = 1:n_trials

    xs.norm(:, :, trial) = zscore(xs.raw(:, :, trial));

    for channel = 1:n_channels
        xs.filtered(channel, :, trial) = filtfilt(filt_params.coefficients, 1, squeeze(xs.norm(channel, :, trial)));
    end

    xs.hilbert(:, :, trial) = hilbert(xs.filtered(:, :, trial));
    xs.phase(:, :, trial) = angle(xs.hilbert(:, :, trial));

end
toc
```

Elapsed time is 29.720960 seconds.

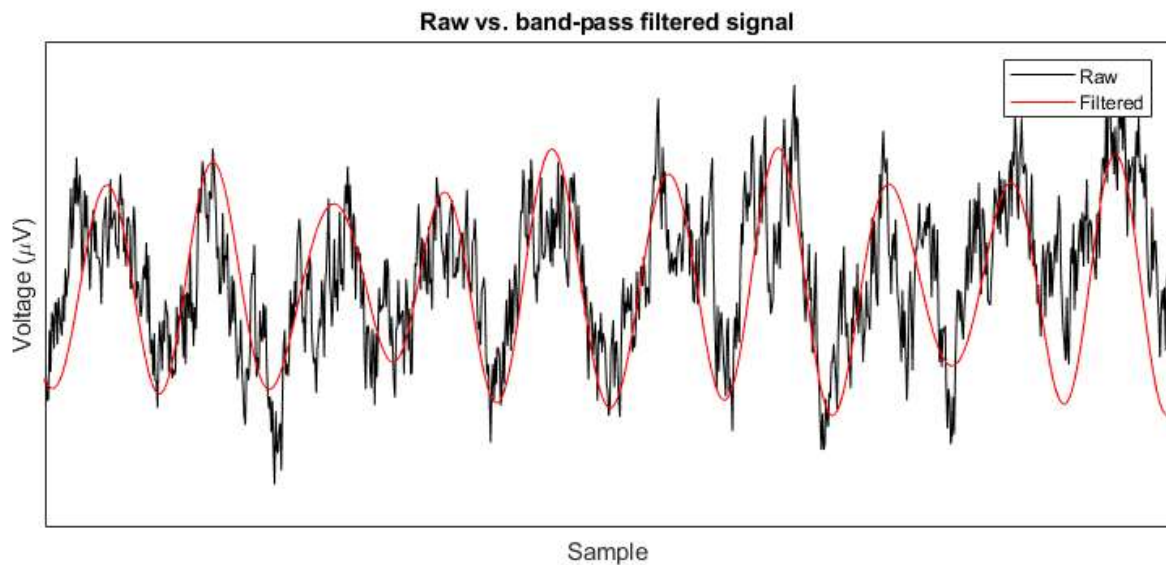
The structure 'xs' (time-series) includes all the intermediate processing steps: the raw signals, the z-scored signals, the band-pass filtered signals, the hilbert-transformed signals, and the instantaneous phase signals.

```
disp(xs)
```

```
raw: [64x6000x70 double]
norm: [64x6000x70 double]
filtered: [64x6000x70 double]
hilbert: [64x6000x70 double]
phase: [64x6000x70 double]
```

Let's plot the raw signal against the band-pass filtered signal to ensure everything worked correctly.

```
figure('Position',[505 428 895 369]);
plot(zscore(xs.raw(1, :, 1)), 'k')
hold on; plot(zscore(xs.filtered(1, :, 1)), 'color', 'red')
xlim([650, 1650])
title('Raw vs. band-pass filtered signal')
legend('Raw', 'Filtered')
xlabel('Sample')
ylabel('Voltage ( $\mu V$ )')
set(gca, 'XTick', [], 'YTick', [])
```



Implement the across-trial phase-lag index.

Now let's go ahead and implement the analyses. First, let's define an in-line function that calculates the phase-lag index. See Stam et al. (2007) for details: <https://pubmed.ncbi.nlm.nih.gov/17266107/>

```
pli_fx = @(phase_set_i,phase_set_j) abs(mean(sign(phase_set_i - phase_set_j),2));
```

Now let's test the function by calculating the across-trial PLI for channels i and j.

```
i = 8;
j = 32;

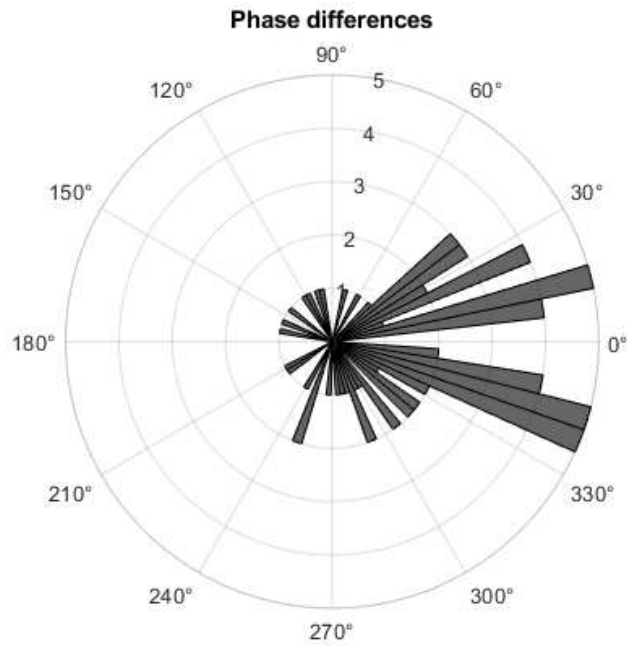
phase_set_i = squeeze(xs.phase(i,:,:));
phase_set_j = squeeze(xs.phase(j,:,:));
```

Let's visualize the angular distribution of the phase differences between channels i and j at a single time-point.

```
% Calculate phase differences.
phase_diffs = phase_set_i - phase_set_j;

% Specify sample of interest.
sample = 1000;

% Plot
figure;
polarhistogram(phase_diffs(sample,:),n_trials,'facecolor','k')
title('Phase differences')
```



The PLI value for this set of phase differences is:

```
pli_fx(phase_set_i(sample,:),phase_set_j(sample,:))
```

```
ans =  
  
    0.1143
```

Now, let's run that calculation for all possible channel pairs, and track the output in the variable 'adjacency_tensor'.

```
tic  
  
adjacency_tensor = zeros(n_channels,n_channels,n_samples);  
channel_pairs = nchoosek(1:n_channels,2);  
  
for ij = 1:length(channel_pairs)  
  
    pair = channel_pairs(ij,:);  
    pli = pli_fx(squeeze(xs.phase(pair(1),:,:)),squeeze(xs.phase(pair(2),:,:)));  
    adjacency_tensor(pair(1),pair(2),:) = pli;  
  
end  
  
% Fill in symmetric values.  
for sample = 1:n_samples  
    adjacency_tensor(:, :, sample) = adjacency_tensor(:, :, sample)' + adjacency_tensor(:, :, sample);  
end  
  
toc
```

Elapsed time is 27.300280 seconds.

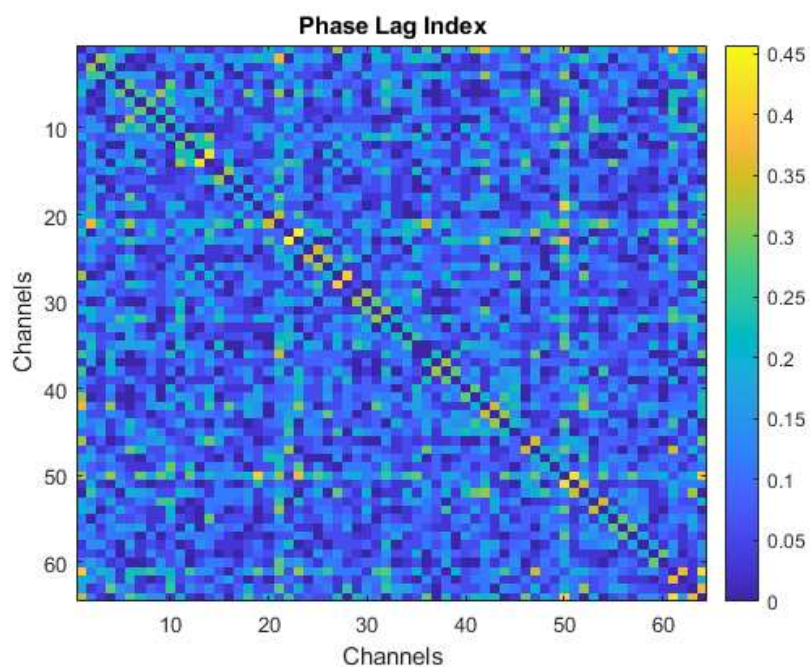
The variable 'adjacency_tensor' is a 3 dimensional array with PLI adjacency matrices over time [Channel x Channel x Sample].

```
fprintf('\nDimensionality of adjacency tensor:')  
disp(size(adjacency_tensor))
```

Dimensionality of adjacency tensor: 64 64 6000

Let's plot the channel-by-channel adjacency matrix at a single time-point.

```
figure;  
imagesc(adjacency_tensor(:,:,1000))  
colorbar  
xlabel('Channels')  
ylabel('Channels')  
title('Phase Lag Index')
```



Now let's see how to implement the same analyses using the function 'across_trial_pli.m'

```
[adjacency_tensor, xs, filt_params] = across_trial_pli(xs.raw,samp_rate,filt_band);  
  
% Plot.  
figure;  
imagesc(adjacency_tensor(:,:,1000))  
colorbar  
xlabel('Channels')  
ylabel('Channels')  
title('Phase Lag Index')
```

Calculating across-trial phase-lag index...
Elapsed time is 57.305452 seconds.

