



Timelockanalysis

- Command: [ft_definetrial](#), [ft_preprocessing](#), [ft_timelockanalysis](#),
[ft_megplanar](#), [ft_combineplanar](#), [ft_singleplotER](#), [ft_multiplotER](#),
[ft_topoplotER](#)

Task:

- Follow the script in timelockanalysis
- <http://www.fieldtriptoolbox.org/tutorial/eventrelatedaveraging>
- You might add the step of reject visual
- Calculate planar gradiometers
- Plot the results
- Compare topography of planar and axial gradiometers



Timelockanalysis

- Command: [ft_redefinetrial](#), [ft_preprocessing](#), [ft_timelockanalysis](#),
[ft_megplanar](#), [ft_combineplanar](#), [ft_singleplotER](#), [ft_multiplotER](#),
[ft_topoplotER](#)

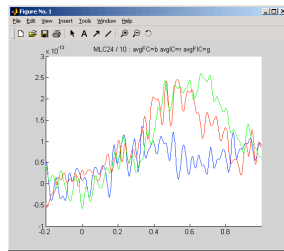
Task:

- Do the timelockanalysis for dataset 04_Inhibition_20150507_14.ds for d2 and d3 single stimulation, with maximal intensity.
- Use artifact rejection.
- Use both planar and axial gradiometers.

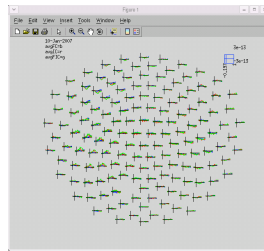


Plotting

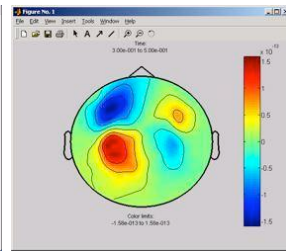
- Command: `ft_singleplotER`, `ft_multiplotER`, `ft_topoplotER`, `ft_singleplotTFR`, `ft_multiplotTFR`, `ft_topoplotTFR`, `ft_clusterplot`



`ft_singleplotER`



`ft_multiplotER`

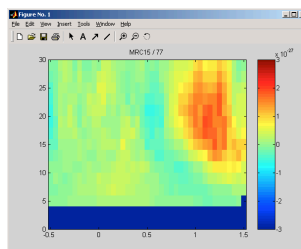


`ft_topoplotER`

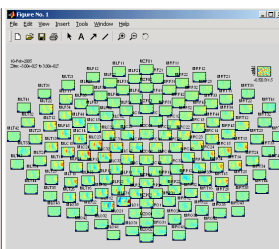


Plotting

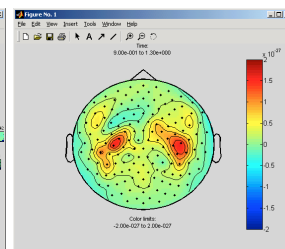
- Command: `ft_singleplotER`, `ft_multiplotER`, `ft_topoplotER`, `ft_singleplotTFR`, `ft_multiplotTFR`, `ft_topoplotTFR`, `ft_clusterplot`



`ft_singleplotTFR`



`ft_multiplotTFR`

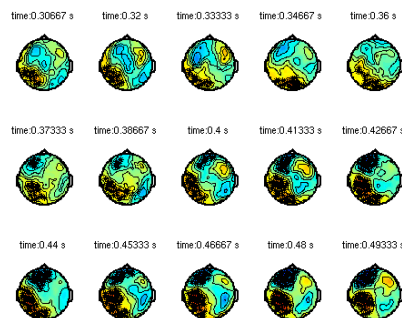


`ft_topoplotTFR`



Plotting

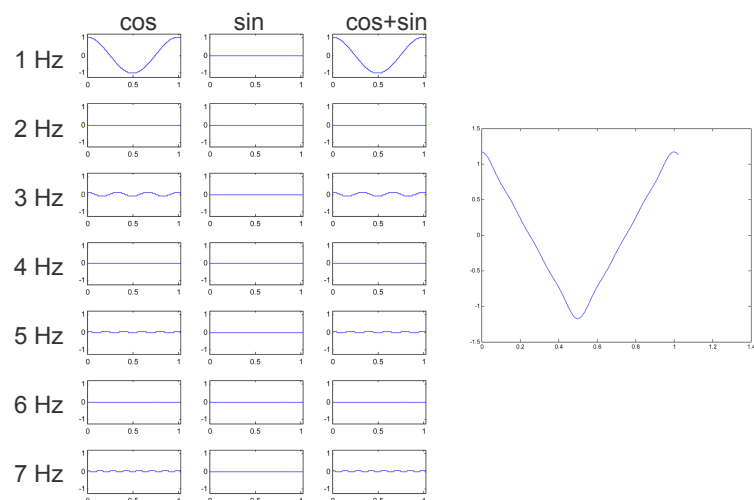
- Command: `ft_singleplotER, ft_multiplotER, ft_topoplotER, ft_singleplotTFR, ft_multiplotTFR, ft_topoplotTFR, ft_clusterplot`



`ft_clusterplot`



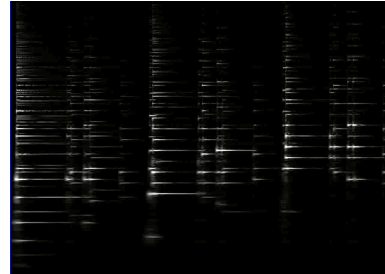
Spectral Analysis



Spectral Analysis

Music is composed of different frequencies.

Using filters and spectral analysis (Fourier transform) the contributing components can be identified.



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SONATE

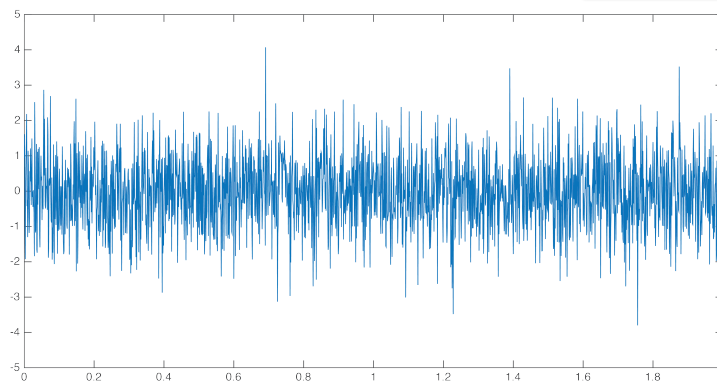
(Palmstige) Op. 13.

Dem Fürsten Carl von Lichnowsky gewidmet.



Spectrum

```
d=randn(2000,1);
t=[0:size(d,1)-1]*0.001;
plot(t,d,'FontSize',16);
ylim([-5 5])
```

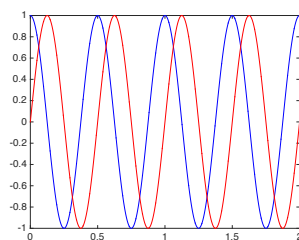




Spectrum

Fourier decomposition any signal can be decomposed in a series of cosine and sine waves with different weights (fourier coefficients)

$$d(t) = A_0 + \sum_n A_n \cos(2\pi nft) + \sum_n B_n \sin(2\pi nft) \quad \text{with } f = \frac{1}{T}$$



```
figure
plot(t,cos(2*pi*2*t),'b')
hold on
plot(t,sin(2*pi*2*t),'r')
ax=gca
ax.FontSize=16;
```



Spectrum

What will happen if you try out the following series

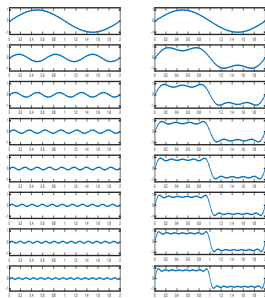
$$A_n = 0 \quad B_n = \begin{cases} \frac{1}{n}, & n \text{ is odd} \\ 0, & n \text{ is even} \end{cases}$$



Spectrum

What will happen if you try out the following series

$$A_n = 0 \quad B_n = \begin{cases} \frac{1}{n}, & n \text{ is odd} \\ 0, & n \text{ is even} \end{cases}$$

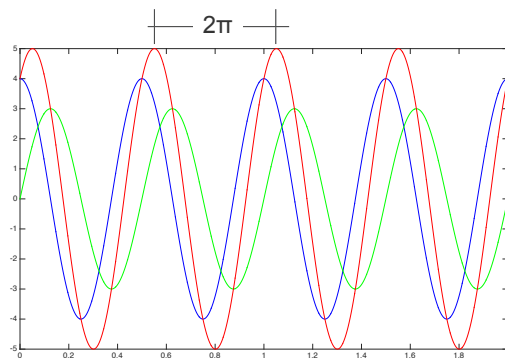


```
x=[0:0.001:2];
f=1/x(end);
y=zeros(size(x));
m=15
for n=1:2:m
    y_tmp=sin(2*pi*n*f*x)/n;
    y=y+y_tmp
    subplot((m+1)/2,2,n)
    plot(x,y_tmp)
    ylim([-1.2,1.2]);
    subplot((m+1)/2,2,n+1)
    plot(x,y)
    ylim([-1.2,1.2]);
end
```



Spectrum: Complex representations

What will happen if you try out the following series

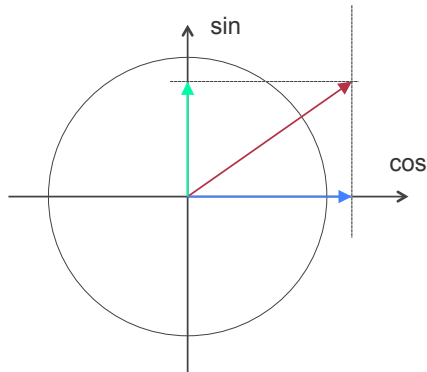


```
t=[0:0.001:2];
a=4*cos(2*pi*2*t)
b=3*sin(2*pi*2*t)
plot(t,a,'b')
hold on
plot(t,b,'g')
plot(t,a+b,'r');
ax=gca;
ax.FontSize=16
```



Spectrum: Complex representations

What will happen if you try out the following series



```
t=[0:0.001:2];
f=1/x(end);
y=zeros(size(x));
m=15
for n=1:2:m
    y_tmp=sin(2*pi*n*f*x)/n;
    y=y+y_tmp
    subplot((m+1)/2,2,n)
    plot(x,y_tmp)
    ylim([-1.2,1.2]);
    subplot((m+1)/2,2,n+1)
    plot(x,y)
    ylim([-1.2,1.2]);
end
```



Time-Frequency Analysis

[Video](#)



Time-Frequency Analysis

ft_freqanalysis

ft_freqanalysis performs frequency and time-frequency analysis
on time series data over multiple trials

cfg.method = different methods of calculating the spectra

- **'mtmfft'**, analyses an entire spectrum for the entire data length, implements multitaper frequency transformation
- **'mtmconvol'**, implements multitaper time-frequency transformation based on multiplication in the frequency domain.
- **'wavelet'**, implements wavelet time frequency transformation (using Morlet wavelets) based on multiplication in the frequency domain.
- **'tfr'**, implements wavelet time frequency transformation (using Morlet wavelets) based on convolution in the time domain.
- **'mvar'**, does a fourier transform on the coefficients of an estimated multivariate autoregressive model, obtained with FT_MVARANALYSIS. In this case, the output will contain a spectral transfer matrix, the cross-spectral density matrix, and the covariance matrix of the innovation noise.



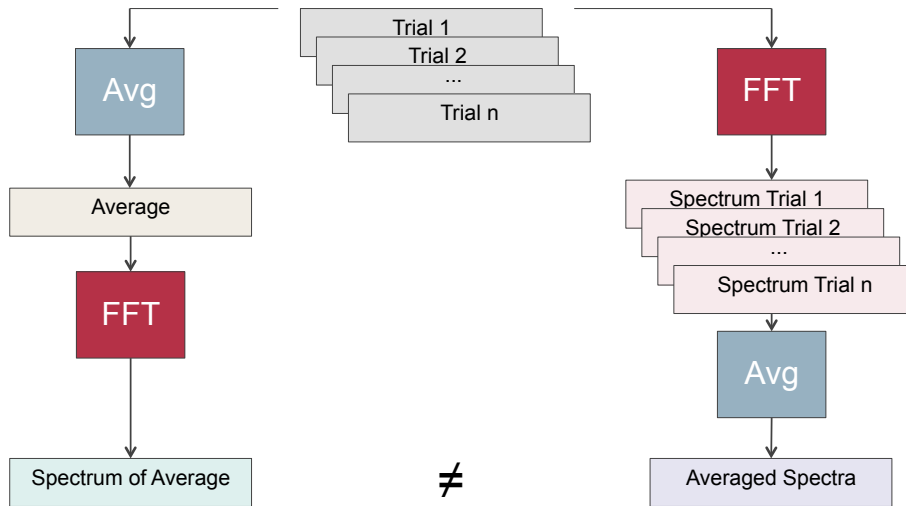
Time-Frequency Analysis

- Command: [ft_freqanalysis](#)

Task:

- Tutorial: Time-frequency analysis
(<http://www.fieldtriptoolbox.org/tutorial/timefrequencyanalysis>)
- Use dataset Subject01
- Try different methods and compare the results

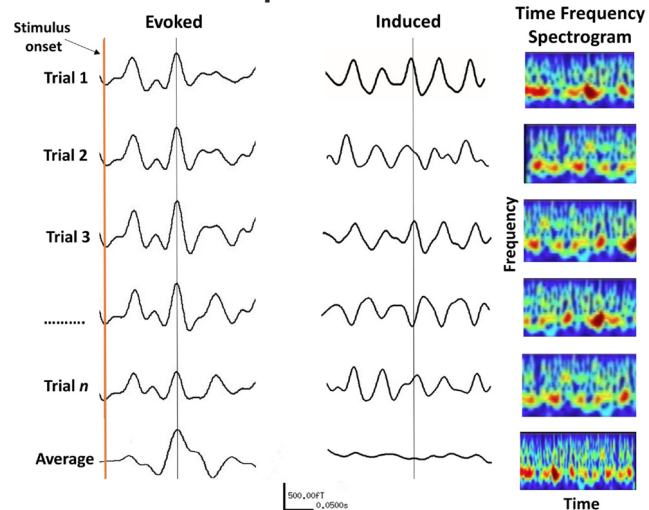
Evoked and induced responses



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Evoked and induced responses



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Spectrum: Complex representations

If cosine and sine wave are combined the complex representation gives an easy interpretation:

$$\begin{aligned}\alpha + i\beta &= ae^{i\varphi} \\ &= \sqrt{\alpha^2 + \beta^2} e^{i\varphi}\end{aligned}$$

$$\varphi = \arctan\left(\frac{\beta}{\alpha}\right)$$

It holds that: $e^{i\varphi} = \cos(\varphi) + i\sin(\varphi)$
