



EEG/MEG 3:

Time-Frequency Analysis Olaf Hauk

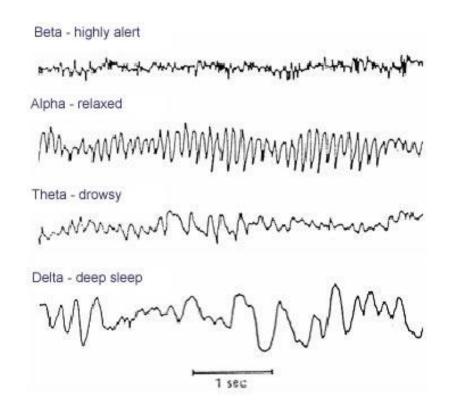
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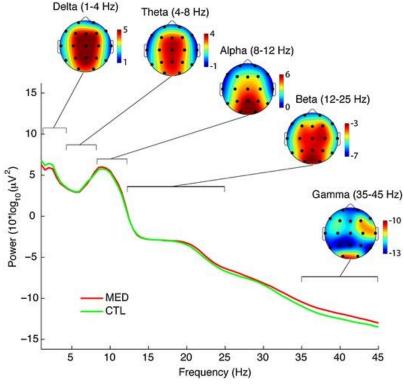
"Brain Rhythms" and "Oscillations"

Time course and topography may differ among different frequency bands

(and may depend on task, environment, subject group etc.)

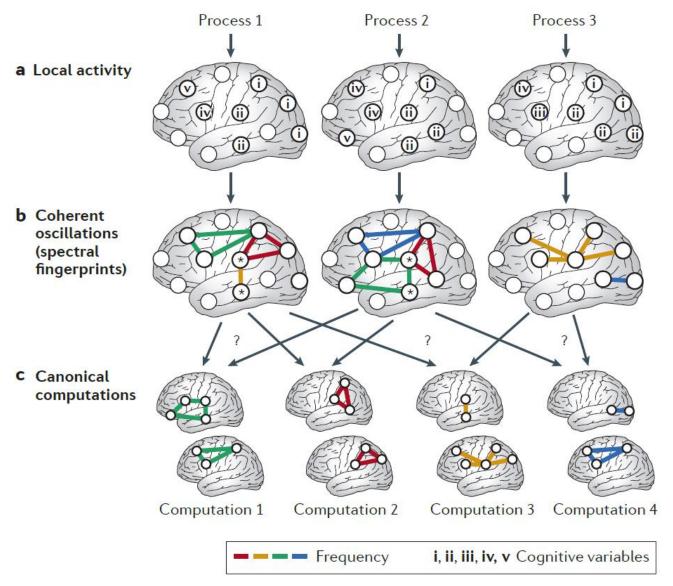
=> Different frequency "bands" may reflect different processes/computations, systems/networks, etc.





Cahn et al., Cogn Proc 2010, http://link.springer.com/article/10.1007%2Fs10339-009-0352-1/

"Brain Rhythms" and "Oscillations"



Polar Representation Of Periodic Signals

Euler's Formula

"Complex" numbers can capture the two axes of the coordinate system for the circle around which the vector rotates periodically – this is rather abstract but helps the notation enormously.

$$e^{-i\theta} = \cos(\theta) + i * \sin(\theta)$$
 $i=\sqrt{-1}$
Therefore:
 $\cos(\theta) = real(e^{-i\theta})$
 $\sin(\theta) = imag(e^{-i\theta})$

An oscillation at a particular frequency can be described in a "polar representation":

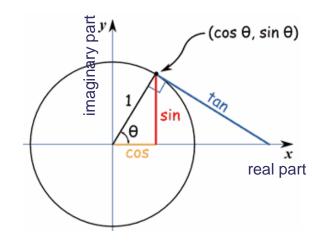
$$a * e^{-i2\pi ft}$$

a: amplitude

 2π : circumference of unit circle

f: frequency

t: time



The Polar Representation Of Periodic Signals

Convenient To Compare Periodic Signals

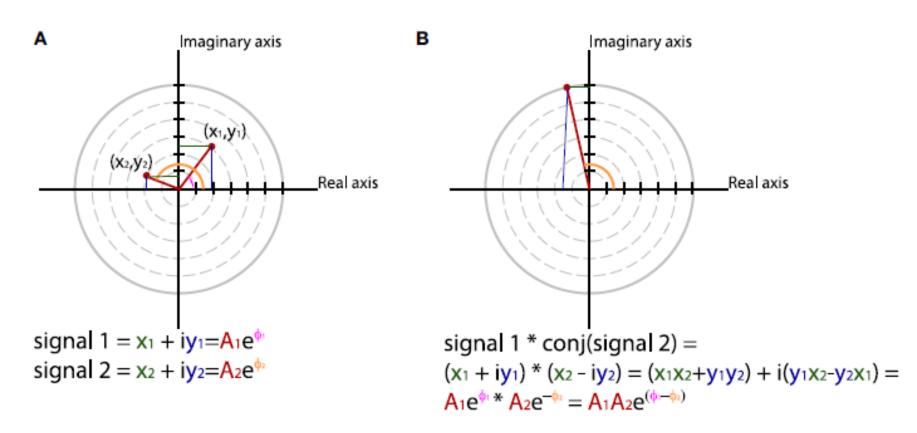
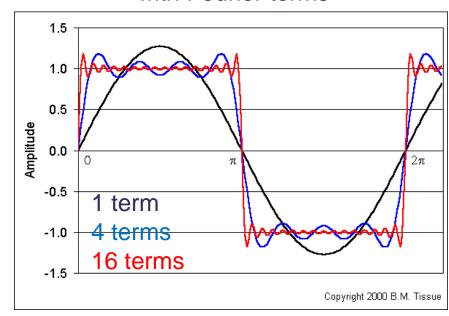


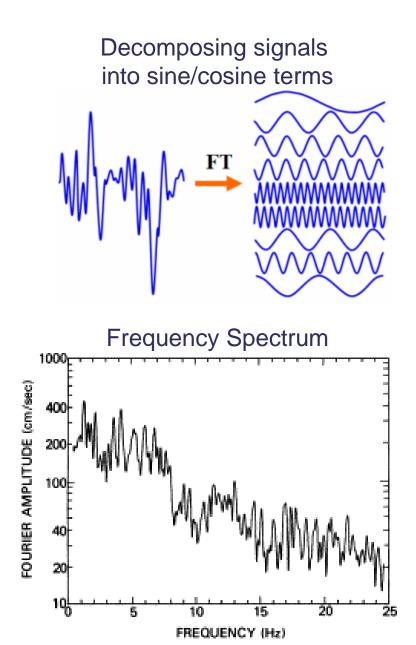
FIGURE 2 | Using polar coordinates and complex numbers to represent signals in the frequency domain. (A) The phase and amplitude of two signals. (B) The cross-spectrum between signal 1 and 2, which corresponds to multiplying the amplitudes of the two signals and subtracting their phases.

The Fourier Decomposition

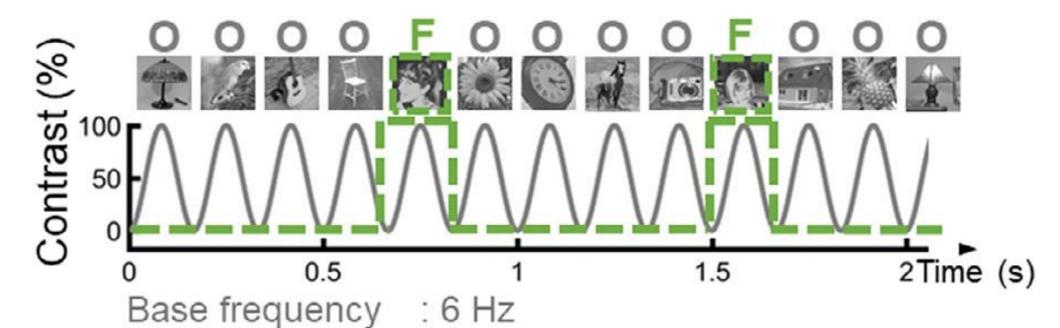
The Fourier (De-)Composition





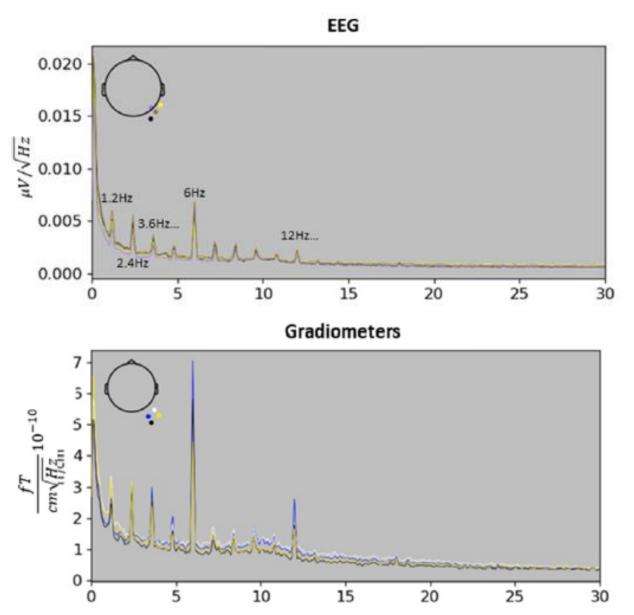


Example: Fast Periodic Visual Stimulation (FPVS)



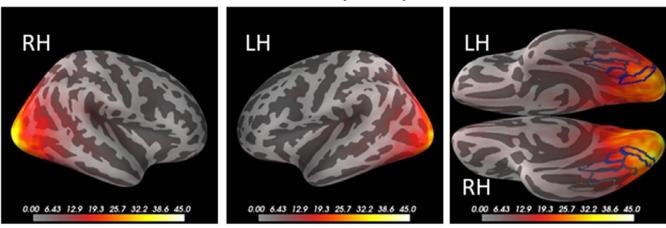
Oddball frequency: 6 Hz/5 = 1.2 Hz

Fast Periodic Visual Stimulation (FPVS)

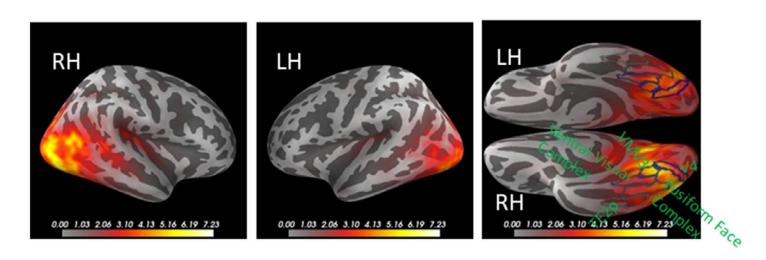


Fast Periodic Visual Stimulation (FPVS)

Base Frequency



Face-selective Frequency

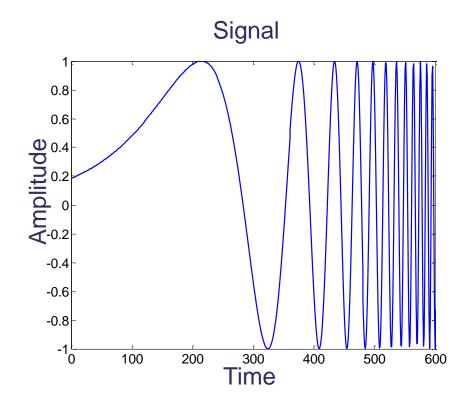


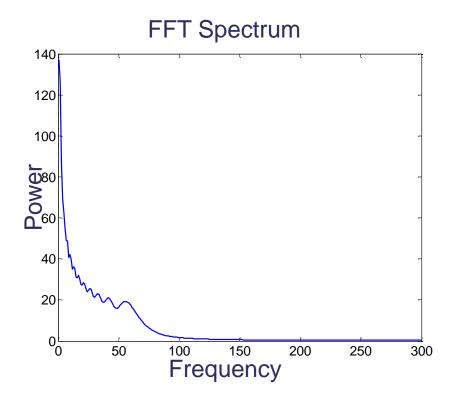
Time-Frequency Analysis

Motivation for Time-Frequency Analysis

Fourier Transform assumes sines and cosines with constant amplitudes across the whole time series ("stationarity").

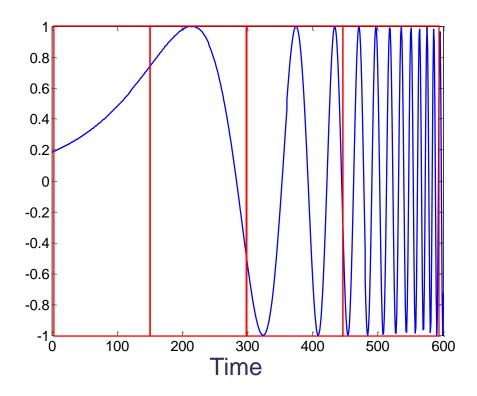
What are we going to do with signals where activity in frequency bands changes over time, e.g. with a signal like this?





Motivation for Time-Frequency Analysis

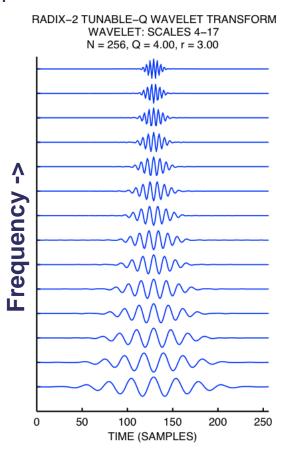
You could run separate FFTs for different (sliding) time windows:



But different window sizes are more or less optimal for different frequencies. Run different FFTs with different window sizes for different frequency ranges? Ouff.

Time-Frequency Analysis: Wavelets ("little waves")

Wavelets provide an optimal trade-off between frequency and time resolution.



Wavelets are getting "broader" with decreasing frequency

=>

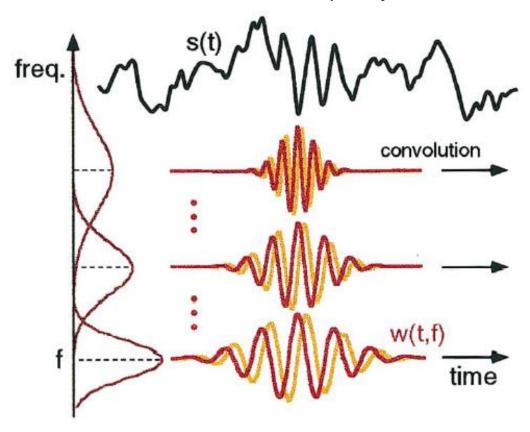
Time resolution decreases as frequency decreases

Wavelets are convolved with the data to give instantaneous amplitude and phase estimates for different frequency ranges.

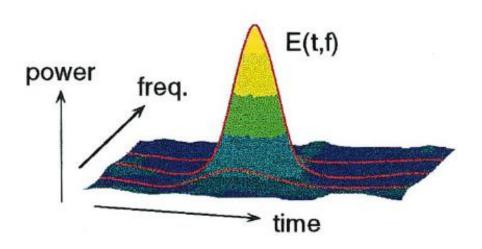
Time-Frequency Analysis: Wavelets

Wavelet Transform

Trade-off between time and frequency resolution



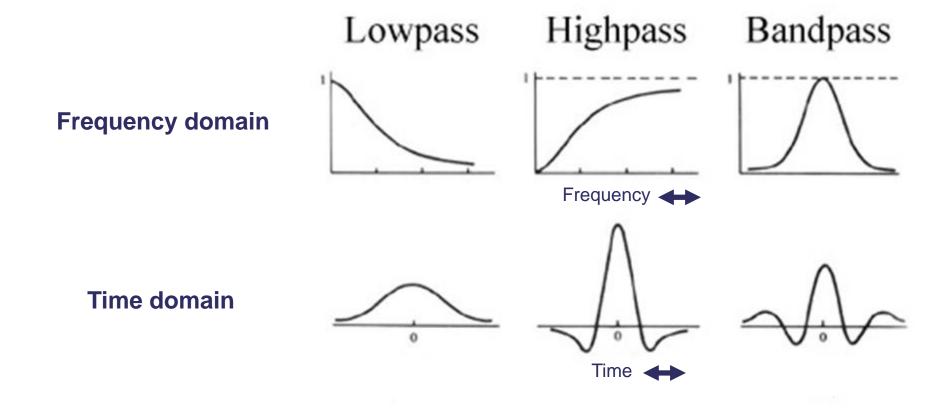
Time-Frequency Power



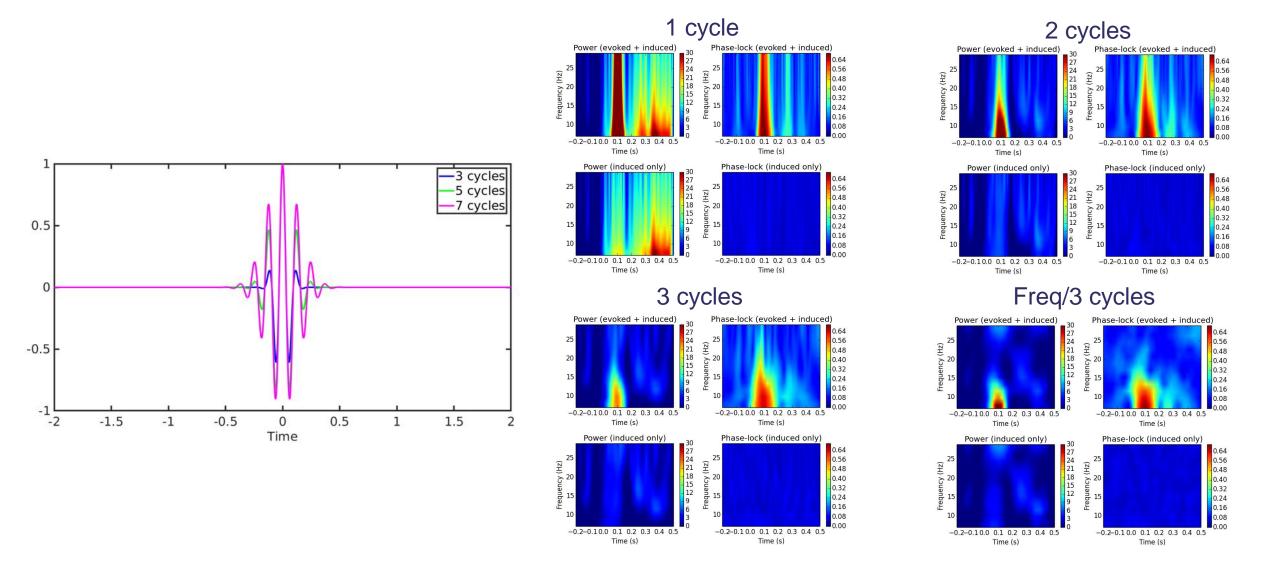
Basic Principals of Frequency Filtering

Time-domain and frequency-domain filtering are two sides of the same coin:

One type of frequency-domain filtering corresponds to one type of time-domain filtering.

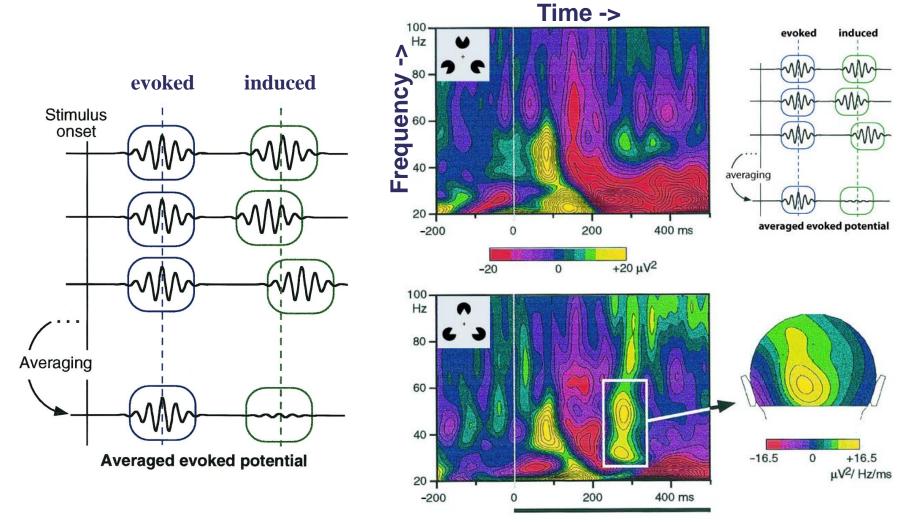


Effect of Number of Cycles

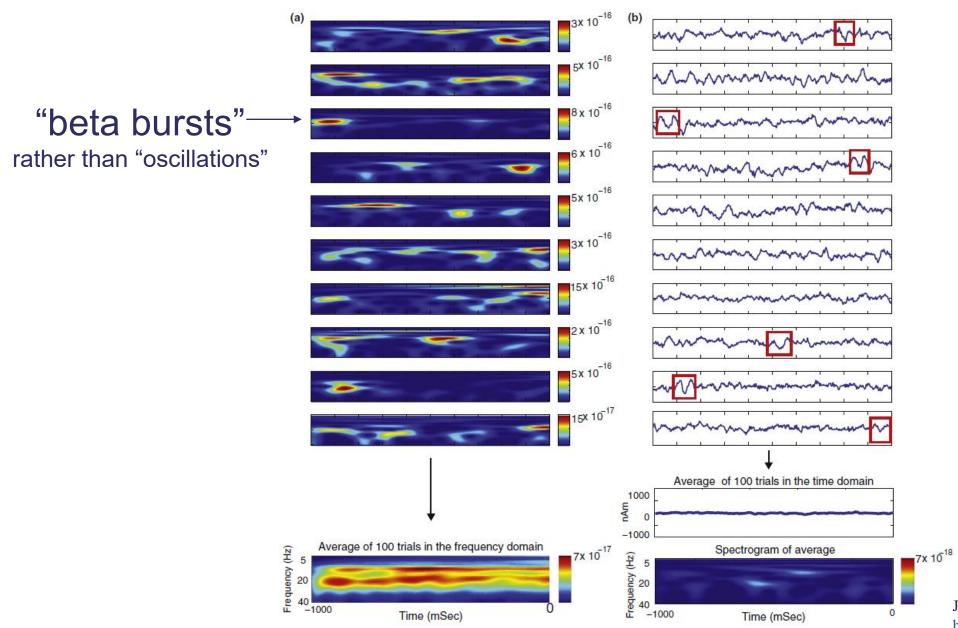


Rule of thumb: For low frequencies (<~10Hz), n=2 or 3; for higher frequencies n=f/3.

Evoked and Induced Rhythmic Activity

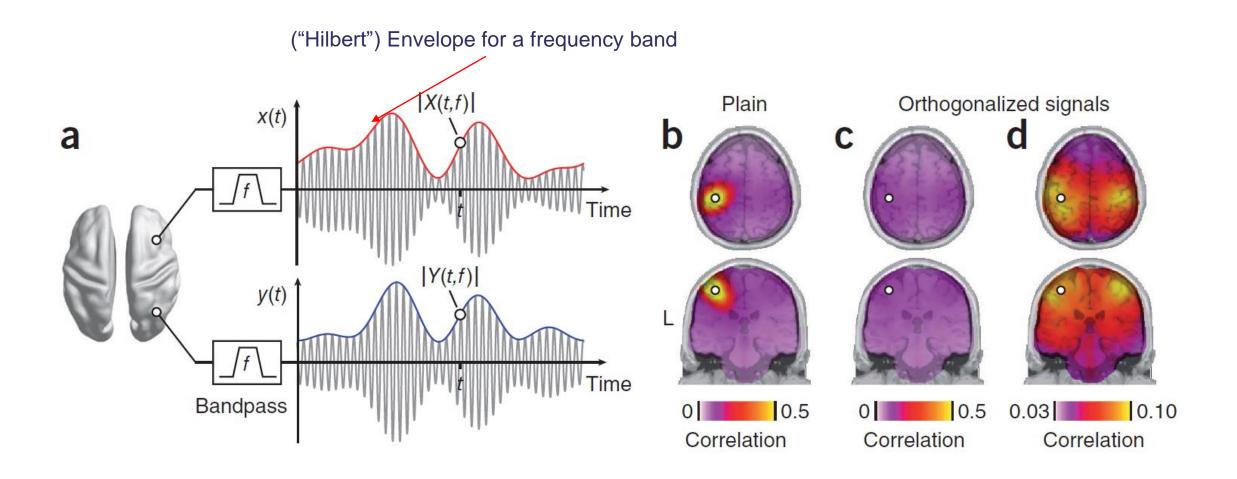


When brain rhythms aren't "rhythmic" – the example of beta "oscillations"



Jones et al., Curr Op Neurobiol 2016 https://pubmed.ncbi.nlm.nih.gov/27400290/

Alternative to wavelets: Hilbert Transform







Thank you

