

# Question of the Day

Can you have a container without boundaries?

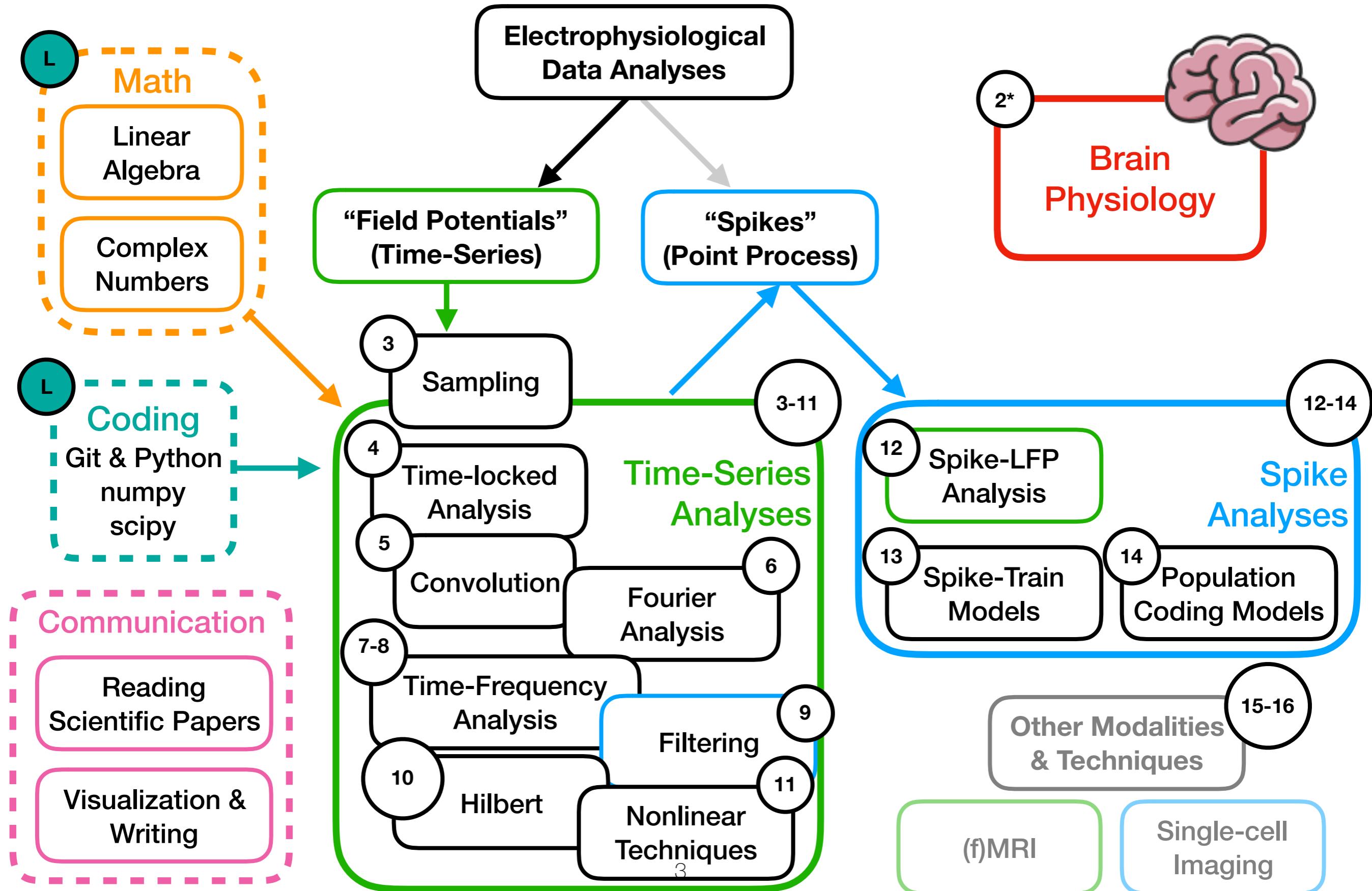


## Time-Frequency Analysis & Windowing

Lecture 8  
July 16, 2019



# Course Outline: Road Map

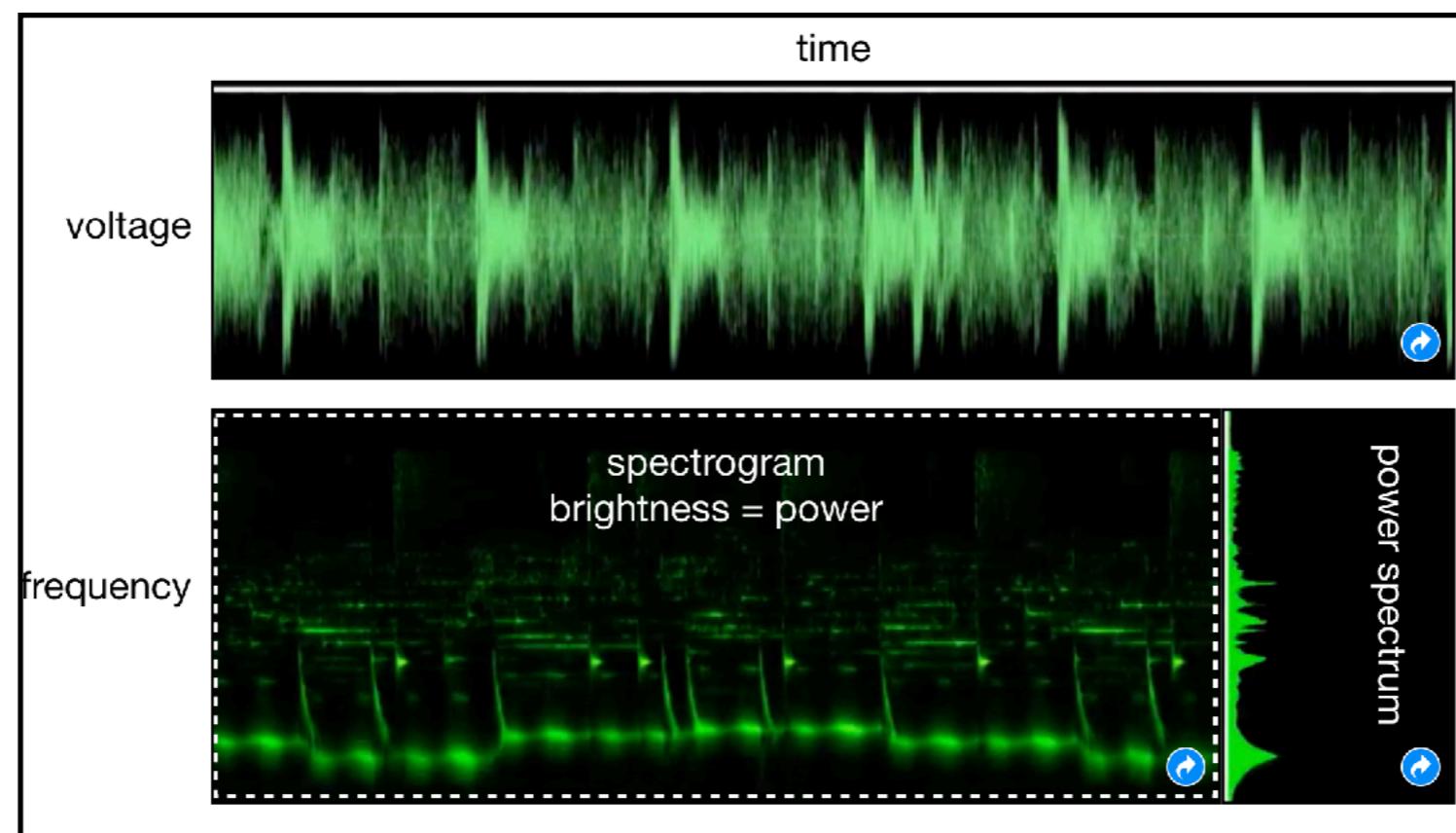


1. Evaluate parameter choices for time-frequency analysis
2. List & understand transformations of STFT
3. Motivate Windowing



# Time-Frequency Analysis

Looking at the DFT or power spectrum is like averaging over all the pictures in a movie, or all the notes in a song.



We need **time-frequency analysis**.

**5 min:** come up with 3 examples of signals with time-dependent frequency content, their dominant frequency(s), and their speed of change.



# Time-Frequency Analysis

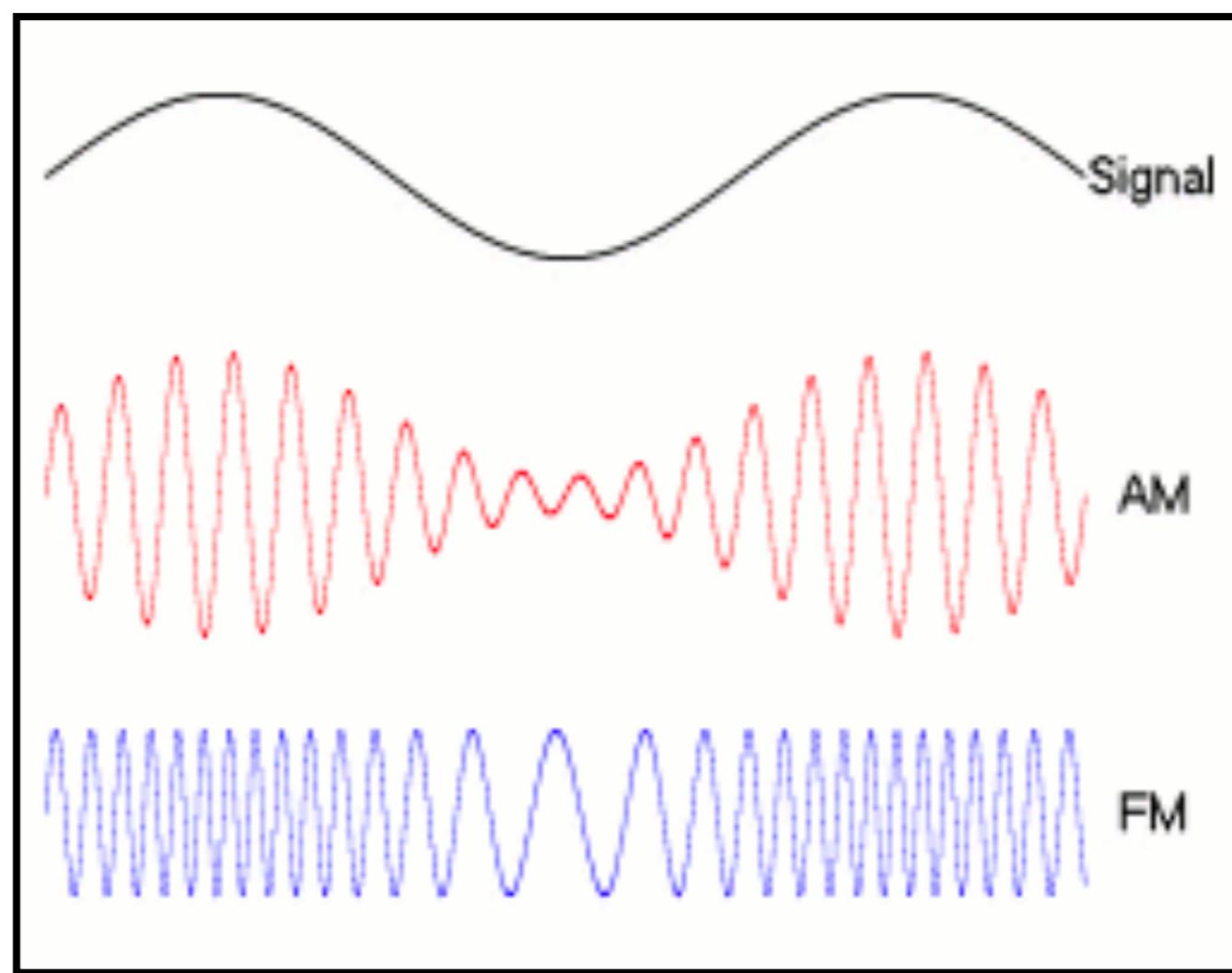
**5 min:** come up with 3 examples of signals with time-dependent frequency content, their dominant frequency(s), and their speed of change.

Example	Frequencies	Timescale of Change	STFT Parameters
<b>seismic waves (earthquakes)</b>	0.2Hz-10Hz 20Hz	minutes	
<b>heart rate</b>	1 - 2Hz	seconds	
<b>color of the sky (from fixed point on earth)</b>	500 tHz	hours	



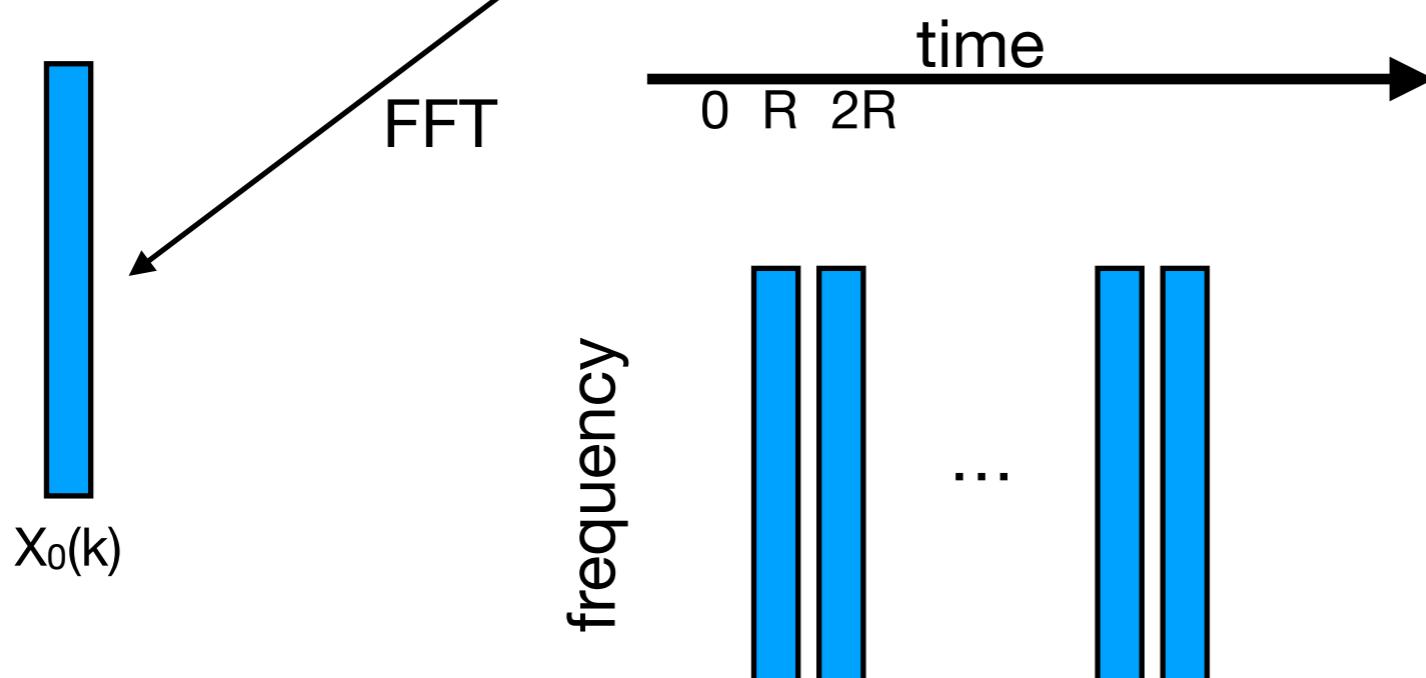
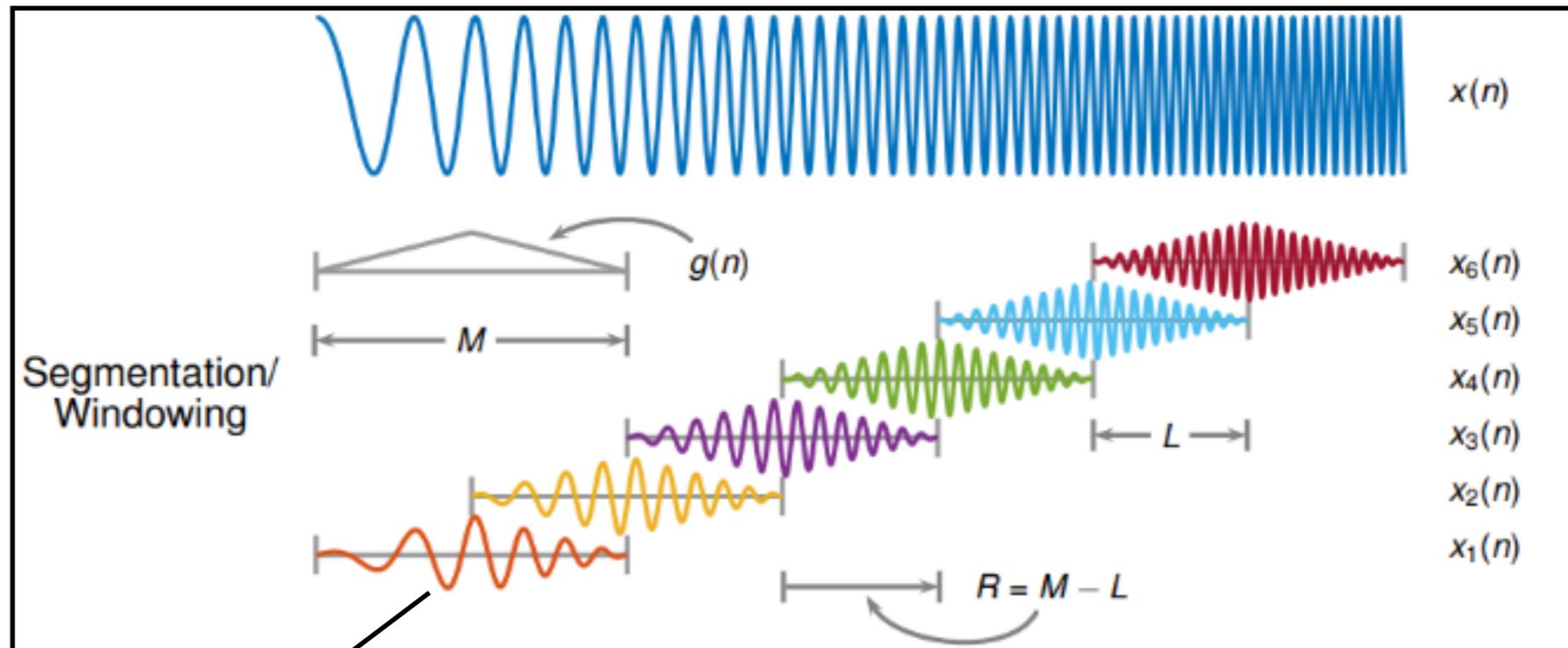
## Frequency modulation

From Wikipedia, the free encyclopedia



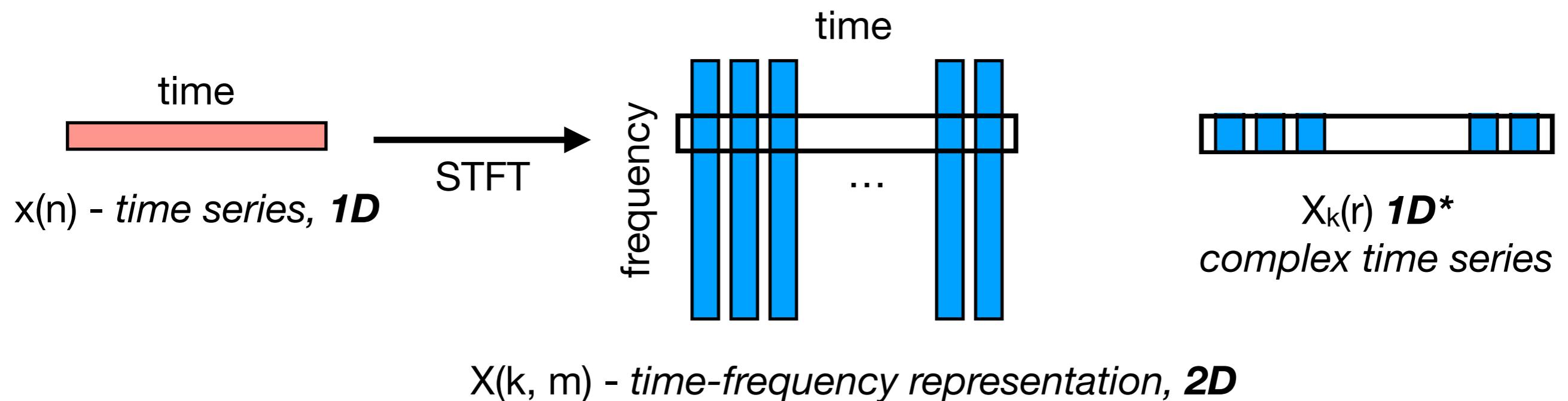
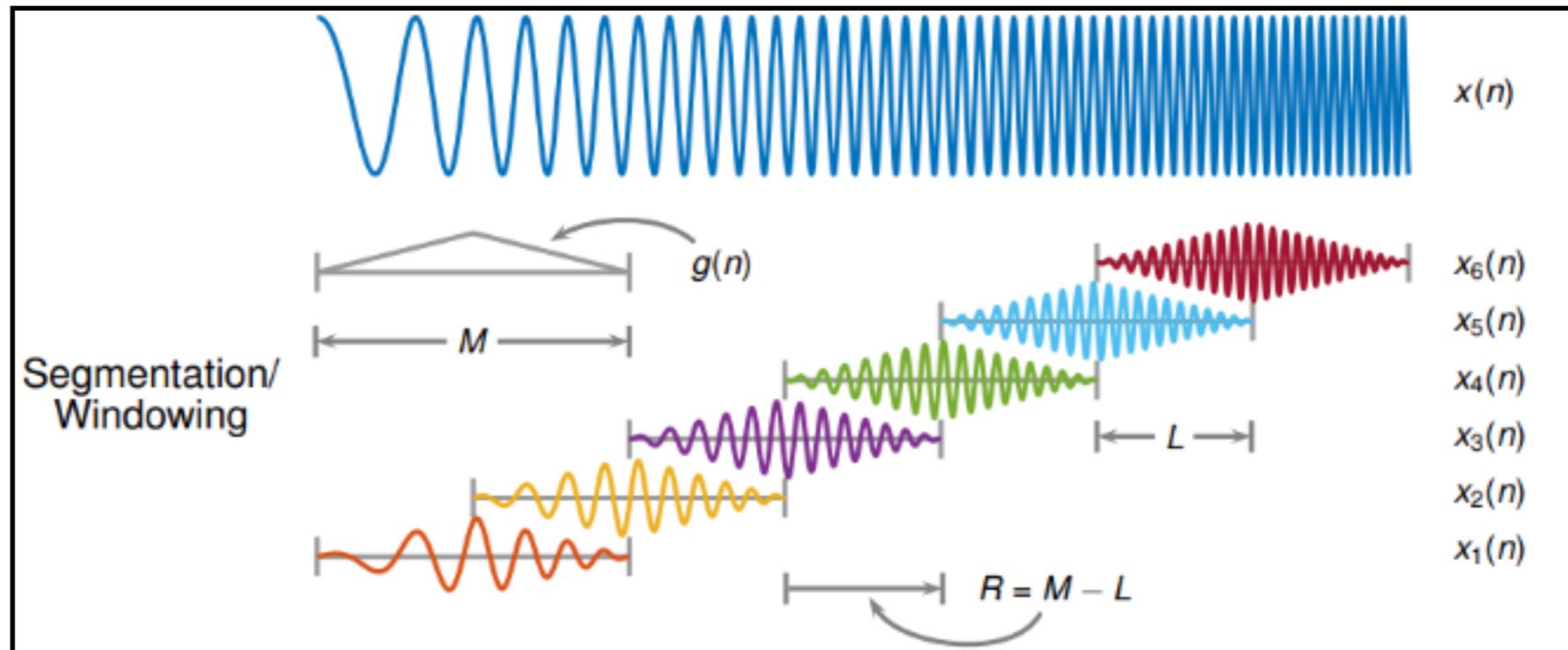
# Short Time Fourier Transform

## Short Time Fourier Transform (STFT)

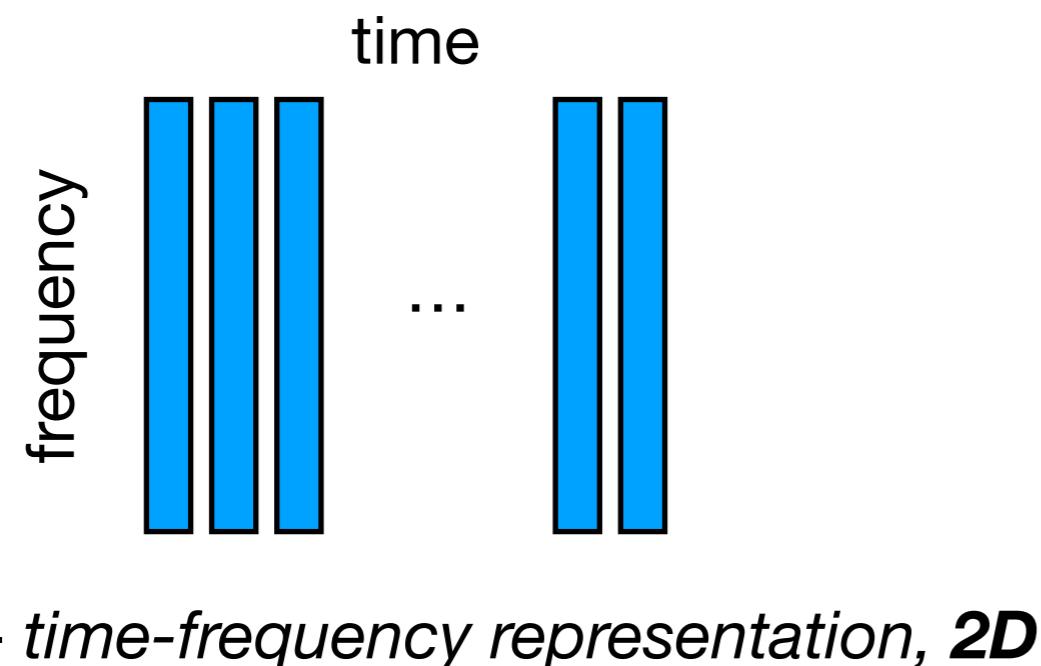


# Short Time Fourier Transform

## Short Time Fourier Transform (STFT)



# Matrix View



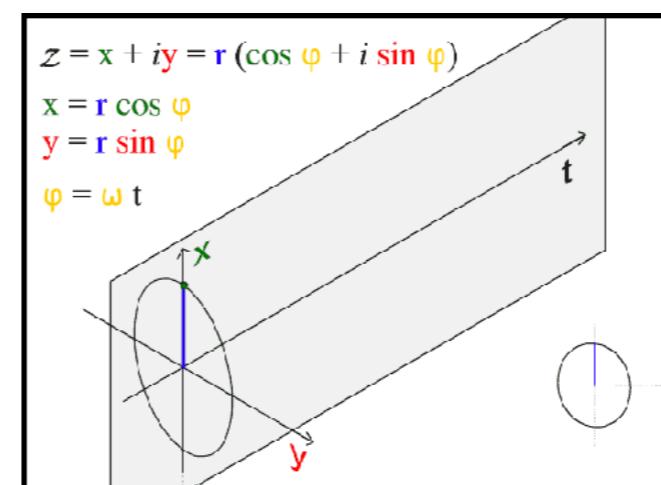
# Matrix of complex numbers over time and frequency.

$$X_{2,1} = X(k=2, r=1) \\ = a + ib$$

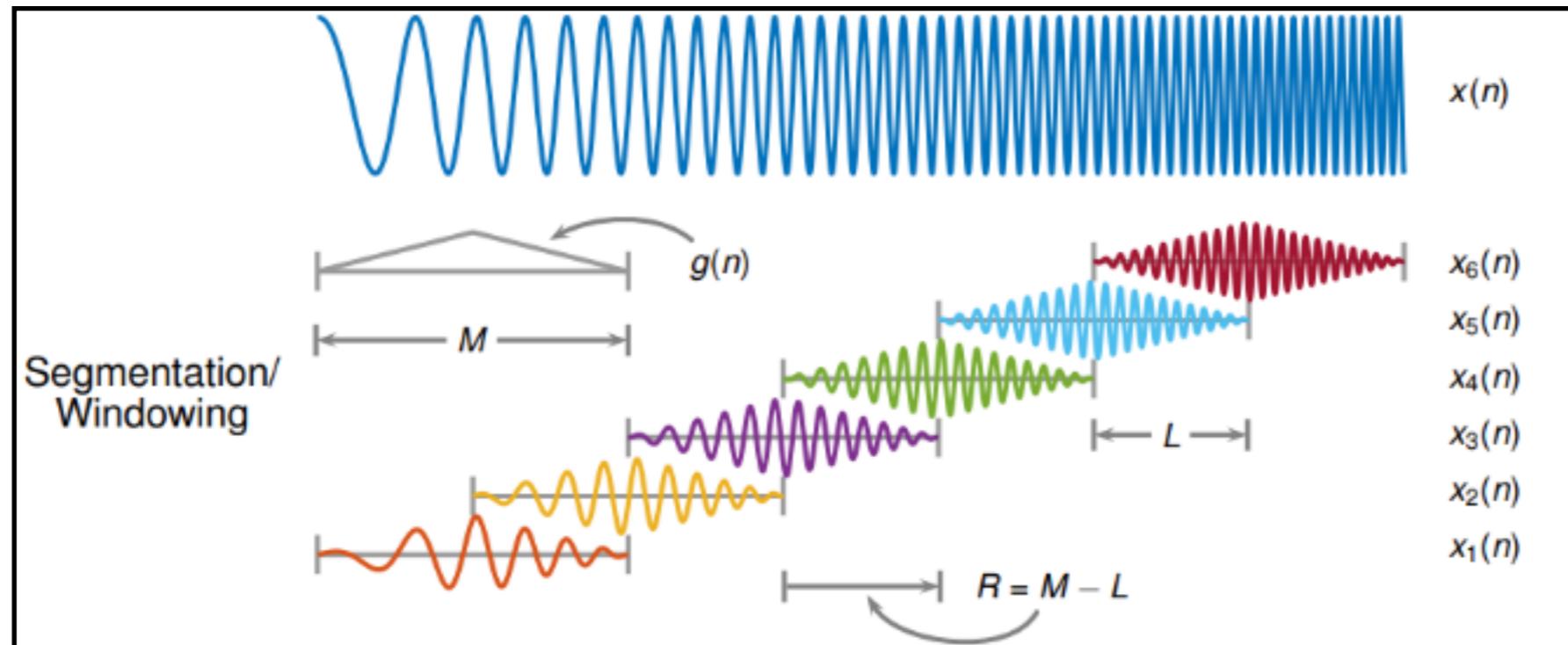


# $X_k(r)$ **1D\***

## *complex time series*



# Parameter Housekeeping



## Givens

$T$  = total signal length (s)

$fs$  = sampling rate (Hz)

$$X_k = \sum_{n=0}^{N-1} x_n \cdot e^{-\frac{i2\pi}{N} kn}$$

## Computable

$dt$  = sampling interval (s)

$N$  = # of samples

$df$  = frequency resolution

$f(k)$  = frequency (Hz)

## STFT Parameters (Choice)

$M$  = window length (s)

$R$  = step length (s)

$L$  = overlap length (s)

$$* M = R + L$$



# Parameter Housekeeping

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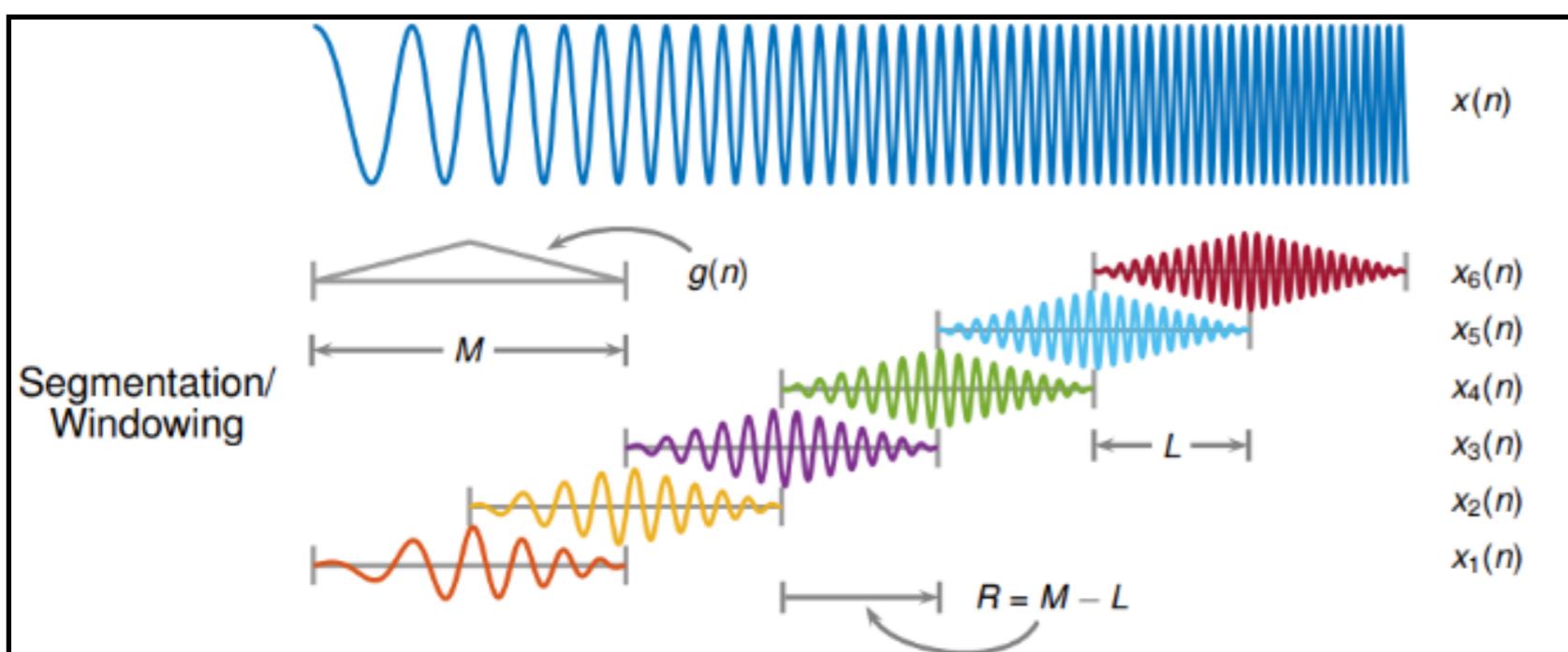
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$$* M = R + L$$



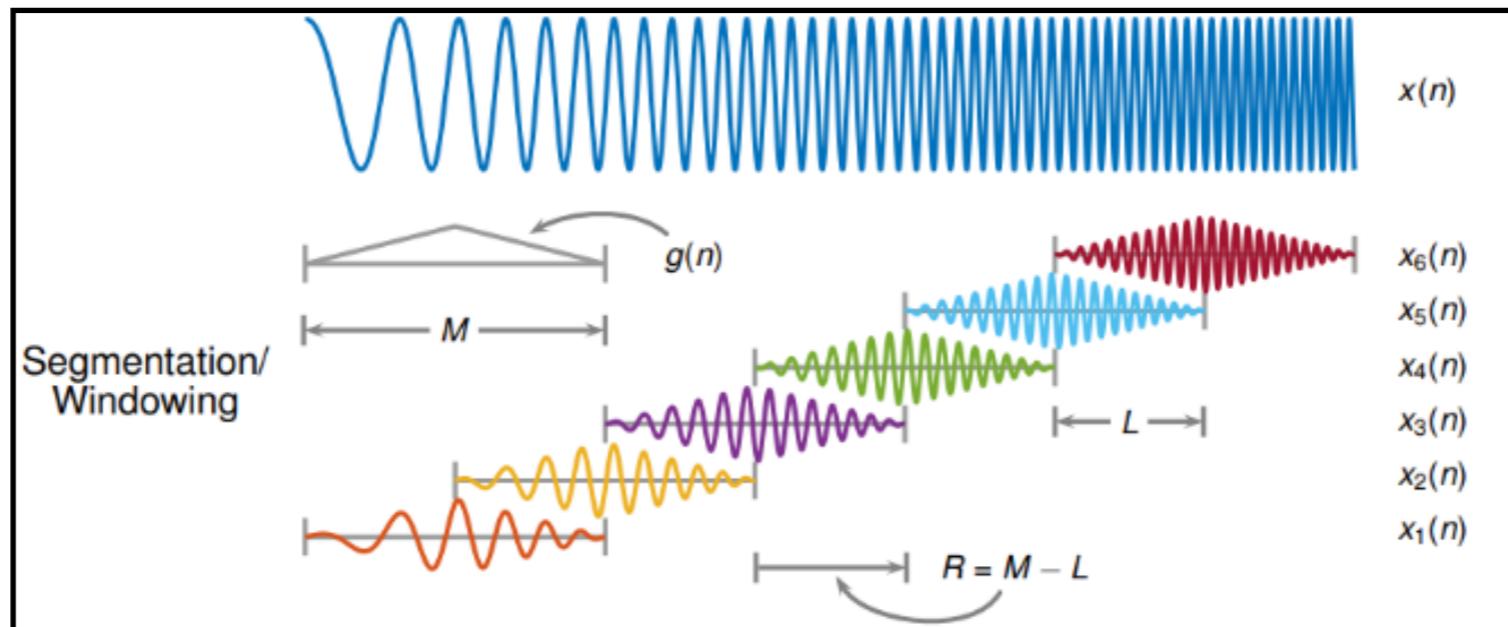
$$df_{FT} = ? \quad 1/T$$

$$df_{STFT} = ? \quad 1/M$$

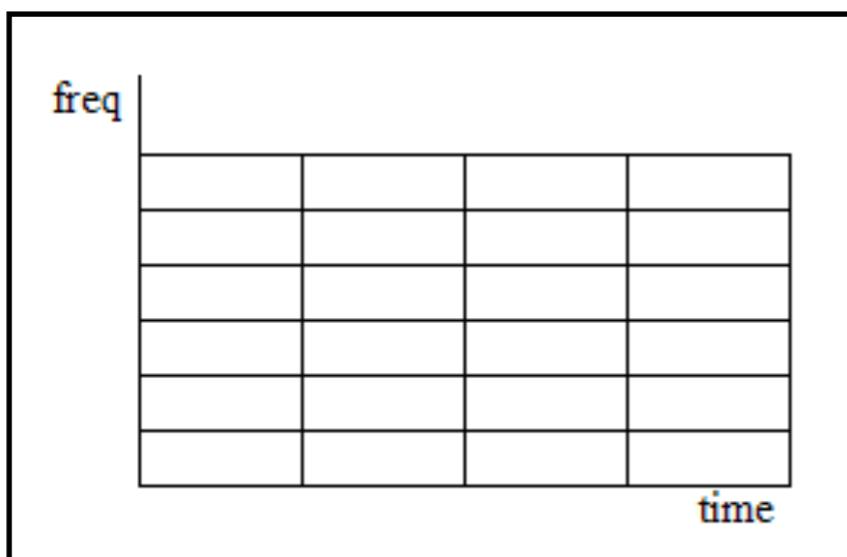


# Time-Frequency Resolution Tradeoff

To get more precise frequency estimates, you lose temporal resolution (where it is in time), and vice versa.

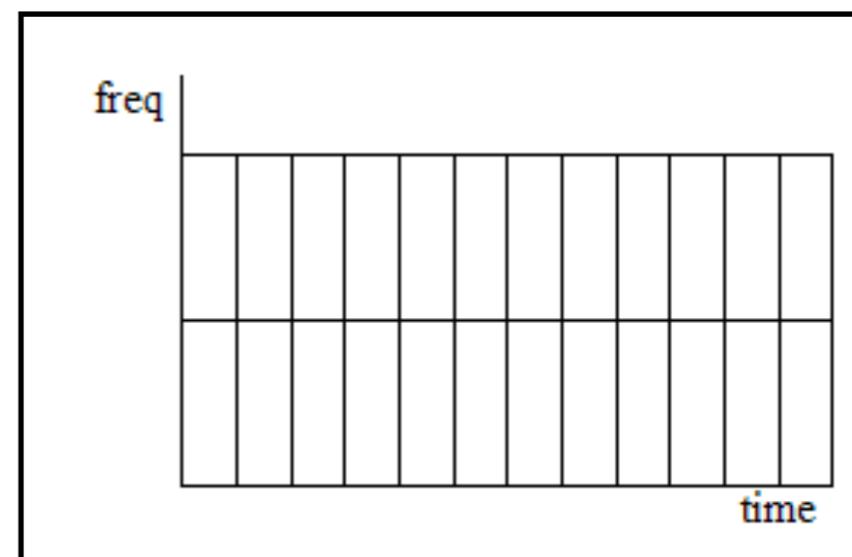


Longer Windows (big M, small df)



$$M \cdot df = 1$$

Shorter Windows (small M, big df)



# Time-Frequency Analysis

**Rule of thumb:** depends on whether you need more certainty in time or frequency.

Example	Frequencies	Timescale of Change	STFT Parameters
<b>seismic waves (earthquakes)</b>	0.2Hz-10Hz 20Hz	minutes	$M = 5s$ or $M=0.1s$
<b>heart rate</b>	1 - 2Hz	seconds	$M=1$ or $0.5s$
<b>color of the sky (from fixed point on earth)</b>	500 tHz	hours	



1. Evaluate parameter choices for time-frequency analysis
2. List & understand transformations of STFT
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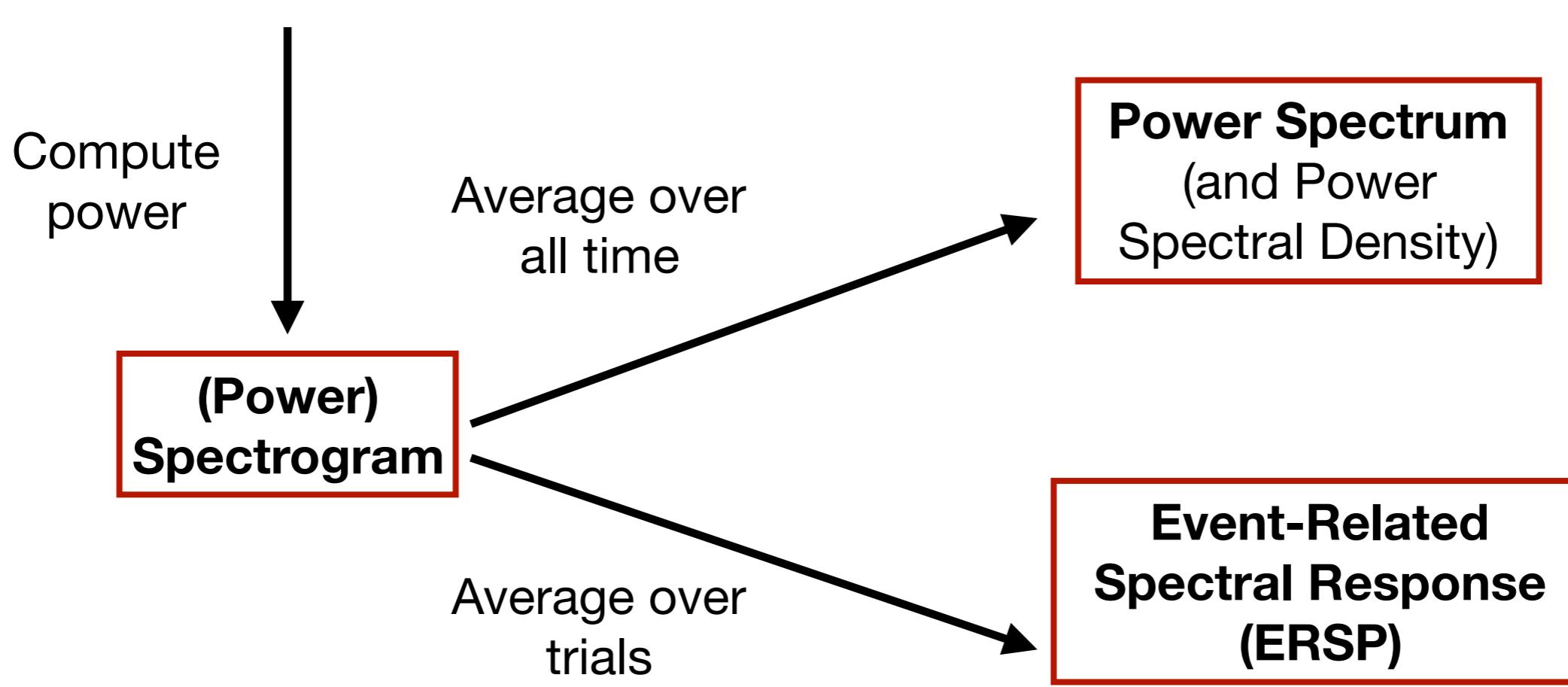


# STFT Transformations: Power



$$X(k, r) = a + ib$$

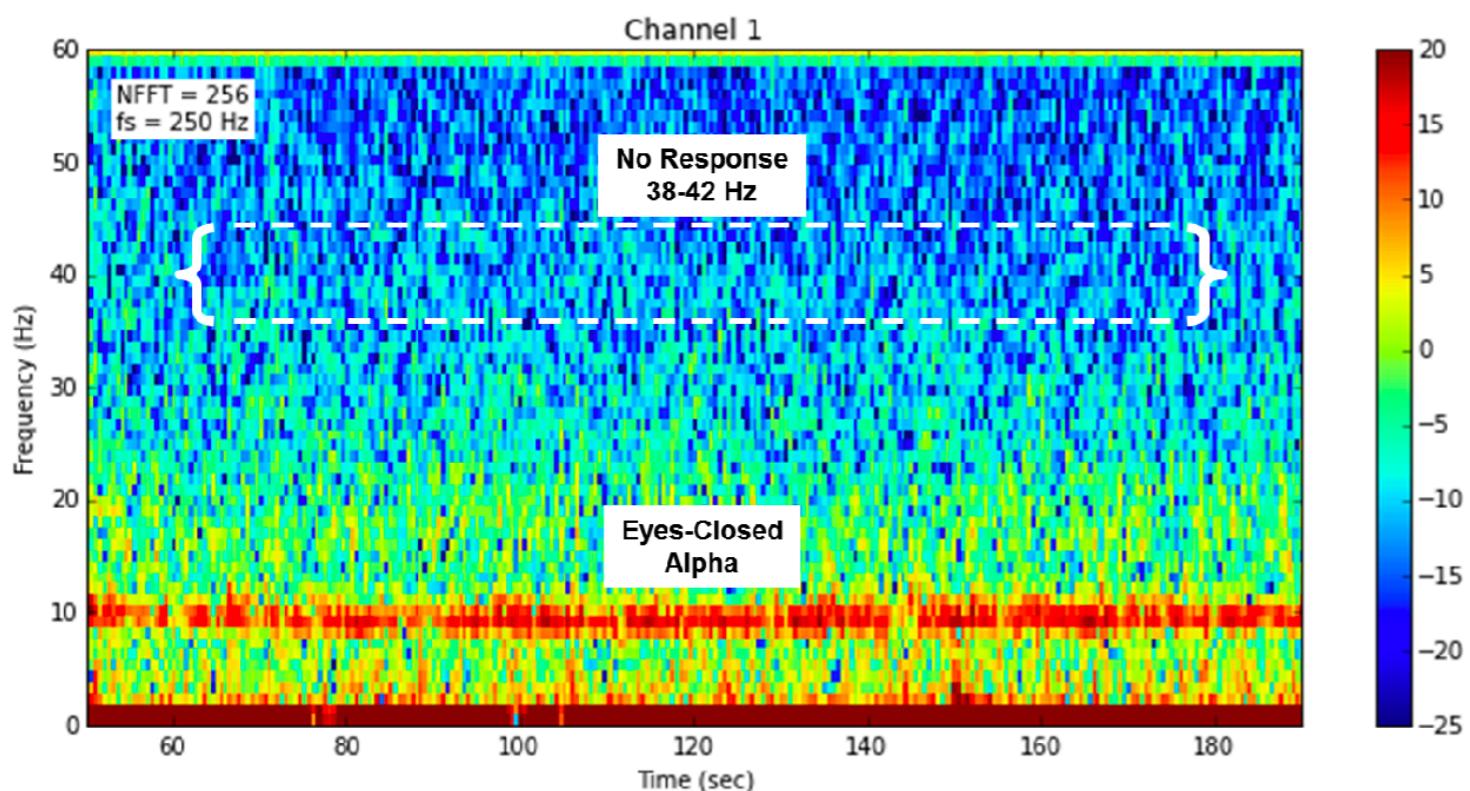
$X(k, r)$  - time-frequency representation, 2D



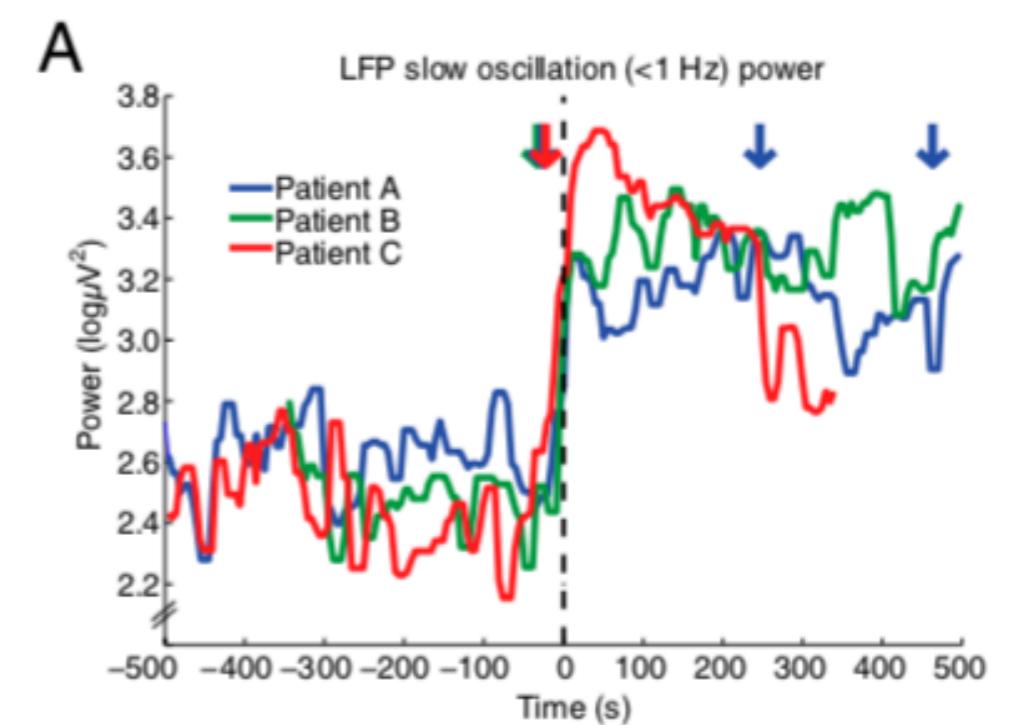
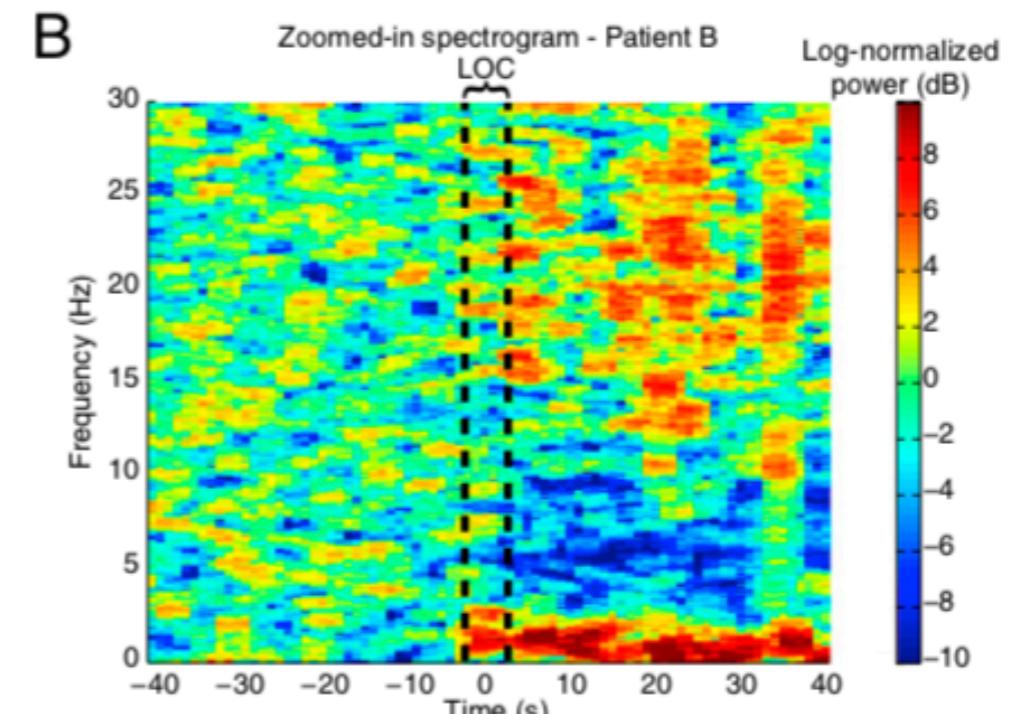
# Spectrogram

$$power(k) = |X_k|^2 = a_k^2 + b_k^2$$

## Alpha Oscillation (10Hz)



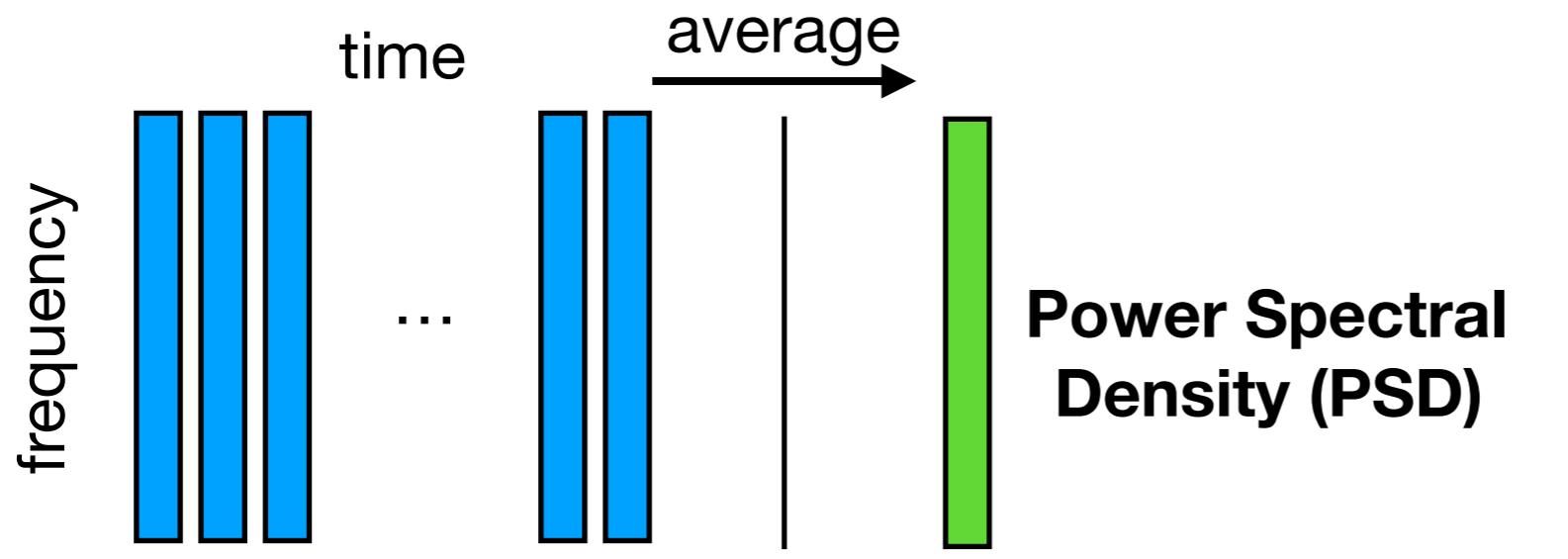
## Sleep Oscillations (0.5Hz)



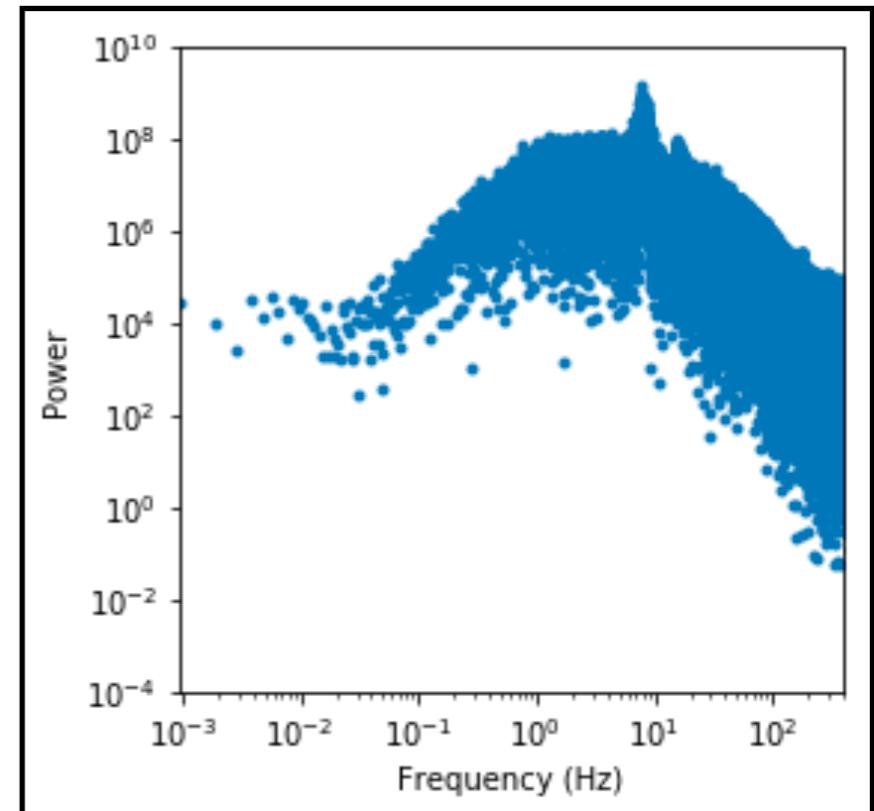
P(t): power over time



# Power Spectrum/Spectral Density

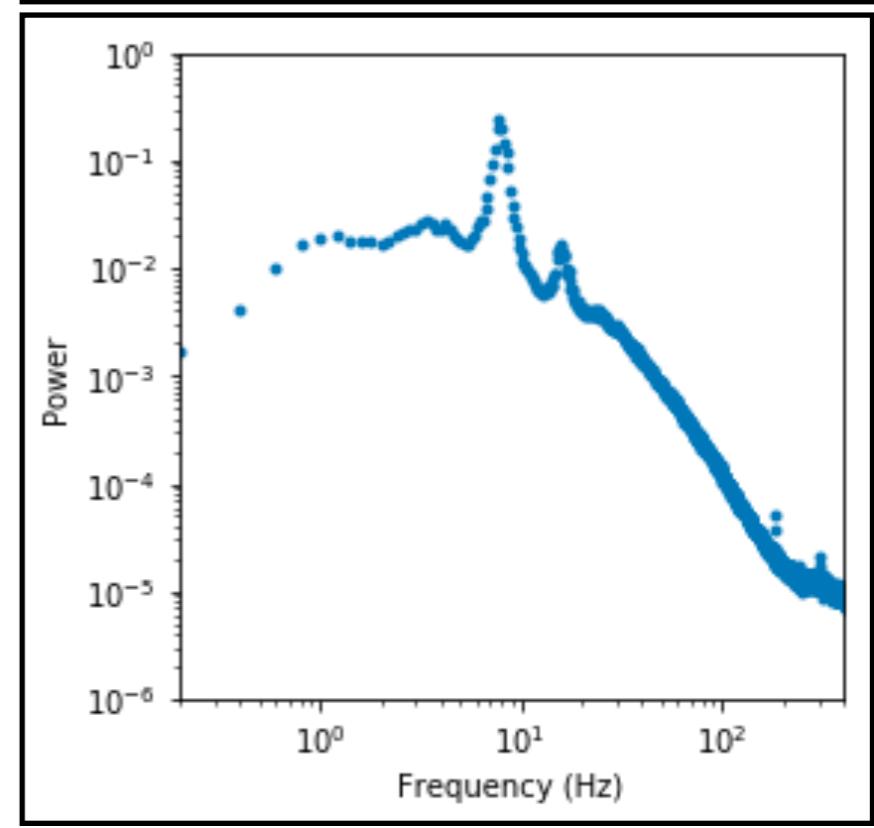


No averaging (directly from DFT):  
**Periodogram**



Averaging (no overlap,  $R = M$ ):  
**Bartlett's Method**

Averaging (half overlap,  $R = M/2$ ):  
**Welch's Method**

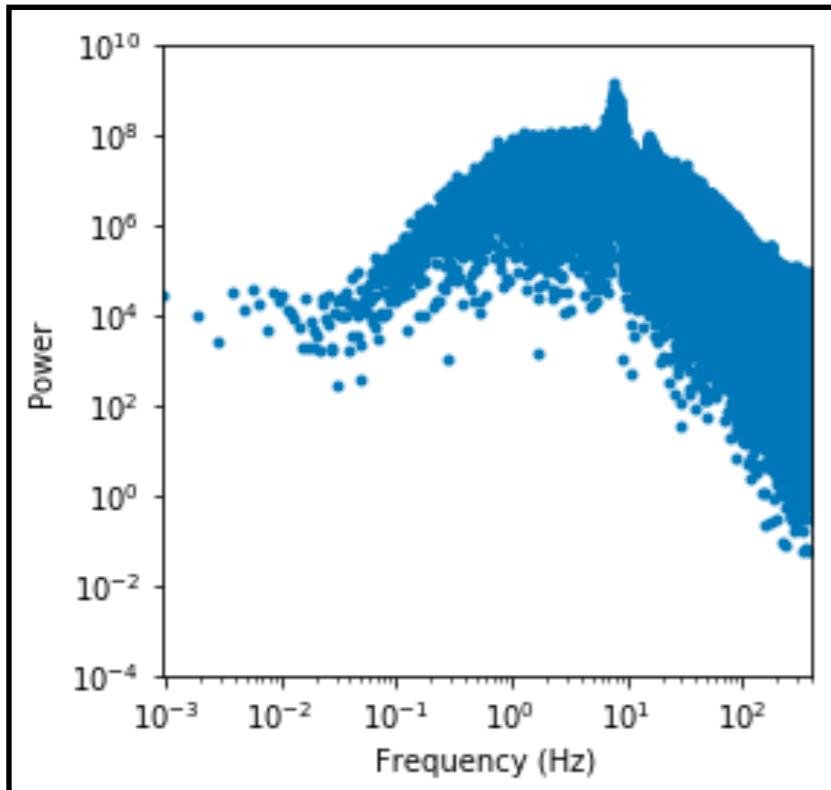


Reduce noise in estimates (similar to ERP)

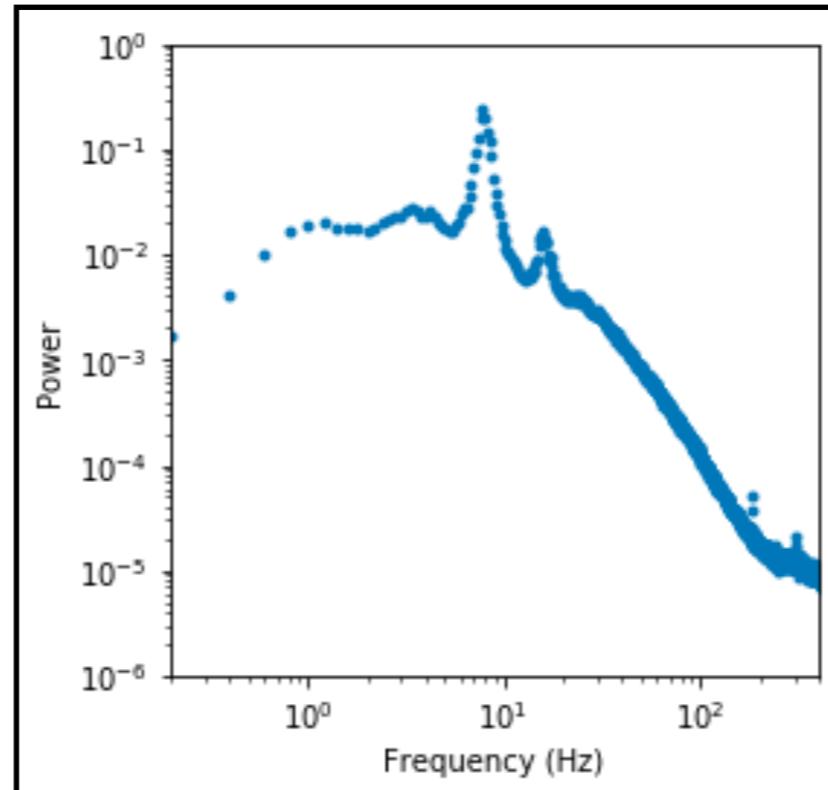


# Why Power Spectral “Density”

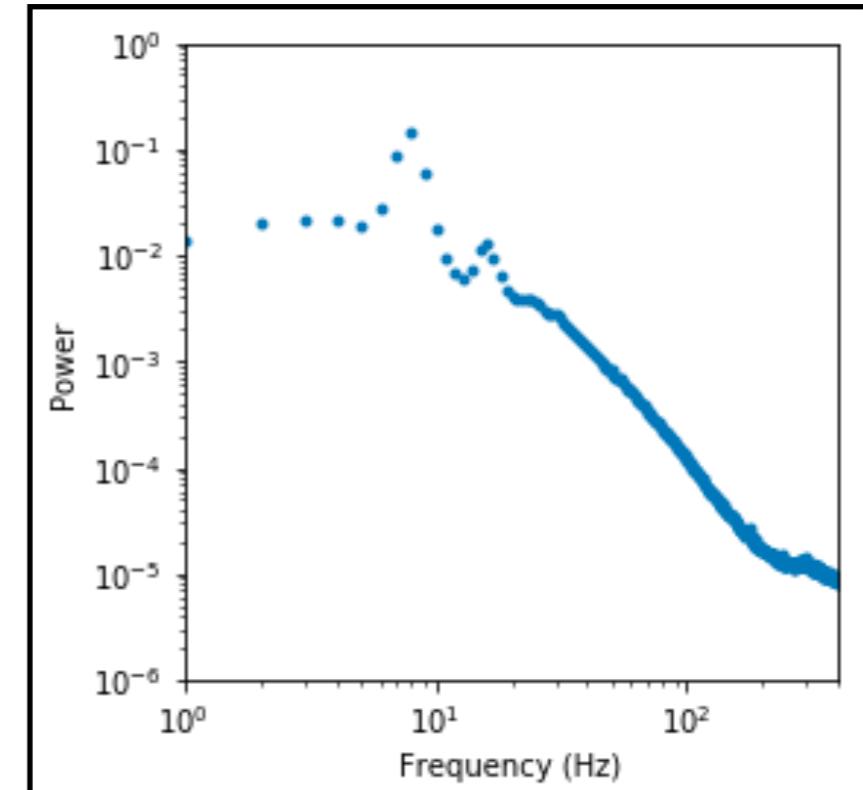
## M = N, unnormalized



$$M = 5s * fs$$



$$M = 1s * fs$$



$$\sum_{n=0}^{N-1} |x_n|^2 = \frac{1}{N} \sum_{k=0}^{N-1} |X_k|^2$$

# Parseval's Theorem

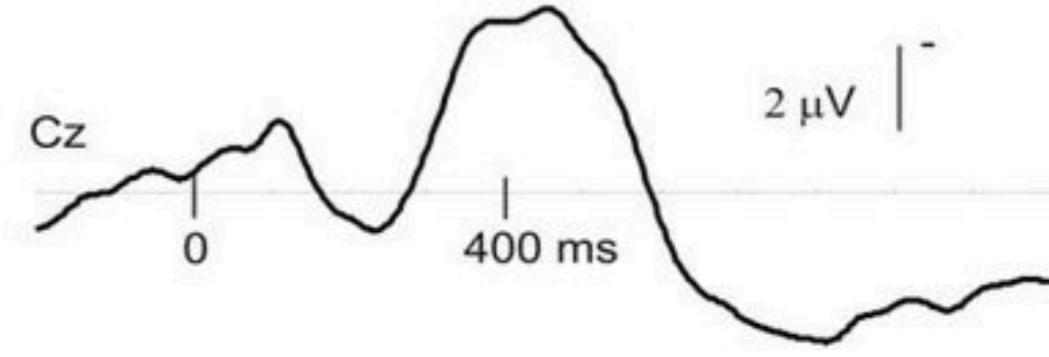
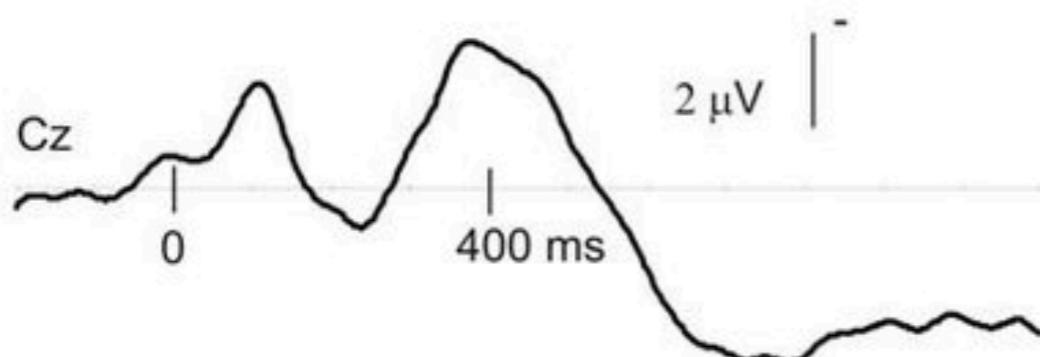
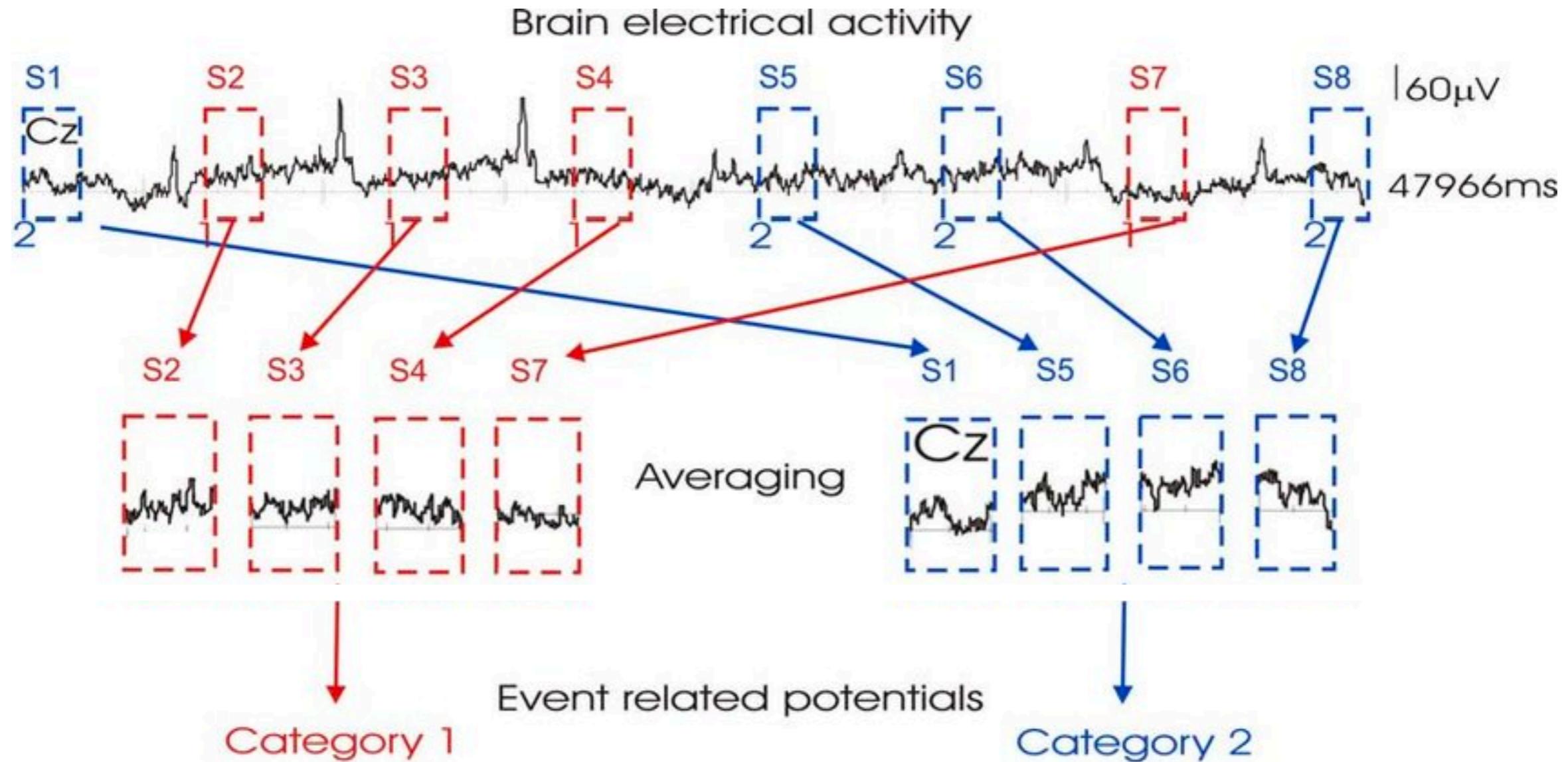
We don't want power estimate to change based on how long of a signal we use

so normalize by M

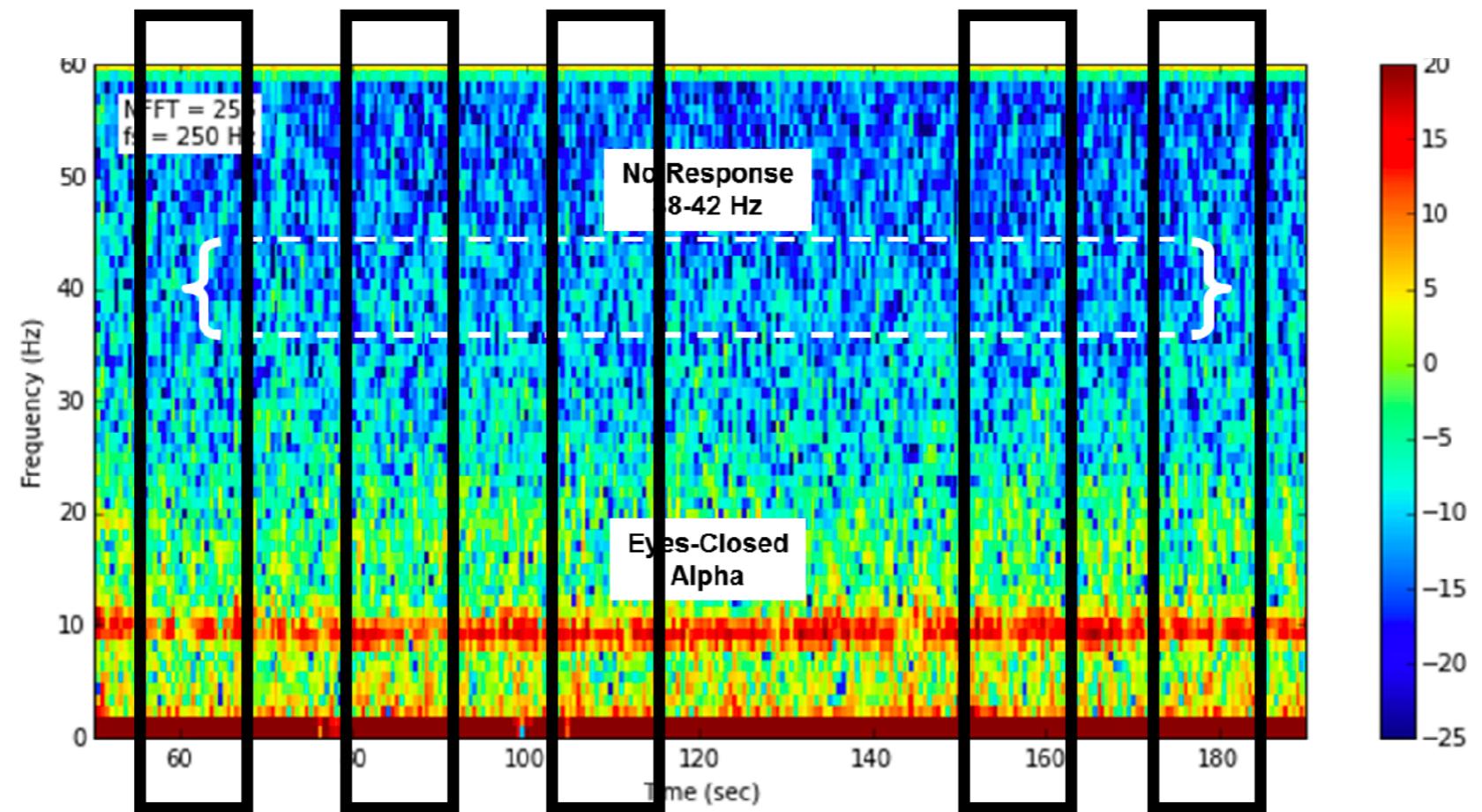
**density** -> power per frequency ( $V^2/Hz$ )



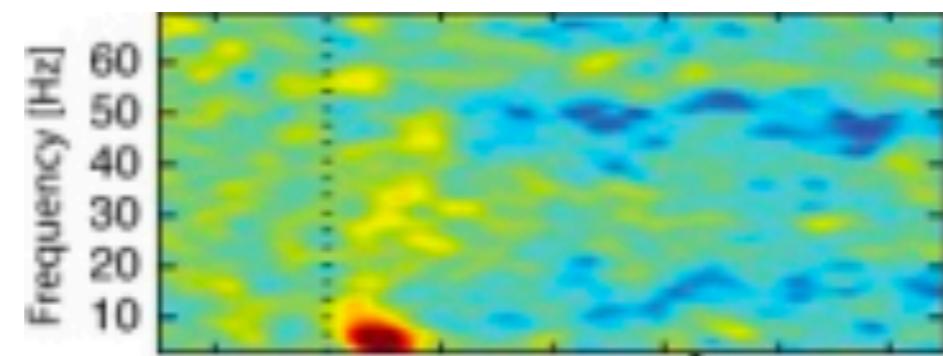
# Review: Event-Related Potential



# Event-Related Spectral Response



## Collect & Average over trials

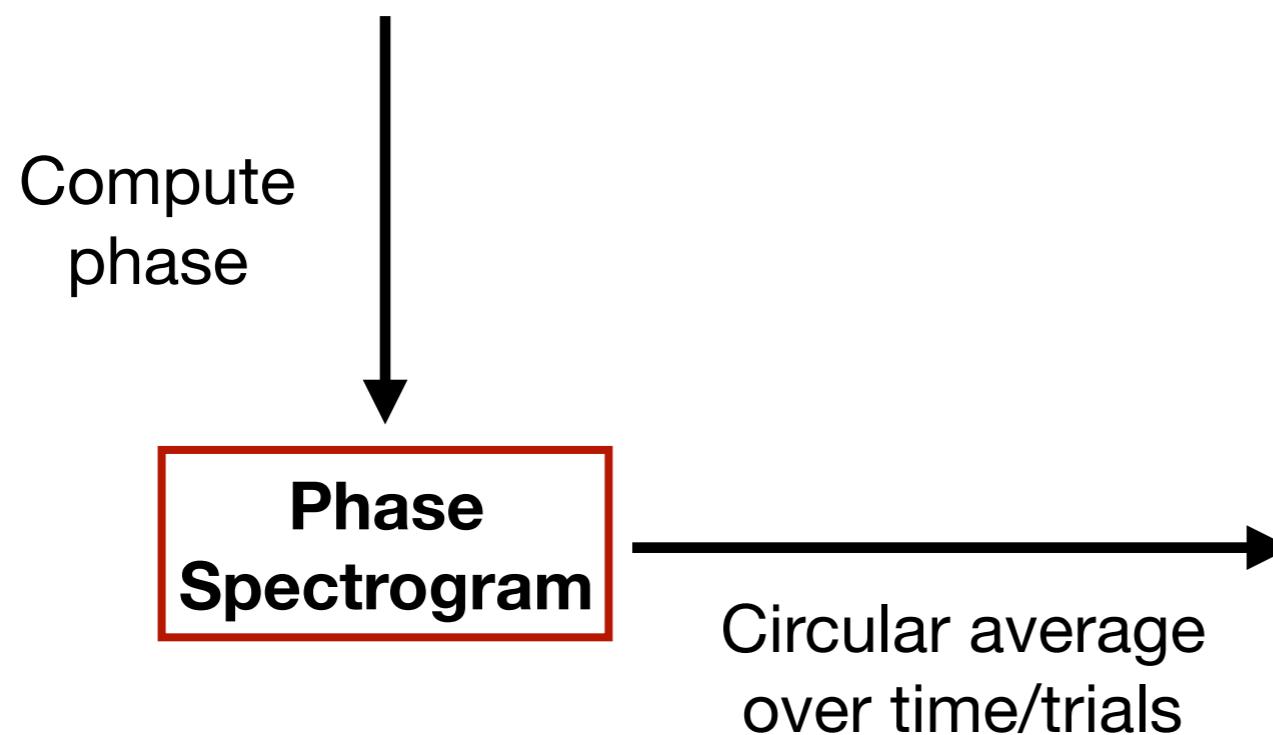


# STFT Transformations: Phase

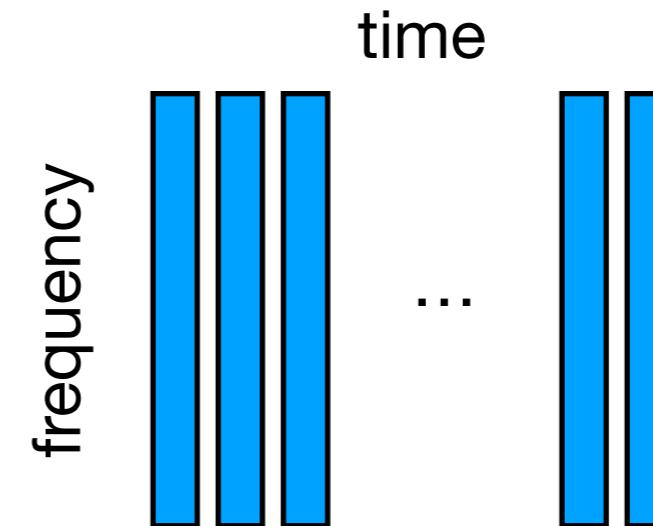
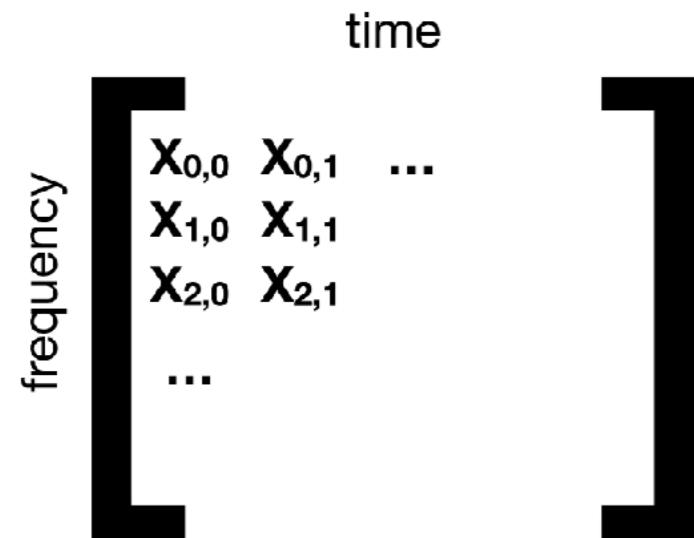


$$X(k, r) = a + ib$$

$X(k, r)$  - time-frequency representation, 2D



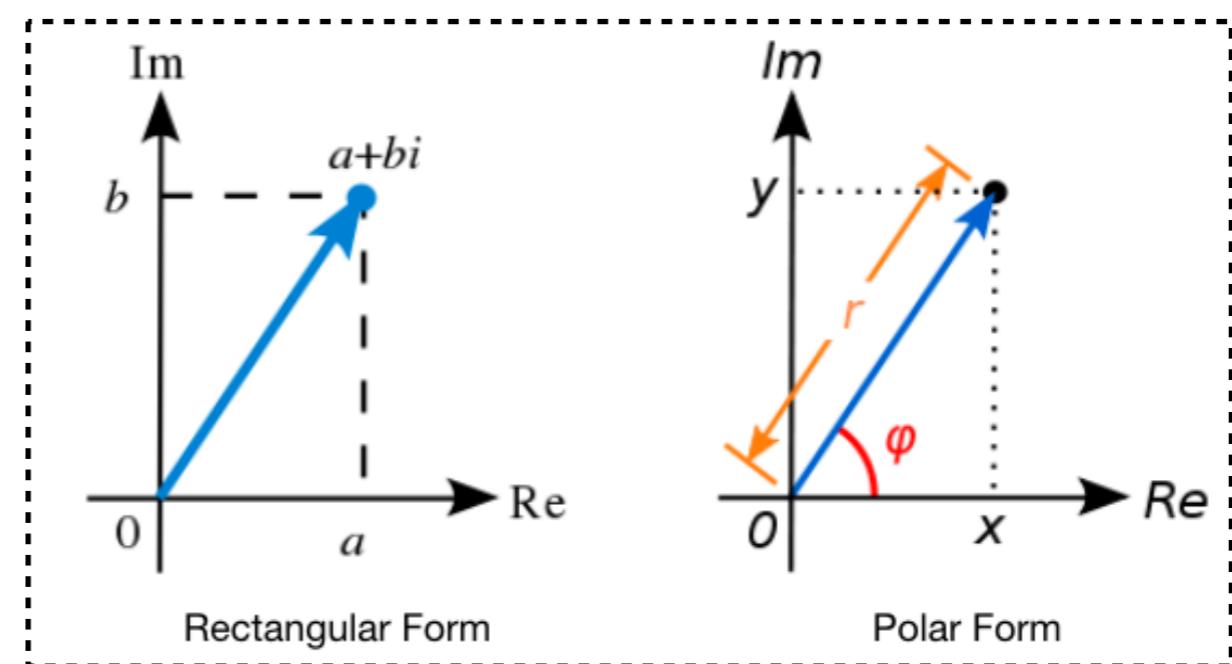
# STFT Transformations: Phase



$$X(k, r) = a + ib$$

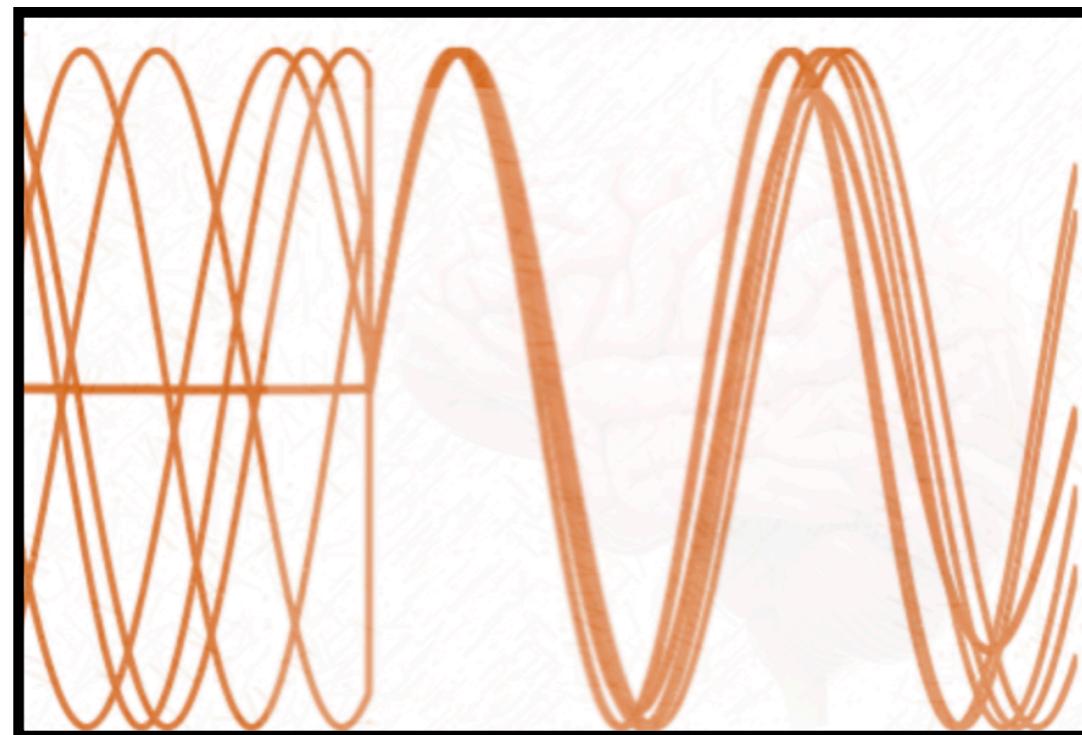
$$\text{phase}(k) = \tan^{-1} \frac{b_k}{a_k}$$

$X(k, r)$  - time-frequency representation, 2D

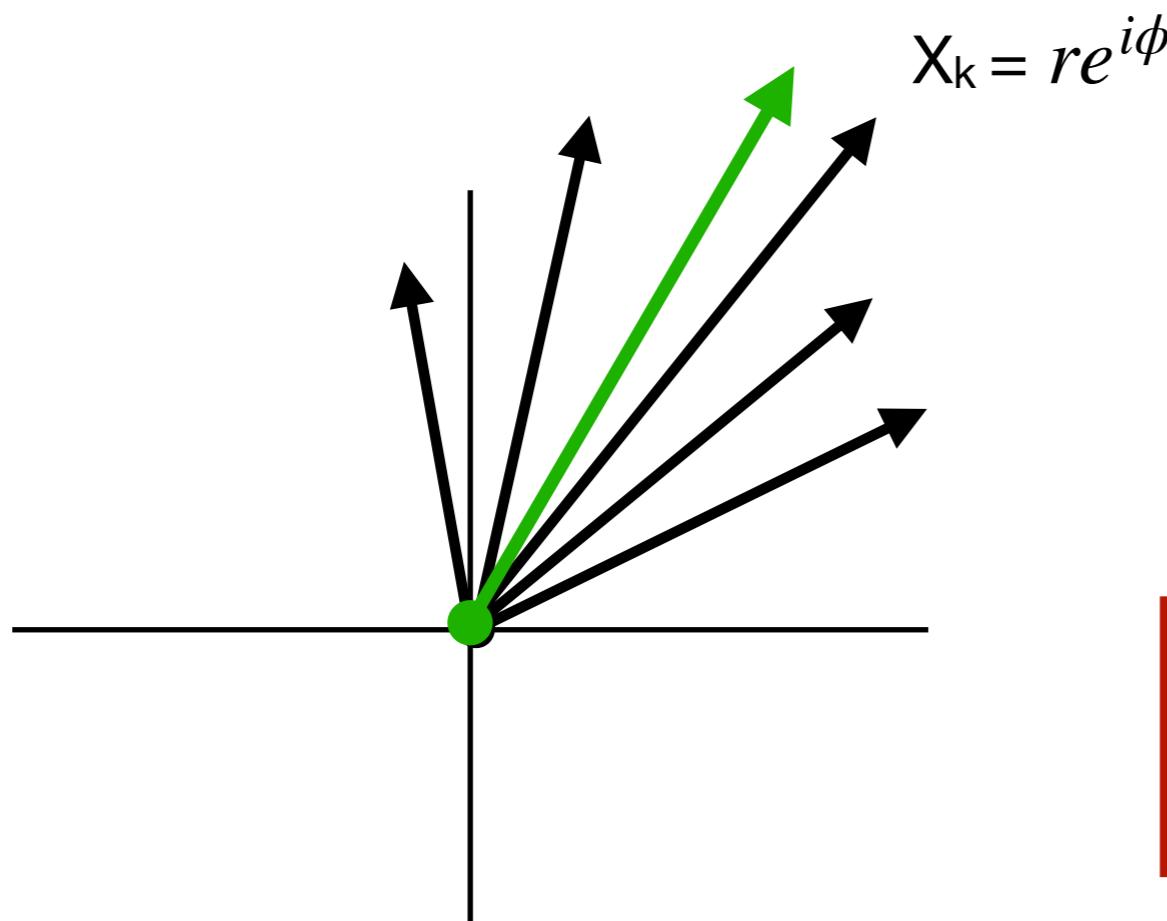


# Synchrony & Phase Locking

# Incoherent (Desynchronized)



# Coherent (Synchronized)



Average over phase angles  
(Compute the normalized vector average)

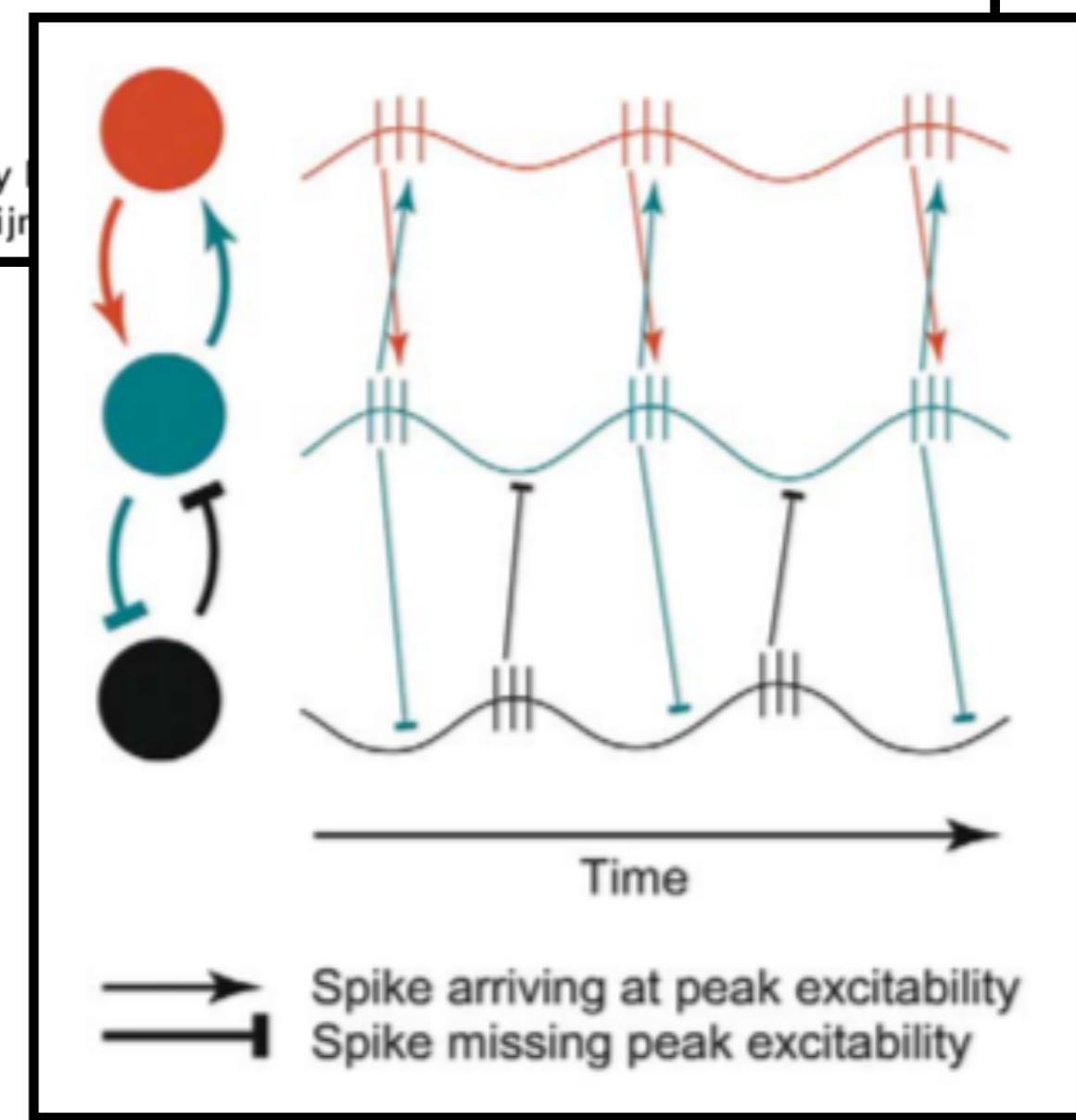
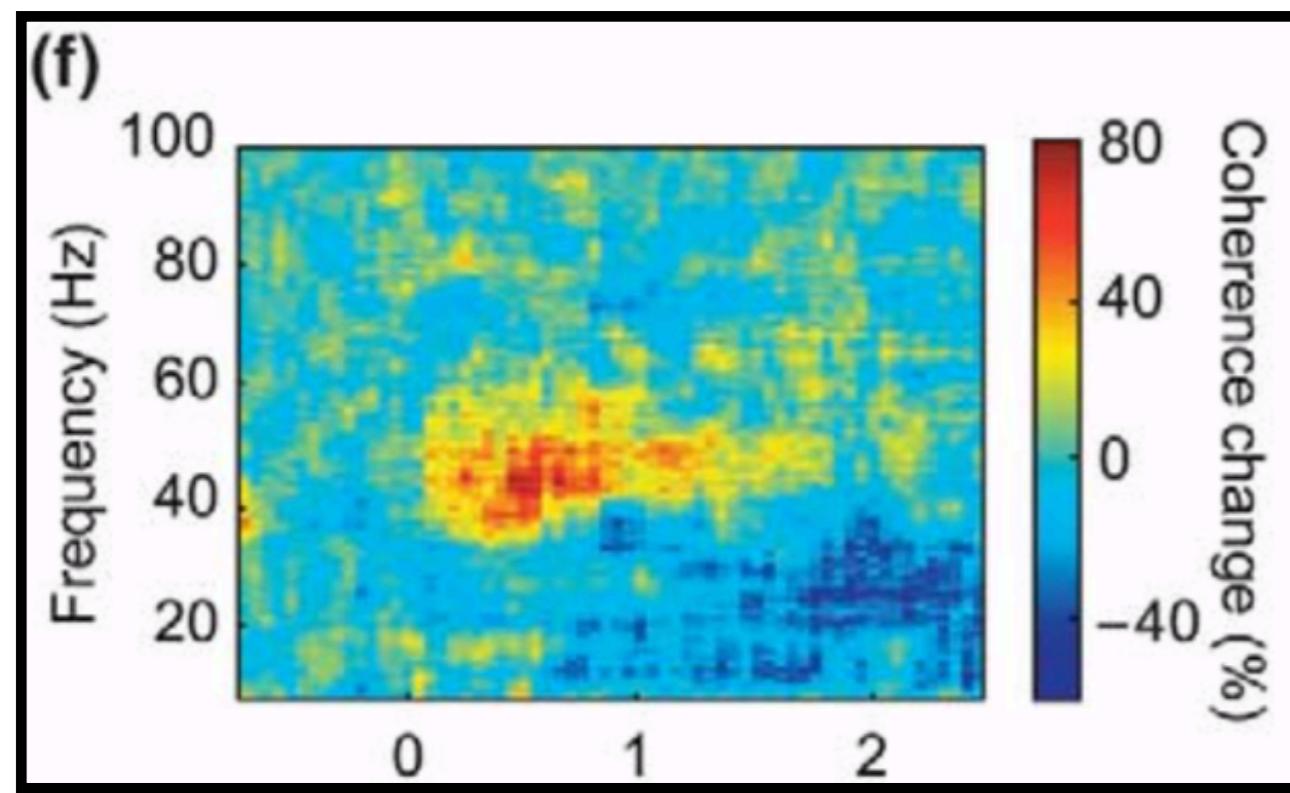
What would the vectors (black) and vector average (green) look like if the phases are incoherent?

## A mechanism for cognitive dynamics: neuronal communication through neuronal coherence

Pascal Fries<sup>1,2</sup>

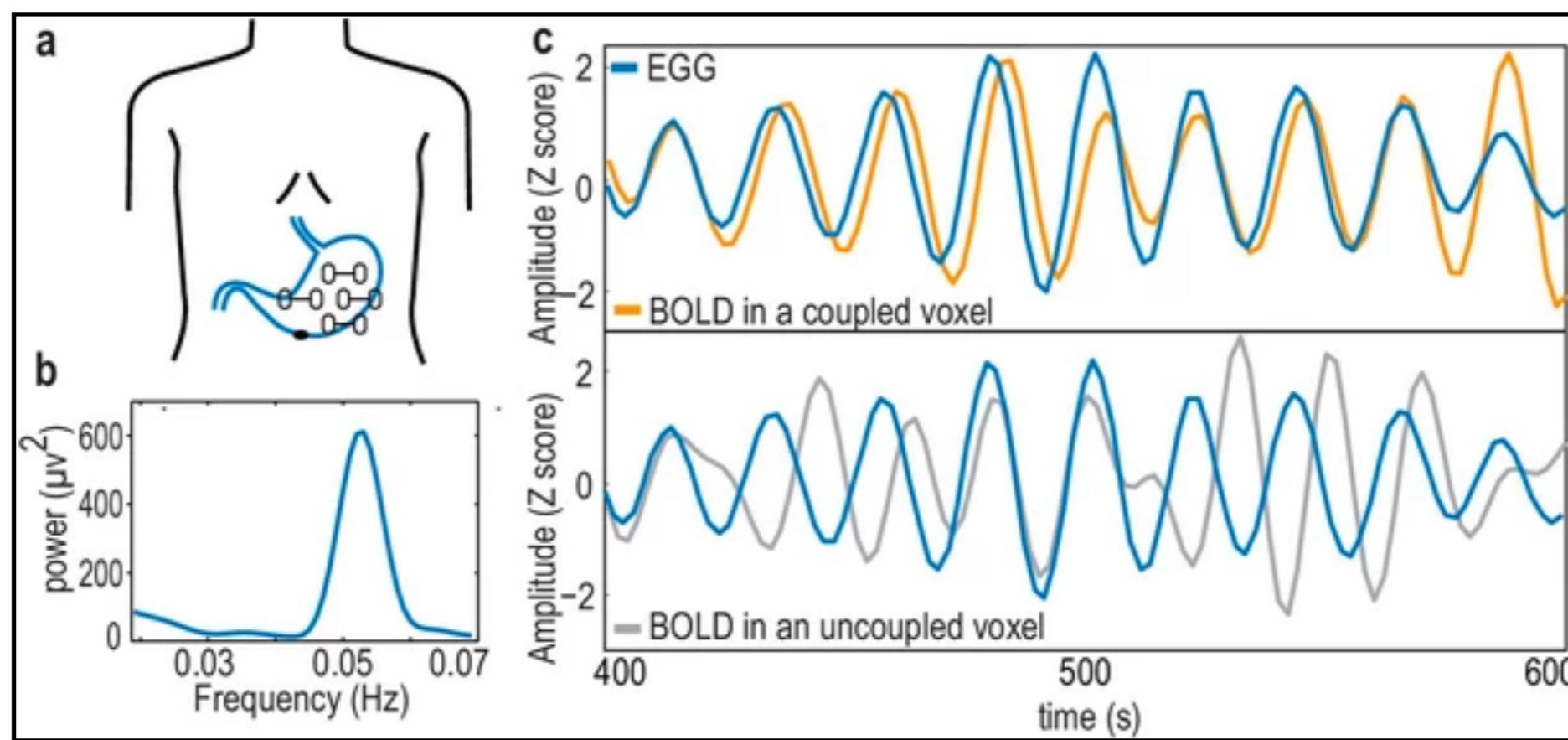
<sup>1</sup>F.C. Donders Centre for Cognitive Neuroimaging, Radboud University

<sup>2</sup>Department of Biophysics, Radboud University Nijmegen, 6525 EZ Nijmegen



# Stomach-Brain Synchrony?

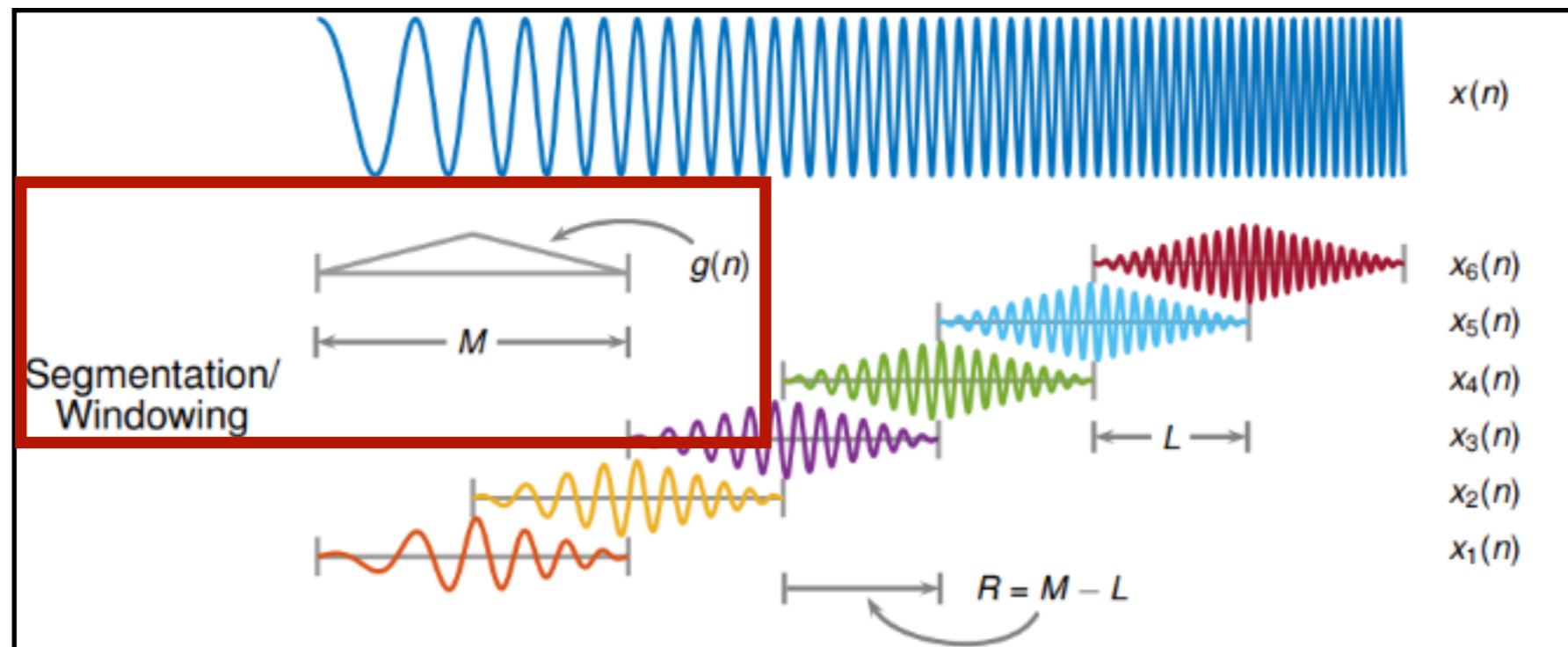
**Stomach-brain synchrony reveals a novel,  
delayed-connectivity resting-state network  
in humans**



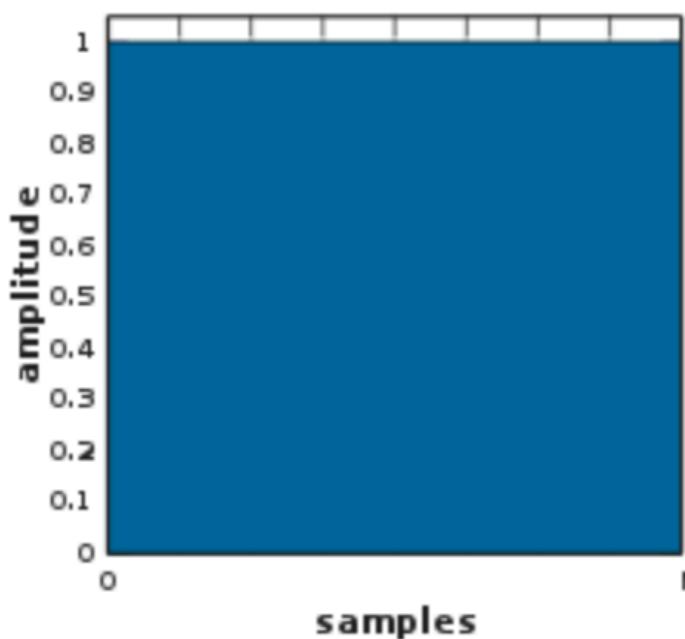
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# Windowing



Rectangular window



Simply grabbing a window of data is like **multiplying** by a rectangular window.

1 inside the window, 0 everywhere else

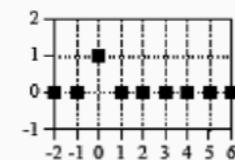
**Intuition:** this introduces sharp edge artifacts that are not present in the window.

# Multiplication & Convolution

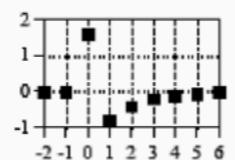
## (Circular) Convolution Theorem

Convolve  
frequency  
h:

Delta  
Function

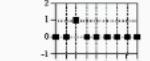


Impulse  
Response



IRF: [2,-1,-1,0]

Delta  
Function

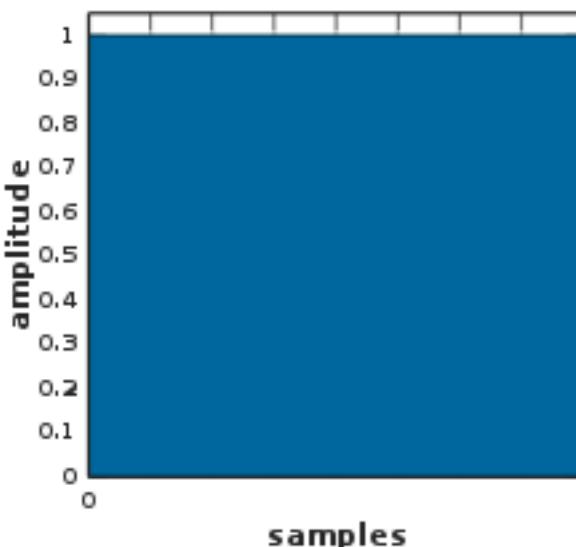


Input:  
 $x(n) = [1,0,3,7,2,5]$

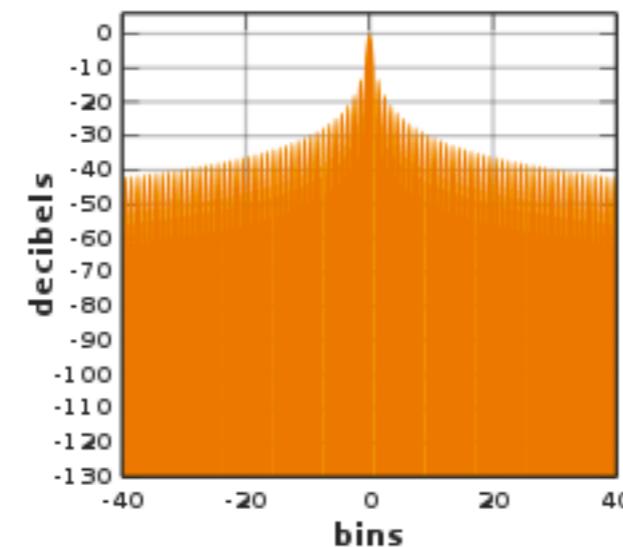
n =	0	1	2	3	4	5	6	7	8	9	10
1	2	-1	-1	0	0	0	0	0	0	0	0
0	0	0	0	0	0						
3	0	0	6	-3	-3	0					
7	0	0	0	14	-7	-7	0				
2	0	0	0	0	4	-2	-2	0			
5	0	0	0	0	0	10	-5	-5	0	0	
	2	-1	5	11	-6	1	-7	-5	0	0	0

Input:	n =	0	1	2
x(n) = [1,0,3,7,2,5]	1	2	-1	-1
	0	0	0	0
	3	0	0	6
	7	0	0	0
	2	0	0	0
	5	0	0	0
		2	-1	5

Rectangular window



Fourier transform



**“Spectral Leakage”:** making it appear as if there is power at a frequency when there isn’t.

Multiplying by a window is equivalent to **convolving by the windows frequency response**.

“Smearing” your signal’s spectrum.

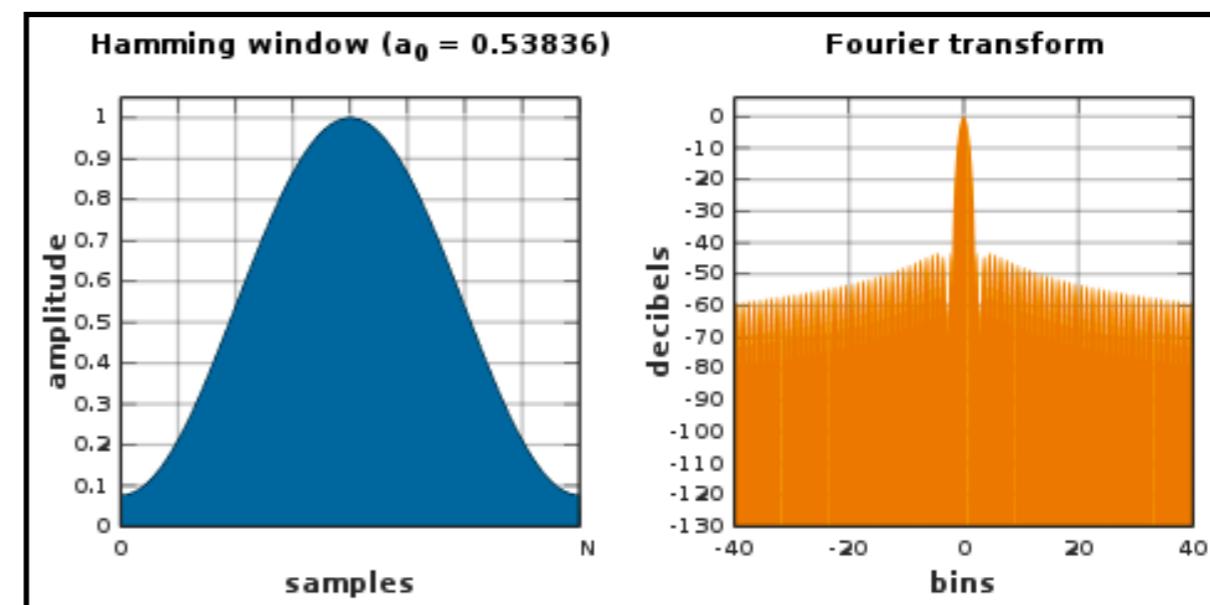
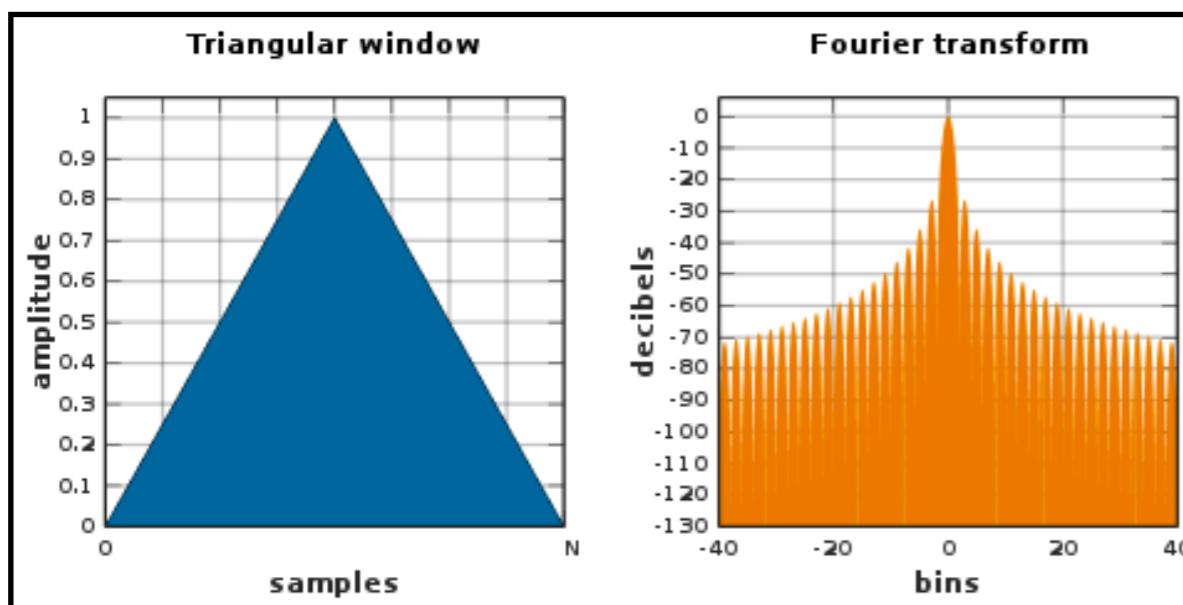
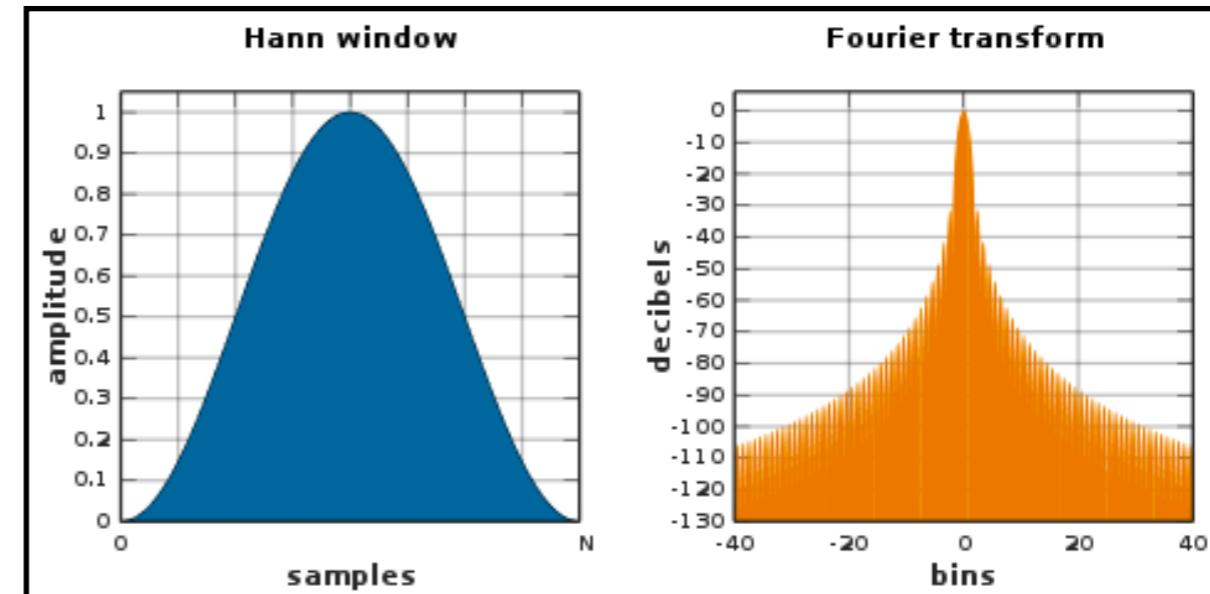
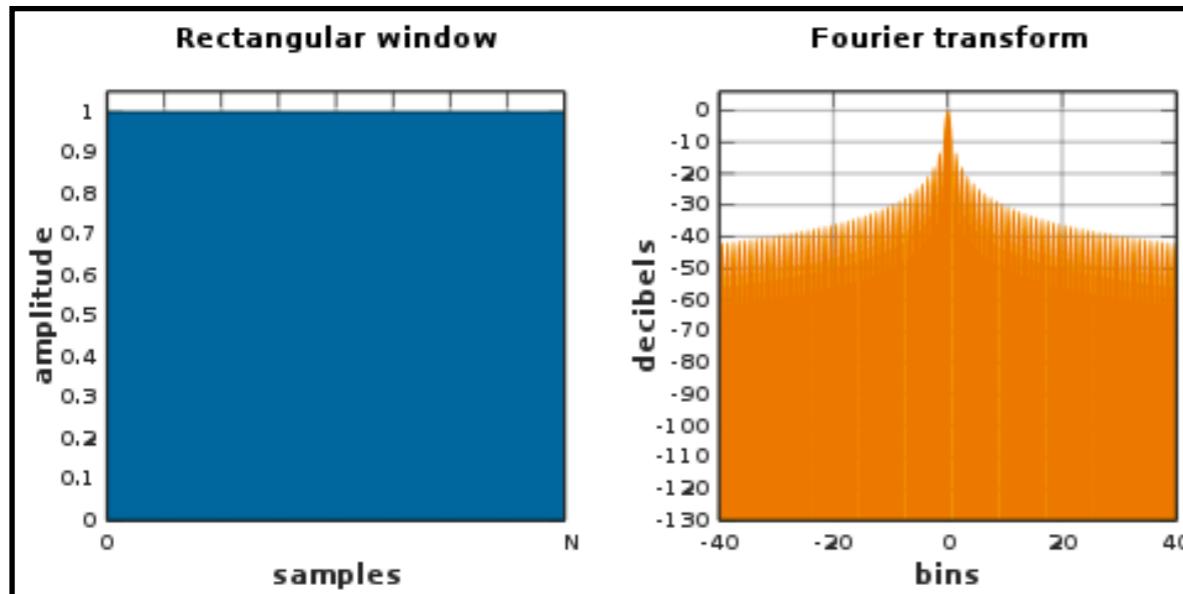
You’re **always** applying a window, even when you think you aren’t.



# Windowing Functions

You're **always** applying a window, even when you think you aren't.

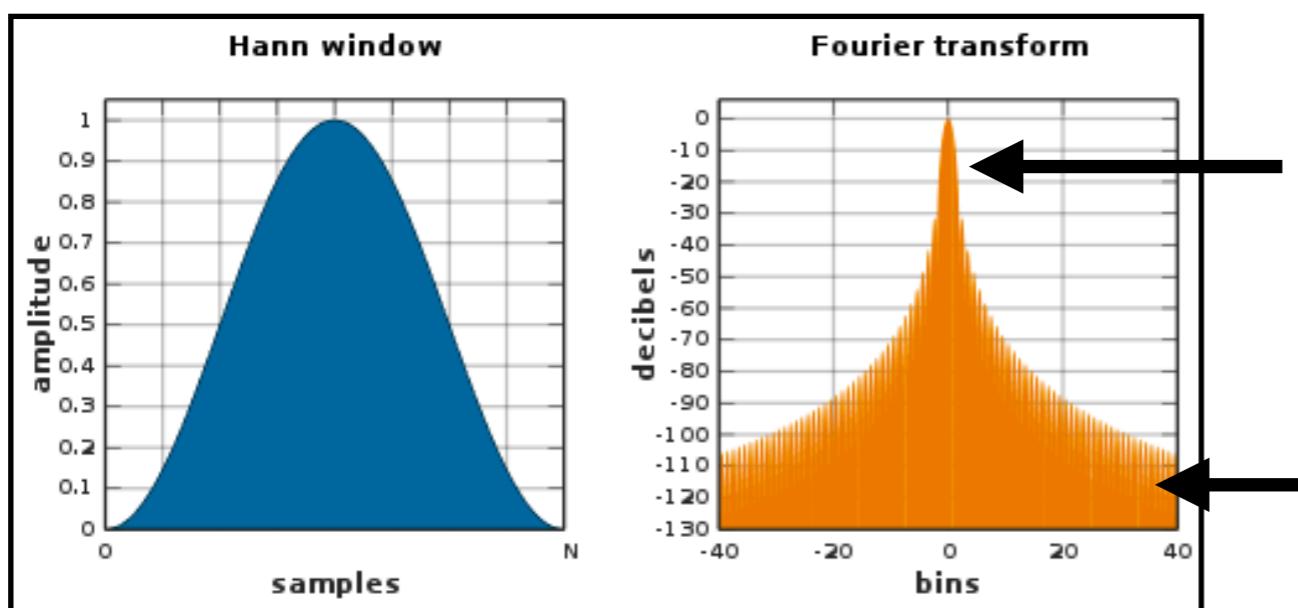
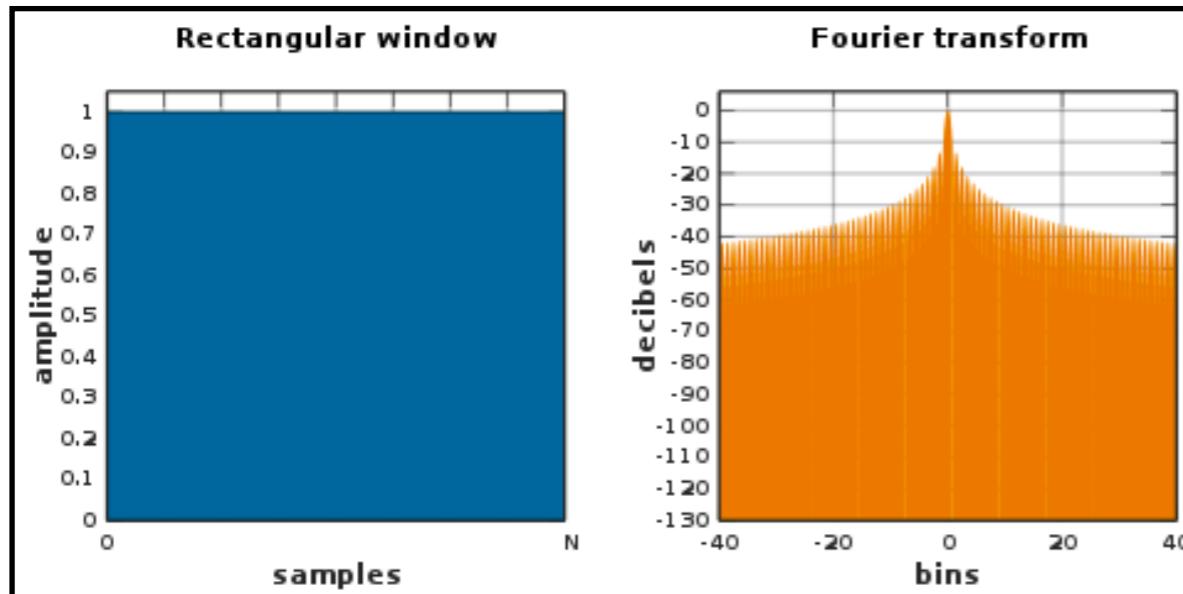
So we explicitly pick a windowing function to optimize frequency response.



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So we explicitly pick a windowing function to optimize frequency response.



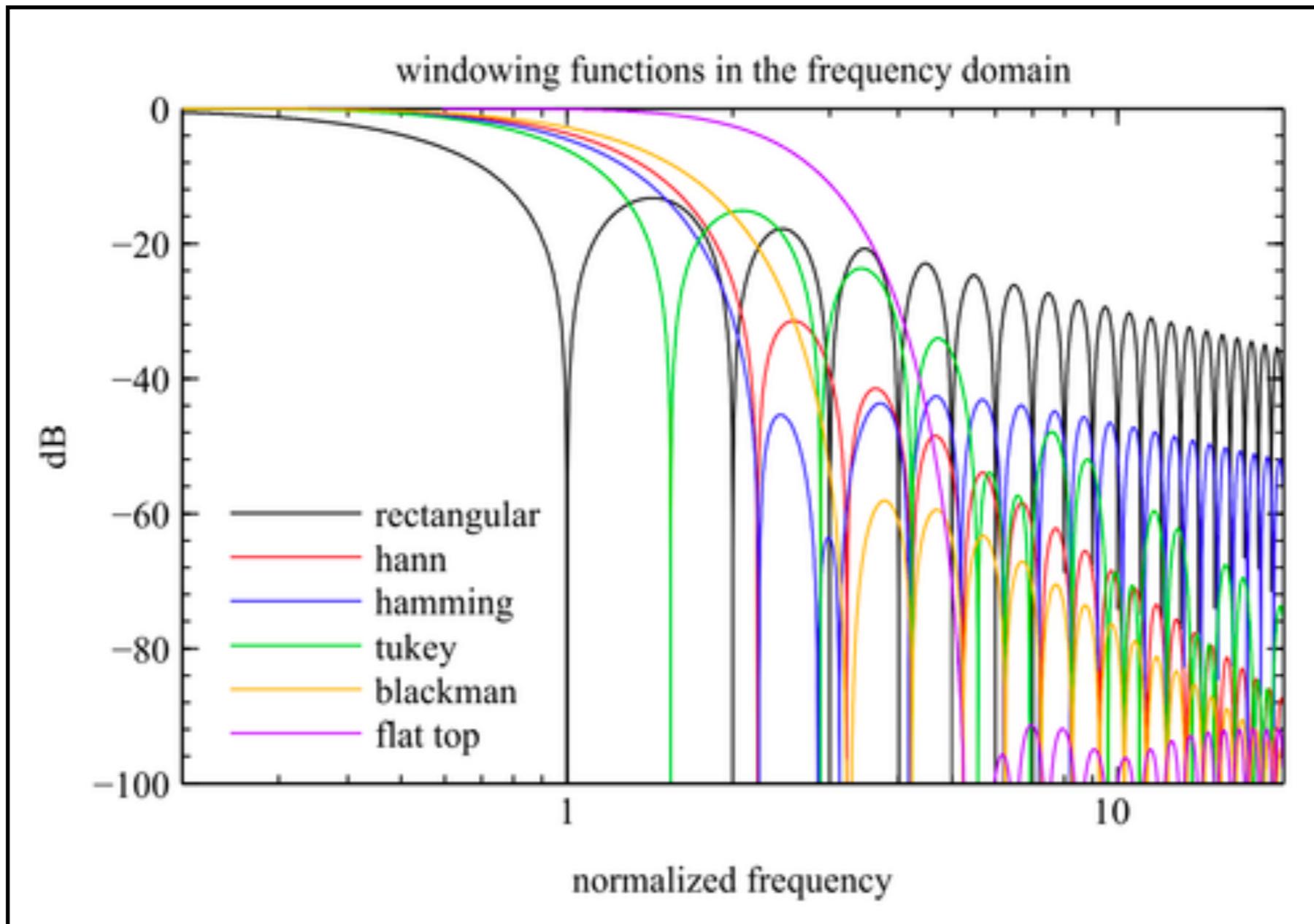
minimize this width, maximize  
its height  
**(main lobe)**

minimize this height and  
maximize the drop-off  
**(side lobe)**

# Windowing Functions

You're **always** applying a window, even when you think you aren't.

So we explicitly pick a windowing function to optimize frequency response.



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<https://tinyurl.com/cogs118c-att>

