

# **TIMESERIES 4**

10.28.2020

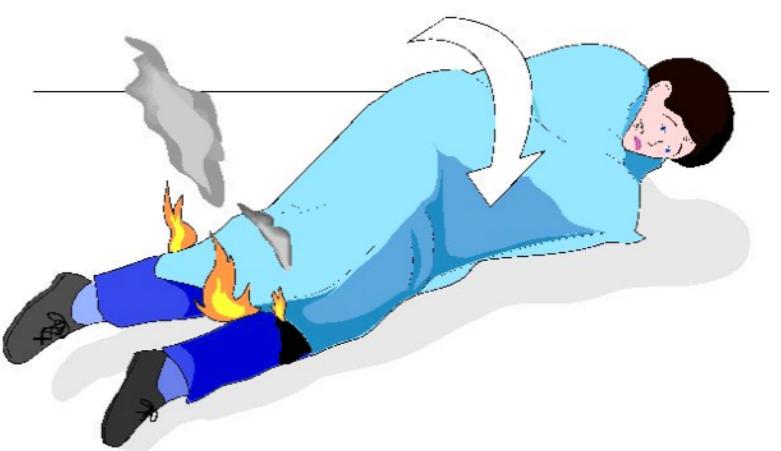
# PROBLEM SET 4

\* is due Friday!



# PROBLEM SET 5

- \* will be posted Friday
- \* but you will have 3 weeks to finish it

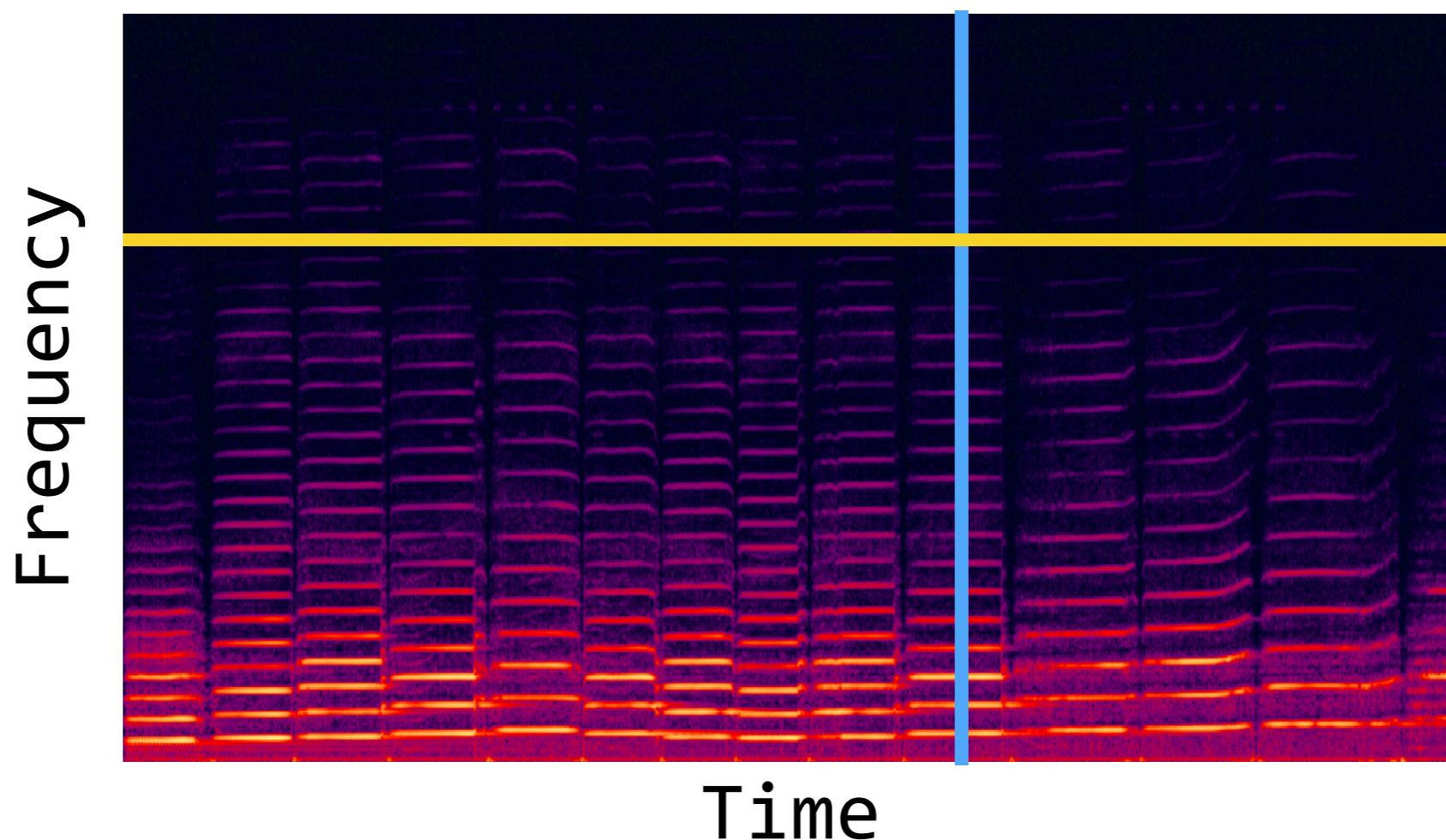


# RECAP: SPECTROGRAM

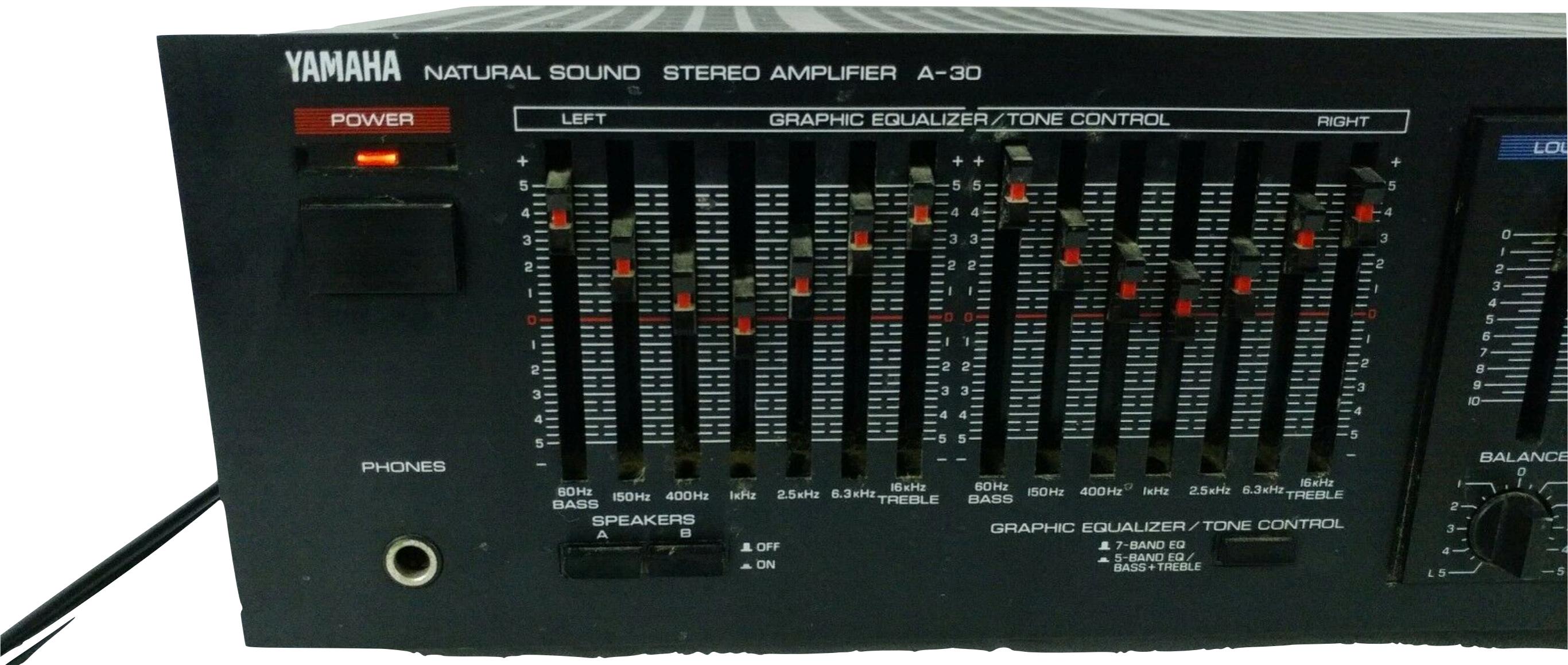
- \* if we compute the fourier transform for small snippets of time and then stack them together into an array
- \* this is the **spectrogram**
- \* it shows which frequencies are present in a timeseries at each point in time
- \* *you should know how to read a spectrogram*

# THE SPECTROGRAM

- \* each **column** is the fourier transform of a short snippet
- \* what about each **row**? what does one row mean?



# FILTERING



# FILTERING

- \* **filtering** is a process that removes some frequencies from a timeseries and lets others remain (or even amplifies them)
- \* this is accomplished by convolving your timeseries with a **filter**, a small array that is designed to have a specific effect

# FILTERING

- \* **low-pass filter:** removes high frequencies, allows low frequencies through
- \* **high-pass filter:** removes low frequencies, allows high frequencies through
- \* **band-pass filter:** removes all frequencies except for a specific band (the “pass band”)

# FILTERING

- \* **low-pass filter:** removes high frequencies, allows low frequencies through
- \* **high-pass filter:** removes low frequencies, allows high frequencies through
- \* **band-pass filter:** removes all frequencies except for a specific band (the “pass band”)

# FILTERING

- \* back to the spectrogram:
  - \* one row of a spectrogram is a lot like a **band-pass filtered** version of a timeseries

# FILTERING

- \* suppose we have some EEG data from a human subject and we want to filter it so that only alpha-band oscillations remain
  - \* (this is a band-pass filter)
- \* how do you make a filter that has the properties you want?

# FILTERING

- \* **scipy.signal** is a module in scipy that contains lots of useful functions for filter design
- \* **scipy.signal.firwin** creates “finite impulse response” filters with desired properties

# ANALYZING A FILTER

- \* **scipy.signal.freqz** is a great function that tells you what the *frequency response* of your filter looks like
  - \* i.e. it tells you what the filter is going to do to your signal

# **RECALL:**

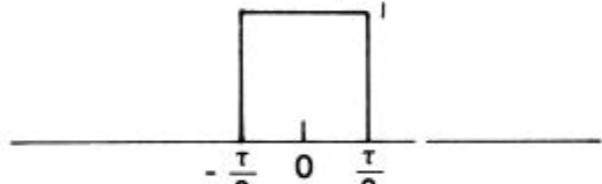
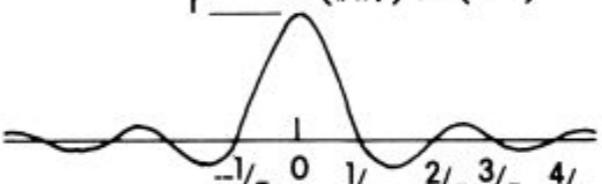
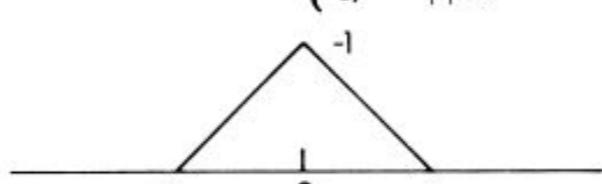
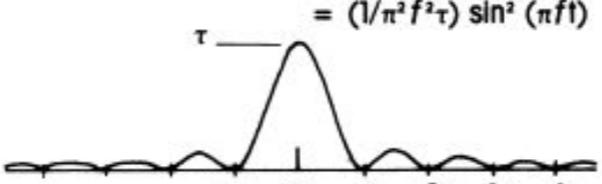
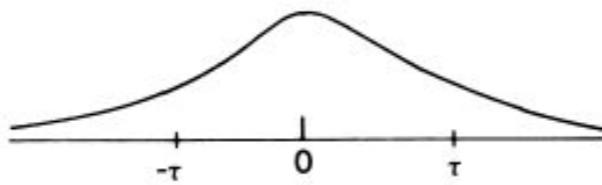
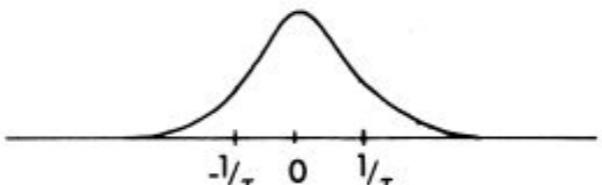
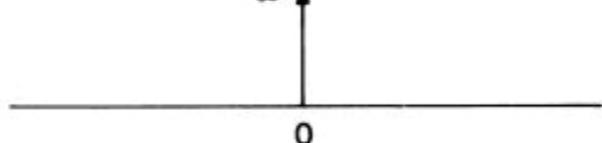
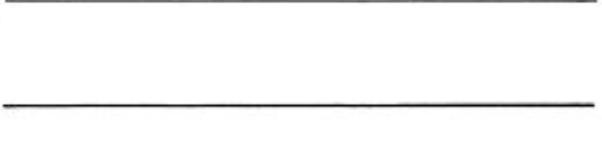
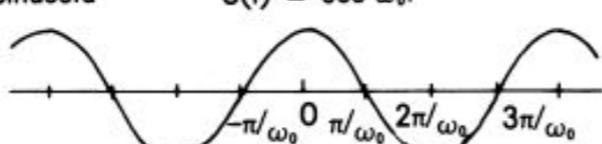
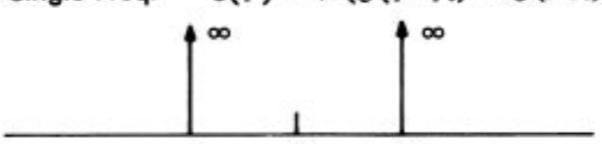
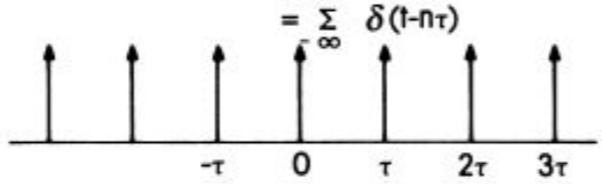
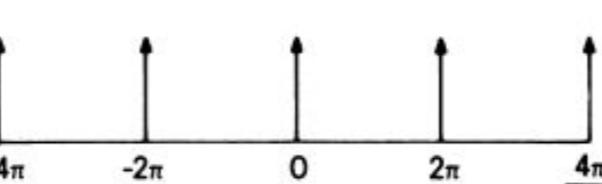
# **FOURIER ANALYSIS**

- \* fourier transforms have an interesting property related to convolution:
- \* given two timeseries,  $f$  and  $g$ , the fourier transform of their convolution = the element-wise product of their fourier transforms

$$\text{FT}(f \star g) = F \cdot G$$

- \* the reverse is also true:

$$F \star G = \text{FT}(f \cdot g)$$

Time Function	Frequency Function
<b>Boxcar</b> $G(t) = \begin{cases} 1, &  t  < \tau/2 \\ 0, &  t  > \tau/2 \end{cases}$ 	<b>Sinc</b> $S(f) = \tau \operatorname{sinc}(f\tau)$ $= (1/\pi f) \sin(\pi f\tau)$ 
<b>Triangle</b> $G(t) = \begin{cases} 1- t /\tau, &  t  < \tau \\ 0, &  t  > \tau \end{cases}$ 	<b>Sinc<sup>2</sup></b> $S(f) = \tau \operatorname{sinc}^2(f\tau)$ $= (1/\pi^2 f^2 \tau) \sin^2(\pi f\tau)$ 
<b>Gaussian</b> $G(t) = e^{-1/2 t^2}$ 	<b>Gaussian</b> $S(f) = \tau(2\pi)^{1/2} e^{-(\pi f\tau)^2}$ 
<b>Impulse</b> $G(t) = \delta(t) = 0, \quad t \neq 0$ 	<b>DC Shift</b> $S(f) = 1$ 
<b>Sinusoid</b> $G(t) = \cos \omega_0 t$ 	<b>Single Freq.</b> $S(f) = 1/2 (\delta(f+f_0) + \delta(f-f_0))$ 
<b>Comb.</b> $G(t) = \text{comb}(t) = \sum_{n=-\infty}^{\infty} \delta(t-n\tau)$ 	<b>Comb.</b> $S(f) = \sum_{n=-\infty}^{\infty} \delta(f-n/\tau)$ 

**END**