TIMESERIES 2

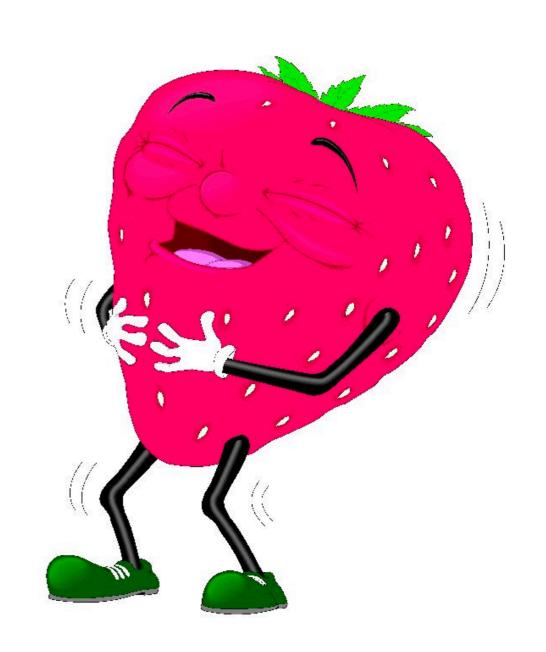
11.2.2018

PROBLEM SET 3

* was due today! which means you're probably done!

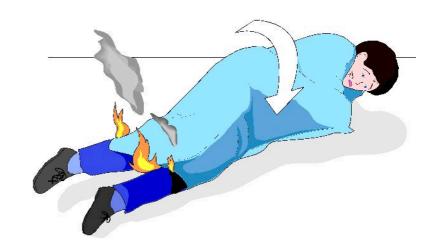
PROBLEM SET 3

- * please take a moment to pat yourself on the back for being great
- * because you are great



PROBLEM SET 4

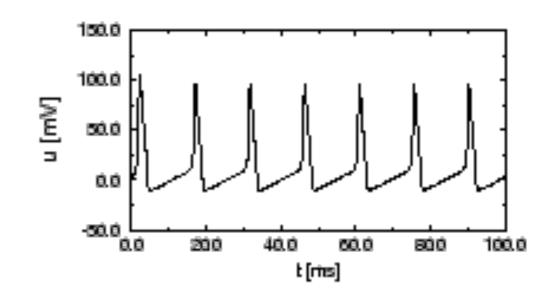
* will be posted end of next week



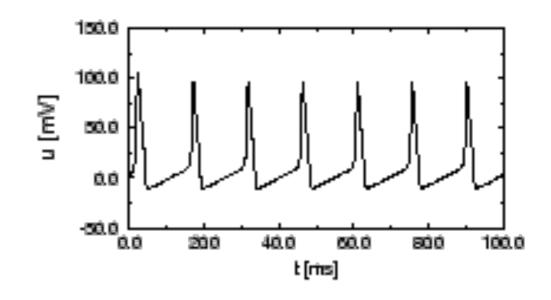
* why do things oscillate/vibrate/wobble/ wiggle/whatever?

- * feedback cycles
 - * A causes B causes C causes ... causes A
- * (but not all feedback cycles cause oscillations)

- * some feedback cycles are complicated, involving lots of variables that are related in non-linear ways
- * like the Hodgkin-Huxley equations that (mostly) govern how action potentials work in neurons



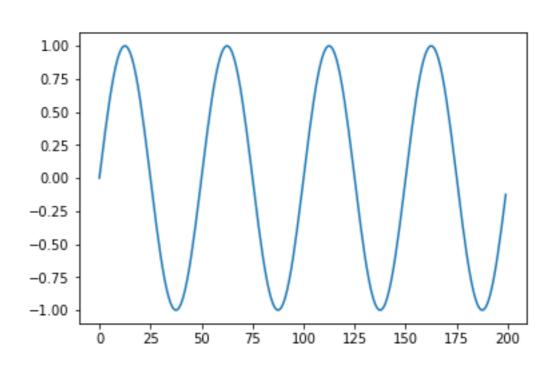
- * these complicated feedback cycles can generate periodic outputs
- * but they tend to be weird looking (like action potentials)



- * but many feedback cycles are quite simple
- * a common type is the harmonic oscillator
 - * these appear wherever acceleration (or force) is negatively proportional to location
 - * e.g. spring, rubber band, pendulum, most things bouncy or springy

* instead of complicated, weird looking outputs, harmonic oscillators always generate very nice and simple outputs:

* sine waves



- * for this (and other, more mathematical) reason(s), it's often useful to think of timeseries as the sum of a bunch of sine waves with different frequencies
- * this is called fourier analysis

- * the fourier transform is a function that figures out how to represent your timeseries as a sum of sine waves
- * every possible timeseries has a fourier transform
- * (although it might take infinitely many sine waves)



Joey Fourier

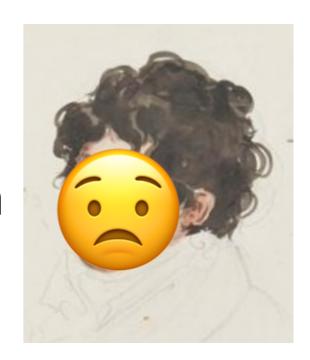


- * the fourier transform of a timeseries f is often written F
- * if the units of f are seconds, then the units of F are (1/seconds) or hertz (Hz)

- * to take a fourier transform of an array you can use np.fft.fft
 - * (fft is the "fast fourier transform" algorithm invented by Cooley & Tukey)
- * but you *almost never* want to use this directly
- * (unless you really know what you are doing)

THE PROBLEM WITH FOURIER TRANSFORMS

- * for the fourier transform to be invertible, its input and output have the same dimensionality
- * that means the fourier transform of a 1-million-point timeseries gives you 1 million frequencies



* this makes fourier transforms noisy, unwieldy, and unreliable

SPECTRAL ANALYSIS

- * if you want to know which frequencies make up a timeseries, you should probably compute the power spectrum or power spectral density (psd)
- * common psd methods (such as welch's periodogram) behave much more nicely than plain fourier transforms in many situations

SPECTRAL ANALYSIS

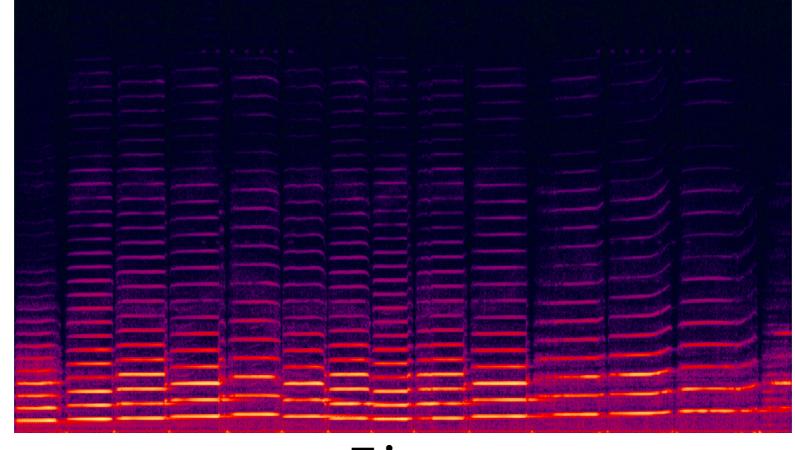
- * spectral density estimators work by taking the fourier transforms of many small snippets (aka windows) of the signal, and then averaging the results
- * thus the psd can have many fewer points than the original signal
- * which means that it's better behaved, and less sensitive to noise, etc.

THE SPECTROGRAM

- * what if we took the fourier transform of many small snippets of our timeseries, and then just looked at them instead of averaging them together?
- * this is called a **spectrogram**
- * a spectrogram tells you which frequencies are present in a timeseries at each time

THE SPECTROGRAM

* spectrograms are 2-dimensional arrays with time on the x-axis (columns) and frequency on the y-axis (rows)



Frequency

Time

THE SPECTROGRAM

* matplotlib provides an excellent method for computing spectrograms: plt.specgram

GOOGLE SPECTROGRAM

* https://musiclab.chromeexperiments.com/ Spectrogram/

CORTEX VORTEX

* http://changlabucsf.github.io/cortexvortex/build/index.html

END