





before you preform an analysis

This could mean scaling the range of the data or mean

centering it









It is important to normalize data specially if you are

comparing measures that do not have a defined

range (i.e., a personality scale vs neural activity)

Normalize data if you are comparing two measures

















#### **Data Interpolation**

If there are large outliers in your dataset, you may consider interpolation (i.e., removing outliers and guessing what the value should have been)

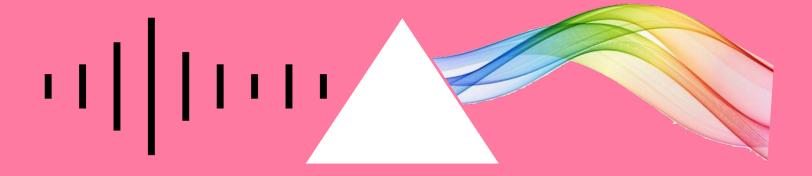
**Filloutliers()** is a useful function to identify and replace outliers in your data



#### **Spectral analysis**

Attempts to break down a wave or signal into its underlying frequency components

Answers the question how much of each frequency is there in my data





#### **Fourier Transform**

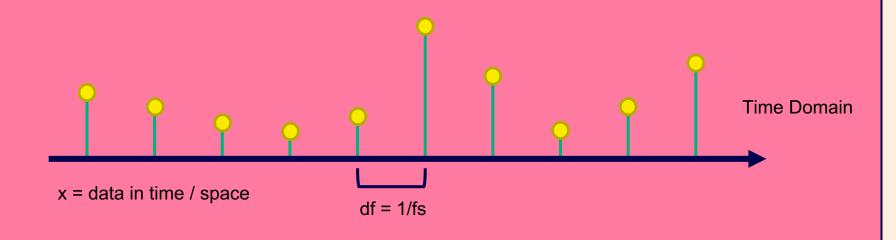
The Fast Fourier Transform is one way we can mathematically express and solve this problem.

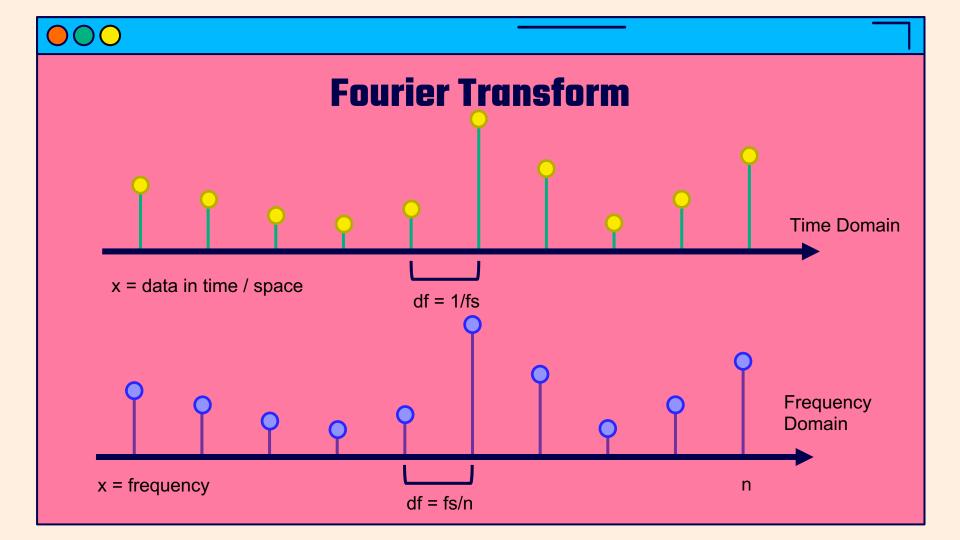
It returns the **spectrum** of **frequencies** that make up a signal

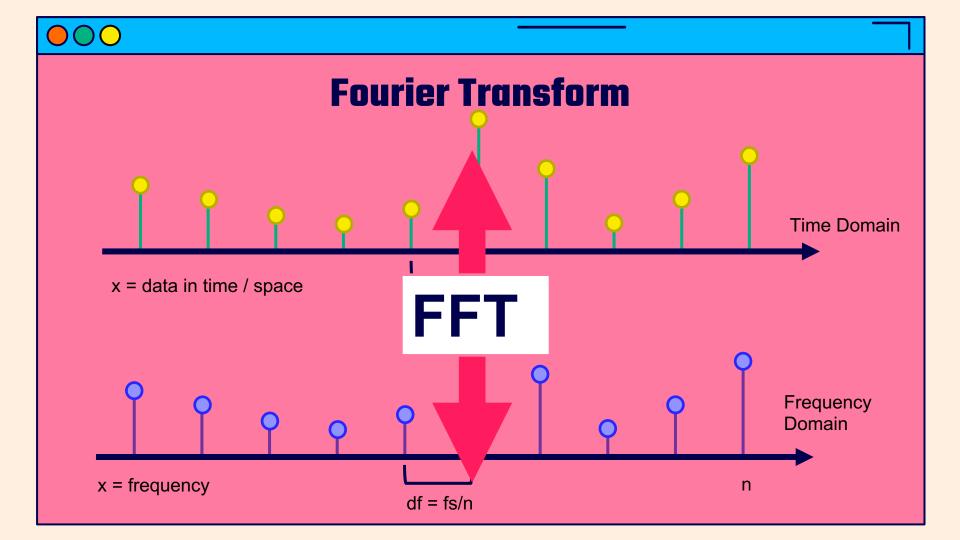


#### **Fourier Transform**

Let us imagen we have the following timeseries: it has a sampling rate and a frequency profile









#### **Fourier Transform**

In MATLAB, the function **fft()** returns the spectral estimate of a timeseries

There are many methods of spectral decomposition, the **Welch** method is one often used to increase the SNR of your output.

Here your signal is broken down into smaller windows, an FFT of these windows is computed and then averaged



## What is a Filter (no not Instagram)

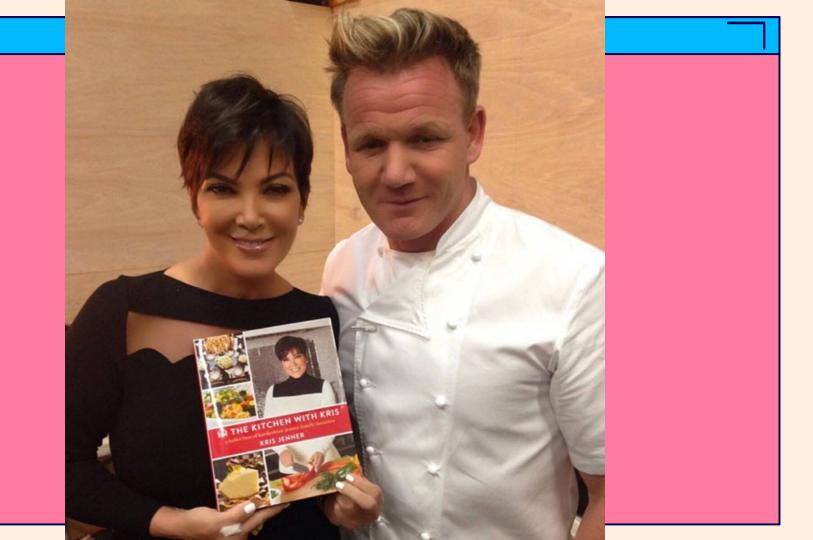
A collection of methods to remove unwanted signals/artifacts/aspects of your data

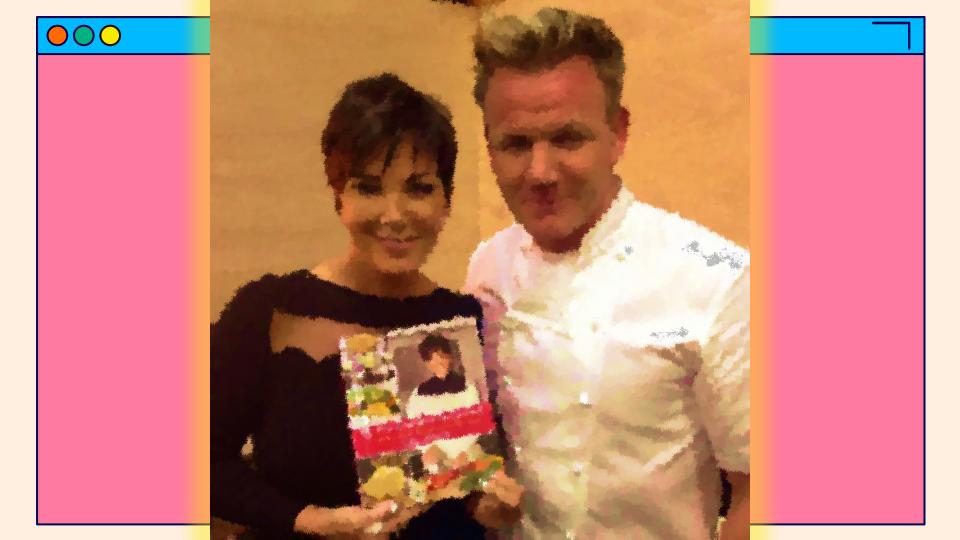


The **smooth** function (i.e., moving average, can be considered a type of filter).

Similarly, removing outliers can also act as a *Low Pass* filters, removing 'spikes' in your data

Filters come in many shapes and forms, but ultimately all aim to 'clean' your data...... but there is such thing as too much of a good thing









Filters can be considered either:

**Lowpass:** only keeps information below the defined frequency

**Highpass:** only keeps the information above the defined frequency

**Bandpass:** only keeps the information within the defined range



There are **digital** and analog **filters**, for the most part you will work with digital filters because your data is discrete (i.e., has a sampling frequency)

Infinite Impulse Response **IIR** are filters whose impulse response do not fall to zero (infinite series)

Finite Impulse Response **FIR** are filters whose impulse response are finite and will reach zero



Infinite Impulse Response **IIR** work faster and take up less memory on a computer while executing

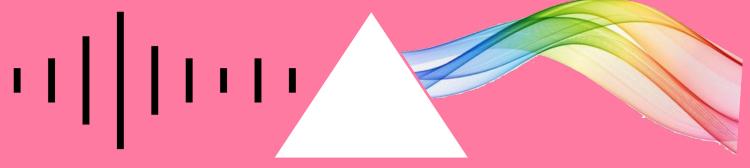
Finite Impulse Response **FIR** do not introduce nonlinear phase responses (i.e., all frequencies are shifted by the same amount in phase, creating no phase distortions)



## **Spectral analysis cont**

FFT and the Welch method are not the only options for spectral analysis, we can compute changes in frequency across time.

Yet, there is an inherent tradeoff between frequency and time resolution





## **Frequency Analysis**

A note on frequency analysis:

While there are many methods to compute power (e.g., FFT, Morlet wavelets, Hilbert, TRSP, etc) every measure is limited by a tradeoff between accuracy in **time** vs **frequency** 

This **tradeoff** means that you can either be precise in your measure of frequency or time but never both



## **Frequency Analysis**

This **tradeoff** means that you can either be precise in your measure of frequency or time but never both

For example, the **Hilbert** can return instantaneous power, yet is poor in frequency resolution

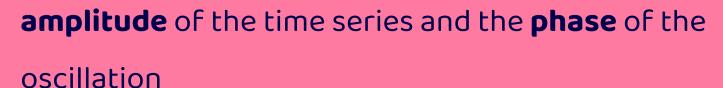
Whereas the Welch method has the opposite problem



#### **Hilbert Transform**

Transformation that takes a real timeseries and returns the

analytic signal represented by the instantaneous

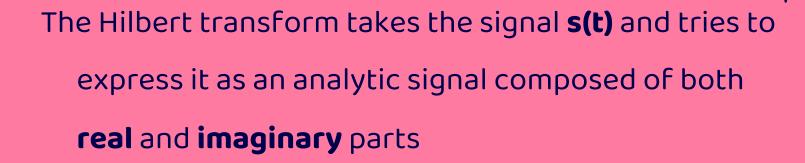


This is one way of computing spectral power over time!





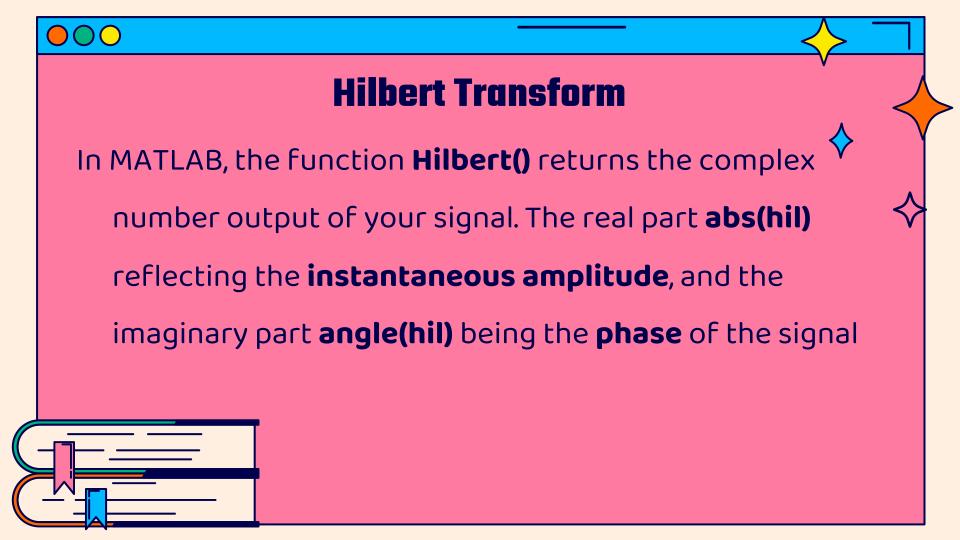
#### **Hilbert Transform**

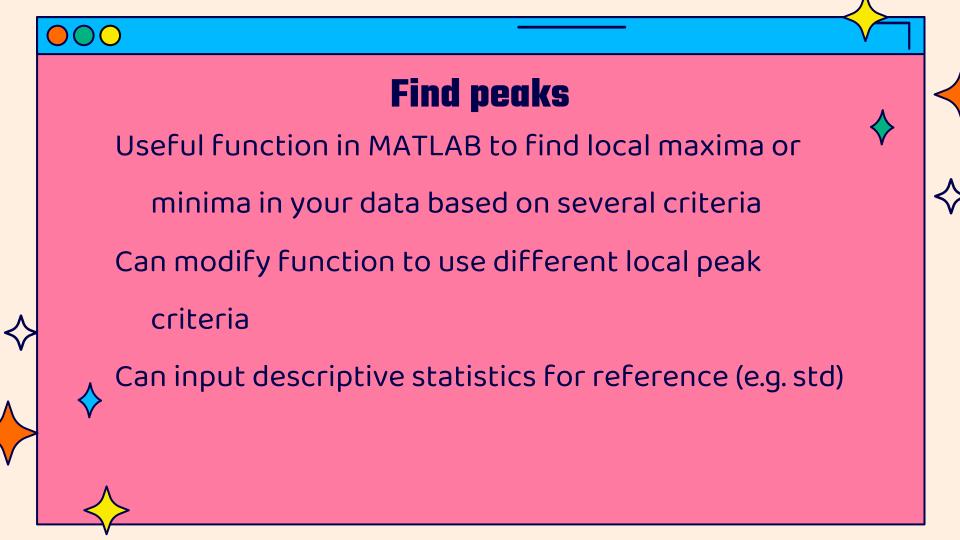


$$s_A(t) = A(t)e^{j\psi(t)}$$









# Find peaks 'MinPeakDistance' –distance between peaks

- 'Npeaks' -maximum number of peaks
- 'MinPeakHeight' –minimum height of peak
- 'MinPeakProminence' -minimum prominence of peak
- 'Threshold' –minimum difference between peak and neighbours



