

1. Introduction to the Fourier Transform

Review: Complex Numbers

$$i = \sqrt{-1}$$

$$i^2 = -1$$

$$z = \underbrace{a}_{\text{Real}} + \underbrace{bi}_{\text{Imaginary Part}}$$

$$|z| = \sqrt{a^2 + b^2}$$

$$z^* = a - bi \quad \text{"complex conjugate"}$$

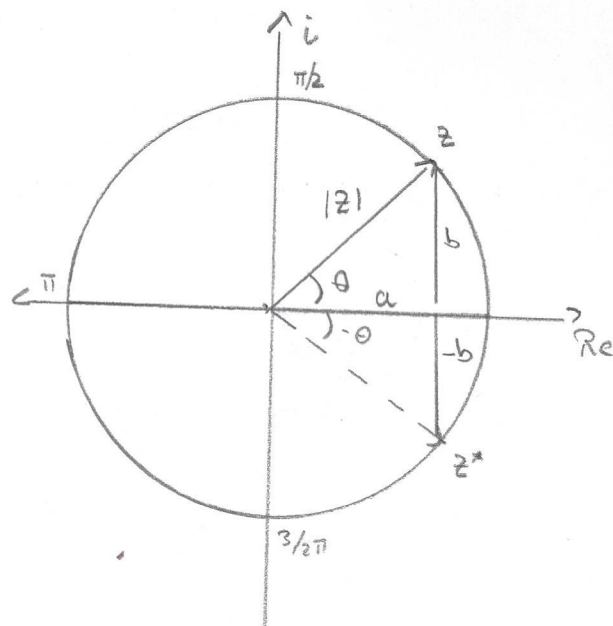
$$a = \cos(\theta)$$

$$b = \sin(\theta)$$

$$\tan(\theta) = \frac{b}{a}$$

$$\Rightarrow \theta = \arctan\left(\frac{b}{a}\right)$$

$\uparrow$
 $\tan^{-1}$



Euler's Identity

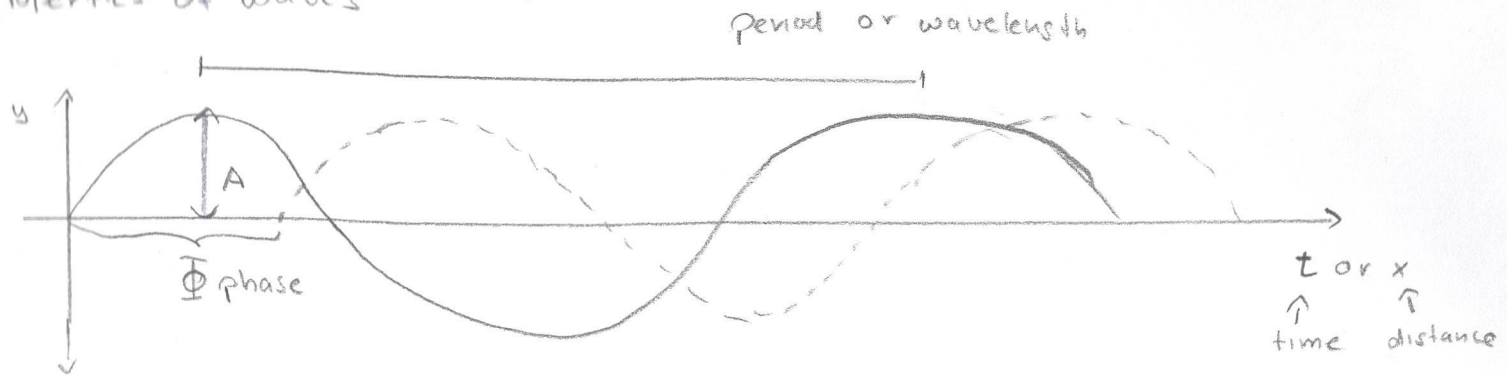
$$\exp(i\theta) = \cos(\theta) + i\sin(\theta)$$

$$\exp(i\pi) = \underbrace{\cos(\pi)}_{-1} + i \underbrace{\sin(\pi)}_0 = -1$$

$$z = a + bi = |z| \exp(i\theta)$$

$$z^* = a - bi = |z| \exp(-i\theta)$$

Properties of waves



A : Amplitude

T : period $[s]$

f : frequency $= \frac{1}{T}$ $[\frac{1}{s}]$ $[Hz]$

λ : wavelength $[m]$

ν : wave number $\frac{1}{\lambda}$ $[\frac{1}{m}]$

$$y(t) = A \cos(2\pi f t + \phi) \text{ Time domain}$$

$$y(x) = A \cos(2\pi \nu x + \phi) \text{ Space domain}$$

Sum of angles

$$\cos(a+b) = \cos(a)\cos(b) - \sin(a)\sin(b)$$

$$y(t) = A \cos(2\pi f t + \phi)$$

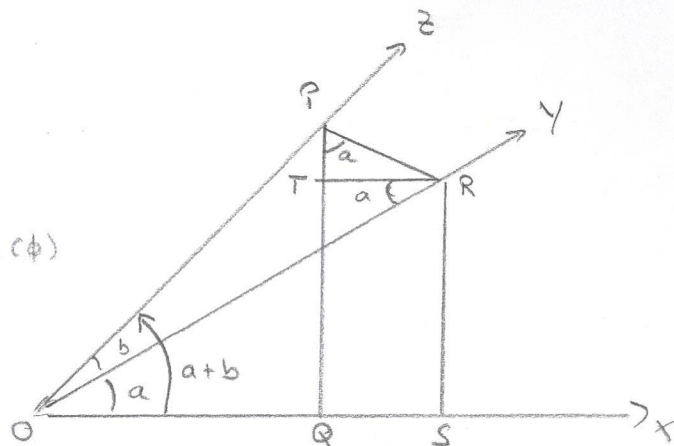
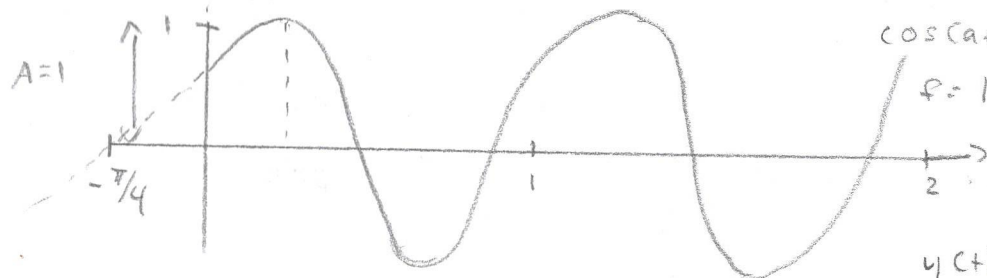
$$y(t) = A \cos(2\pi f t) \cos(\phi) - A \sin(2\pi f t) \sin(\phi)$$

$$y(t) = a_n \cos(x) + b_n \sin(x)$$

$$a_n = A \cos(\phi)$$

$$b_n = -A \sin(\phi)$$

$$x = 2\pi f t \quad \phi = -\pi/4$$



$$\cos(a+b) = \frac{OQ}{OP}$$

$$\cos(a+b) = \cos(a)\cos(b) - \sin(a)\sin(b)$$

$$f = 1s^{-1}$$

$$t \in [5]$$

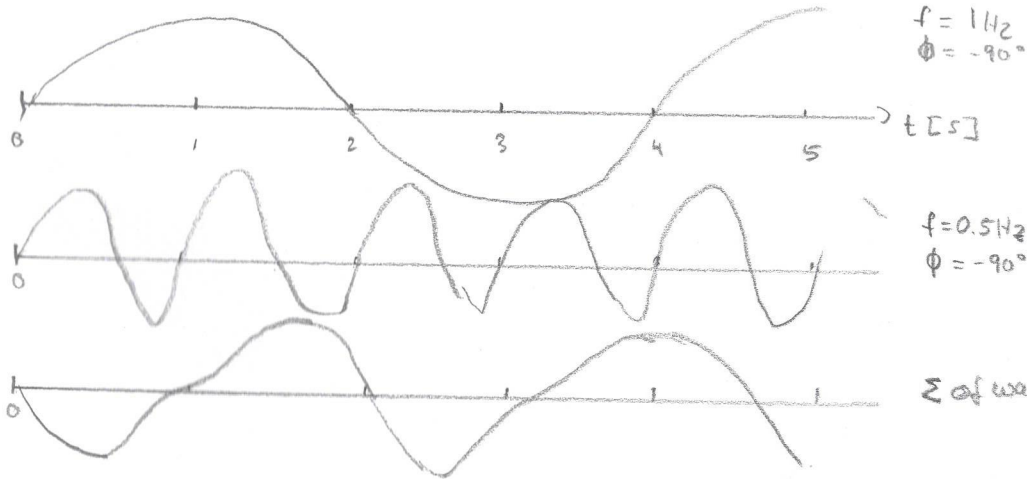
$$y(t) = \cos(2\pi t - \pi/4)$$

$$y(t) = A \cos(\pi/4) \cos(2\pi t) - A \sin(\pi/4) \sin(2\pi t)$$

Fourier Series: Expansion of a periodic function into a sum of trigonometric functions

$$y(t) = \text{mean} + \sum_{k=1}^{\infty} A_k \cos\left(2\pi \frac{k}{N} t + \phi_k\right)$$

$$y(t) = \frac{1}{2} a_0 + \sum_{k=1}^{\infty} a_k \cos\left(2\pi \frac{k}{N} t\right) + b_k \sin\left(2\pi \frac{k}{N} t\right)$$



N : Period

$\frac{k}{N}$: Frequency

Σ of waves $\Rightarrow$ mean zero
 • periodic function

"represent periodic function with period N using Fourier series"

$\frac{1}{N} \left(\frac{1}{5s} \right)$ lowest frequency
 $\frac{8}{N} \left(\frac{8}{5s} \right)$ highest frequency

Fourier Series in the Complex Domain

$$y(t) = \frac{1}{2} a_0 + \sum_{k=1}^{\infty} \underbrace{a_k \cos(kx) + b_k \sin(kx)}$$

$$\exp(ix) = \cos(x) + i \sin(x)$$

$$x = 2\pi \frac{k}{N} t$$

$$\exp(ix) = \cos(x) + i \sin(x)$$

$$\exp(-ix) = \cos(x) - i \sin(x)$$

$$\frac{1}{2} (a_k - i b_k) \exp(ix) = \frac{1}{2} a_k \cos(x) + \frac{1}{2} a_k i \sin(x) - \frac{1}{2} i b_k \cos(x) + \frac{1}{2} b_k \sin(x)$$

$$+ \frac{1}{2} (a_k + i b_k) \exp(-ix) = \frac{1}{2} a_k \cos(x) - \frac{1}{2} a_k i \sin(x) + \frac{1}{2} i b_k \cos(x) + \frac{1}{2} b_k \sin(x)$$

$$X_k^* \exp(ix) + X_k \exp(-ix) = a_k \cos(x) + b_k \sin(x)$$

$$X_k = \frac{1}{2} (a_k + i b_k)$$

$$X_k^* = \frac{1}{2} (a_k - i b_k)$$

$$\frac{1}{2} \sqrt{X \cdot X^*} = A \text{ (Amplitude)}$$

$$\tan^{-1} \left(\frac{b_k}{a_k} \right) = \tan^{-1} \left(\frac{\text{Im}(X_k)}{\text{Re}(X_k)} \right) = \phi \text{ (phase angle)}$$

$$y(t) = \sum_{-\infty}^{\infty} X_k \exp(-ikx)$$

$$\text{note: } \frac{1}{2} \sqrt{X_0 X_0^*} = \frac{1}{2} a_0 = \text{mean}$$

each real frequency (e.g. $k=1$)

has two X coefficients

Discrete Fourier Transform: how to find X_k ?

Fourier Series

$$y(t) = \sum_{-\infty}^{\infty} X_n \exp(-inx)$$

$$x = 2\pi \frac{k}{N} t$$

Infinite series representing an

N-periodic function in the time domain

$\Rightarrow$ infinite X_k

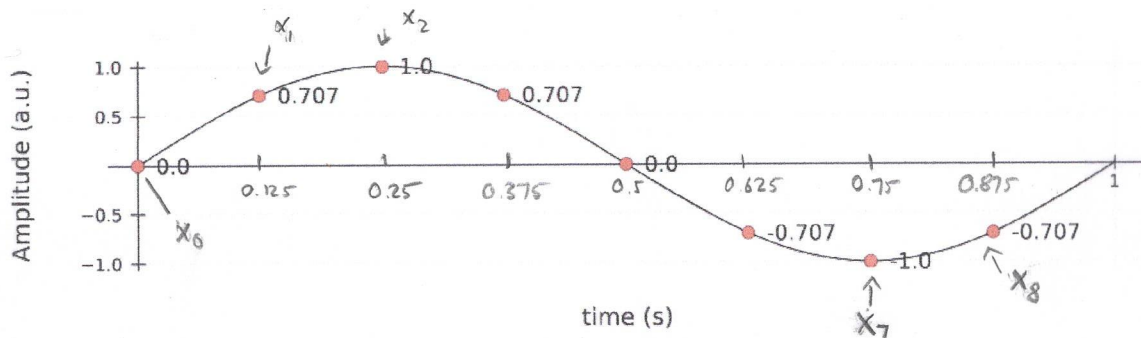
Discrete Fourier Series

$$X_n = \frac{1}{N} \sum_{k=0}^{N-1} X_k \exp(-inx)$$

$$x = 2\pi \frac{k}{N} n$$

Finite sequence of N equally spaced

samples of a function $y(t)$



$$y(t) = \sin(2\pi t)$$

$f = 8 \text{ Hz}$ "sample frequency"

$$\Delta t = \frac{1}{8} = 0.125 \text{ s}$$

$N = 8$ points

$$y_0 = 0$$

$\vdots$

$$y_2 = 1.0$$

How to find X_k ?

$$X_k = \sum_{n=0}^{N-1} x_n \exp(-i x)$$

$$x = 2\pi \frac{k}{N} n$$

$\Rightarrow$ N Fourier coefficients X_k for N sample points

$$X_0 = \sum_{n=0}^{N-1} x_n$$

$$\Rightarrow \text{sample mean} = \frac{1}{N} \sum_{n=0}^{N-1} x_n = \frac{1}{N} X_0$$

k	t	f	X_k	A
0	0	0 Hz	0	mean
1	0.125	1 Hz	$0 - 4i$	1
2	0.25	2	0	0
3	1	3	0	0
4	1	4	0	0
5	1	5	0	unsolvable
6	1	6	0	
7	0.875	7	$0 + 4i$	

$$f_s = \frac{1}{\Delta t} = 8 \text{ Hz}$$

$$f = \frac{k}{N} f_s$$

$$A = \frac{2}{N} \sqrt{X_k X_k^*}$$

$$f_s/2 = N_{Ny}$$

"Nyquist
frequency"

$$X_1 = X_7^*$$

How to find X_k ?

$$X_k = \sum_{n=0}^{N-1} x_n \exp(-i k x) \quad \text{each } X_k \text{ requires } N \text{ additions}$$

thus need $N \times N = N^2$ operations

DFT = $O(N^2)$ $\Rightarrow$ slow if N is large

FFT: Fast Fourier Transform

- implemented as `fft(x)` in most languages
- requires 2^N data points 16, 32, 64, ...
- $O(N \log N)$

Example Reconstruction: DFT of 3 waves

$$y(t) = A \cos(2\pi f t + \phi)$$

← Sum of three waves

$$\begin{aligned} A_1 &= 1 & f_1 &= 2 \text{ Hz} & \phi_1 &= 15^\circ \\ A_2 &= 0.5 & f_2 &= 10 \text{ Hz} & \phi_2 &= 77^\circ \\ A_3 &= 2 & f_3 &= 40 \text{ Hz} & \phi_3 &= 270^\circ \end{aligned}$$

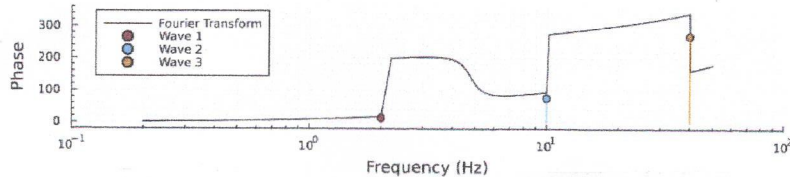
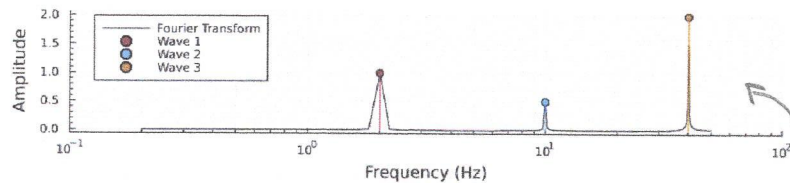
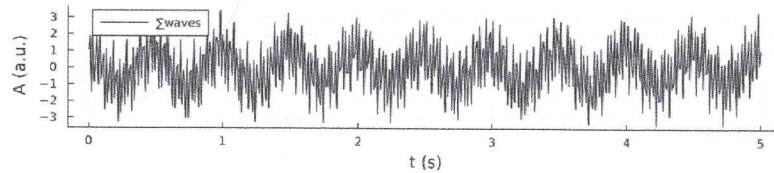
$$N = 500 \text{ points}$$

$$\Delta t = 0.01 \text{ s}$$

$$f_s = 100 \text{ Hz}$$

Amplitude Spectrum

- plot of f vs. A
- peaks recover f_i
- smeared due to resolution (leakage)
- peaks recover A_i



Phase spectrum

- f vs. ϕ
- typically noisy

Fourier Table for 3 wave example

fft(x)

	k	f	X	A	Θ
1	0	0.0	1.0784+0.0im	0.00431361	0.0
2	1	0.2	1.08794+0.0382485im	0.00435444	2.01351
3	2	0.4	1.11773+0.0785207im	0.00448195	4.01843
4	3	0.6	1.17173+0.123278im	0.0047128	6.00599
5	4	0.8	1.25807+0.176076im	0.00508134	7.96716
6	5	1.0	1.39262+0.242862im	0.00565457	9.89243
7	6	1.2	1.60812+0.33512im	0.00657065	11.7716
8	7	1.4	1.98165+0.479166im	0.00815503	13.5933
9	8	1.6	2.74408+0.753014im	0.0113821	15.345
10	9	1.8	5.03318+1.53995im	0.021054	17.0121
: more					
251	250	50.0	-6.03357+0.0im	0.0241343	180.0

$$f = k \frac{f_s}{N}$$

$$k=1 \Rightarrow f = \frac{100 \text{ Hz}}{500} = 0.2 \text{ Hz}$$

lowest resolved frequency

$$f_{Ny} = \frac{f_s}{2} = 50 \text{ Hz} \quad (k=250)$$

$$A = \frac{2}{N} \sqrt{x_k x_k^*}$$

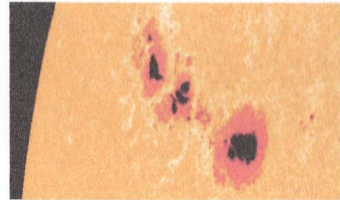
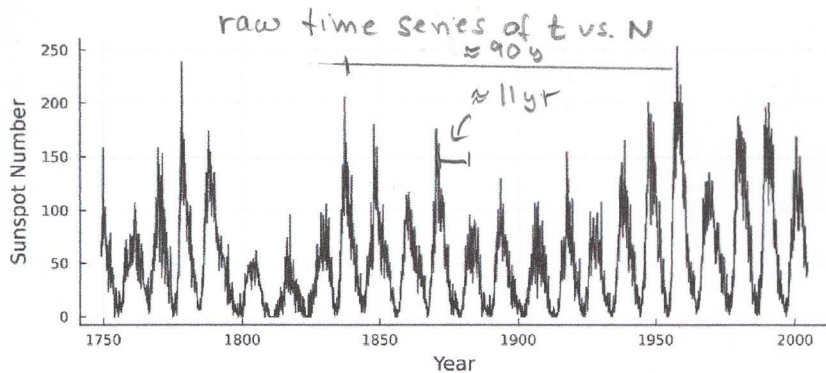
$$\phi = \text{atan} \left(\frac{\text{Im}(x_k)}{\text{Re}(x_k)} \right)$$

typically in radians

$$\hookrightarrow \Theta = \left(\phi \frac{180}{\pi} + 360 \right) \bmod 360$$

↑
modulo is
the remainder
of division

Example application: Sunspot Cycle



— check if dataset is clean (equal Δt , fix missing data)

— check resolution $\Delta t \approx 0.08 \text{ y}$

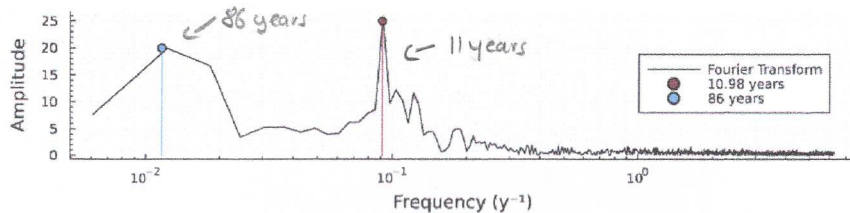
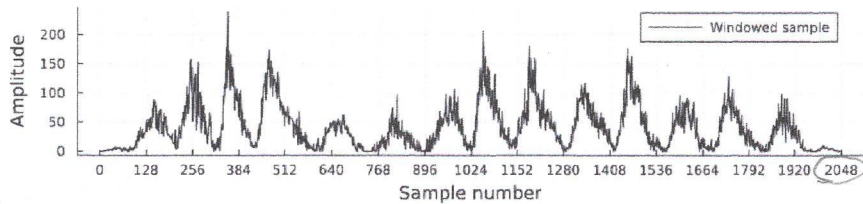
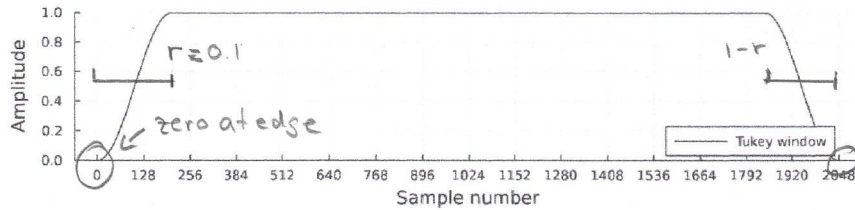
$$f = \frac{1}{\Delta t} = 12.5 \text{ y}^{-1}$$

$$f_{ny} = 6.25 \text{ y}^{-1}$$

— evaluate time series by plotting

— evaluate length of data: 3067 data points

DFT analysis



(1) FFT requires 2^n points
 $2^{11} = 2048$ $2^{12} = 4096$

record len 3067

→ pick 2048 points

(2) DFT/FFT is for N periodic functions

→ data may not be for selected subset

window function

$$w = \begin{cases} \frac{1}{2} [1 + \cos(\frac{2\pi}{r} [x - \frac{r}{2}])] & 0 \leq x < \frac{r}{2} \\ 1 & \frac{r}{2} \leq x < 1 - \frac{r}{2} \\ \frac{1}{2} [1 + \cos(\frac{2\pi}{r} [x - 1 + \frac{r}{2}])] & 1 - \frac{r}{2} \leq x < 1 \end{cases}$$

r is a fraction

data used in $\text{fft}(x)$ is $x[1:2048] \cdot w(x)$ (windowed sample)

DFT table for sunspot example

	k	f	X	A	θ
1	0	0.0	83709.7+0.0im	40.8739	0.0
2	1	0.00610352	-7833.07-89.1831im	7.64998	180.652
3	2	0.012207	-10183.3-17659.2im	19.9072	240.03
4	3	0.0183105	-15789.9+6492.85im	16.6726	157.647
5	4	0.0244141	183.586+3533.23im	3.45507	87.0256
6	5	0.0305176	-5398.21+399.728im	5.28612	175.765
7	6	0.0366211	-1474.74-5275.08im	5.34897	254.381
8	7	0.0427246	-2306.99+3858.64im	4.39033	120.874
9	8	0.0488281	83.3539+5241.94im	5.11973	89.089
10	9	0.0549316	-2809.03-2938.31im	3.96974	226.289
: more					
1025	1024	6.25	-541.718+0.0im	0.529022	180.0

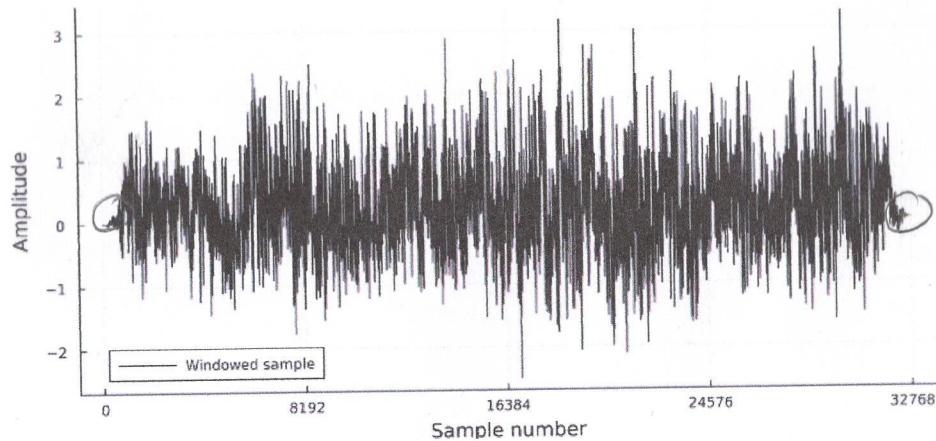
mean of subsampled and
windowed data

- check against calculating
mean directly

$$\mu = \frac{1}{N} \sum x_i$$

$$f_{Ny} = 6.25 \text{ y}^{-1}$$

DTT / FFT of turbulence data



$N = 2^{15} = 32768$ points

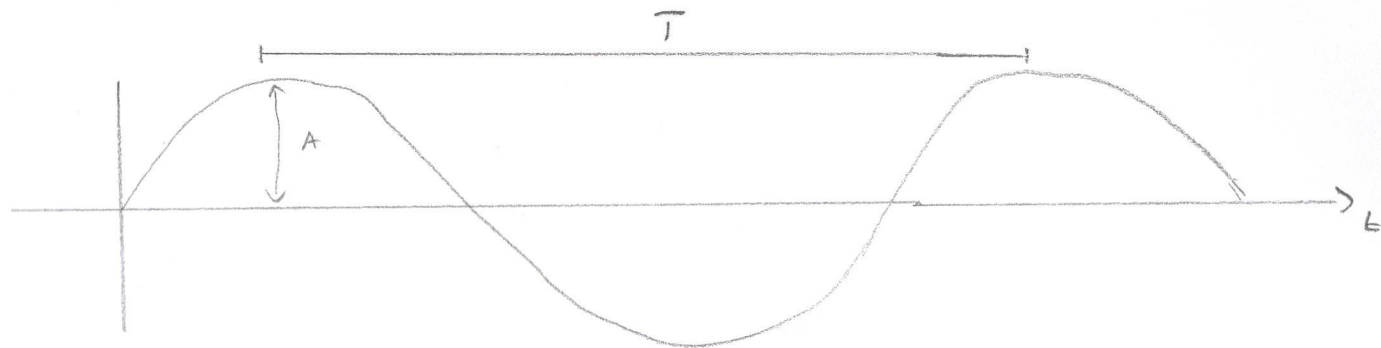
$\Delta t = 0.1s$

$f_s = 10Hz$ $f_{NB} = 5Hz$

A has units! m/s

data is windowed

Power of a wave



waves in Physics

- string wave wavelength

$$E = \frac{1}{2} \mu \omega^2 \lambda A^2$$

↑ Energy ↑ mass ↑ frequency Amplitude

- Ocean wave density

$$E = \frac{1}{8} \rho g H^2$$

↑ gravity wave height

- Light wave

$$I = \frac{1}{2} c \epsilon_0 E_0^2$$

↑ Intensity ↑ speed of light electric wave amplitude

Energy of a pulse

$$E = \int_{-\infty}^{\infty} |g(t)|^2 dt$$

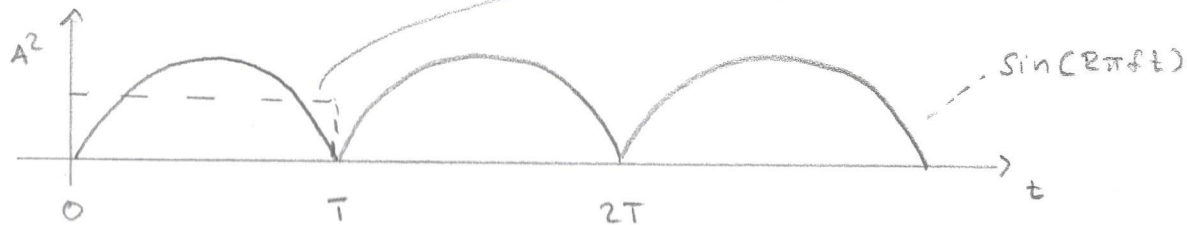
↑ arbitrary signal

$$\text{Power} = \frac{\text{Energy}}{\text{time}}$$

$$P = \frac{1}{T} \int_0^T g^2(t) dt$$

power of a cyclical signal with period T

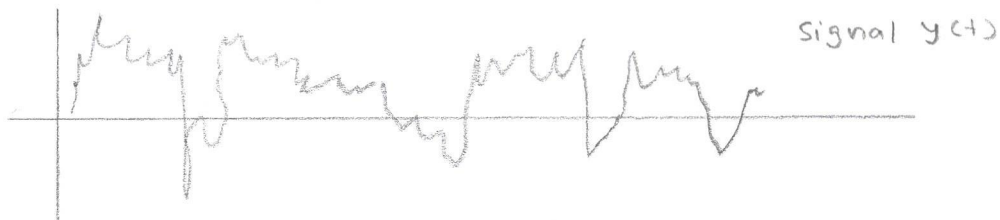
$$P = \frac{1}{T} \int_0^T [A \sin(2\pi f t)]^2 dt = \frac{1}{2} A^2 = A_{\text{avg}}^2$$



Random Process

$$P = E(x^2) = \text{variance of signal}$$

$\uparrow$
 expectation value Random variable



$$P_1 = \frac{1}{2} A^2$$



$$P_2 = \frac{1}{2} A^2$$

$$\sum_{k=1}^N P_k = \sum \frac{1}{2} A_k^2 = \text{Var}$$

$$\sum_{k=1}^{\infty} P_k = \text{Var}$$

DFT Table of vertical velocity data

A: [m/s]

P: [m²/s²]

	k	f [Hz]	X	A [m/s]	P [m ² /s ²]
				mean	
1	0	0.0 } Δf	6605.28+0.0im	0.201577	0.0812666
2	1	0.000305176	321.645+187.304im	0.0227177	0.000258047
3	2	0.000610352	516.103+887.738im	0.0626746	0.00196405
4	3	0.000915527	-644.831-49.2019im	0.0394718	0.00077901
5	4	0.0012207	168.242+209.318im	0.016391	0.000134332
6	5	0.00152588	772.533+97.5288im	0.047526	0.00112936
7	6	0.00183105	-1675.72-1007.32im	0.119335	0.00712037
8	7	0.00213623	-1114.94-98.3688im	0.0683146	0.00233344
9	8	0.00244141	9.2303+988.143im	0.0603141	0.00181889
10	9	0.00274658	505.846-364.321im	0.0380484	0.000723842
: more					
16385	16384	5.0	-9.86553+0.0im	0.000602144	1.81289e-7

$$P_k = \frac{1}{2} A_k^2$$

$$\sum_{k=1}^{N/2} P_k = \text{Var}(y(t))$$

always check

mean, var from data

=== mean, var from DFT

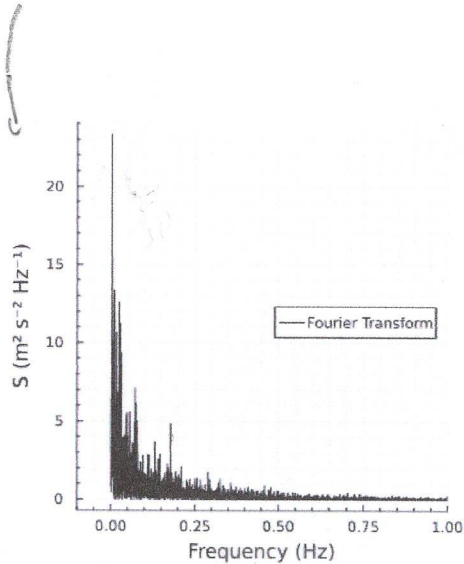
$$S = \frac{P}{\Delta f} \quad \left[\frac{\text{m}^2}{\text{s}^2 \text{ Hz}} \right]$$

↑
power spectral density

$$\text{Var} = \int S df$$

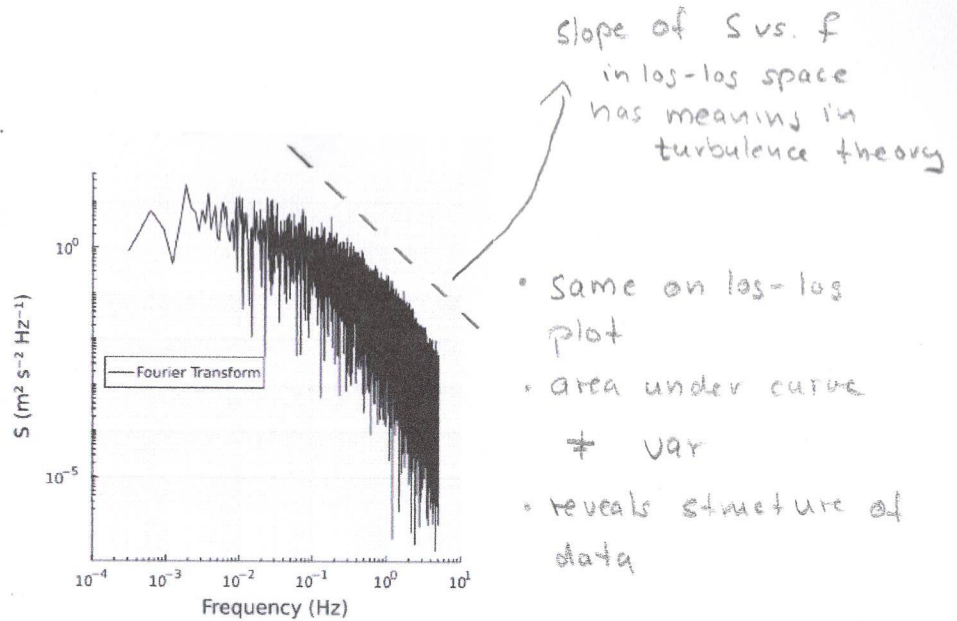
Power Spectrum

Power spectral density



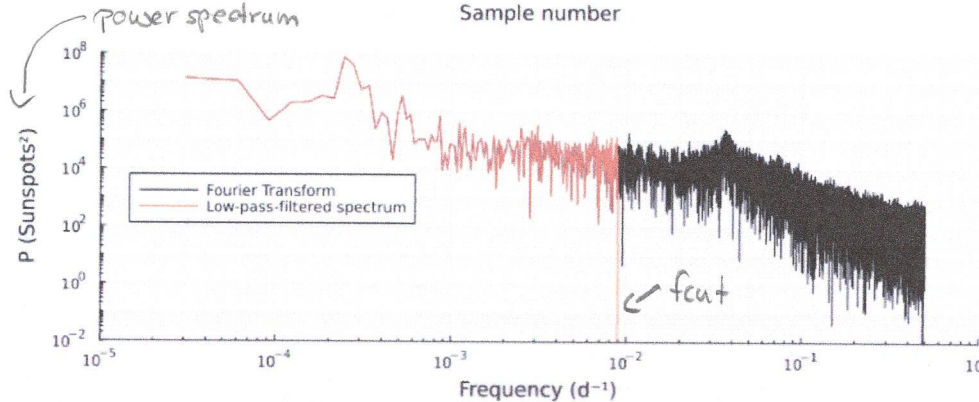
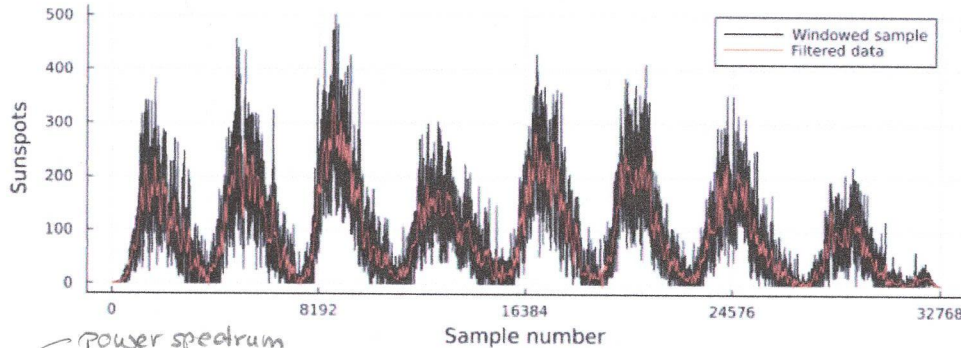
plot of S vs. f
reveals spectral contribution
to variance

$$\text{var} = \int S df$$



Application: Frequency Filtering of Data

(High frequency sunspot data)



(1) window data y_n

$$(2) X_n = \sum_{n=0}^{N-1} x_n \exp(-ix)$$

$$(3) P_n = \frac{1}{2} X_n X_n^*$$

(4) remove unwanted frequencies

low-pass: zero X_n for $f < f_{cut}$

must zero also X_n^* !

$$(5) x_n = \frac{1}{N} \sum_{n=0}^{N-1} X_n \exp(-ix)$$

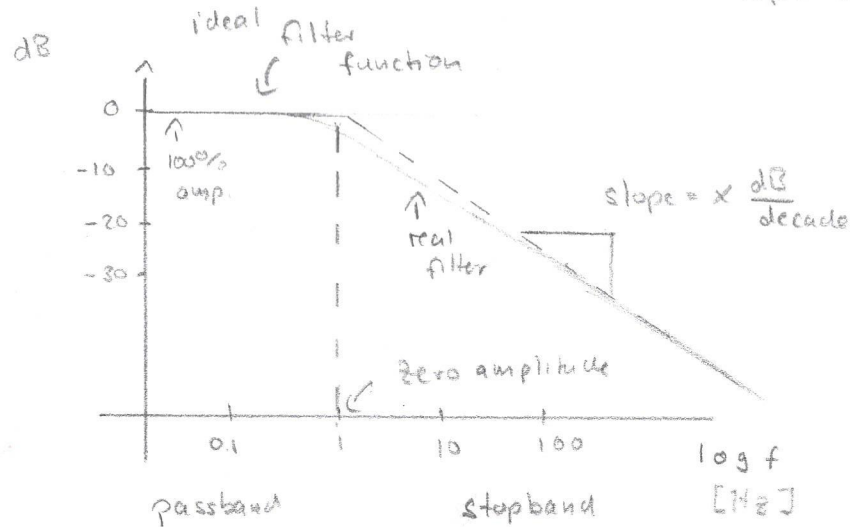
[use ifft]

Filtering Continued

$$\text{Decibel} = 10 \log_{10} (P_i / P_o)$$

gain/loss in power relative
to reference power P_o

note $-10 \text{ dB} =$ order of
magnitude
reduction
in Power



Fourier Transform of Functions

Discrete

$$X_n = \sum_{n=0}^{N-1} x_n \exp(-i 2\pi \frac{k}{N} n)$$

n data $\Rightarrow n$ Fourier coefficient

N : period of data

Fourier Series

$$\bar{f}(k) = \sum_{t=-\infty}^{\infty} f(t) \exp(-i 2\pi \frac{k}{N} t)$$

continuous $f(t) \Rightarrow$ infinite Fourier coefficients

N -periodic function

Fourier Integral: $\lim_{N \rightarrow \infty} \frac{k}{N} = \nu$ "oscillation frequency"

$$\bar{f}(\nu) = \int_{-\infty}^{\infty} f(t) \exp(-i 2\pi \nu t) dt$$

continuous $f(t) \Rightarrow$ continuous ν
non-periodic, converges
to zero flux

$$f(t) = \int_{-\infty}^{\infty} \bar{f}(\nu) \exp(i 2\pi \nu t) d\nu$$

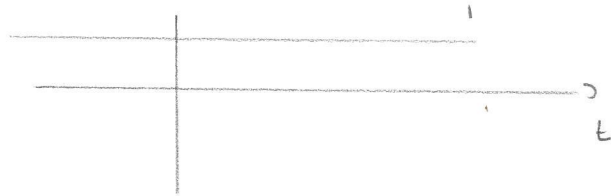
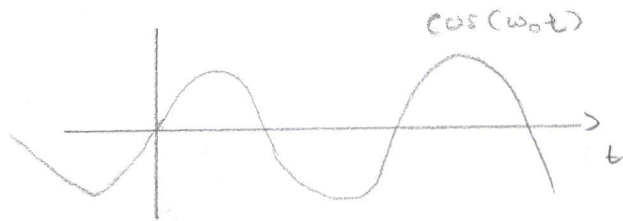
$\omega = 2\pi \nu$ "angular frequency"

$$\bar{f}(\omega) = \int_{-\infty}^{\infty} f(t) \exp(-i \omega t) dt$$

$$f(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \bar{f}(\omega) \exp(i \omega t) d\omega$$

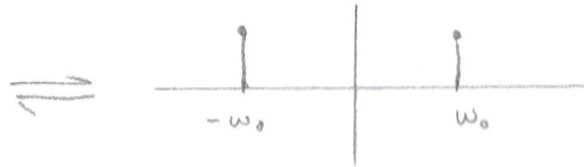
Example Transforms

$f(t)$



t -domain

$F(\omega)$



$$\delta(t) = \begin{cases} 0 & t \neq 0 \\ \infty & t = 0 \end{cases}$$

$$\int_{-\infty}^{\infty} \delta(t) dt = 1$$

ω -domain

Applications of Fourier Integral

(1) derivatives become algebraic functions

Fourier
transform
of

$$\longrightarrow \mathcal{F}(f'(t)) = -i\omega \mathcal{F}(f) = -i\omega \bar{F}(\omega)$$

(2) convolutions become multiplication

$$\mathcal{F}(f * g) = \mathcal{F}(f) \cdot \mathcal{F}(g)$$

$$f * g = \int f(\tilde{t}) g(t - \tilde{t}) d\tilde{t} \quad \Rightarrow \text{session 3}$$

(3) linear combination

$$\mathcal{F}(a f(t) + b g(t)) = a \bar{F}(k) + b \bar{G}(k)$$

Laplace Transform

Fourier $\mathcal{F}(f(t)) = \int_{-\infty}^{\infty} f(t) \exp(-i\omega t) dt$

Laplace $\mathcal{L}(f(t)) = \int_{-\infty}^{\infty} f(t) \exp(-st) dt$

$s = a + bi \Rightarrow$ Fourier transform is special case of Laplace transform

more functions have Laplace transform $\rightarrow$ need not converge at ∞



Properties of Fourier transform apply to Laplace transform