

1. Introduction to the Fourier Transform

Review: Complex Numbers

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

$$i^2 = -1$$

$$z = a + bi$$

Real Imaginary Part

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

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

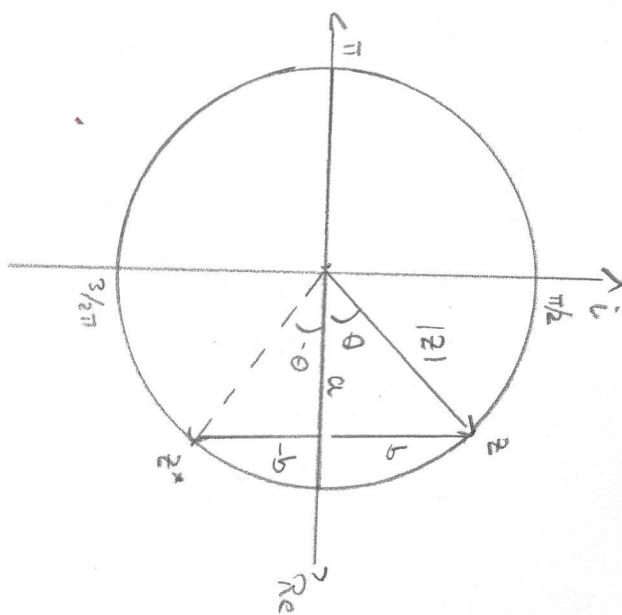
$$a = \cos(\theta)$$

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

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

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

\uparrow
 \tan^{-1}



Euler's Identity

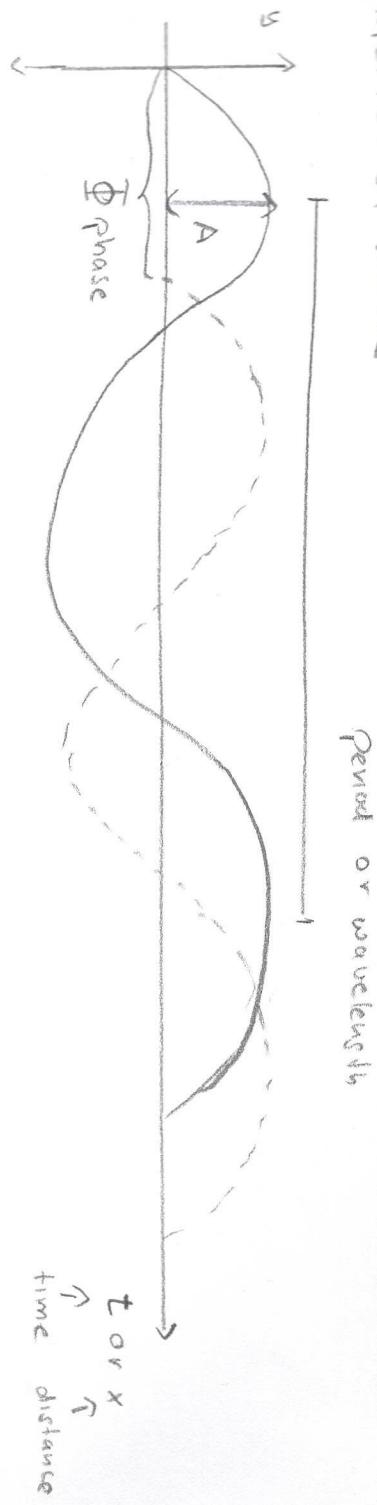
$$\exp(i\theta) = \cos(\theta) + i\sin(\theta)$$

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

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

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

Properties of waves



A: Amplitude

T: Period [s]

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

λ : wavelength [m]

v: wavelength $\frac{1}{\lambda}$ [$\frac{1}{m}$]

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

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

Sum of angles

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

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

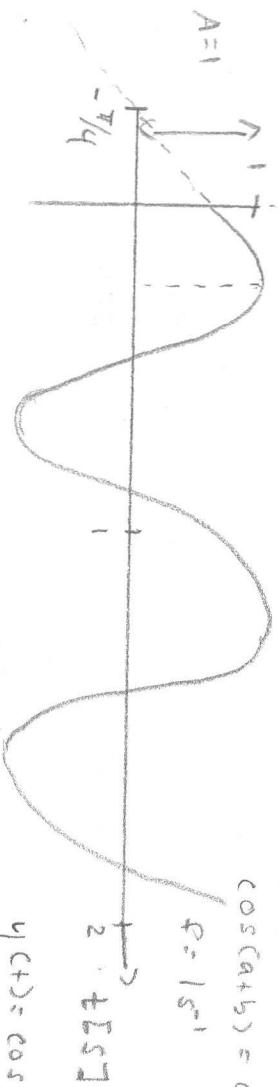
$$y(t) = A \cos(2\pi f t) \cos(\phi) - A \sin(2\pi f t) \sin(\phi)$$

$$y(t) = a \cos(\omega t) + b \sin(\omega t)$$

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

$$= -\Delta \beta_1 S(\Phi)$$

X = 1 + 2 + 3



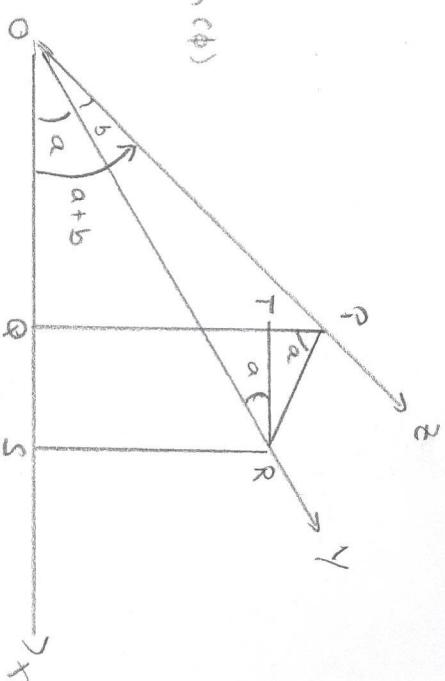
$$\cos(a+b) = \cos(a)\cos(b) - \sin(a)\sin(b)$$

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$$v_1(t) = \cos(2\pi t - \pi/4)$$

$$\begin{aligned}y(4) &= A \cos(\pi/4) \cos(2\pi t) \\&= A \sin(\pi/4) \sin(2\pi f t)\end{aligned}$$



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

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

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

$$f = 1 Hz$$

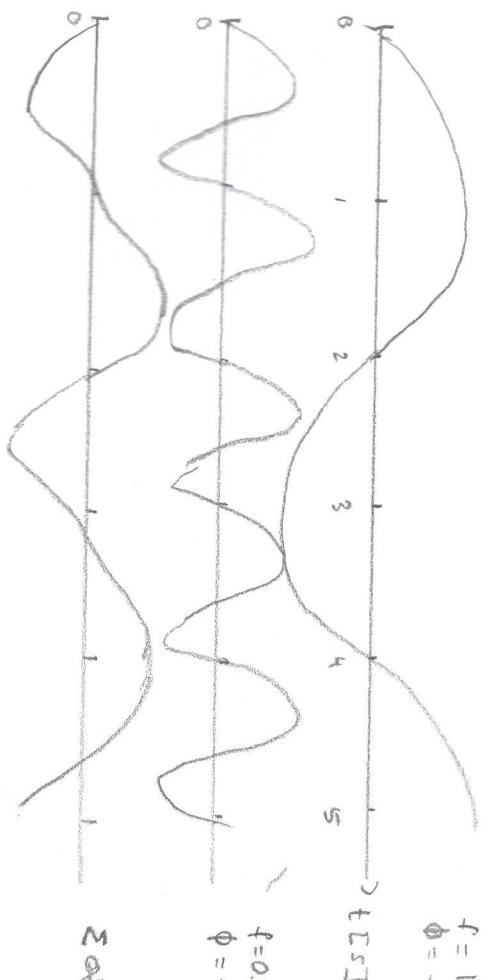
$$\phi = -90^\circ$$

N : Period

$$f = 0.5 Hz$$

$$\frac{k}{N} : \text{frequency}$$

$$\phi = -90^\circ$$



Σ of waves \Rightarrow mean zero
• periodic function

" represent periodic function with period $\frac{1}{N}$ using Fourier series "

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

$$\frac{1}{N} \left(\frac{100}{5s} \right) \text{ highest frequency}$$

Fourier Series in the Complex Domain

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

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

$$\begin{aligned} \exp(ix) &= \cos(x) + i \sin(x) \\ \exp(-ix) &= \cos(x) - i \sin(x) \end{aligned}$$

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

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

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

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

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

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

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

$$y(x) = \sum_{-\infty}^{\infty} X_k \exp(-ix)$$

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

each real frequency (e.g. k=1)
has two X coefficients

Discrete Fourier transform: how to find X_k ?

Fourier Series

$$y(t) = \sum_{n=0}^{\infty} X_n \exp(-inx)$$

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

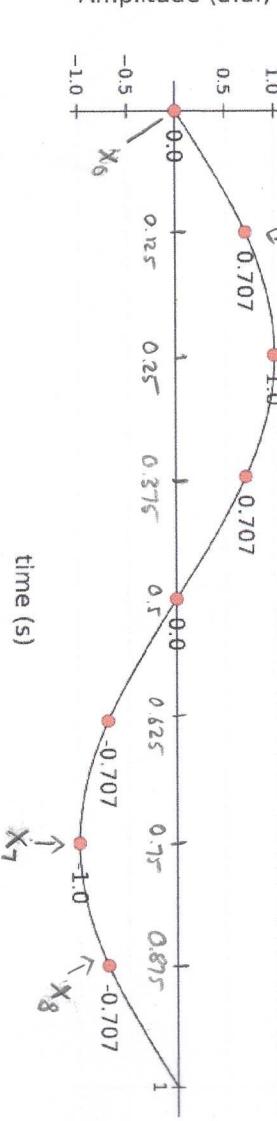
Infinite series representing an N-periodic function in the time domain \Rightarrow infinite X_k

Discrete Fourier Series

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

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

Finite sequence of N equally spaced samples of a function $y(t)$



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

$f = 8$ Hz "sample frequency"

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

$N = 8$ points

$$y_0 = 0$$

$$y_2 = 1.0$$

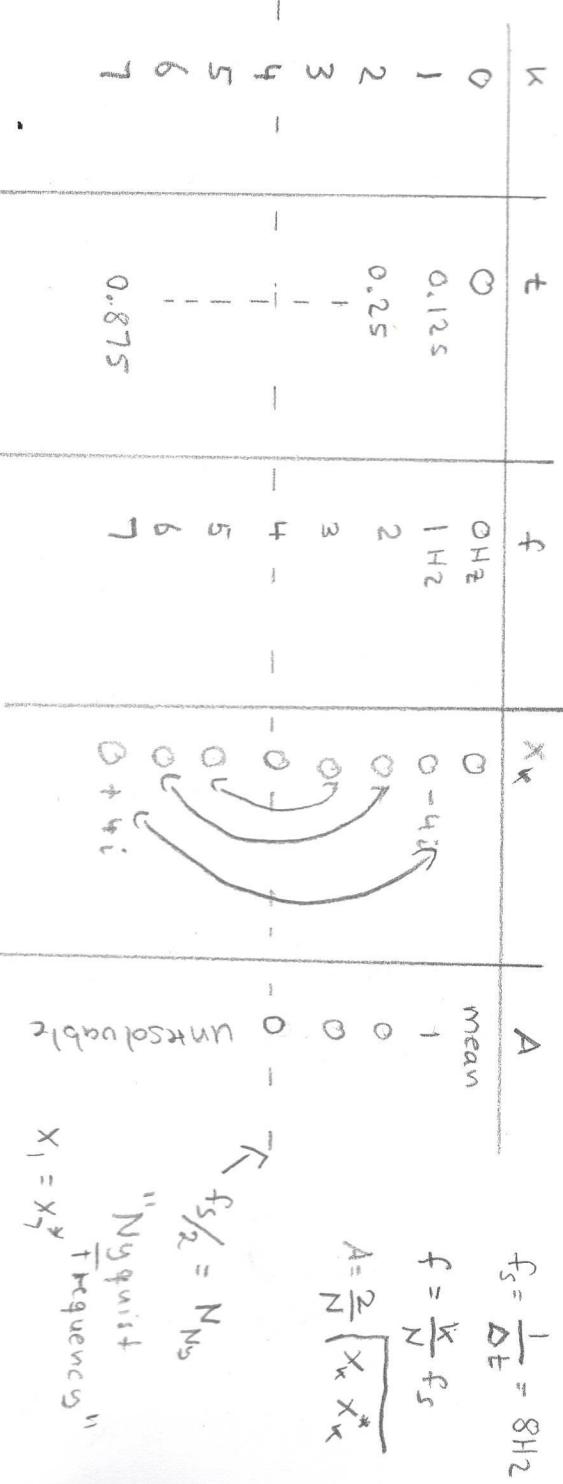
How to find X_k ?

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

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

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

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



$$\nearrow f_s/2 = N_{\text{Ns}}$$

"Nyquist Frequency"

$$X_1 = X_7^*$$

How to find X_k ?

$$X_k = \sum_{n=0}^{N-1} x_n \exp(-ix)$$
 each X_k requires N additions

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

$\mathcal{O}(N^2) = O(N^2)$ \Rightarrow slow if N is large

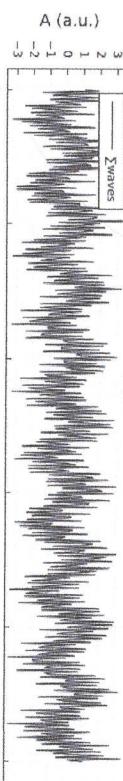
$\overline{\text{FFT}}$: Fast Fourier Transform

- implemented as $\text{fft}(x)$ in most languages
- requires 2^N data points
 $16, 32, 64, \dots, 2^m$
- $O(N \log N)$

Example Reconstruction: Sum of 3 waves

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

Sum of three waves



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

$N = 500$ 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

$\text{fft}(x)$

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

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+0.0785267im	0.00448195 4.01843
4	3	0.6	1.17173+0.123278im	0.0047128 6.00599
5	4	0.8	1.25802+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	0.00815563 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

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

lowest resolved frequency

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

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

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

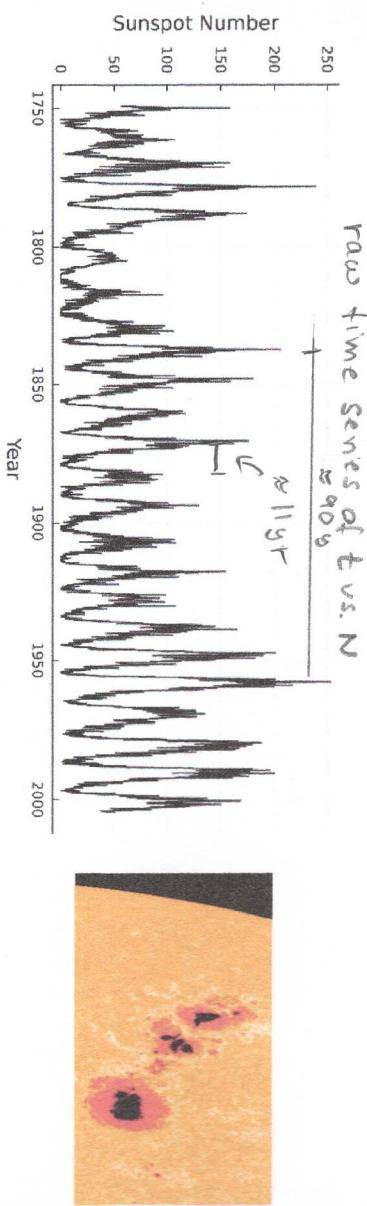
typically in radians

$$\Rightarrow \theta = \left(\phi - \frac{180}{\pi} + 360 \right) \bmod 360$$

modulo is

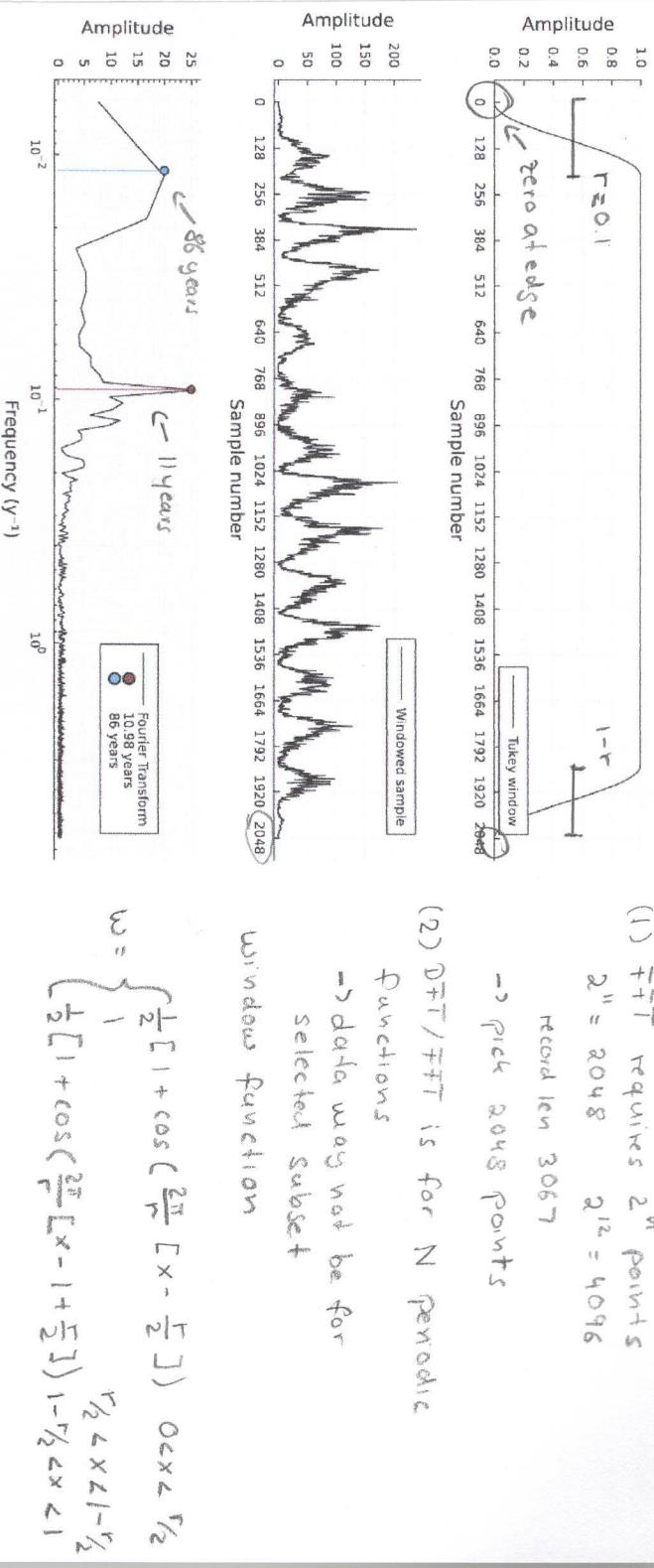
the remainder
of division

Example application: Sunspot Cycle



- check if dataset is clean (equal Δt , fix missing data)
- check resolution $\Delta t \approx 0.08\text{y}$
- $f = \frac{1}{\Delta t} = 12.5\text{ y}^{-1}$
- $f_{\mu_0} = 6.25\text{ y}^{-1}$
- evaluate time series by plotting
- evaluate length of data: 3067 data points

DFT analysis



data used in `fft(x)` is `x[1:2048].w(x)` (windowed sample)

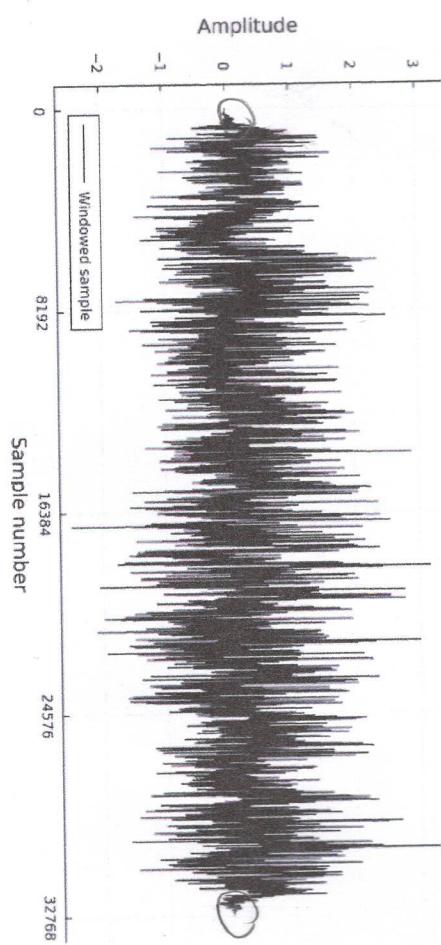
Γ is a fraction

DTT table for sunspot example

k	f	x	A	Θ
1	0	0.0	83709.7+0.0im	<u>40.8739</u> 0.0
2	1	0.00610352	-7833.07-89.1831im	7.64998
3	2	0.012207	-10183.3-17659.2im	19.9072
4	3	0.0183105	-15789.9+6492.85im	16.6726
5	4	0.0244141	183.586-3533.23im	157.647
6	5	0.0305176	-5398.21+399.728im	3.45507
7	6	0.0366211	-1474.74-5275.08im	87.0256
8	7	0.0427246	-2306.99+3858.64im	5.28612
9	8	0.0488281	83.3539+5241.94im	175.765
10	9	0.0549316	-2809.03-2938.31im	5.34897
				$\mu = \frac{1}{N} \sum x_i$
				mean of subsampled and windwed data
				- check against calculation
				mean direction
				$\mu = \frac{1}{N} \sum x_i$
				more
1025	(1024) 6.25	-541.718+0.0im	0.529022	180.0

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

DFT of turbulence data



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

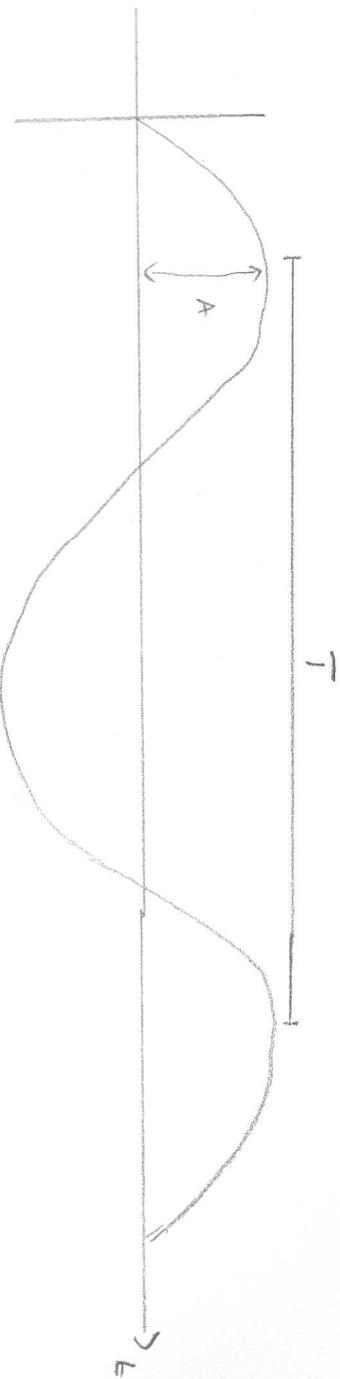
$$\Delta t = 0.1 \text{ s}$$

$$f_s = 10 \text{ Hz} \quad f_{\text{Nyq}} = 5 \text{ Hz}$$

$$A \text{ has units! } \text{m/s}$$

data is windowed

Power of a wave



waves in Physics

- string wave

wavelength

$$\bar{E} = \frac{1}{2} \mu w z \pi A^2$$

\uparrow mass frequency

Energy

\uparrow Amplitude

- Ocean wave

wave density

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

\uparrow gravity

wave height

- Light wave

$$I = \frac{1}{2} c \epsilon_0 \bar{E}_0^2$$

\uparrow electric wave amplitude

\uparrow speed of light

Intensity

Energy of a pulse

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

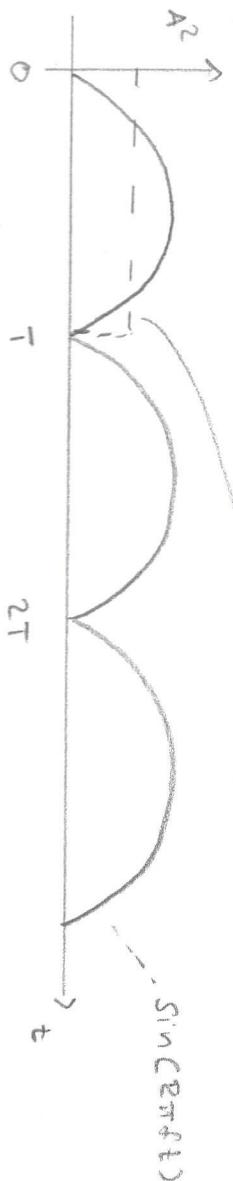
arbitrary signal

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

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

power of a cyclical signal with period T

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

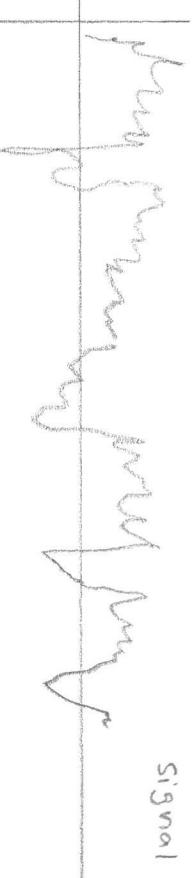


Random Process

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

↑
expectation Random
value

Signal $y(t)$



$$\hat{P}_1 = \frac{1}{2} A^2$$

$$\hat{P}_2 = \frac{1}{2} A^2$$

$$\sum_{k=1}^N P_k = \sum_{i=1}^{n/2} \frac{1}{2} A_i^2 = \text{Var}$$
$$\sum_{k=1}^N P_k = \text{Var}$$

DFT Table of vertical velocity data.

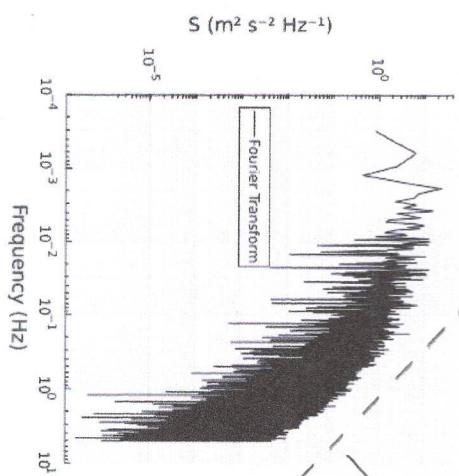
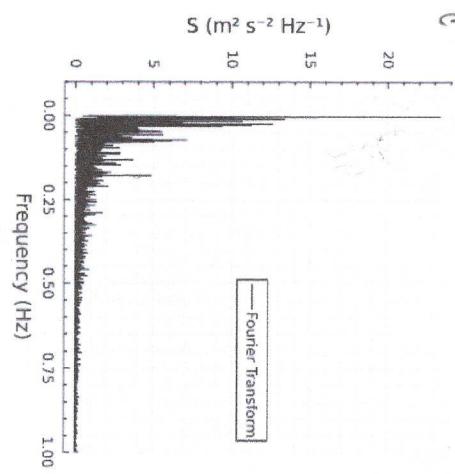
$$A: [m/s]$$

$$\rho: [m^2/s^2]$$

k	f [Hz]	X	A [m/s]	P [m²/s²]	$\hat{P}_k = \frac{1}{2} A_k^2$
meas	0.201577	0.0812666	$\sum_{k=1}^{N/2} \hat{P}_k = \text{Var}(y(t))$		
1	0.0	0.0 Δf	6605.28+0.0im	0.0012666	
2	1	0.0003050176	321.645+187.304im	0.0227177	
3	2	0.000610352	516.103+887.738im	0.0626746	
4	3	0.000915527	-644.831-49.2019im	0.0394718	
5	4	0.0012207	168.242+209.318im	0.016391	
6	5	0.00152588	772.533+97.5288im	0.047526	
7	6	0.00183105	-1675.72-1007.32im	0.119335	
8	7	0.00213623	-1114.94-98.3688im	0.0683146	
9	8	0.00244141	9.2303+988.143im	0.0603141	
10	9	0.00274658	505.846-364.321im	0.0380484	
more					
16385	16384	5.0	-9.86553+0.0im	0.000602144	$\text{Var} = \int S df$
					$S = \frac{\hat{P}}{\Delta f} \quad [\frac{m^2}{s^2 \cdot Hz}]$
					Power spectral density

Power Spectrum

Power spectral density



- Slope of S vs. f in log-log space has meaning in turbulence theory
- Same on log-log plot
- Area under curve \neq Var
- Reveals structure of data

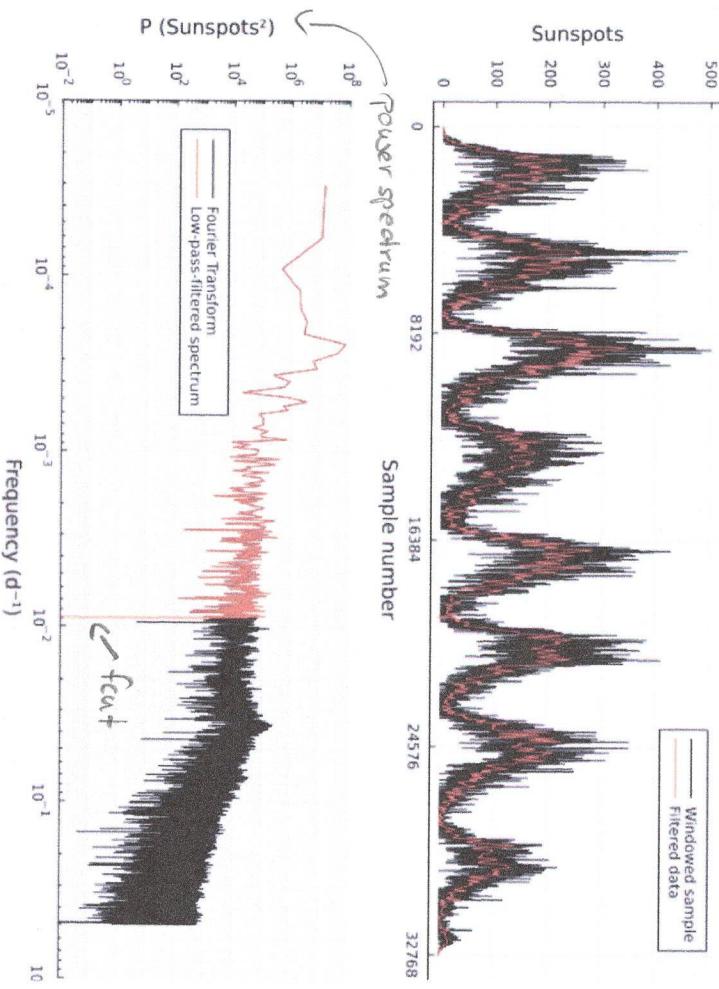
Plot of S vs. f

reveals spectral combination to variance

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

Application: Frequency Filtering of Data

(High-frequency sunspot data)



(1) window data y_n

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

$$(3) \hat{r}_k = \frac{1}{2} X_k X_k^*$$

(4) remove unwanted frequencies

low-pass: zero X_k for

$$f < f_{cut}$$

must zero also X_k^* !

$$(5) \hat{x}_n = \frac{1}{N} \sum_{n=0}^{N-1} X_k \exp(-inx)$$

[use ifft]

Filtrering continued

$$\text{Decibel} = 10 \log_{10} \left(\frac{P_o}{P_0} \right)$$

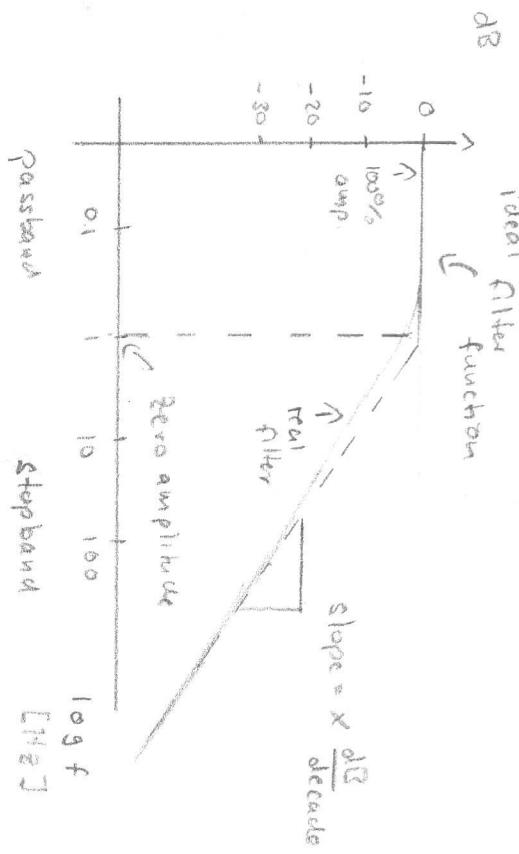
gain / loss in power relative
to reference Power P_0

Note - $10 \text{dB} =$ Order of

magnitude

reduction

in Power



Fourier Transform of Functions

Discrete

$$X_k = \sum_{n=0}^{N-1} x_n \exp(-i 2\pi \frac{k}{N} n) \quad n \text{ data} \rightarrow n \text{ Fourier coefficients}$$

N : Period of data

Fourier Series

$$\tilde{f}(w) = \sum_{t=-\infty}^{\infty} f(t) \exp(-i 2\pi \frac{w}{N} t) \quad \text{continuous } f(t) \Rightarrow \text{infinite Fourier coefficients}$$

N -periodic function

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

$$\tilde{f}(w) = \int_{-\infty}^{\infty} f(t) \exp(-i 2\pi w t) dt \quad \text{continuous } f(t) \Rightarrow \text{continuous } w$$

 $f(t) = \int_{-\infty}^{\infty} \tilde{f}(w) \exp(i 2\pi w t) dw \quad \text{to zero if } f(t)$

$w = 2\pi v$ "angular frequency"

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

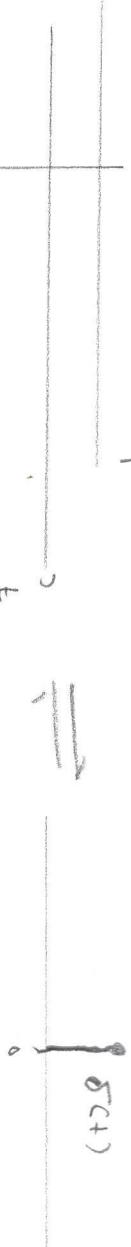
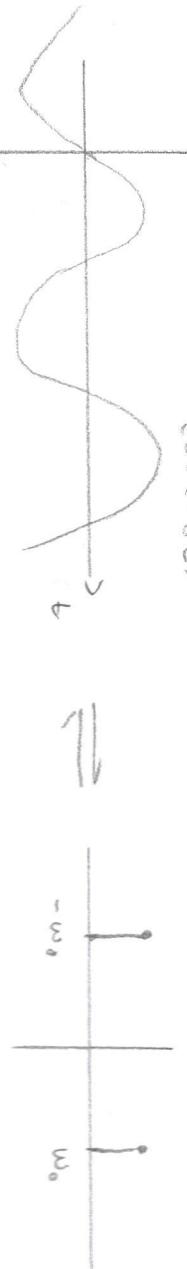
$$f(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \tilde{f}(w) \exp(i \omega t) dw$$

Example transforms

FTs

$\cos(\omega_0 t)$

$F(\omega)$



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

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

t - domain

ω - domain

Applications of Fourier Integral

(1) derivatives become algebraic functions

$$\text{Fourier Transform} \rightarrow \mathcal{F}(f'(c_+)) = -i\omega \mathcal{F}(w)$$

(2) convolution becomes multiplication

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

$$f \circ g = \int f(c) g(t^{-1}) \, dc \Rightarrow \text{separable}$$

(3) linear combination

$$\mathcal{F}(a f(c_+) + b g(c_+)) = a \mathcal{F}(w) + b \mathcal{G}(w)$$

Laplace Transform

Fourier

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

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

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

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

t-domain



s-domain



Properties of Fourier transform apply to Laplace transform