

# Computational Fourier Analysis

## Mathematics, Computing and Nonlinear Oscillations

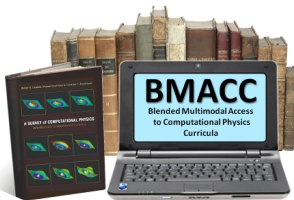
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Based on *A Survey of Computational Physics* by Landau, Páez, & Bordeianu

with Support from the National Science Foundation

Course: **Computational Physics II**



# Outline

# Applied Math: Approximate Fourier Integral

## Numerical Integration

$$\text{Transform } Y(\omega) = \int_{-\infty}^{+\infty} dt \, y(t) \frac{e^{-i\omega t}}{\sqrt{2\pi}} \quad (1)$$

$$\simeq \sum_{i=0}^N h y(t_i) \frac{e^{-i\omega t_i}}{\sqrt{2\pi}} \quad (2)$$

- Approximate Fourier integral  $\rightarrow$  finite Fourier series
- Consequences to follow

# Experimental Constraints Too!

Transform, Spectral: 
$$Y(\omega) = \int_{-\infty}^{+\infty} dt y(t) \frac{e^{-i\omega t}}{\sqrt{2\pi}} \quad (3)$$

Inverse, Synthesis: 
$$y(t) = \int_{-\infty}^{+\infty} d\omega Y(\omega) \frac{e^{+i\omega t}}{\sqrt{2\pi}} \quad (4)$$

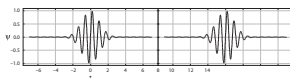
## Real World: Data Restrict Us

- Measured  $y(t)$  only @  $N$  times ( $t_i$ 's)
- Discrete **not** continuous & **not**  $-\infty \leq t \leq +\infty$
- Can't measure enough data to determine  $Y(\omega)$
- The **inverse problem** with incomplete data
- DFT: one possible solution

# Algorithm with Discrete & Finite Times

## Measure: $N$ signal values

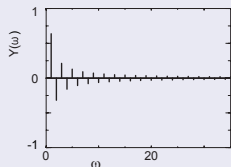
- Uniform time steps  $\Delta t = h$ ,  $t_k = kh$
- $y_k = y(t_k)$ ,  $k = 0, 1, \dots, N$
- **Finite**  $T \Rightarrow$  ambiguity
- Integrate over all  $t$ ;  $y(t < 0)$ ,  $y(t > T) = ?$
- Assume periodicity  $y(t + T) = y(t)$   
(removes ambiguity)
- $\Rightarrow Y(\omega)$  at  $N$  discrete  $\omega_j$ 's
- $\Rightarrow y_0 \equiv y_N$  repeats!
- $\Rightarrow N + 1$  values,  $N$  independent



# “A” Solution to Indeterminant Problem

Discrete  $y(t_i)$ s  $\Rightarrow$  Discrete  $\omega_j$

- $N$  independent  $y(t_i)$  measured
- $\Rightarrow N$  independent  $Y(\omega_j)$
- $\omega_j = ?$
- Total time  $T = Nh \Rightarrow \min \omega_1$ :



$$\omega_1 = \frac{2\pi}{T} = \frac{2\pi}{Nh} \quad (5)$$

- Only  $N$   $\omega_j$ s (+1)

$$\omega_k = k\omega_1, \quad k = 0, 1, \dots, N \quad (6)$$

$$\omega_0 = 0 \times \omega_1 = 0 \quad (\text{DC}) \quad (7)$$

# Algorithm: Discrete Fourier Transform (DFT)

Trapezoid Rule:  $\int f(t)dt \simeq \sum hf(t_i)$

$$Y(\omega_n) = \int_{-\infty}^{+\infty} dt \frac{e^{-i\omega_n t}}{\sqrt{2\pi}} y(t) \simeq \int_0^T dt \frac{e^{-i\omega_n t}}{\sqrt{2\pi}} y(t) \quad (8)$$

$$\simeq \sum_{k=1}^N h y(t_k) \frac{e^{-i\omega_n t_k}}{\sqrt{2\pi}} \quad (9)$$

Symmetrize notation, substitute  $\omega_n$ :

$$Y_n = \frac{1}{h} Y(\omega_n) = \sum_{k=1}^N y_k \frac{e^{-2\pi i k n / N}}{\sqrt{2\pi}} \quad (10)$$

# Discrete Inverse Transform (Function Synthesis)

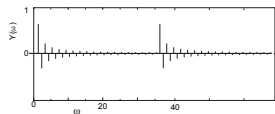
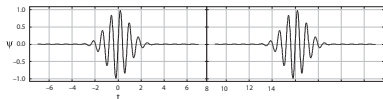
## Trapezoid Rule for Inverse

- Frequency Step  $\Delta\omega = \omega_1 = \frac{2\pi}{T} = \frac{2\pi}{Nh}$

$$y(t) = \int_{-\infty}^{+\infty} d\omega \frac{e^{i\omega t}}{\sqrt{2\pi}} Y(\omega) \simeq \sum_{n=1}^N \frac{2\pi}{Nh} \frac{e^{i\omega_n t}}{\sqrt{2\pi}} Y(\omega_n) \quad (11)$$

- Trig functions  $\Rightarrow$  periodicity for  $y$  and  $Y$ :

$$y(t_{k+N}) = y(t_k), \quad Y(\omega_{n+N}) = Y(\omega_n) \quad (12)$$

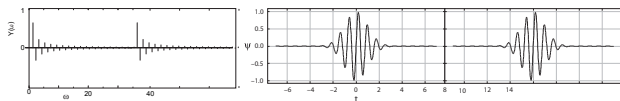




# Consequences of Discreteness

$$Y(\omega_n) \simeq \sum_{k=1}^N h y(t_k) \frac{e^{-i\omega_n t_k}}{\sqrt{2\pi}}$$

$$y(t) \simeq \sum_{n=1}^N \frac{2\pi}{Nh} \frac{e^{i\omega_n t}}{\sqrt{2\pi}} Y(\omega_n)$$



- $\Delta\omega = \omega_1 = \frac{2\pi}{T} = \frac{2\pi}{Nh}$
- Finer  $\omega \Leftrightarrow$  larger  $T = Nh$
- Finer  $\omega \rightarrow$  smoother  $Y(\omega)$
- “Pad”  $y(t) \Rightarrow$  smoother  $Y(\omega)$

- **Ethical question**
- Synthetic  $y(t) =$  bad  $t \rightarrow T$
- Periodicity  $\downarrow$  as  $T \rightarrow \infty$
- Aliases and ghosts:  
special lecture

# Concise & Efficient DFT Computation

## Compute 1 Complex number

$$Z = e^{-2\pi i/N} = \cos \frac{2\pi}{N} - i \sin \frac{2\pi}{N} \quad (13)$$

$$Y_n = \frac{1}{\sqrt{2\pi}} \sum_{k=1}^N Z^{nk} y_k \quad (\text{Transform}) \quad (14)$$

$$y_k = \frac{\sqrt{2\pi}}{N} \sum_{n=1}^N Z^{-nk} Y_n \quad (\text{Synthesis, TF}^{-1}) \quad (15)$$

- $Z$  rotates  $y$  into **complex**  $Y$  and *visa versa*
- Compute only powers of  $Z$  (basis of FFT)

$$Z^{nk} \equiv (Z^n)^k = \cos \frac{2\pi nk}{N} - i \sin \frac{2\pi nk}{wN} \quad (16)$$

# DFT Fourier Series: $y(t) = \sum_{n=0}^{\infty} a_n \cos(n\omega t) \simeq ?$

Discrete: Sample  $y(t)$  @  $N$  times

$$y(t = t_k) \equiv y_k, \quad k = 0, 1, \dots, N \quad (17)$$

- Repeat period  $T = Nh \Rightarrow y_0 = y_N$
- $\Rightarrow N$  independent  $a_n$ s
- Use trapezoid rule,  $\omega_1 = 2\pi/Nh$ :

$$a_n = \frac{2}{T} \int_0^T dt \cos(k\omega t) y(t) \simeq \frac{2}{N} \sum_{k=1}^N \cos(nk\omega_1) y_k \quad (18)$$

- Truncate sum:

$$y(t) \simeq \sum_{n=1}^N a_n \cos(n\omega_1 t) \quad (19)$$

# Code Implementation: DFTcomplex.py

## Example

```
N = 1000;
twopi = 2.*math.pi;
sq2pi = 1./math.sqrt(twopi);

def fourier(dftz):
    for n in range(0, N):
        zsum = complex(0.0, 0.0)
        for k in range(0, N):
            zexpo = complex(0, twopi*k*n/N)
            zsum += signal[k]*exp(-zexpo)
        dftz[n] = zsum * sq2pi
```

# Assessment: Test Where know Answers

## Simple Analytic Cases

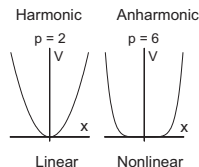
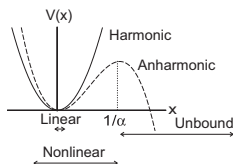
$$\text{e.g. } y(t) = 3 \cos(\omega t) + 2 \cos(3\omega t) + \cos(5\omega t) \quad (20)$$

- 1 Known:  $Y_1 : Y_2 : Y_3 = 3 : 2 : 1$  (9 : 4 : 1 power spectrum)
- 2 Resum to input signal? (> graph, idea of error)
- 3 Effect of: time step  $h$ , period  $T = Nh$
- 4 Sample mixed signal:

$$y(t) = 4 + 5 \sin(\omega t + 7) + 2 \cos(3\omega t) + \sin(5\omega t) \quad (21)$$

- 5 Effects of varying 4 & 7?

# Physics Assessment



## Determine Spectrum and Check Inversion

### 1 Nonlinearly perturbed oscillator:

$$V(x) = \frac{1}{2}kx^2 \left(1 - \frac{2}{3}\alpha x\right) \quad (1)$$

### 2 Determine when $> 10\%$ higher harmonics ( $b_{n>1} \geq 10\%$ )

### 3 Highly nonlinear oscillator:

$$V(x) = kx^{12} \quad (2)$$

### 4 Compare to sawtooth.

# Summary

- Represent periodic or nonperiodic functions with DFT.
- Finiteness of measurements  $\rightarrow$  ambiguities ( $T$ )
- Infinite series or integral not practical algorithm or in experiment.
- Approximate integration  $\rightarrow$  simplicity & approximations
- Better high frequency components: smaller  $h$ , same  $T$ .
- Smoother transform: larger  $T$ , same  $h$  (padding).
- Less periodicity: more measurements.
- DFT is simple, elegant and powerful.
- Rotation between signal and transform space.
- $(e^{i\phi})^n \rightarrow$  Fast Fourier Transform.