

COM303: Digital Signal Processing

Lecture 5: The Discrete Fourier Transform

Overview

- ▶ the Fourier basis for \mathbb{C}^N (recap)
- ▶ the DFT: definition and examples
- ▶ interpreting a DFT plot

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The canonical (time) basis for \mathbb{C}^N

- ▶ in "signal" notation: $\delta_k[n] = \delta[n-k],$ n, k = 0, 1, ..., N-1
- ▶ in vector notation: $\{\delta^{(k)}\}_{k=0,1,...,N-1}$ with $\delta^{(k)}_n = \delta[n-k]$

The canonical (time) basis for \mathbb{C}^N

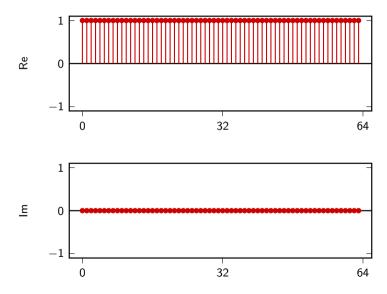
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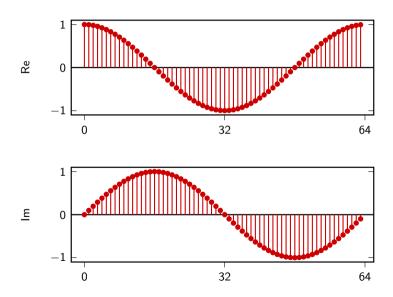
The Fourier (frequency) basis for \mathbb{C}^N

- ightharpoonup in "signal" notation: $w_k[n] = e^{j\frac{2\pi}{N}nk}, \qquad n,k=0,1,\ldots,N-1$
- ▶ in vector notation: $\{\mathbf{w}^{(k)}\}_{k=0,1,\dots,N-1}$ with $w_n^{(k)}=e^{j\frac{2\pi}{N}nk}$

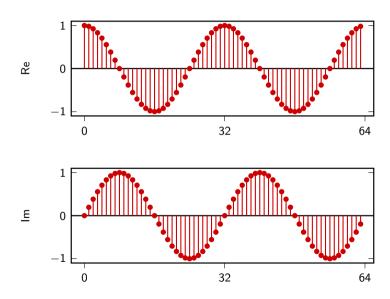
The Fourier (frequency) basis for \mathbb{C}^N

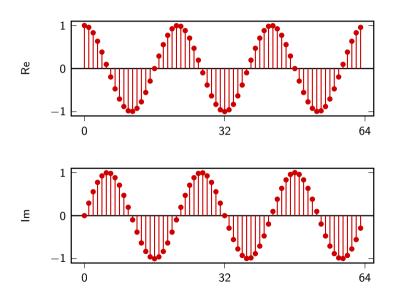
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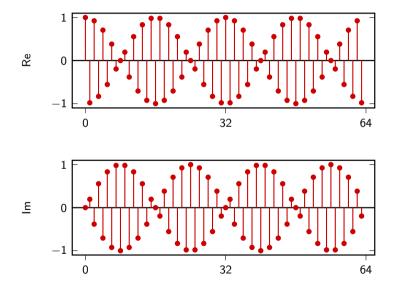


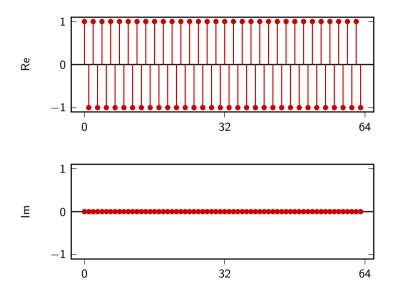


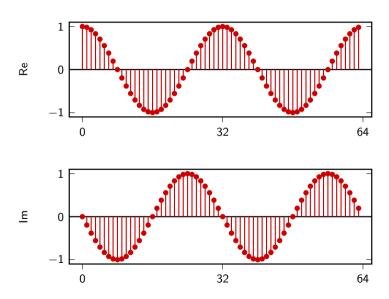
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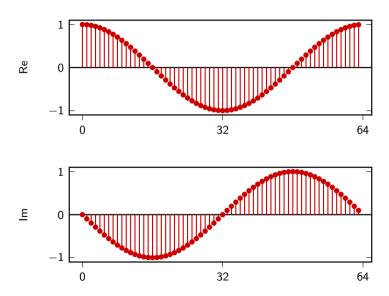












Proof of orthogonality

$$\begin{split} \langle \mathbf{w}^{(k)}, \mathbf{w}^{(h)} \rangle &= \sum_{n=0}^{N-1} (e^{j\frac{2\pi}{N}nk})^* e^{j\frac{2\pi}{N}nh} \\ &= \sum_{n=0}^{N-1} e^{j\frac{2\pi}{N}(h-k)n} \\ &= \begin{cases} N & \text{for } h = k \\ \frac{1 - e^{j2\pi(h-k)}}{1 - e^{j\frac{2\pi}{N}(h-k)}} = 0 & \text{otherwise} \end{cases} \end{split}$$

The Fourier Basis for \mathbb{C}^N

- ightharpoonup N orthogonal vectors \longrightarrow basis for \mathbb{C}^N
- \blacktriangleright vectors are not ortho*normal*. Normalization factor would be $1/\sqrt{N}$
- ▶ will keep normalization factor explicit in DFT formulas

Basis expansion

Analysis formula:

$$X_k = \langle \mathbf{w}^{(k)}, \mathbf{x} \rangle$$

Synthesis formula:

$$\mathbf{x} = \frac{1}{N} \sum_{k=0}^{N-1} X_k \mathbf{w}^{(k)}$$

Basis expansion (signal notation)

Analysis formula:

$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-j\frac{2\pi}{N}nk}, \qquad k = 0, 1, \dots, N-1$$

N-point signal in the frequency domain

Synthesis formula:

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N-point signal in the "time" domain

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N-point signal in the "time" domain

Change of basis in matrix form

Define
$$W_N = e^{-j \frac{2\pi}{N}}$$
 (or simply W when N is evident from the context)

Change of basis matrix **W** with $W[n, m] = W_N^{nm}$:

$$\mathbf{W} = \begin{bmatrix} 1 & 1 & 1 & 1 & \dots & 1 \\ 1 & W^1 & W^2 & W^3 & \dots & W^{N-1} \\ 1 & W^2 & W^4 & W^6 & \dots & W^{2(N-1)} \\ & & & & \dots & & \\ 1 & W^{N-1} & W^{2(N-1)} & W^{3(N-1)} & \dots & W^{(N-1)^2} \end{bmatrix}$$

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Change of basis in matrix form

Analysis formula:

$$\mathbf{X} = \mathbf{W}\mathbf{x}$$

Synthesis formula:

$$\mathbf{x} = \frac{1}{N} \mathbf{W}^H \mathbf{X}$$

DFT Matrix

$$W_N^m = W_N^{(m \mod N)}$$

e.g.
$$W_8^{11} = W_8^3$$

DFT Matrix

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Small DFT matrices: N = 2, 3

$$W_2 = e^{-jrac{2\pi}{2}} = -1$$
 $\mathbf{W}_2 = egin{bmatrix} 1 & 1 \ 1 & W \end{bmatrix} = egin{bmatrix} 1 & 1 \ 1 & -1 \end{bmatrix}$

$$\mathbf{W}_{3} = e^{-j\frac{2\pi}{3}} = -(1+j\sqrt{3})/2$$

$$\mathbf{W}_{3} = \begin{bmatrix} 1 & 1 & 1\\ 1 & W & W^{2}\\ 1 & W^{2} & W^{4} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1\\ 1 & W & W^{2}\\ 1 & W^{2} & W \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1\\ 1 & -(1+j\sqrt{3})/2 & -(1-j\sqrt{3})/2\\ 1 & -(1-j\sqrt{3})/2 & (1-j\sqrt{3})/2 \end{bmatrix}$$

Small DFT matrices: N = 4

$$W_4 = e^{-j\frac{2\pi}{4}} = e^{-j\frac{\pi}{2}} = -j$$

$$\mathbf{W}_4 = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & W & W^2 & W^3 \\ 1 & W^2 & W^4 & W^6 \\ 1 & W^3 & W^6 & W^9 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & W & W^2 & W^3 \\ 1 & W^2 & 1 & W^2 \\ 1 & W^3 & W^2 & W \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix}$$

Small DFT matrices: N = 5

$$\mathbf{W}_{5} = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & W & W^{2} & W^{3} & W^{4} \\ 1 & W^{2} & W^{4} & W^{6} & W^{8} \\ 1 & W^{3} & W^{6} & W^{9} & W^{12} \\ 1 & W^{4} & W^{8} & W^{12} & W^{16} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & W & W^{2} & W^{3} & W^{4} \\ 1 & W^{2} & W^{4} & W & W^{3} \\ 1 & W^{3} & W & W^{4} & W^{2} \\ 1 & W^{4} & W^{3} & W^{2} & W \end{bmatrix}$$

Small DFT matrices: N = 6

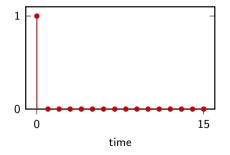
$$\boldsymbol{W}_{6} = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & W & W^{2} & W^{3} & W^{4} & W^{5} \\ 1 & W^{2} & W^{4} & W^{6} & W^{8} & W^{10} \\ 1 & W^{3} & W^{6} & W^{9} & W^{12} & W^{15} \\ 1 & W^{4} & W^{8} & W^{12} & W^{16} & W^{20} \\ 1 & W^{5} & W^{10} & W^{15} & W^{20} & W^{25} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & W & W^{2} & W^{3} & W^{4} & W^{5} \\ 1 & W^{2} & W^{4} & 1 & W^{2} & W^{4} \\ 1 & W^{3} & 1 & W^{3} & 1 & W^{3} \\ 1 & W^{4} & W^{2} & 1 & W^{4} & W^{2} \\ 1 & W^{5} & W^{4} & W^{3} & W^{2} & W \end{bmatrix}$$

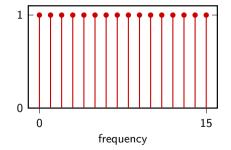
DFT is obviously linear

$$\mathsf{DFT}\left\{\alpha\,x[\mathbf{n}] + \beta\,y[\mathbf{n}]\right\} = \alpha\,\mathsf{DFT}\left\{x[\mathbf{n}]\right\} + \beta\,\mathsf{DFT}\left\{y[\mathbf{n}]\right\}$$

DFT of $\delta[n] \in \mathbb{C}^N$

$$\sum_{n=0}^{N-1} \delta[n] e^{-j\frac{2\pi}{N}nk} = 1 \quad \forall k$$





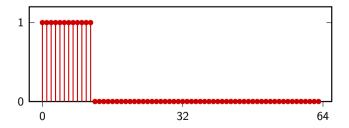
DFT of $\boldsymbol{\delta}^{(m)} = \delta[n-m] \in \mathbb{C}^N$

$$\sum_{n=0}^{N-1} \delta[n-m] e^{-j\frac{2\pi}{N}nk} = e^{-j\frac{2\pi}{N}mk}$$

$$\mathsf{DFT}\left\{\boldsymbol{\delta}^{(m)}\right\} = (\mathbf{w}^{(m)})^*$$

DFT of length-M step in \mathbb{C}^N

$$x[n] = \sum_{h=0}^{M-1} \delta[n-m], \quad n = 0, 1, \dots, N-1$$



DFT of length-M step in \mathbb{C}^N

$$X[k] = \sum_{m=0}^{M-1} \mathsf{DFT} \left\{ \delta[n-m] \right\} = \sum_{m=0}^{M-1} e^{-j\frac{2\pi}{N}mk}$$

$$= \frac{1 - e^{-j\frac{2\pi}{N}kM}}{1 - e^{-j\frac{2\pi}{N}k}}$$

$$= \frac{e^{-j\frac{\pi}{N}kM} \left[e^{j\frac{\pi}{N}kM} - e^{-j\frac{\pi}{N}kM} \right]}{e^{-j\frac{\pi}{N}k} \left[e^{j\frac{\pi}{N}k} - e^{-j\frac{\pi}{N}k} \right]}$$

$$= \frac{\sin\left(\frac{\pi}{N}Mk\right)}{\sin\left(\frac{\pi}{N}k\right)} e^{-j\frac{\pi}{N}(M-1)k}$$

DFT of length-M step in \mathbb{C}^N

$$\begin{split} X[k] &= \sum_{m=0}^{M-1} \mathsf{DFT} \left\{ \delta[n-m] \right\} = \sum_{m=0}^{M-1} e^{-j\frac{2\pi}{N}mk} \\ &= \frac{1 - e^{-j\frac{2\pi}{N}kM}}{1 - e^{-j\frac{2\pi}{N}k}} \\ &= \frac{e^{-j\frac{\pi}{N}kM} \left[e^{j\frac{\pi}{N}kM} - e^{-j\frac{\pi}{N}kM} \right]}{e^{-j\frac{\pi}{N}k} \left[e^{j\frac{\pi}{N}k} - e^{-j\frac{\pi}{N}k} \right]} \\ &= \frac{\sin \left(\frac{\pi}{N}Mk \right)}{\sin \left(\frac{\pi}{N}k \right)} e^{-j\frac{\pi}{N}(M-1)k} \end{split}$$

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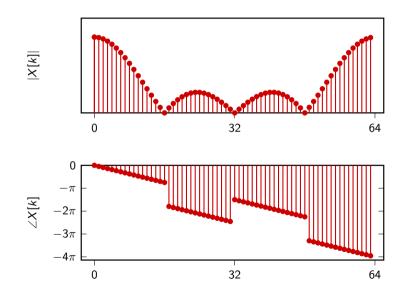
- ightharpoonup X[0] = M, from the definition of the sum
- ▶ X[k] = 0 if Mk/N integer $(0 \le k < N)$
- $ightharpoonup \angle X[k]$ linear in k (except at sign changes for the real part)

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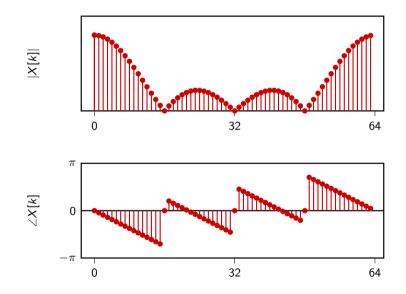


Wrapping the phase

Often the phase is displayed "wrapped" over the $[-\pi, \pi]$ interval.

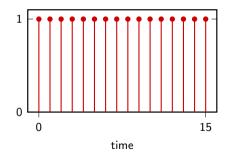
- most numerical packages return wrapped phase
- \blacktriangleright phase can be unwrapped by adding multiples of 2π

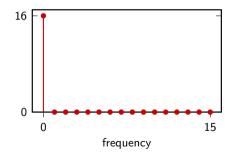
DFT of length-4 step in \mathbb{C}^{64} (phase wrapped)



DFT of $x[n] = 1, \quad x[n] \in \mathbb{C}^N$

$$X[k] = \sum_{n=0}^{N-1} e^{-j\frac{2\pi}{N}nk} = N\delta[k]$$





DFT of harmonic oscillations

$$\mathsf{DFT}\left\{e^{jrac{2\pi}{N}mn}
ight\} = \langle \mathbf{w}^{(k)}, \mathbf{w}^{(m)}
angle = N\delta[m-k]$$

$$x[n] = 3\cos\left(\frac{2\pi}{16}n\right)$$

$$= 3\cos\left(\frac{2\pi}{64}4n\right)$$

$$= \frac{3}{2}\left[e^{j\frac{2\pi}{64}4n} + e^{-j\frac{2\pi}{64}4n}\right]$$

$$= \frac{3}{2}\left[e^{j\frac{2\pi}{64}4n} + e^{j\frac{2\pi}{64}60n}\right]$$

$$= \frac{3}{2}(w_4[n] + w_{60}[n])$$

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$$= \frac{3}{2}(w_4[n] + w_{60}[n])$$

$$X[k] = \langle w_k[n], x[n] \rangle$$

$$= \langle w_k[n], \frac{3}{2} (w_4[n] + w_{60}[n]) \rangle$$

$$= \frac{3}{2} \langle w_k[n], w_4[n] \rangle + \frac{3}{2} \langle w_k[n], w_{60}[n] \rangle$$

$$= \begin{cases} 96 & \text{for } k = 4, 60 \\ 0 & \text{otherwise} \end{cases}$$

DFT of
$$x[n] = 3\cos(2\pi/16 n)$$
, $x[n] \in \mathbb{C}^{64}$

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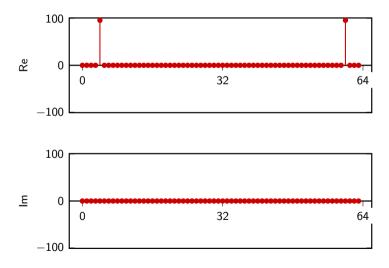
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$$x[n] = 3\cos\left(\frac{2\pi}{16}n + \frac{\pi}{3}\right)$$

$$= 3\cos\left(\frac{2\pi}{64}4n + \frac{\pi}{3}\right)$$

$$= \frac{3}{2}\left[e^{j\frac{2\pi}{64}4n}e^{j\frac{\pi}{3}} + e^{-j\frac{2\pi}{64}4n}e^{-j\frac{\pi}{3}}\right]$$

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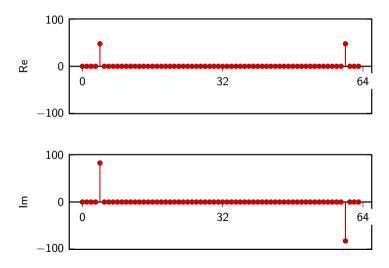
$$= \frac{3}{2}\left[e^{j\frac{2\pi}{64}4n}e^{j\frac{\pi}{3}} + e^{-j\frac{2\pi}{64}4n}e^{-j\frac{\pi}{3}}\right]$$

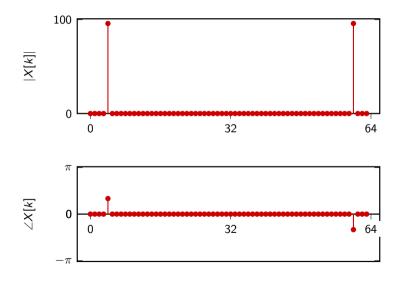
$$= \frac{3}{2}(e^{j\frac{\pi}{3}}w_4[n] + e^{-j\frac{\pi}{3}}w_{60}[n])$$

DFT of
$$x[n] = 3\cos(2\pi/16 n + \pi/3), x[n] \in \mathbb{C}^{64}$$

$$X[k] = \langle w_k[n], x[n] \rangle$$

$$= \begin{cases} 96e^{j\frac{\pi}{3}} & \text{for } k = 4\\ 96e^{-j\frac{\pi}{3}} & \text{for } k = 60\\ 0 & \text{otherwise} \end{cases}$$



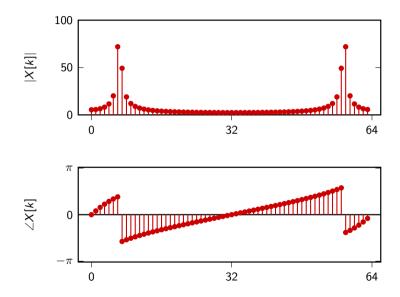


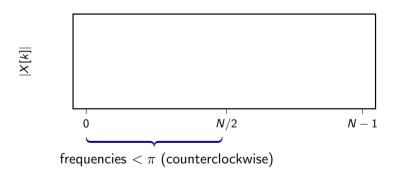
DFT of
$$x[n] = 3\cos(2\pi/10 n)$$
, $x[n] \in \mathbb{C}^{64}$

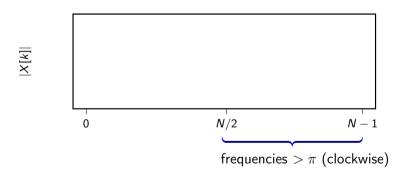
$$\frac{2\pi}{64} \, 6 < \frac{2\pi}{10} < \frac{2\pi}{64} \, 7$$

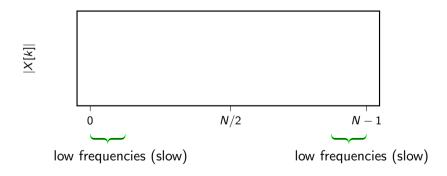
The DFT is an algorithm!

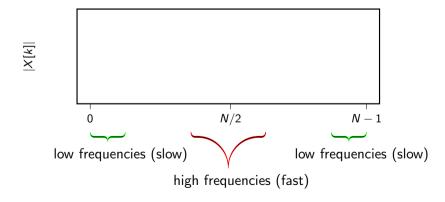
$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-j\frac{2\pi}{N}nk}, \qquad k = 0, 1, \dots, N-1$$

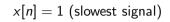


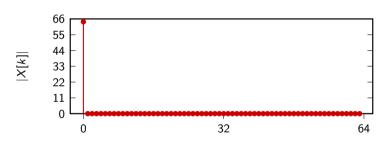






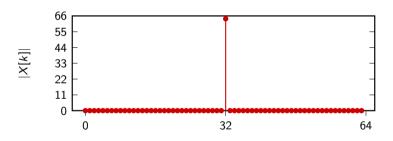






only lowest frequency

$$x[n] = \cos \pi n = (-1)^n$$
 (fastest signal)



only highest frequency

Energy distribution

Recall Parseval's Theorem: $\|\mathbf{x}\|^2 = \sum |\alpha_k|^2$

$$\sum_{n=0}^{N-1} |x[n]|^2 = \frac{1}{N} \sum_{k=0}^{N-1} |X[k]|^2$$

square magnitude of k-th DFT coefficient proportional to signal's energy at frequency $\omega = (2\pi/N) R$

Energy distribution

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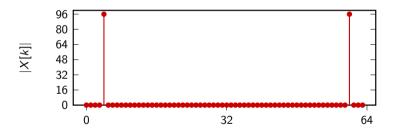
$$\sum_{n=0}^{N-1} |x[n]|^2 = \frac{1}{N} \sum_{k=0}^{N-1} |X[k]|^2$$

square magnitude of k-th DFT coefficient proportional to signal's energy at frequency $\omega=(2\pi/N)k$

47

Interpreting a DFT plot

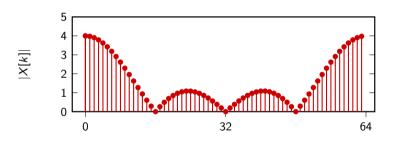
$$x[n] = 3\cos(2\pi/16 n)$$
 (sinusoid)



energy concentrated on single frequency (counterclockwise and clockwise combine to give real signal)

Interpreting a DFT plot

$$x[n] = u[n] - u[n - 4]$$
 (step)

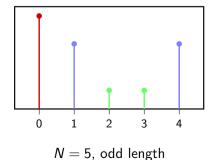


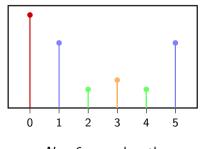
energy mostly in low frequencies

DFT of real signals

For real signals the DFT is "symmetric" in magnitude:

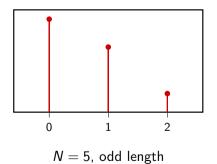
$$|X[k]| = |X[N-k]|$$
 for $k = 1, 2, ..., \lfloor N/2 \rfloor$

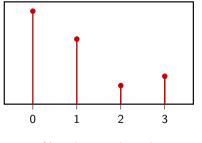




DFT of real signals

For real signals, magnitude plots need only $\lfloor N/2 \rfloor + 1$ points

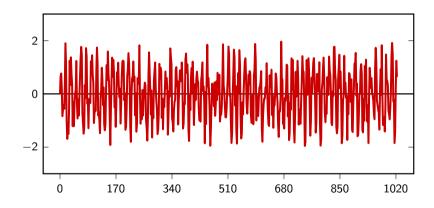


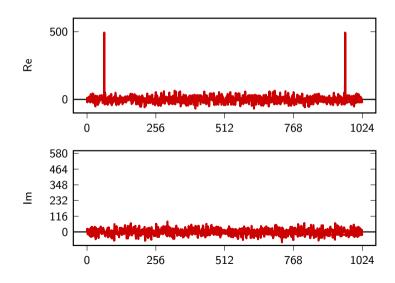


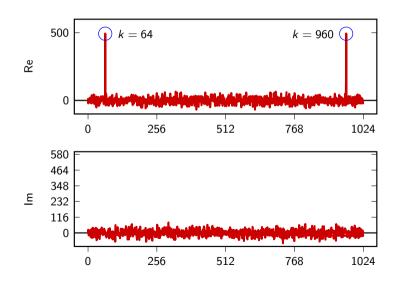


Overview

- ► DFT analysis examples
- ► Labeling the DFT axes







$$\mathbf{x}[\mathbf{n}] = \cos(\omega \mathbf{n} + \phi) + \eta[\mathbf{n}]$$

with

$$\phi = 0$$

$$\omega = \frac{2\pi}{1024} 6^{4}$$

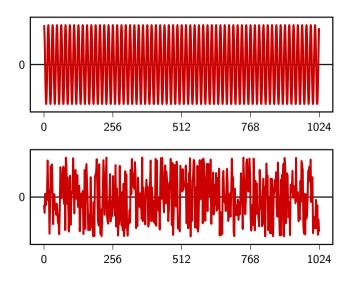
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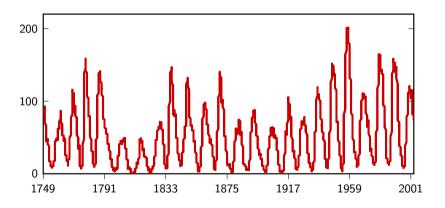
$$\phi = \frac{2\pi}{1024} 64$$

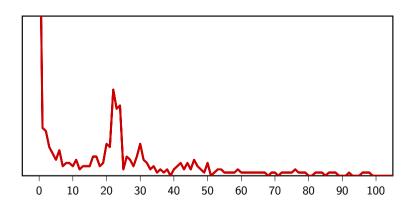
Mystery signal unveiled

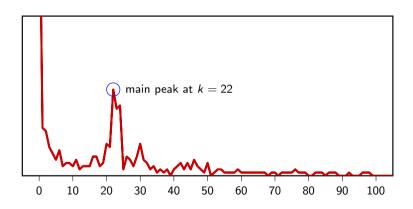


- ▶ sunspot number: $s = 10 \times \#$ of clusters + # of spots
- ▶ data set from 1749 to 2003, 2904 months

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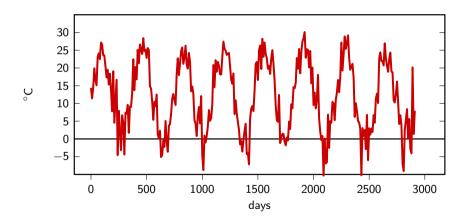


- ▶ DFT main peak for k = 22
- ▶ 22 cycles over 2904 months
- ▶ period: $\frac{2904}{22} \approx 11$ years

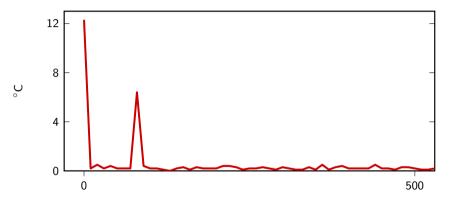
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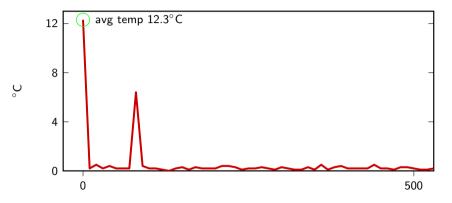
Daily temperature (2920 days)



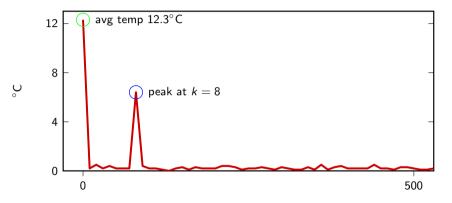
Daily temperature: DFT



Daily temperature: DFT



Daily temperature: DFT



Daily temperature

- ▶ average value (0-th DFT coefficient): 12.3°C
- ▶ DFT main peak for k = 8, value 6.4°C
- ▶ 8 cycles over 2920 days
- period: $\frac{2920}{8} = 365 \text{ days}$
- ▶ temperature excursion: $12.3^{\circ}\text{C} \pm 12.8^{\circ}\text{C}$

Daily temperature

In case you're wondering why $\pm 12.8^{\circ}$:

DFT
$$\left\{ A \cos \left(\frac{2\pi}{N} M n \right) \right\} [k] = \begin{cases} \frac{A}{2} N & \text{for } k = M, N - M \\ 0 & \text{otherwise} \end{cases}$$

- fastest (positive) frequency is $\omega = \pi$
- lacktriangleright sinusoid at $\omega=\pi$ needs two samples to do a full revolution
- ▶ time between samples: $T_s = 1/F_s$ seconds
- \triangleright real-world period for fastest sinusoid: $2T_s$ seconds
- ightharpoonup real-world frequency for fastest sinusoid: $F_s/2$ Hz

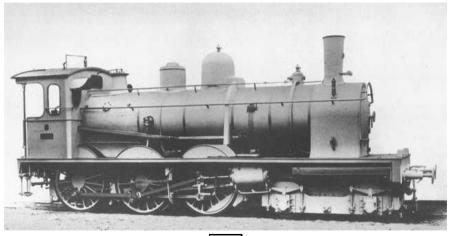
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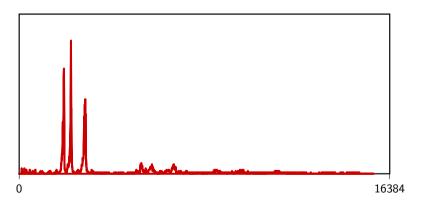
Example: train whistle



Play

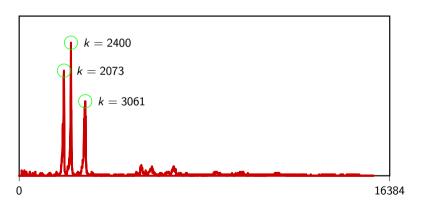
Example: train whistle

32768 samples (the "clock" of the system $F_s = 8000 \mathrm{Hz}$)



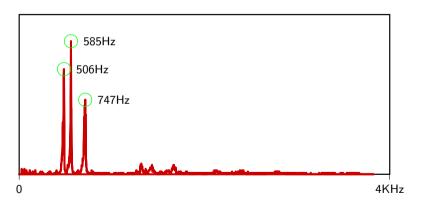
Example: train whistle

32768 samples (the "clock" of the system $F_s = 8000 \text{Hz}$)



Example: train whistle

the "clock" of the system $F_s = 8000 \text{Hz}$



Example: train whistle

if we look up the frequencies:



B minor chord



DFT formulas

Analysis formula:

$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-j\frac{2\pi}{N}nk}, \qquad k = 0, 1, \dots, N-1$$

N-point signal in the frequency domain

Synthesis formula:

$$x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j\frac{2\pi}{N}nk}, \qquad n = 0, 1, \dots, N-1$$

N-point signal in the "time" domain

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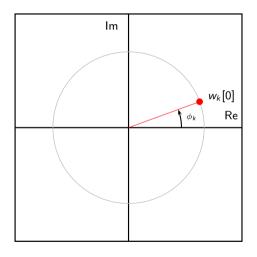
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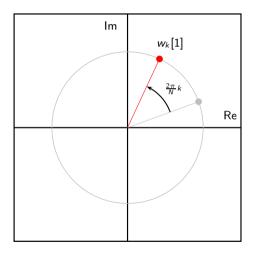
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N-point signal in the "time" domain

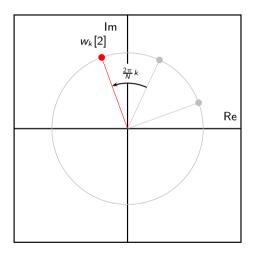
$$w_k[n] = e^{j(\frac{2\pi}{N}kn + \phi_k)}$$



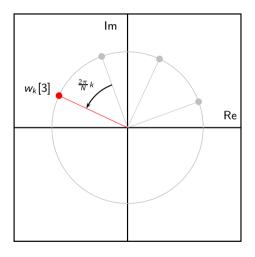
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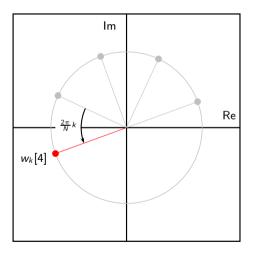
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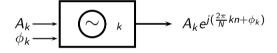


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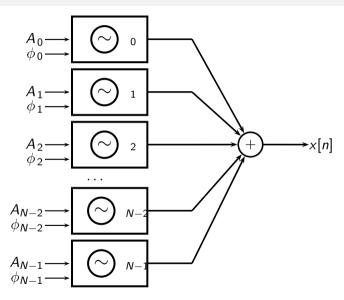


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DFT synthesis formula

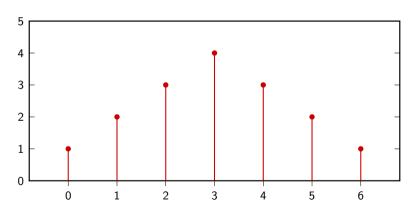


Initializing the machine

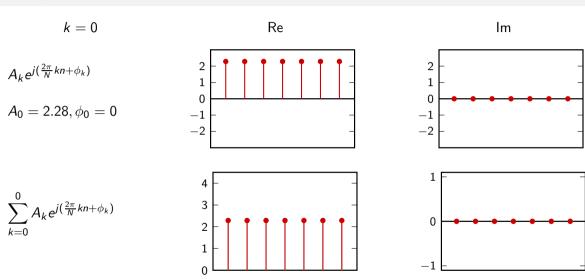
$$A_k = |X[k]|/N$$

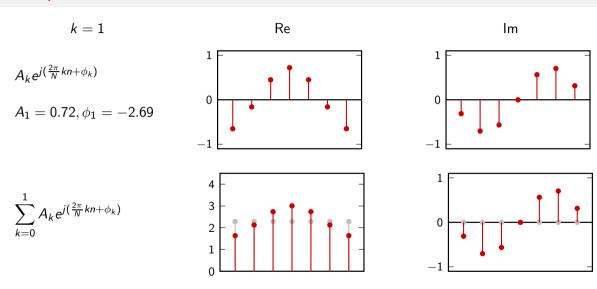
$$\phi_k = \angle X[k]$$

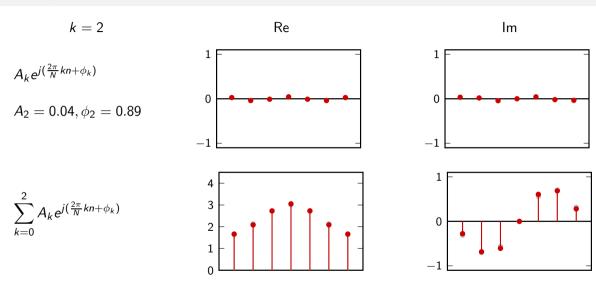
$$\mathbf{x} = [1 \ 2 \ 3 \ 4 \ 3 \ 2 \ 1]^T$$

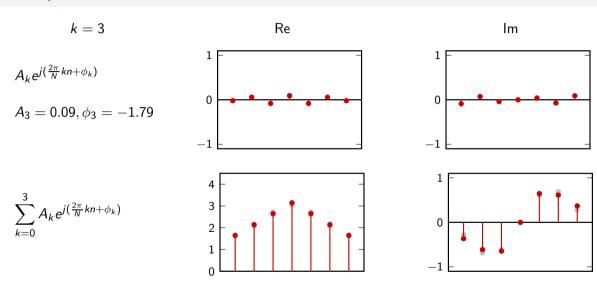


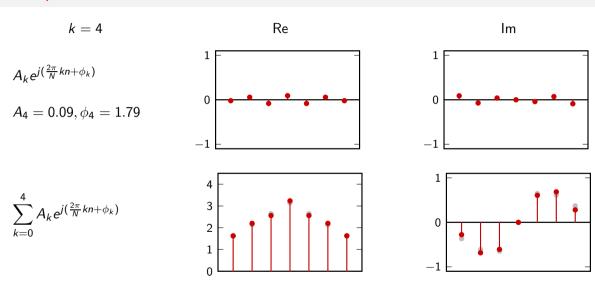
k	A_k	ϕ_{k}
0	2.2857	0.0000
1	0.7213	-2.6928
2	0.0440	0.8976
3	0.0919	-1.7952
4	0.0919	1.7952
5	0.0440	-0.8976
6	0.7213	2.6928

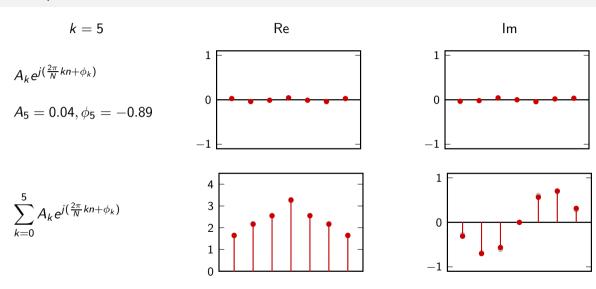


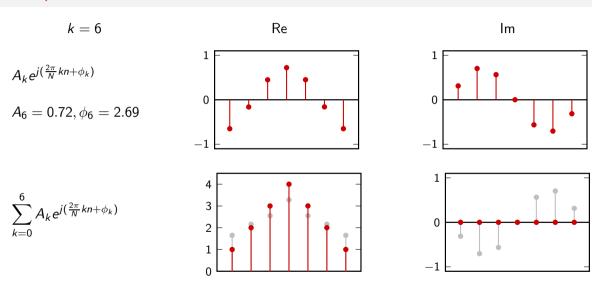




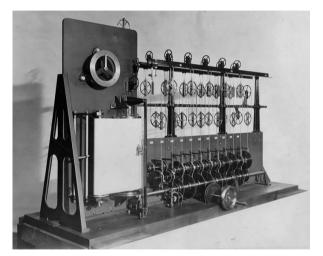








The machine before DSP



tide-predicting machine (originally invented by Lord Kelvin)

Wonderful website

http://jackschaedler.github.io/circles-sines-signals

Running the machine too long...

$$x[n + N] = x[n]$$

output signal is N-periodic!

Inherent periodicities in the DFT

the synthesis formula:

$$x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j\frac{2\pi}{N}nk}, \qquad n = 0, 1, \dots, N-1$$

produces an N-point signal in the time domain

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Discrete Fourier Series (DFS)

DFS = DFT with periodicity explicit

- ▶ the DFS maps an *N*-periodic signal onto an *N*-periodic sequence of Fourier coefficients
- ► the inverse DFS maps an *N*-periodic sequence of Fourier coefficients a set onto an *N*-periodic signal
- ▶ the DFS of an *N*-periodic signal is mathematically equivalent to the DFT of one period

The DFS helps us understand how to define time shifts for finite-length signals.

For an *N*- periodic sequence $\tilde{x}[n]$:

- $ightharpoonup ilde{x}[n-M]$ is well-defined for all $M\in\mathbb{N}$
- ▶ DFS $\{\tilde{x}[n-M]\} = e^{-j\frac{2\pi}{N}Mk}\tilde{X}[k]$ (easy derivation)
- $\blacktriangleright \mathsf{IDFS}\left\{\ \tilde{X}[k]\right\} = \tilde{x}[n-M]$

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a delay in time becomes a linear phase factor in frequency

For an N-point signal x[n]:

- \blacktriangleright x[n-M] is *not* well-defined
- ▶ what is IDFT $\left\{e^{-j\frac{2\pi}{N}Mk} X[k]\right\}$?

For an *N*-point signal x[n]:

- \triangleright x[n-M] is *not* well-defined
- what is IDFT $\left\{e^{-j\frac{2\pi}{N}Mk} X[k]\right\}$?

$$\begin{split} \mathsf{IDFT} \left\{ e^{-j\frac{2\pi}{N}Mk} \, X[k] \right\} [n] &= \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{-j\frac{2\pi}{N}Mk} \, e^{j\frac{2\pi}{N}nk} \\ &= \frac{1}{N} \sum_{k=0}^{N-1} \left(\sum_{m=0}^{N-1} x[m] \, e^{-j\frac{2\pi}{N}mk} \right) \, e^{-j\frac{2\pi}{N}Mk} \, e^{j\frac{2\pi}{N}nk} \\ &= \frac{1}{N} \sum_{m=0}^{N-1} x[m] \sum_{k=0}^{N-1} e^{j\frac{2\pi}{N}(n-M-m)k} \end{split}$$

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We've seen something like this before...

$$\sum_{n=0}^{N-1} e^{-j\frac{2\pi}{N}nk} = \begin{cases} N & \text{if } k \text{ multiple of } N \\ 0 & \text{otherwise} \end{cases}$$

(remember the orthogonality proof for the DFT basis)

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Modulo operator

given $C \in \mathbb{N}$, find m such that $0 \le m < N$ and C - m is a multiple of N

any integer C can be written as $C = pN + (C \mod N)$, $p \in \mathbb{N}$

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shifts for finite-length signals are "naturally" circula

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