

Discrete Fourier Transform (DFT)

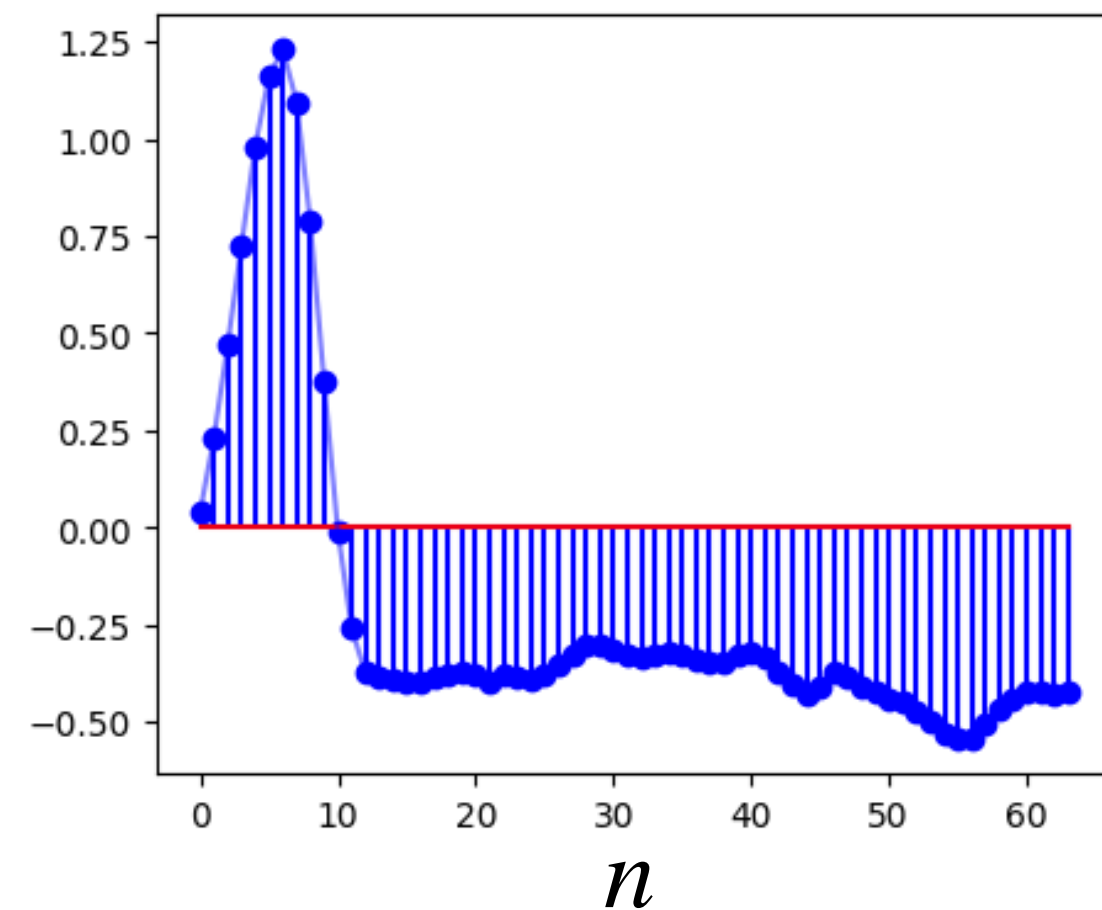
Itthi Chatnuntaweche

Signal Representation

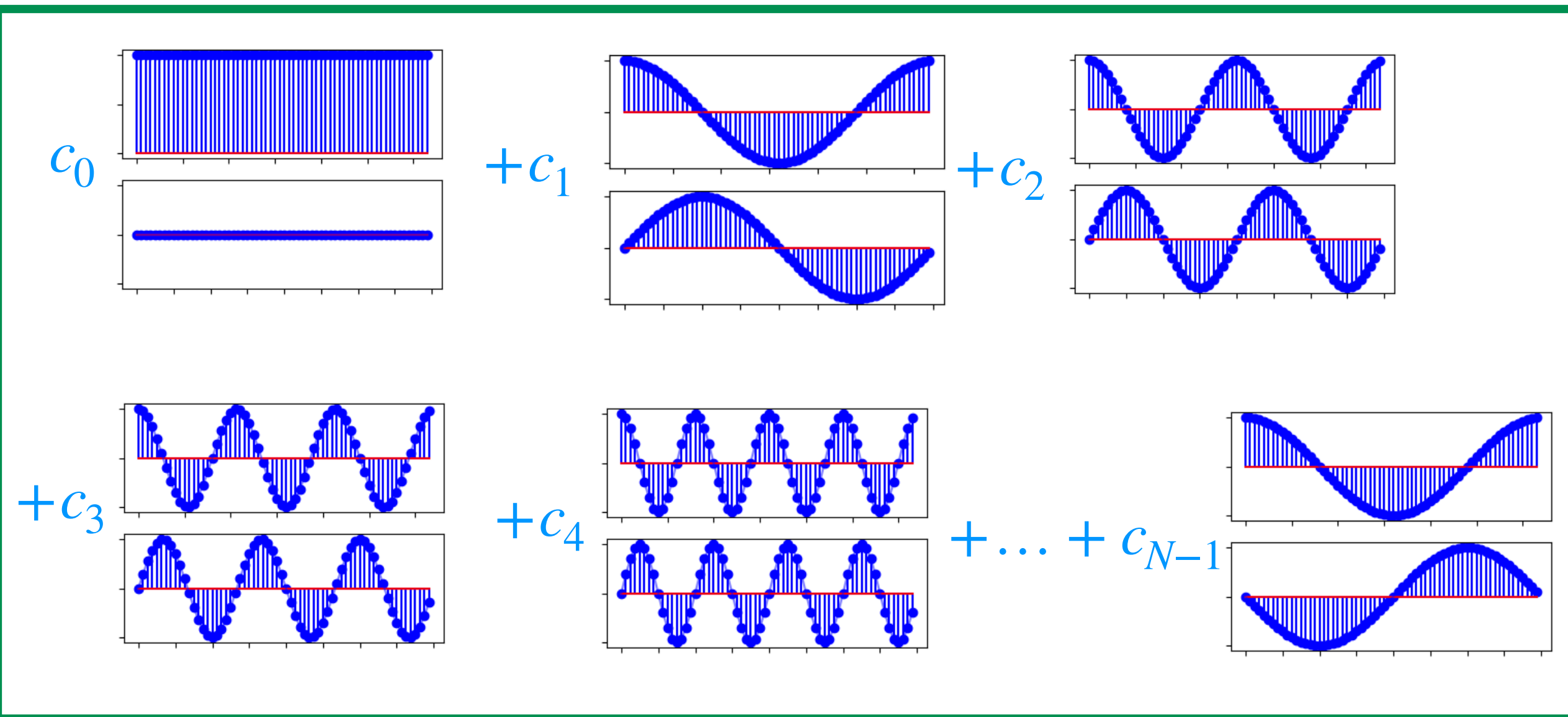
finite-length sequences

Most practical finite-duration, discrete-time signals can be written as a weighted sum of harmonically related complex exponentials

$x[n]$ of length N

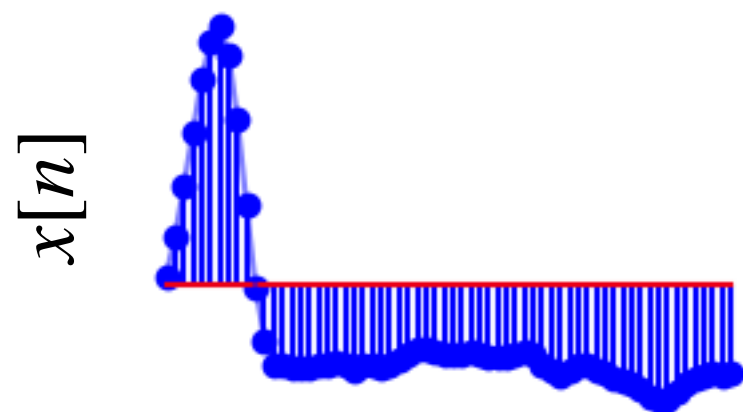


\approx



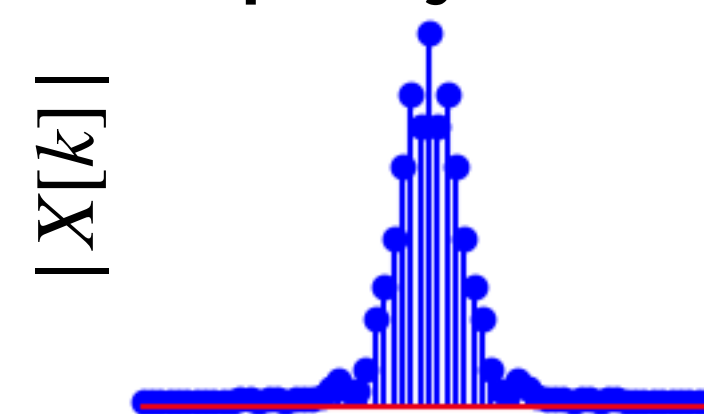
Discrete Fourier Transform (DFT)

Time domain



$$\begin{matrix} x[n] \\ \left[\begin{array}{c} x[0] \\ x[1] \\ \vdots \\ x[N-1] \end{array} \right] \end{matrix} \xleftrightarrow{DFT} \begin{matrix} X[k] \\ \left[\begin{array}{c} X[0] \\ X[1] \\ \vdots \\ X[N-1] \end{array} \right] \end{matrix}$$

Frequency domain



$$x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j(\frac{2\pi}{N})kn}$$

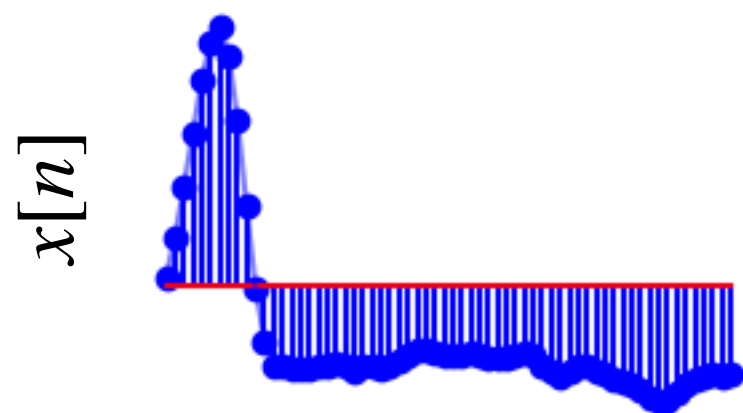
$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-j(\frac{2\pi}{N})kn}$$

Oppenheim, Alan V. Discrete-time signal processing. Pearson Education India, 1999.

Oppenheim, Alan V., et al. Signals and systems. Vol. 2. Upper Saddle River, NJ: Prentice hall, 1997.

Discrete Fourier Transform (DFT)

Time domain



$x[n]$

$$\begin{bmatrix} x[0] \\ x[1] \\ \vdots \\ x[N-1] \end{bmatrix}$$

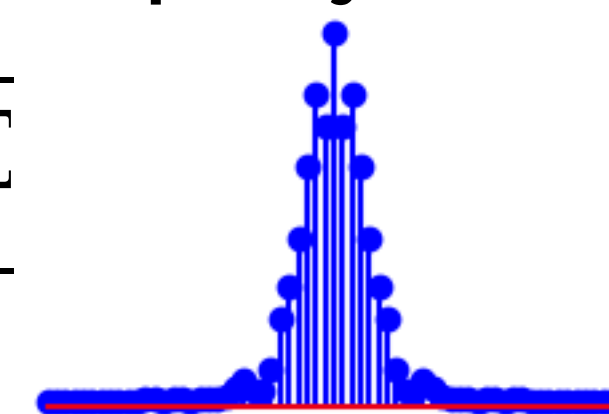
DFT
 \longleftrightarrow

$X[k]$

$$\begin{bmatrix} X[0] \\ X[1] \\ \vdots \\ X[N-1] \end{bmatrix}$$

Frequency domain

$|X[k]|$



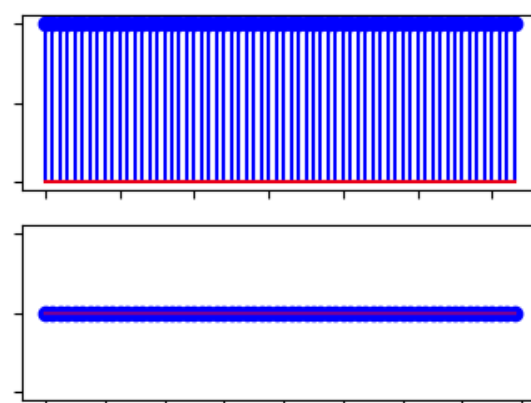
$$x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j(\frac{2\pi}{N})kn}$$

$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-j(\frac{2\pi}{N})kn}$$

$x[n]$

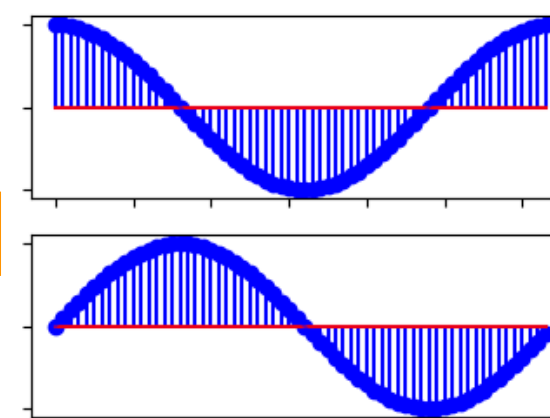
$$= \frac{1}{N} \left($$

$X[0]$



e^{j0}

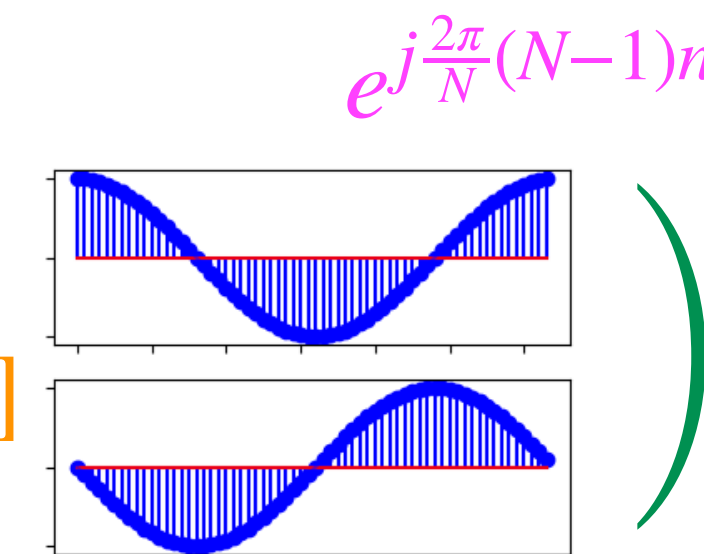
$+ X[1]$



$e^{j\frac{2\pi}{N}n}$

$+ \dots +$

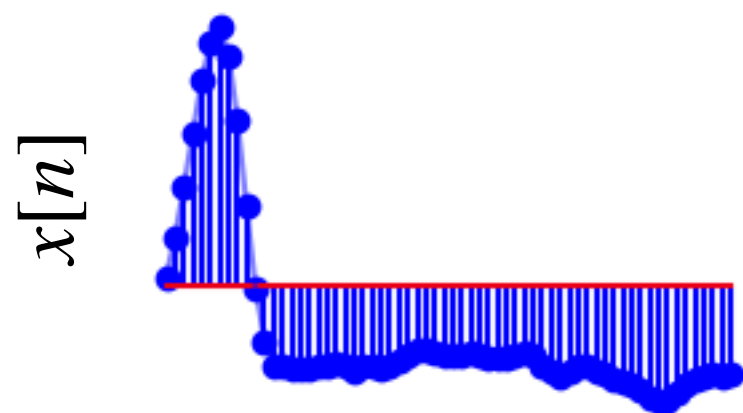
$X[N-1]$



$e^{j\frac{2\pi}{N}(N-1)n}$

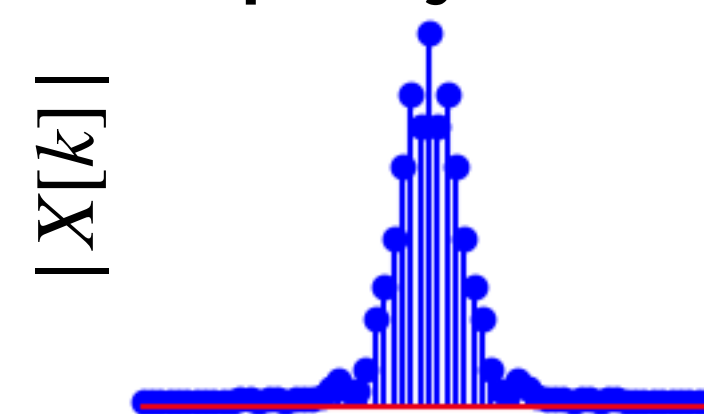
Discrete Fourier Transform (DFT)

Time domain



$$\begin{bmatrix} x[0] \\ x[1] \\ \vdots \\ x[N-1] \end{bmatrix} \xleftrightarrow{DFT} \begin{bmatrix} X[0] \\ X[1] \\ \vdots \\ X[N-1] \end{bmatrix}$$

Frequency domain



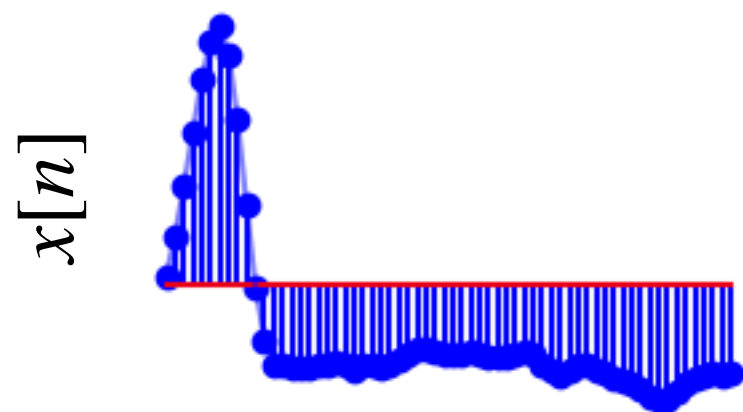
$$x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j(\frac{2\pi}{N})kn}$$

$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-j(\frac{2\pi}{N})kn}$$

$$x[n] = \frac{1}{N} \left(X[0] \begin{bmatrix} \text{Plot of } e^{j0} \\ \text{Plot of } e^{j0} \end{bmatrix} + X[1] \begin{bmatrix} \text{Plot of } e^{j\frac{2\pi}{N}} \\ \text{Plot of } e^{-j\frac{2\pi}{N}} \end{bmatrix} + \dots + X[N-1] \begin{bmatrix} \text{Plot of } e^{j\frac{2\pi(N-1)}{N}} \\ \text{Plot of } e^{-j\frac{2\pi(N-1)}{N}} \end{bmatrix} \right)$$

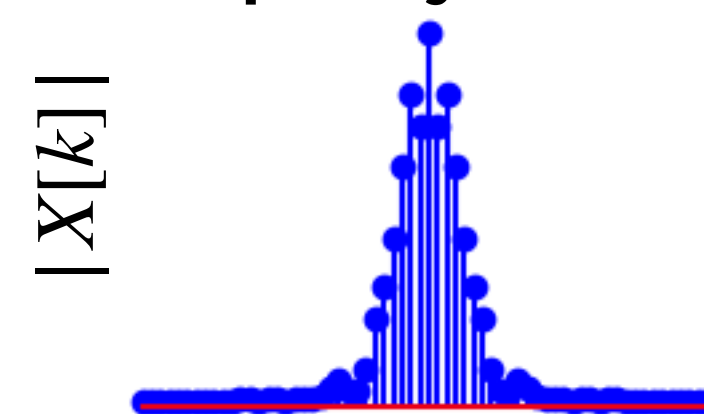
Discrete Fourier Transform (DFT)

Time domain



$$\begin{matrix} x[n] \\ \left[\begin{array}{c} x[0] \\ x[1] \\ \vdots \\ x[N-1] \end{array} \right] \end{matrix} \xleftrightarrow{DFT} \begin{matrix} X[k] \\ \left[\begin{array}{c} X[0] \\ X[1] \\ \vdots \\ X[N-1] \end{array} \right] \end{matrix}$$

Frequency domain



$$x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j(\frac{2\pi}{N})kn}$$

$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-j(\frac{2\pi}{N})kn}$$

scipy.fft.fft

```
scipy.fft.fft(x, n=None, axis=-1, norm=None, overwrite_x=False,
workers=None, *, plan=None)
```

[\[source\]](#)

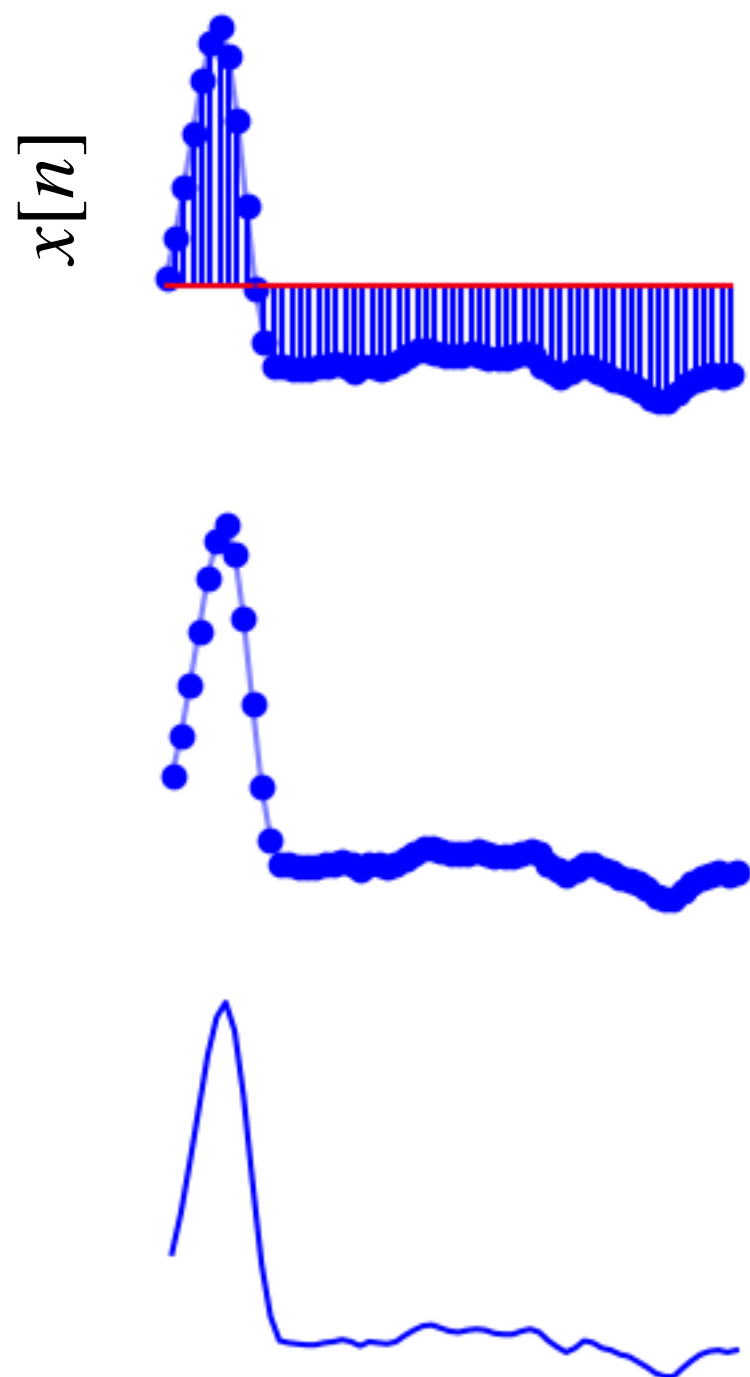
Compute the 1-D discrete Fourier Transform.

This function computes the 1-D n -point discrete Fourier Transform (DFT) with the efficient Fast Fourier Transform (FFT) algorithm [\[1\]](#).

Fast Fourier Transform (FFT) - An efficient algorithm that computes the DFT of a signal

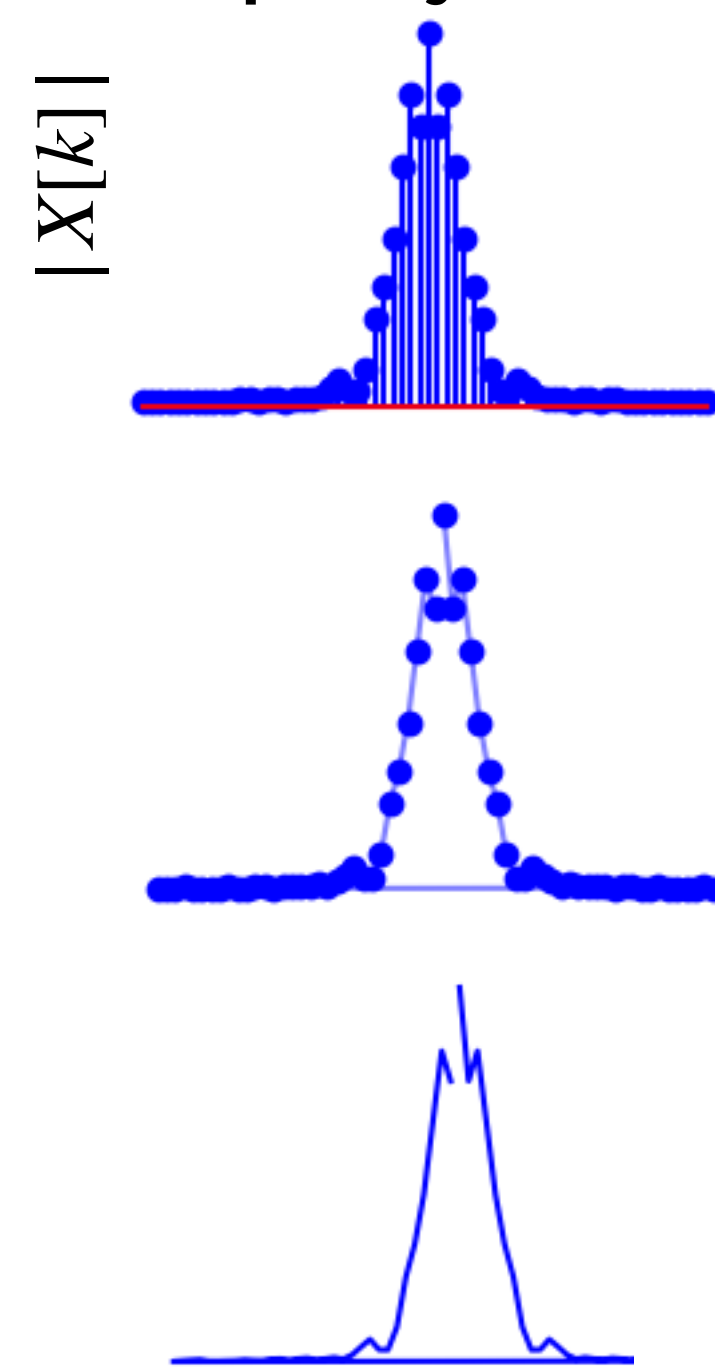
Discrete Fourier Transform (DFT)

Time domain



$$\begin{matrix} x[n] \\ \left[\begin{matrix} x[0] \\ x[1] \\ \vdots \\ x[N-1] \end{matrix} \right] \end{matrix} \xleftrightarrow{DFT} \begin{matrix} X[k] \\ \left[\begin{matrix} X[0] \\ X[1] \\ \vdots \\ X[N-1] \end{matrix} \right] \end{matrix}$$

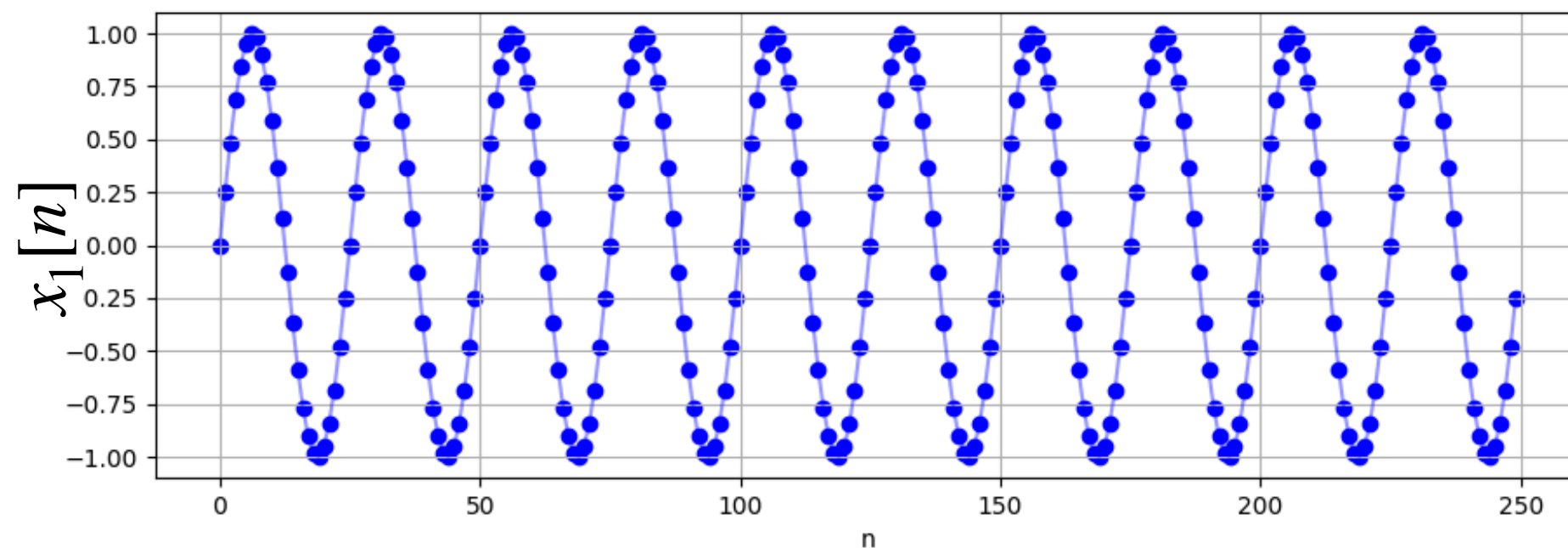
Frequency domain



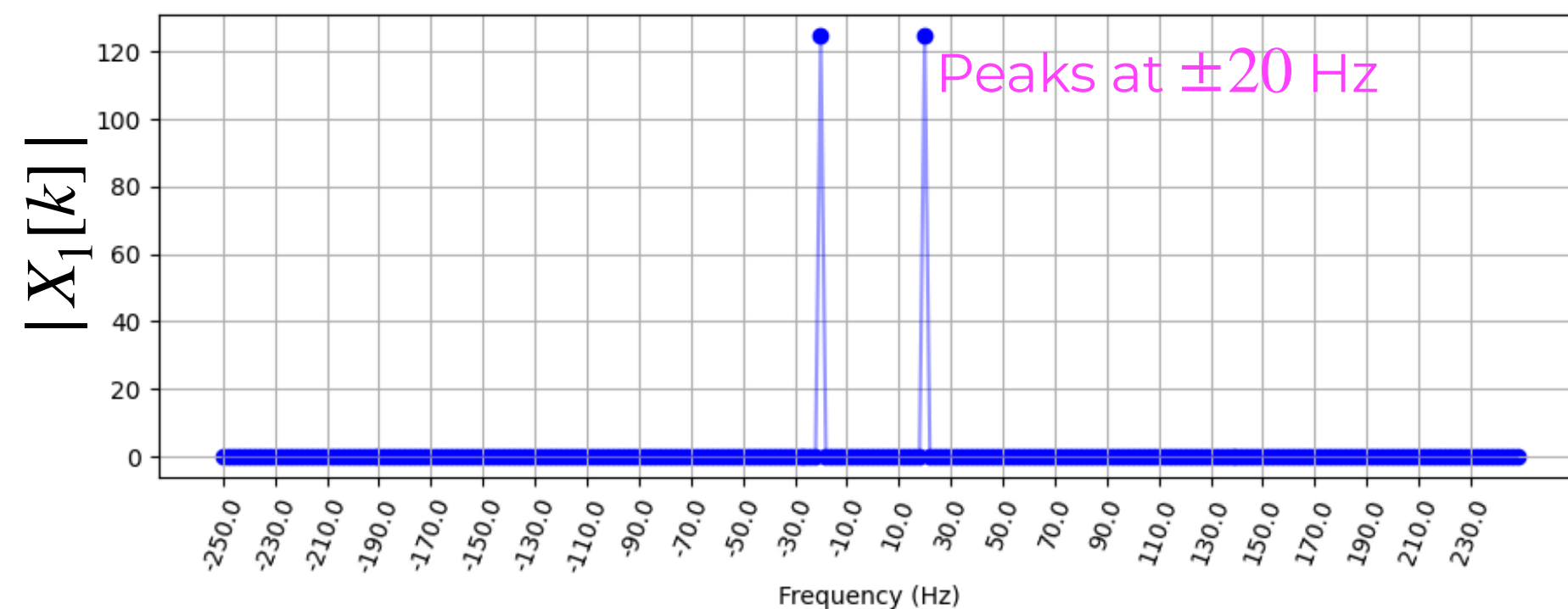
Discrete Fourier Transform (DFT)

Time domain

Samples from $x(t) = \sin(2\pi \times 20t)$



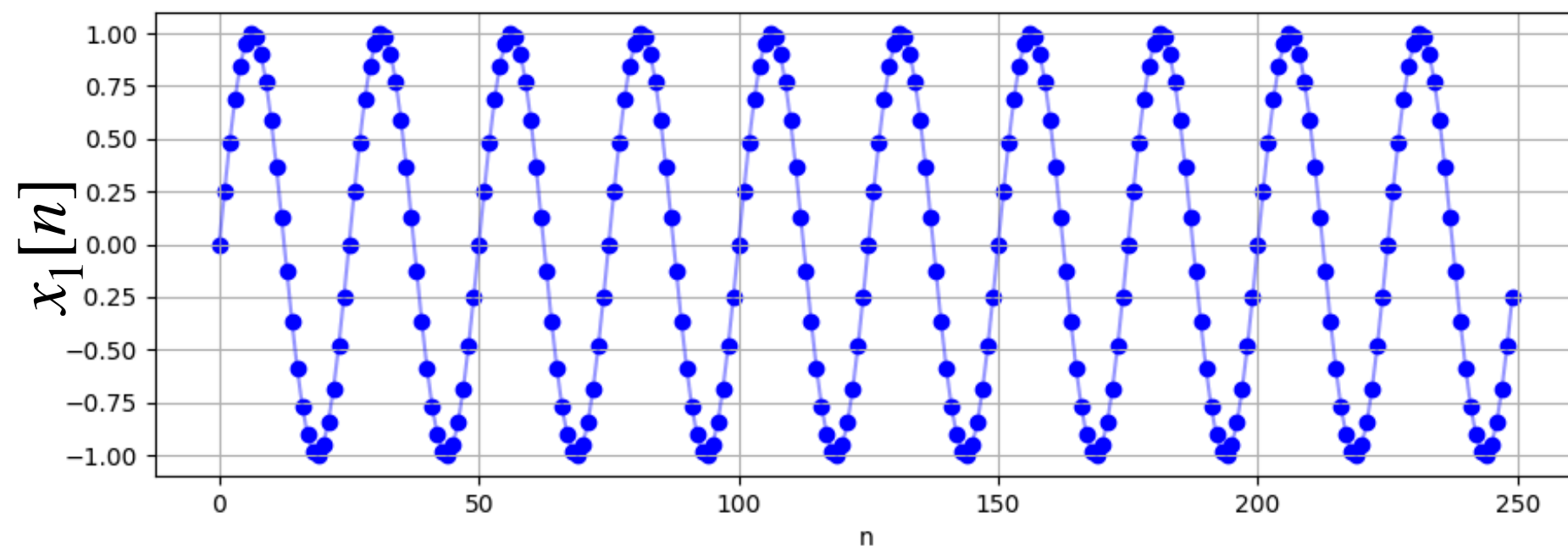
Frequency domain



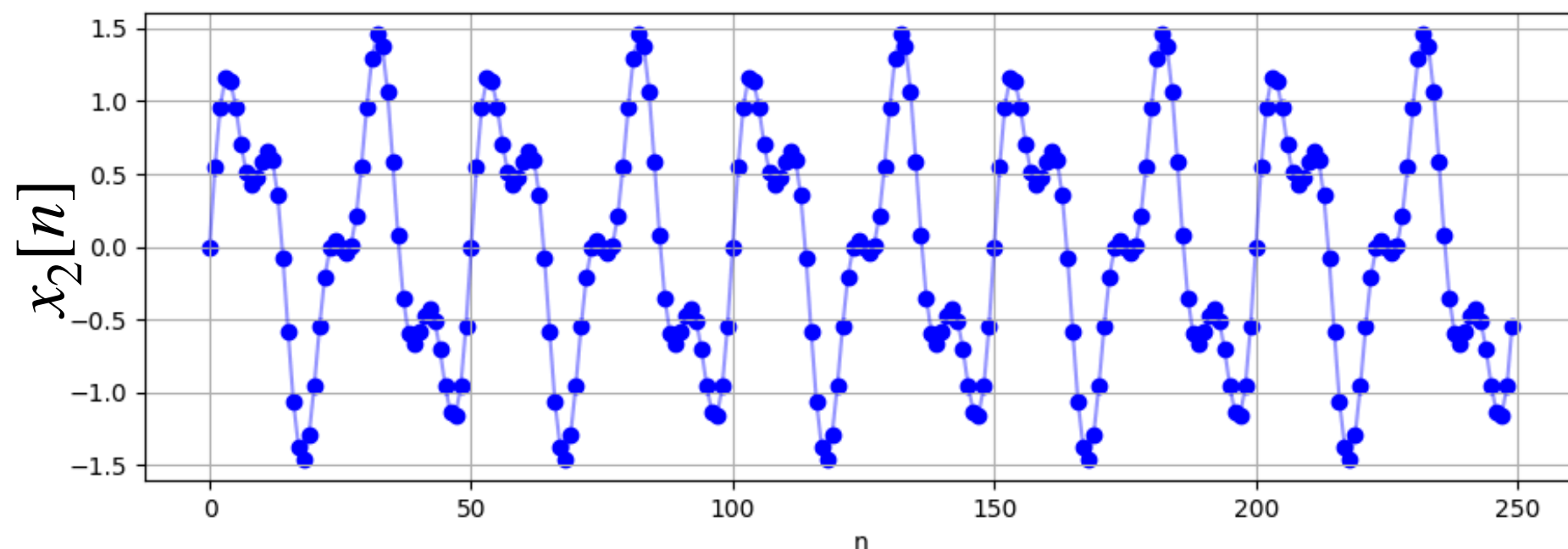
Discrete Fourier Transform (DFT)

Time domain

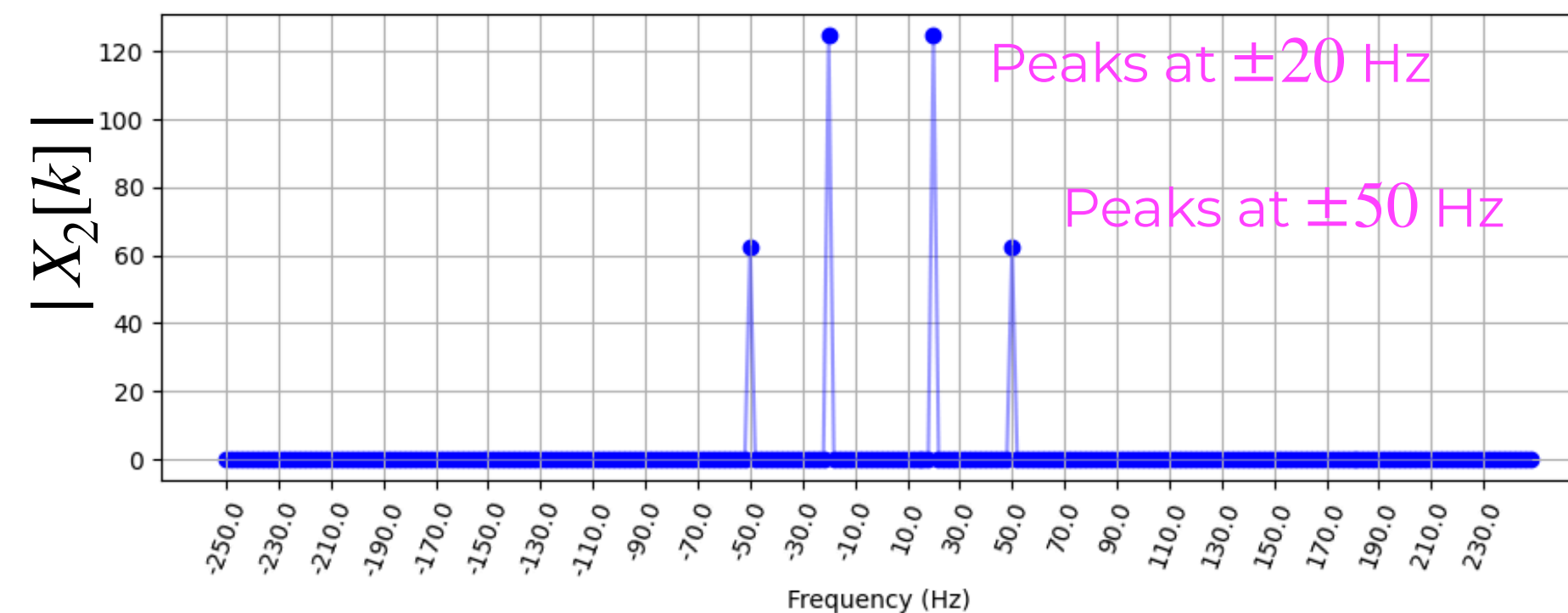
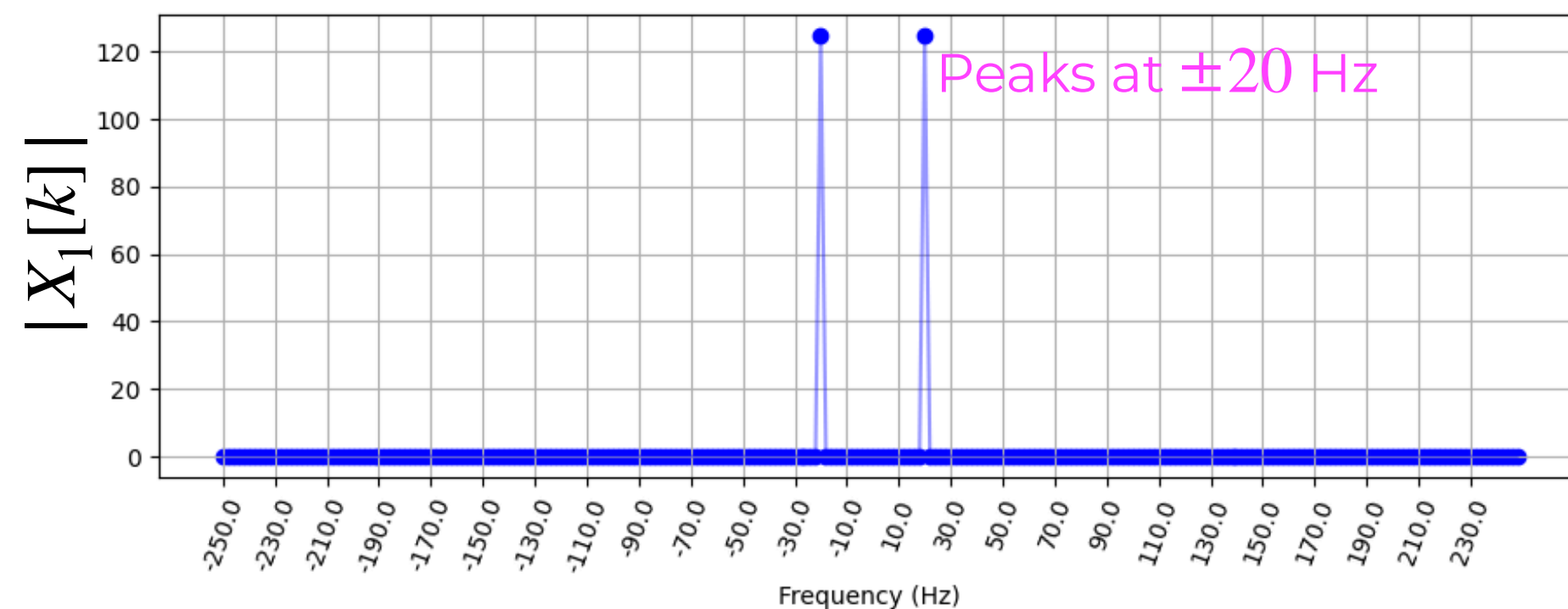
Samples from $x(t) = \sin(2\pi \times 20t)$



Samples from $x(t) = \sin(2\pi \times 20t) + 0.5 \sin(2\pi \times 50t)$



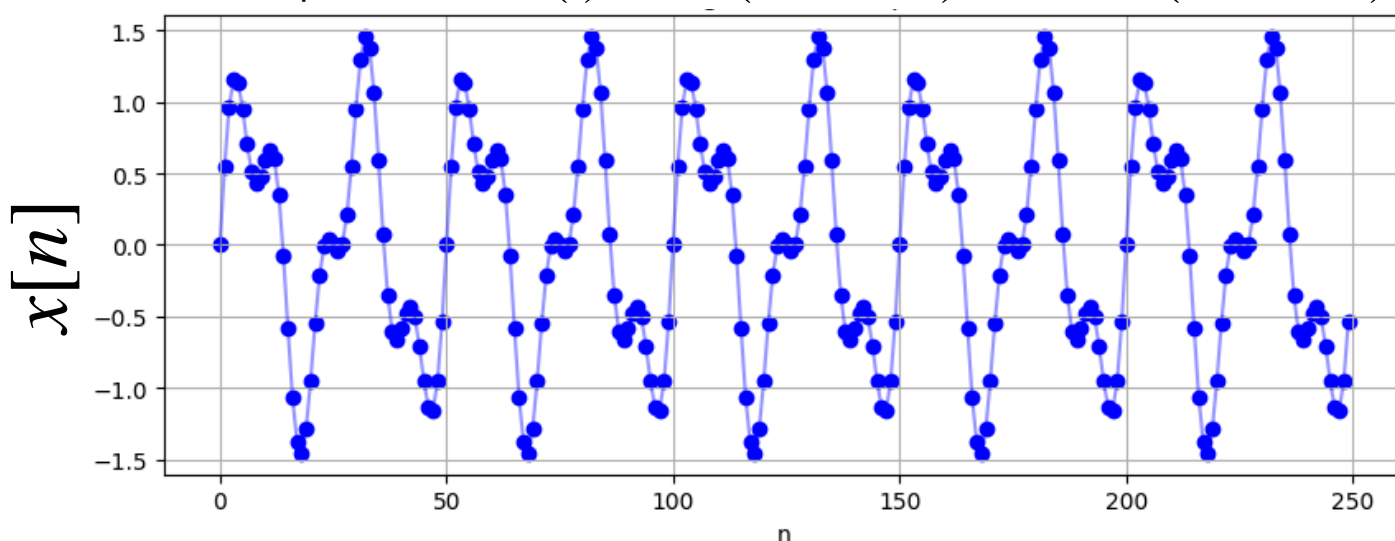
Frequency domain



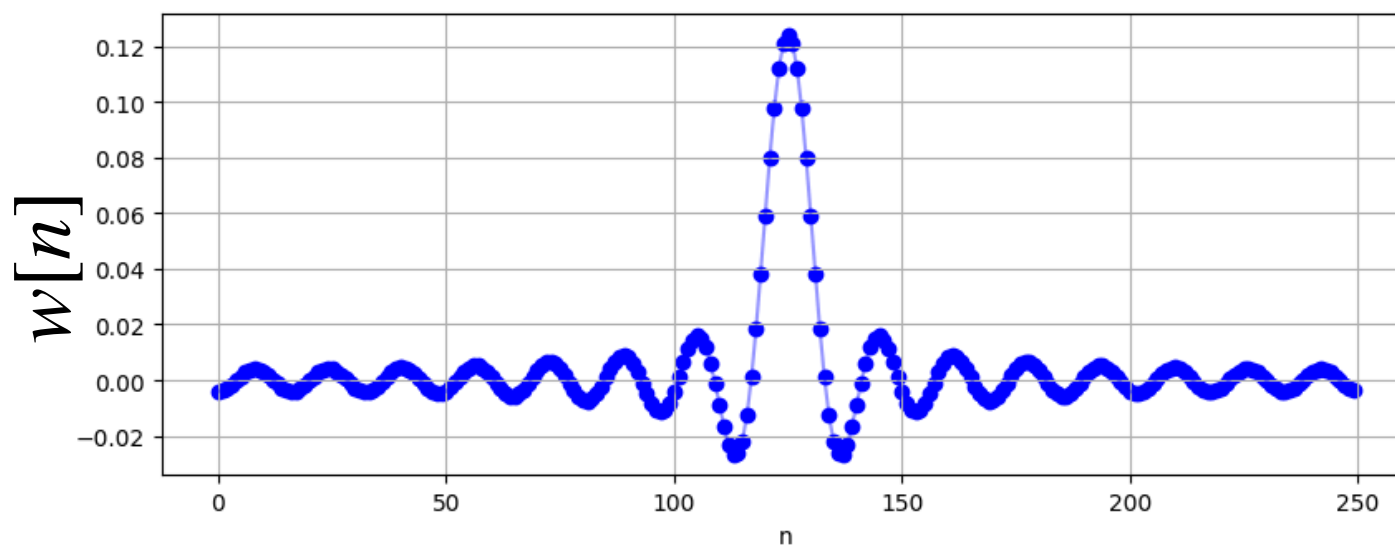
A Simple Filtering

Time domain

Samples from $x(t) = \sin(2\pi \times 20t) + 0.5 \sin(2\pi \times 50t)$

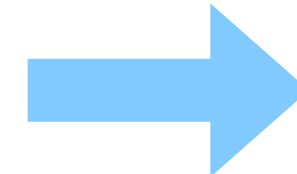


Original
signal

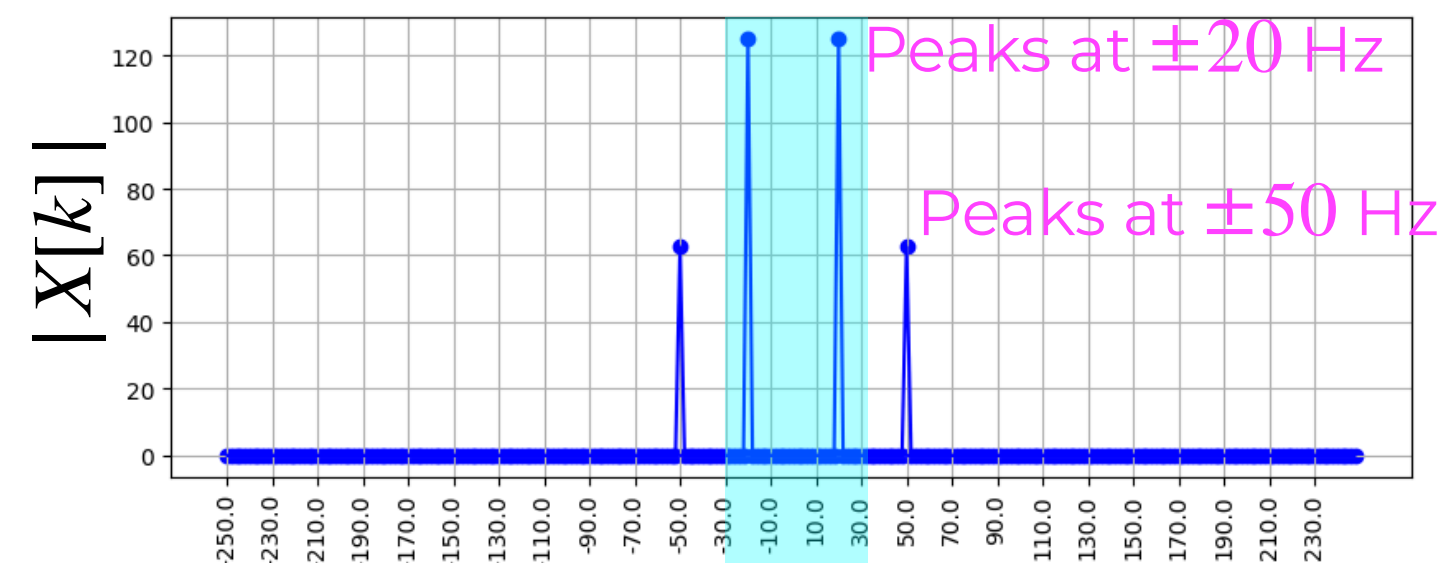


Window

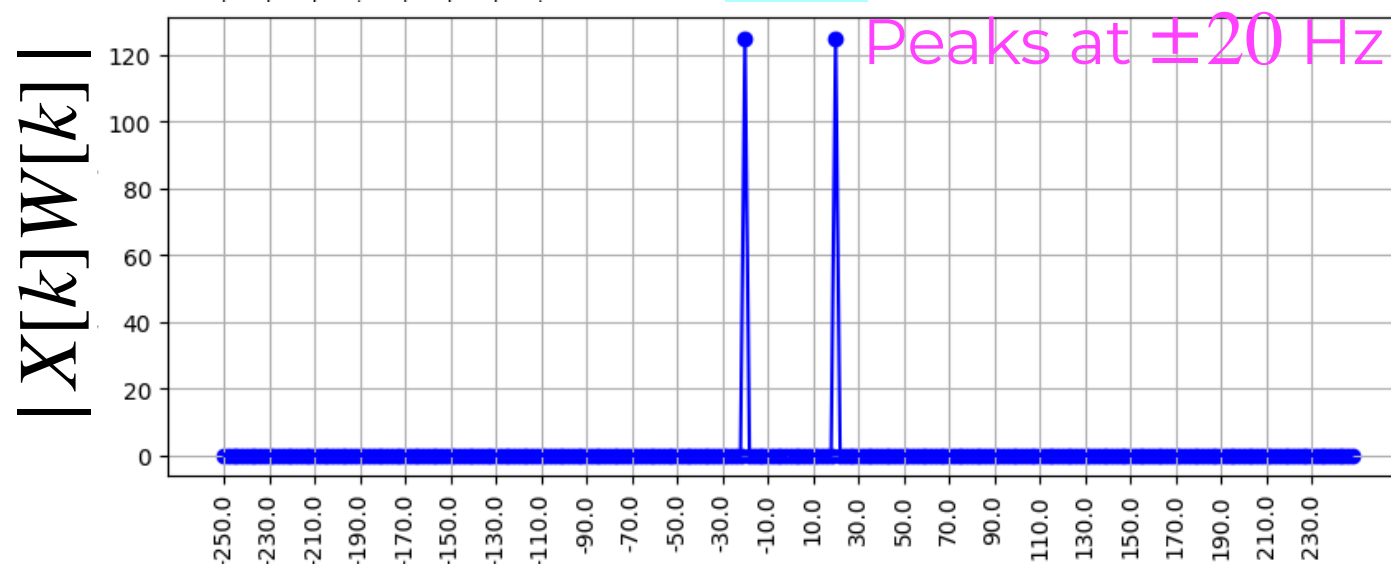
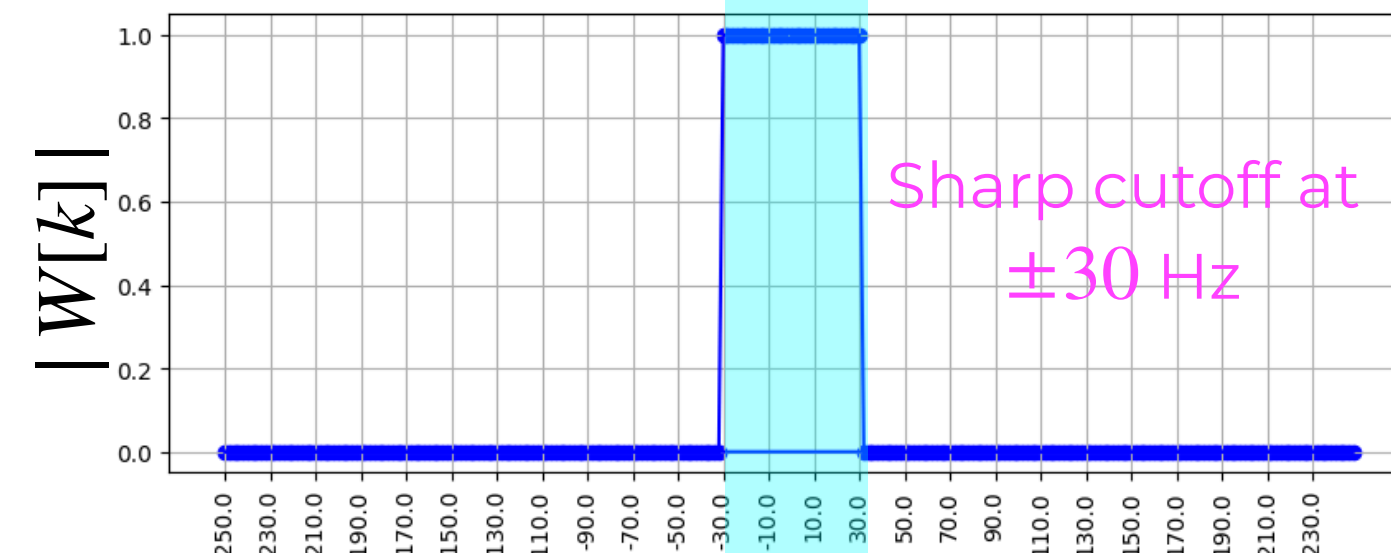
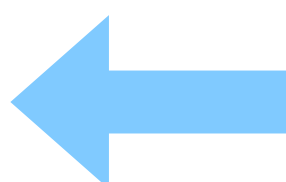
DFT



Frequency domain



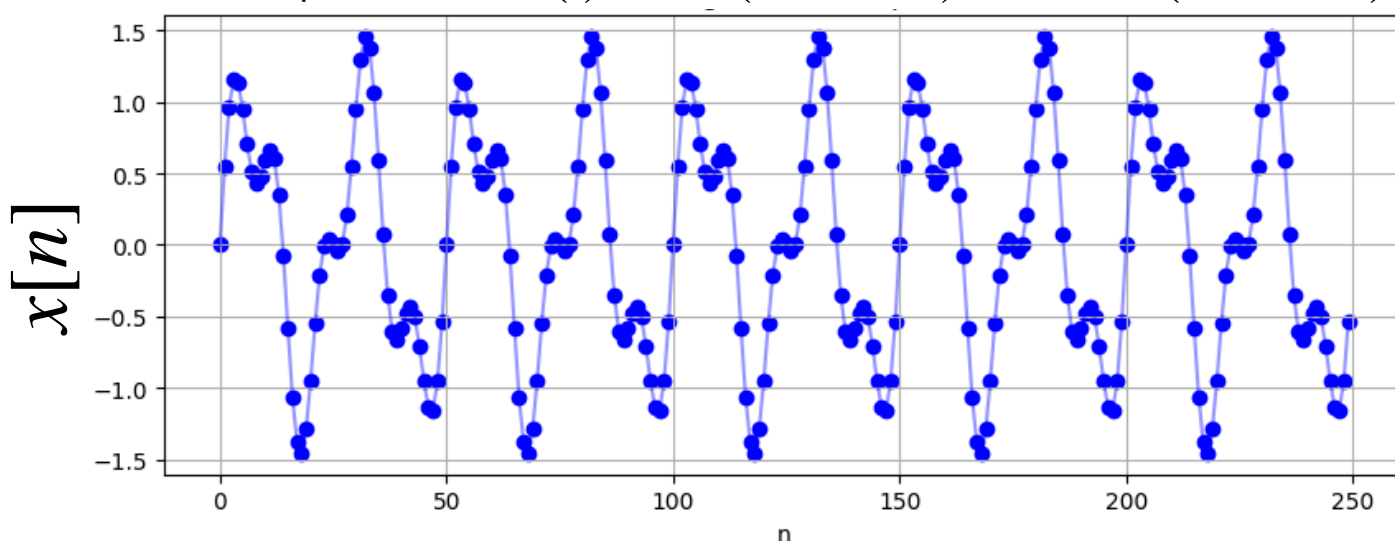
$IDFT$



A Simple Filtering

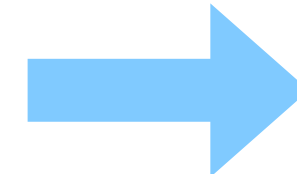
Time domain

Samples from $x(t) = \sin(2\pi \times 20t) + 0.5 \sin(2\pi \times 50t)$

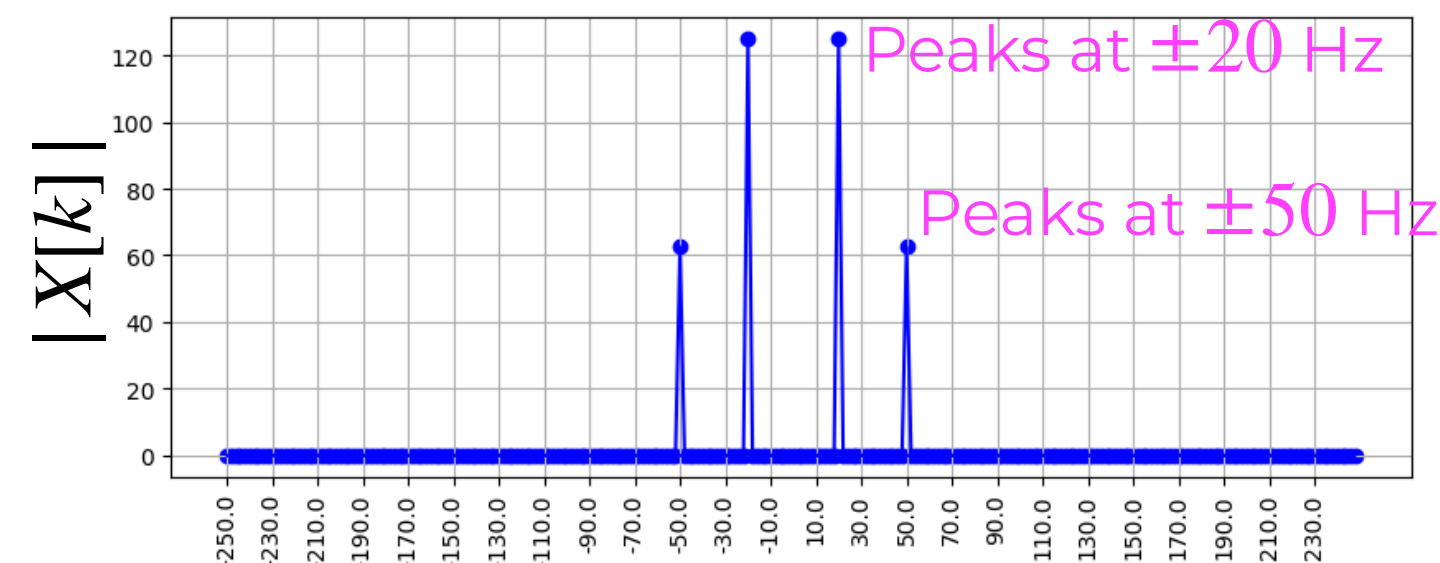


**Original
signal**

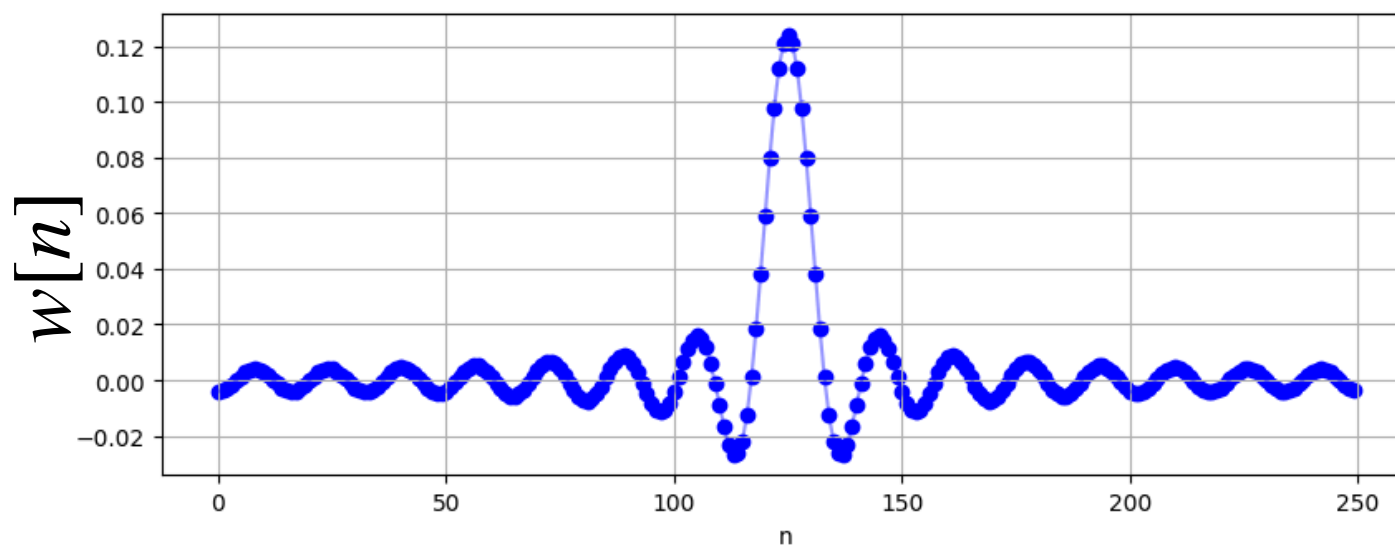
DFT



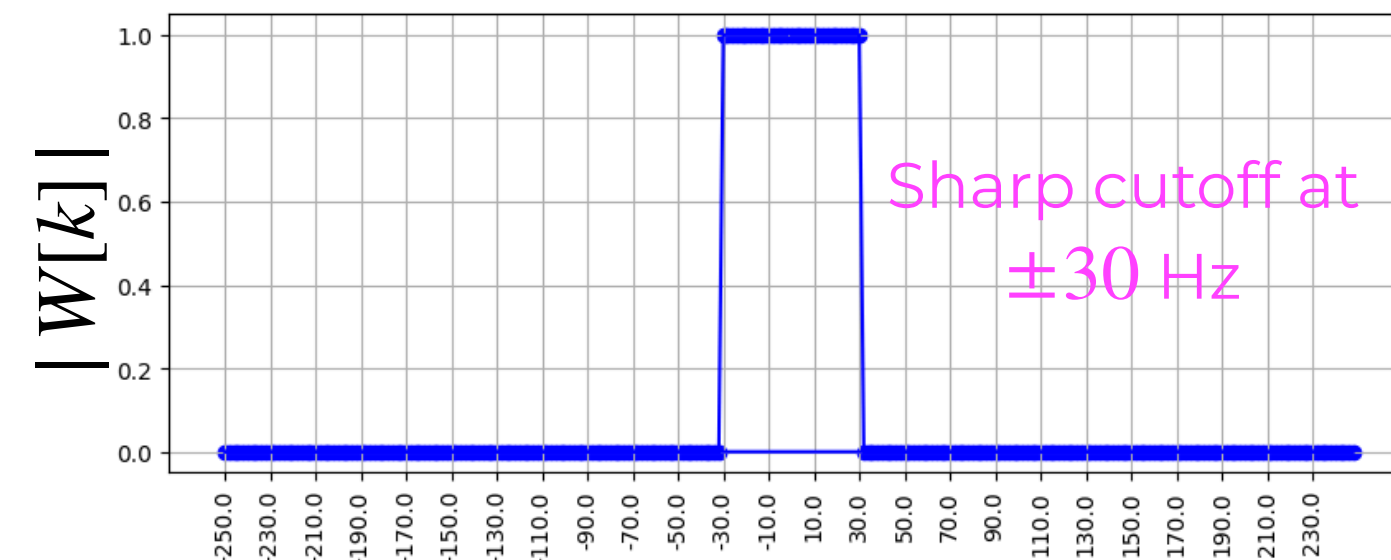
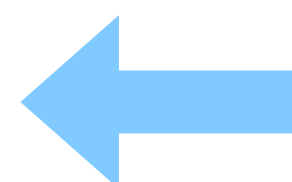
Frequency domain



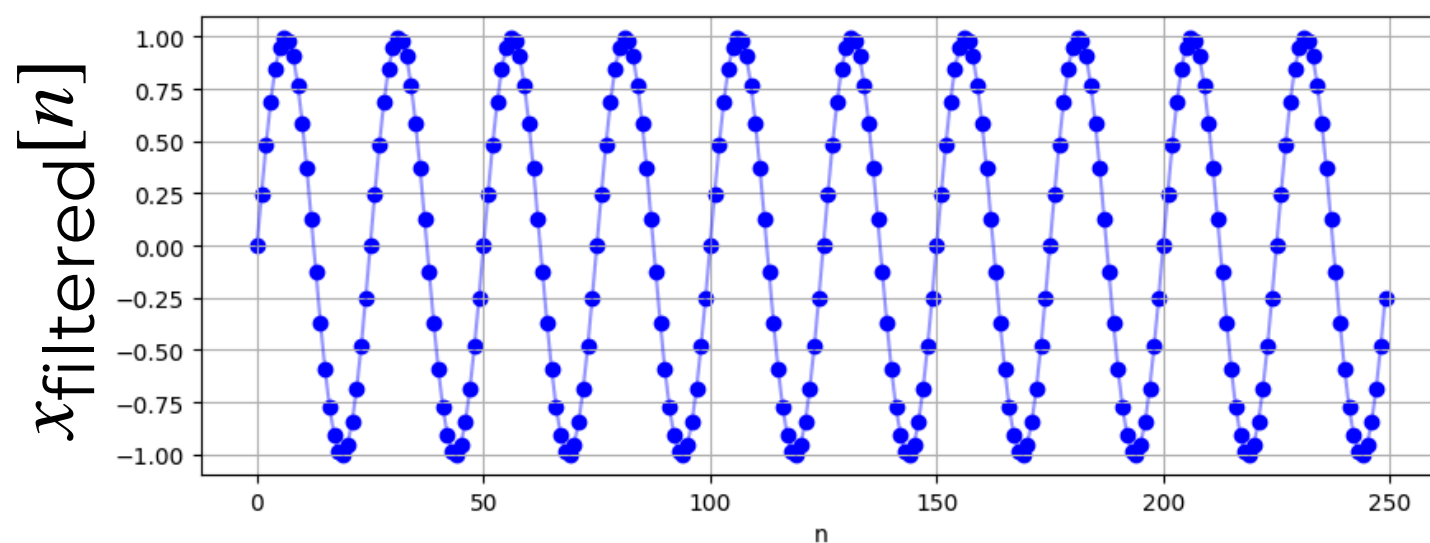
Window



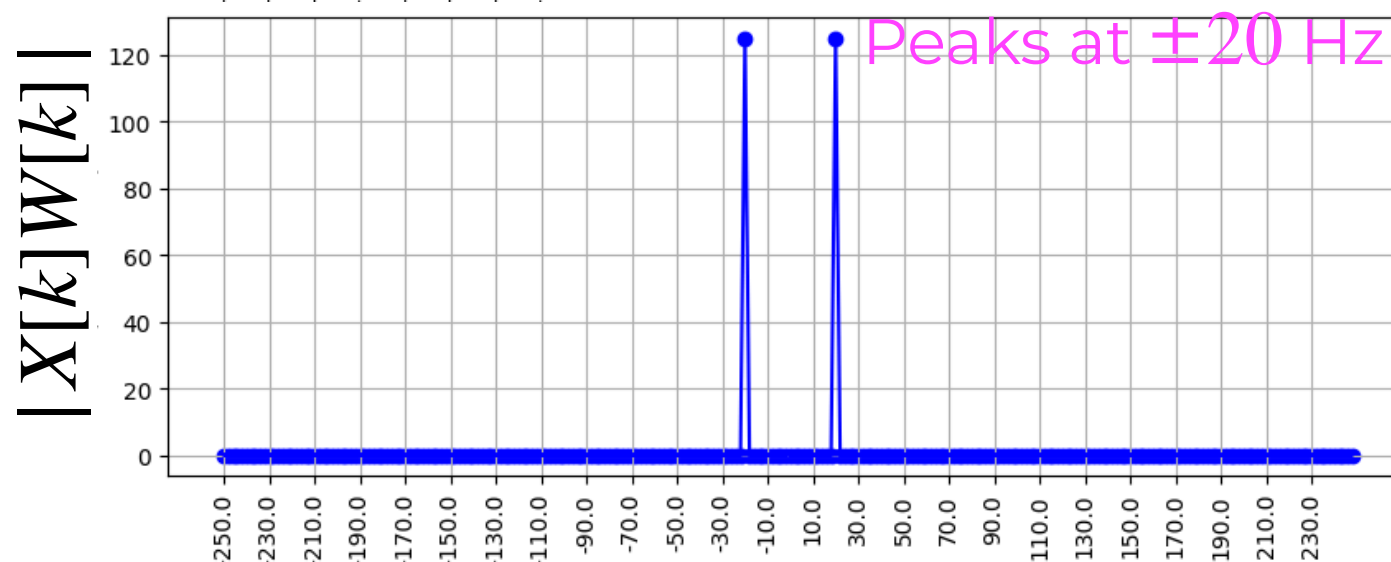
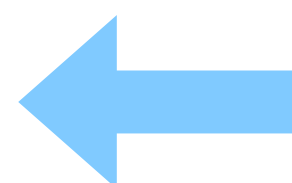
IDFT



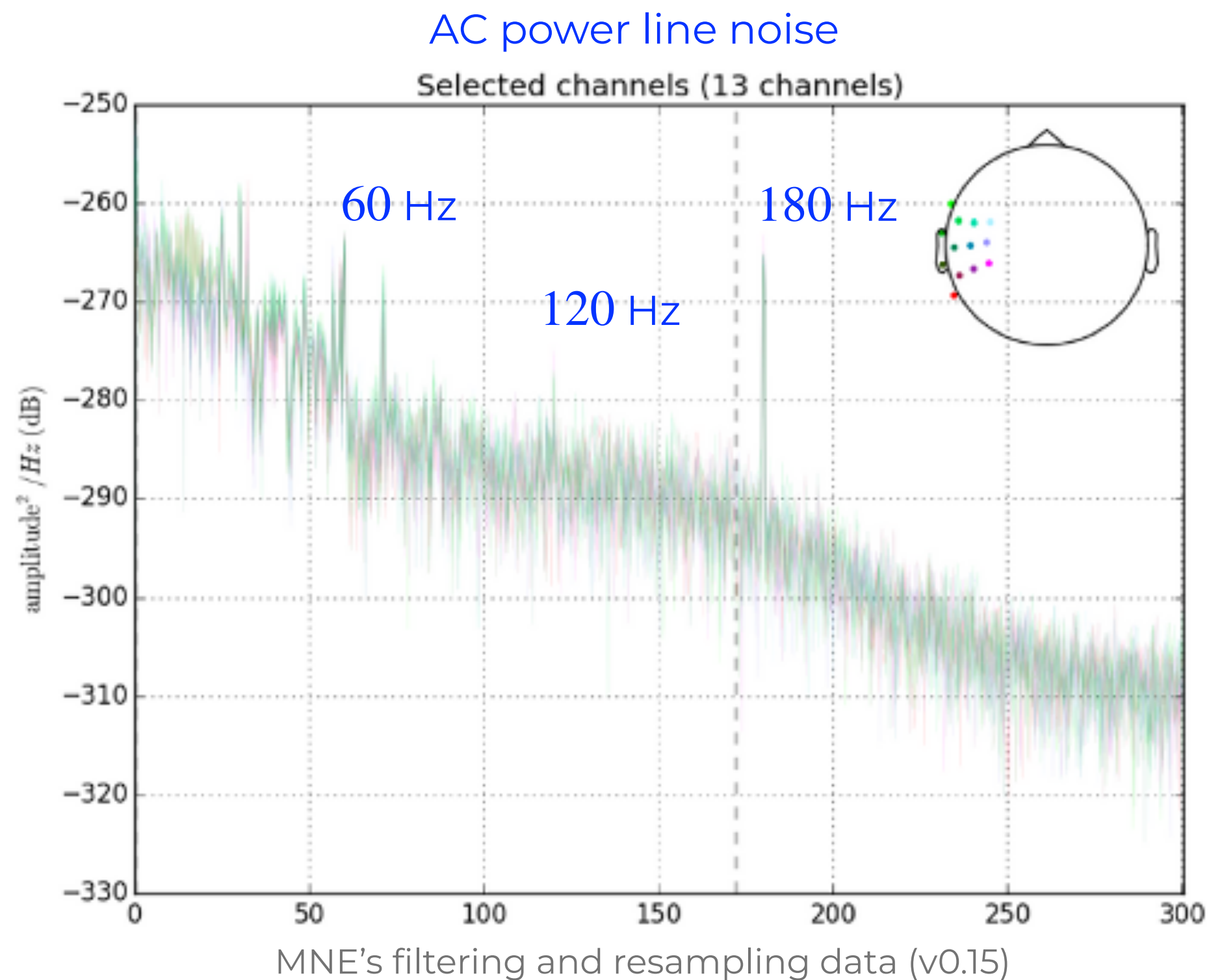
**Filtered
signal**



IDFT



EEG Preprocessing

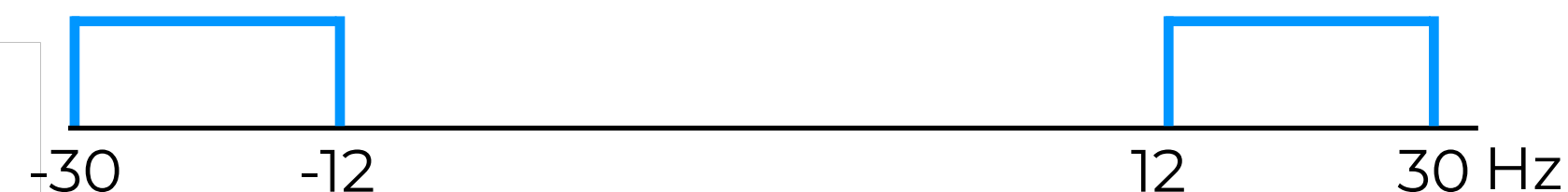


EEG Preprocessing

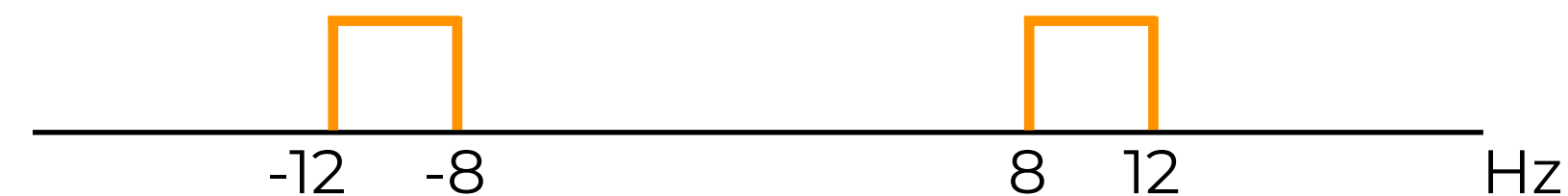
Brain rhythm frequency bands associated with cognitive processes

Ideal windows

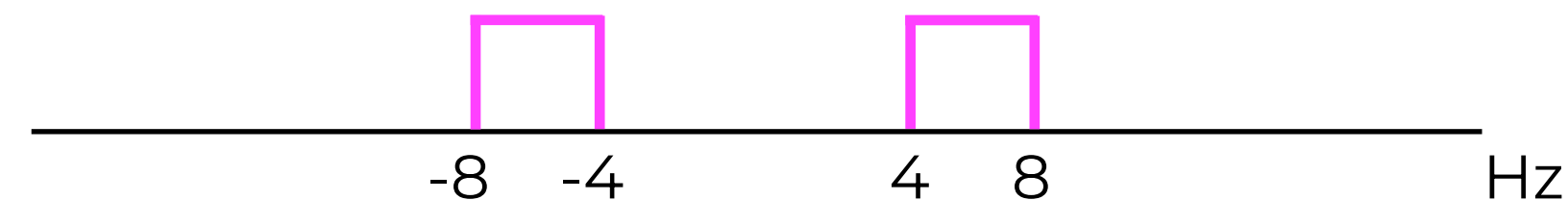
Beta
[12-30 Hz]



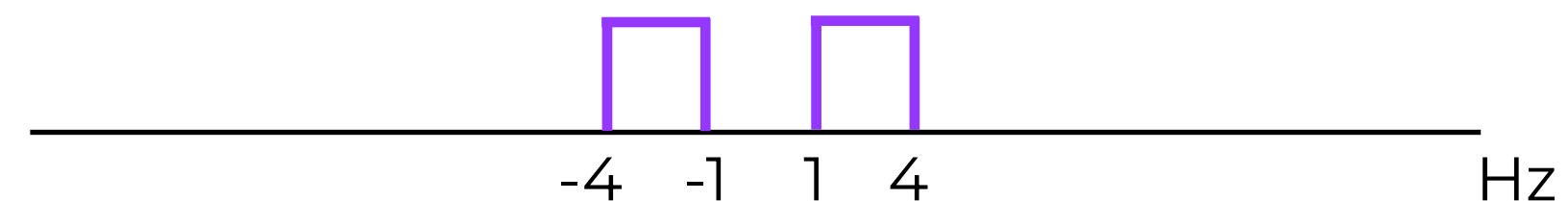
Alpha
[8-12 Hz]



Theta
[4-8 Hz]



Delta
[1-4 Hz]

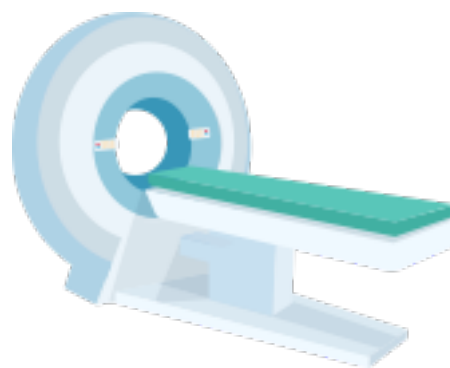


Vallat R. Compute the average bandpower of an EEG signal (2018)

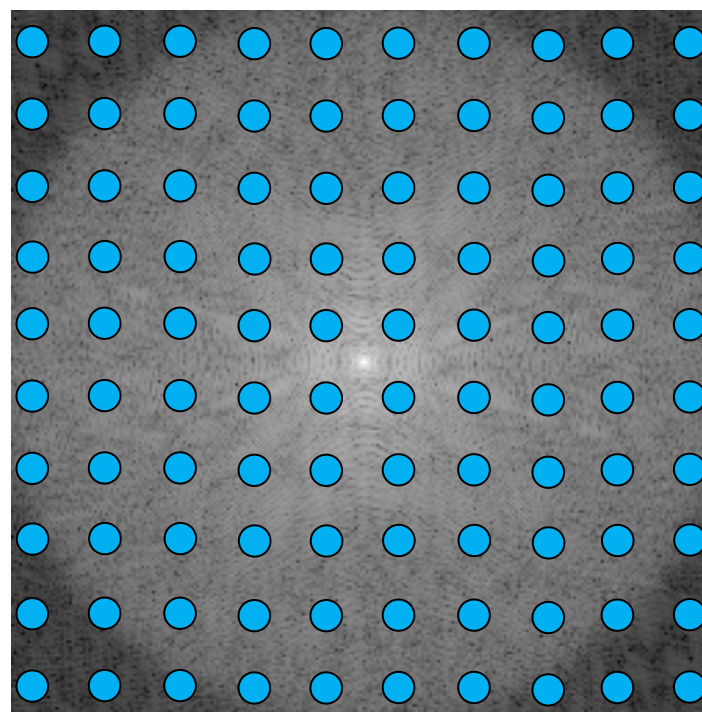
MRI Acquisition and Reconstruction

- The acquired data are the DFT samples of the object being imaged
- If the sampling rate is high enough, the image can be reconstructed by applying the inverse DFT to the k-space data

MRI scanner



Acquired data
(k-space)



2D-IDFT



Reconstructed data
(image-space)

