

Contactless Sensing

Department of Computer Science
and Engineering



INDIAN INSTITUTE OF TECHNOLOGY
KHARAGPUR

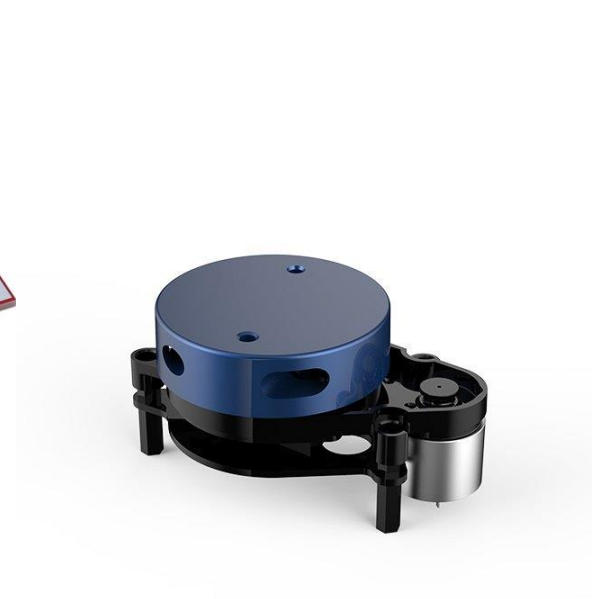
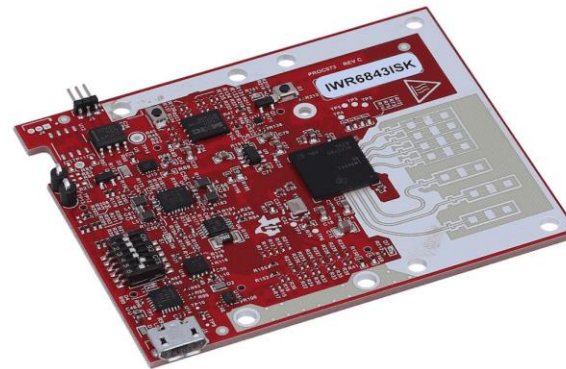
Part I: Sensing with RF: Basic Principles

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Contactless Sensing

- IMU-based sensing needs the subject to wear some devices: smartwatch, earable, smart glass, smart ring, ...
 - Might not be very convenient all the times (ex. Sleep monitoring)
- Contactless sensing are useful when the user does not want to attach the sensors with their body
 - Useful for continuous and passive monitoring of human activities
 - Widely used to monitor objects (materials, liquids, structures, ...)

Sensing Modalities

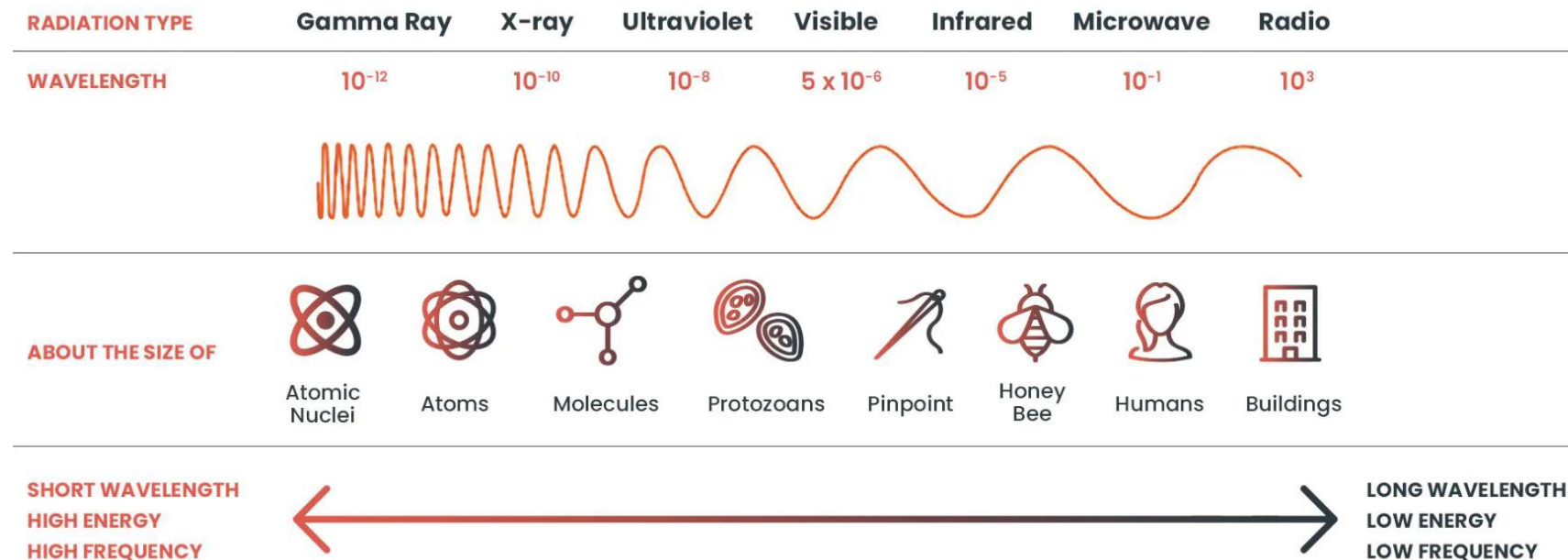


Sensing Modalities

- Primarily two modalities
 - Electromagnetic waves (majority of the sensing devices work on some EM waves)
 - Mechanical waves (acoustic-based sensing)
- The basic operating principles vary depending on the type and the frequency of the waves being used in the sensing applications

Fundamentals of EM Waves

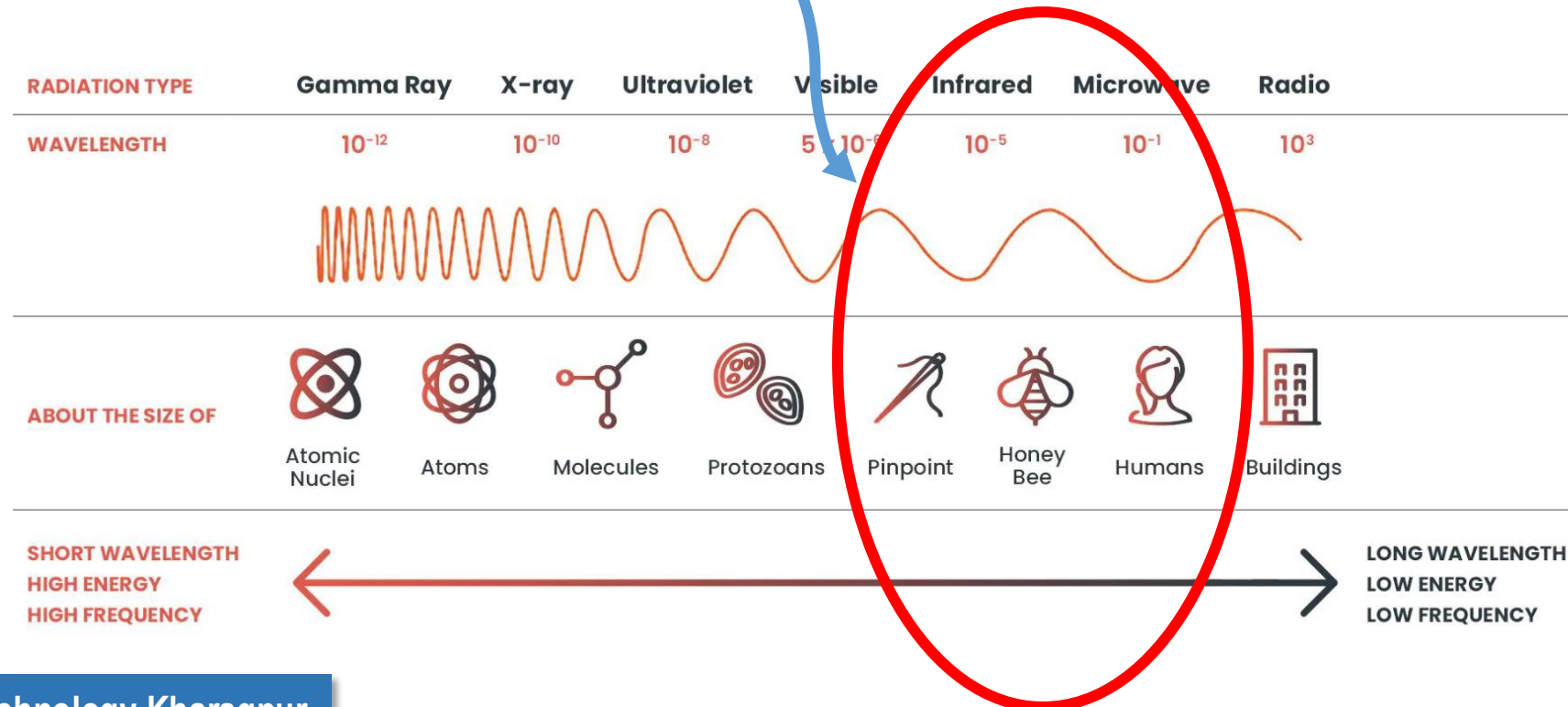
- The existence of EM was predicted by James Clerk Maxwell in 1864
 - Heinrich Hertz confirmed the same in 1887
 - Hertz also demonstrated that EM waves are affected and reflected by solid objects
- The frequency and the wavelength characterizes the waves



Fundamentals of EM Waves

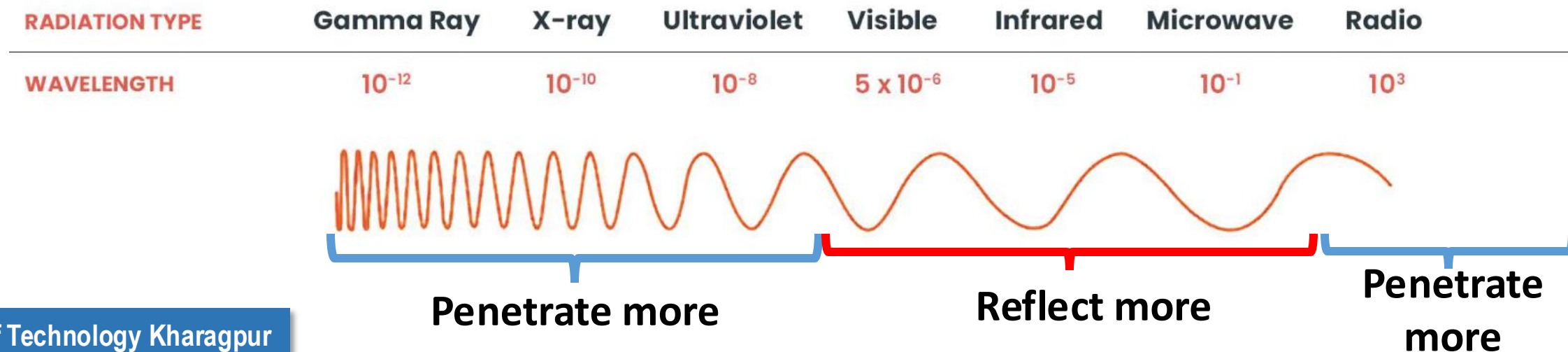
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- The frequency and the wavelength of the waves

The objects we sense



Fundamentals of EM Waves

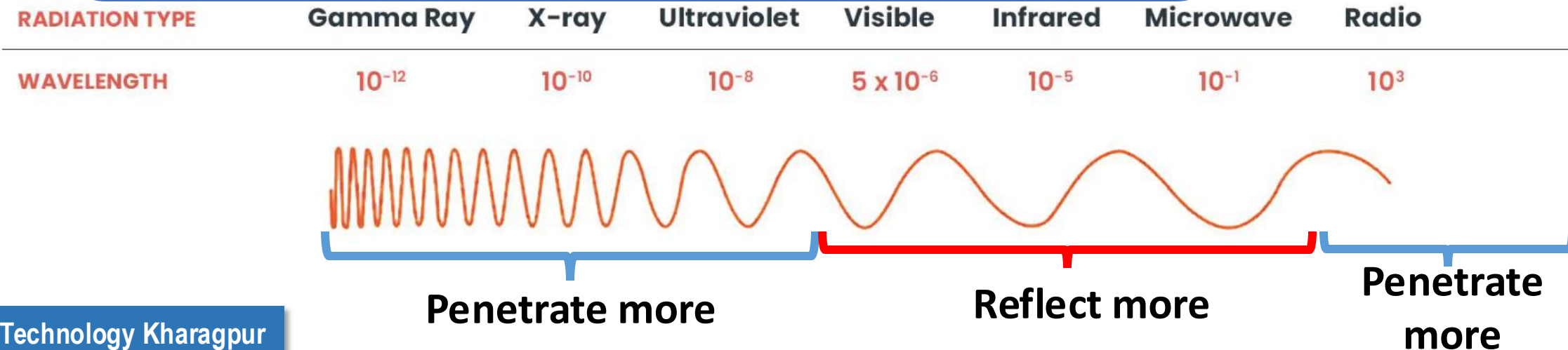
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- The frequency and the wavelength characterizes the waves
 - **Long wavelength, low frequency:** Penetrate more through physical objects
 - **Medium wavelength, medium frequency:** Reflect more from physical objects
 - **Short wavelength, high frequency:** Penetrate more through the objects



Fundamentals of EM Waves

- The existence of EM was predicted by James Clerk Maxwell in 1864
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 - Hertz also demonstrated that EM waves are affected and reflected by solid objects
- The frequency of EM waves affects its penetration and reflection by solid objects
 - Long wavelength waves penetrate more
 - Medium wavelength waves reflect more
 - Short wavelength waves penetrate less

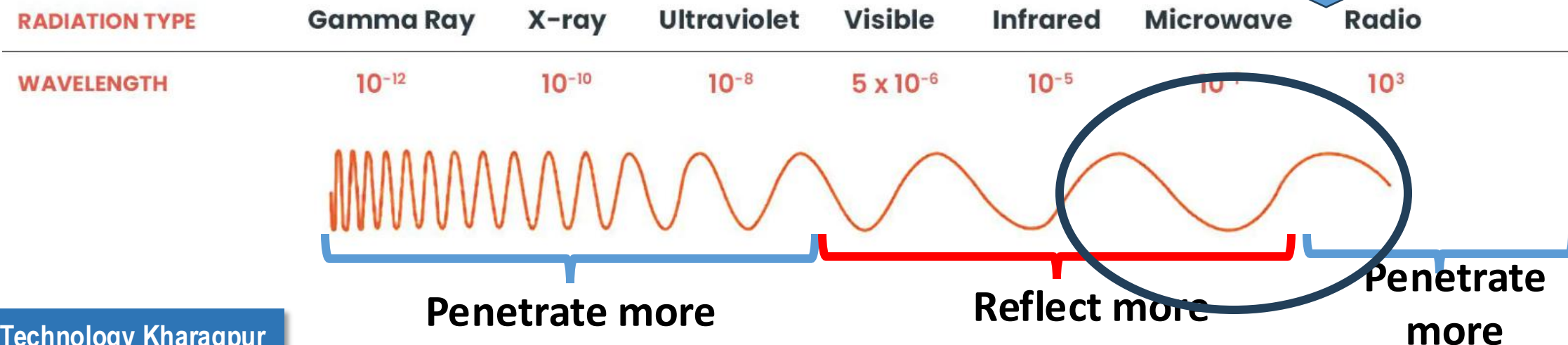
This is just a general idea; the penetration/reflection capability of the signal also depends on its bandwidth and other channel parameters



Fundamentals of EM Waves

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 - **Medium wavelength, medium frequency:** Reflect more from
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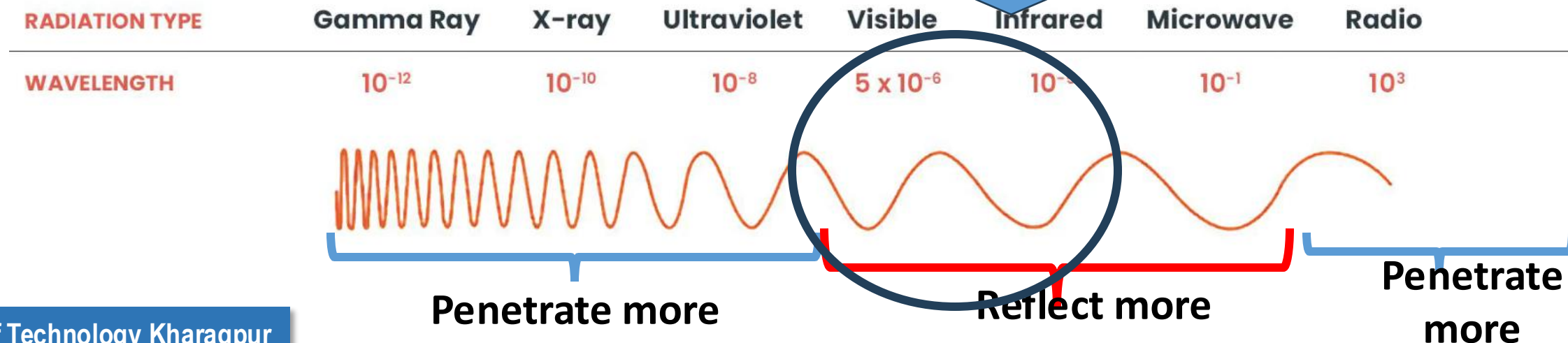
Useful for passive contactless sensing



Fundamentals of EM Waves

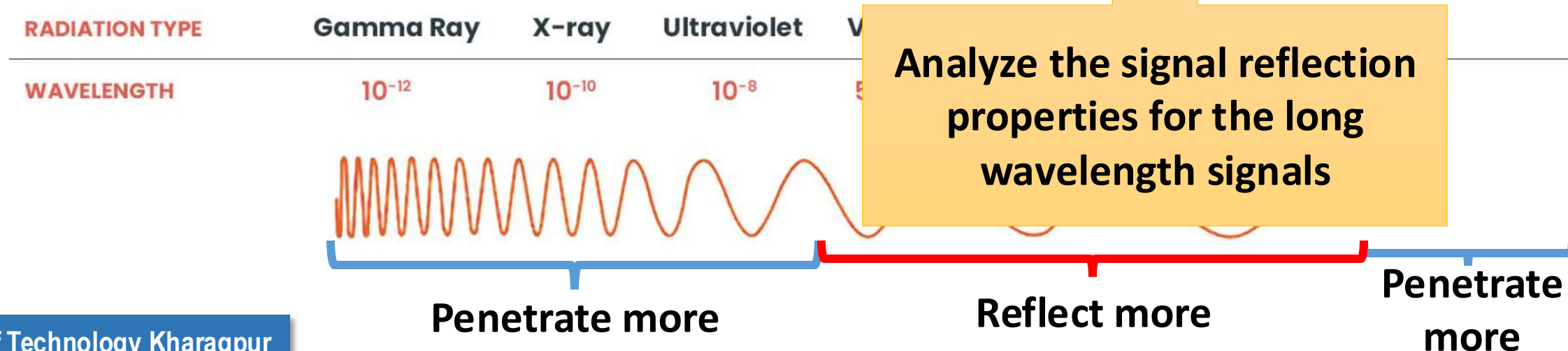
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 - Short wavelength, high frequency:

Used widely for sensing, but privacy is a concern (the vision domain)



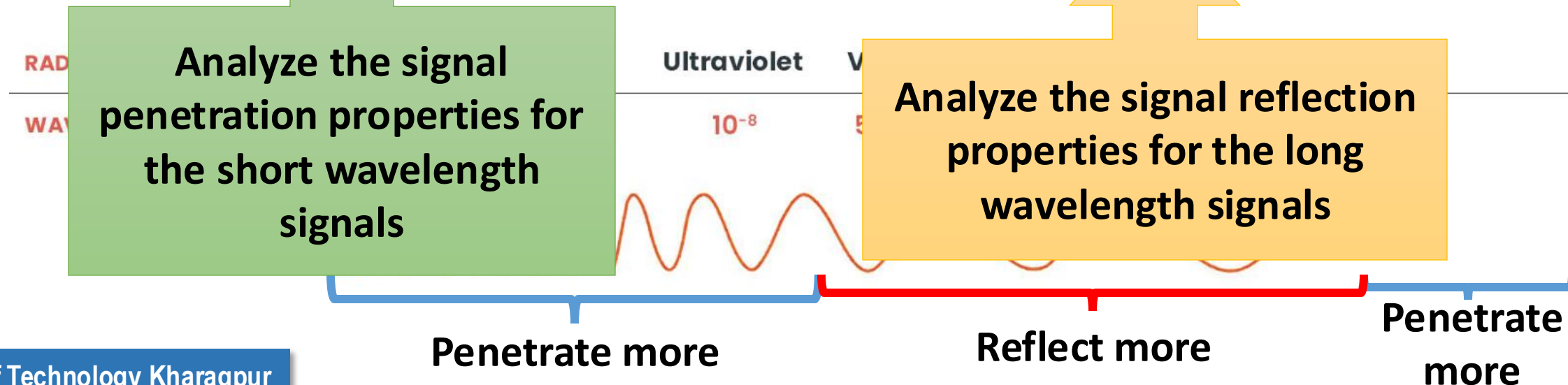
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Radio Waves

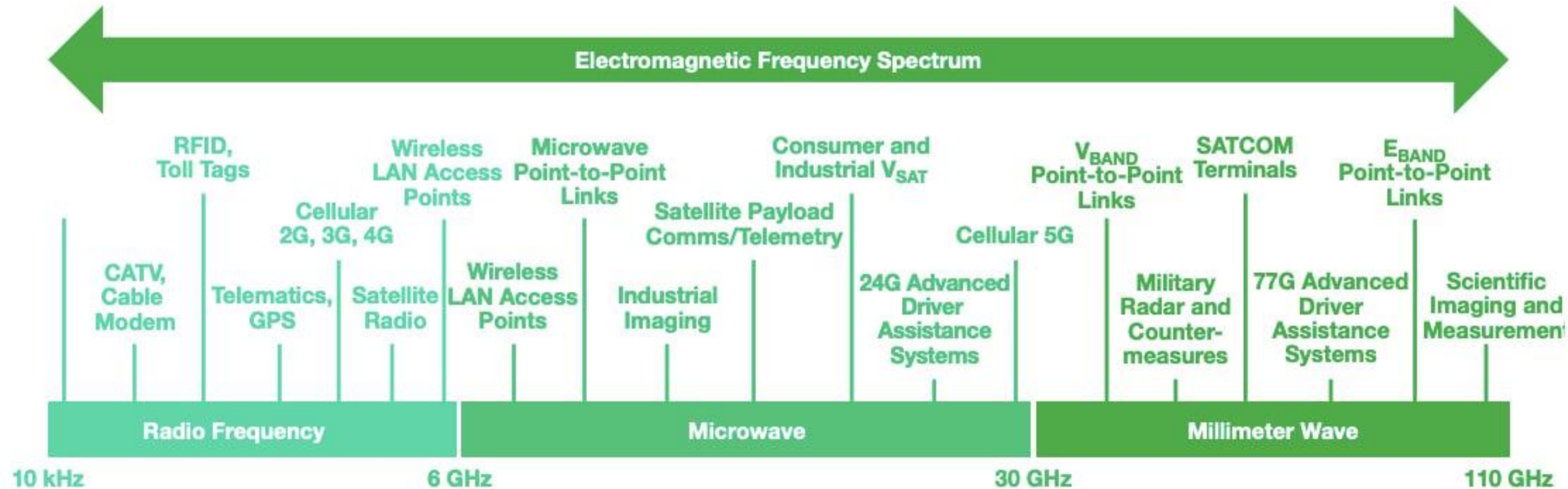


Image Source: <https://www.allaboutcircuits.com/technical-articles/basics-of-millimeter-wave-mmwave-technology/>

Radio Waves

We'll see sensing applications in these ranges

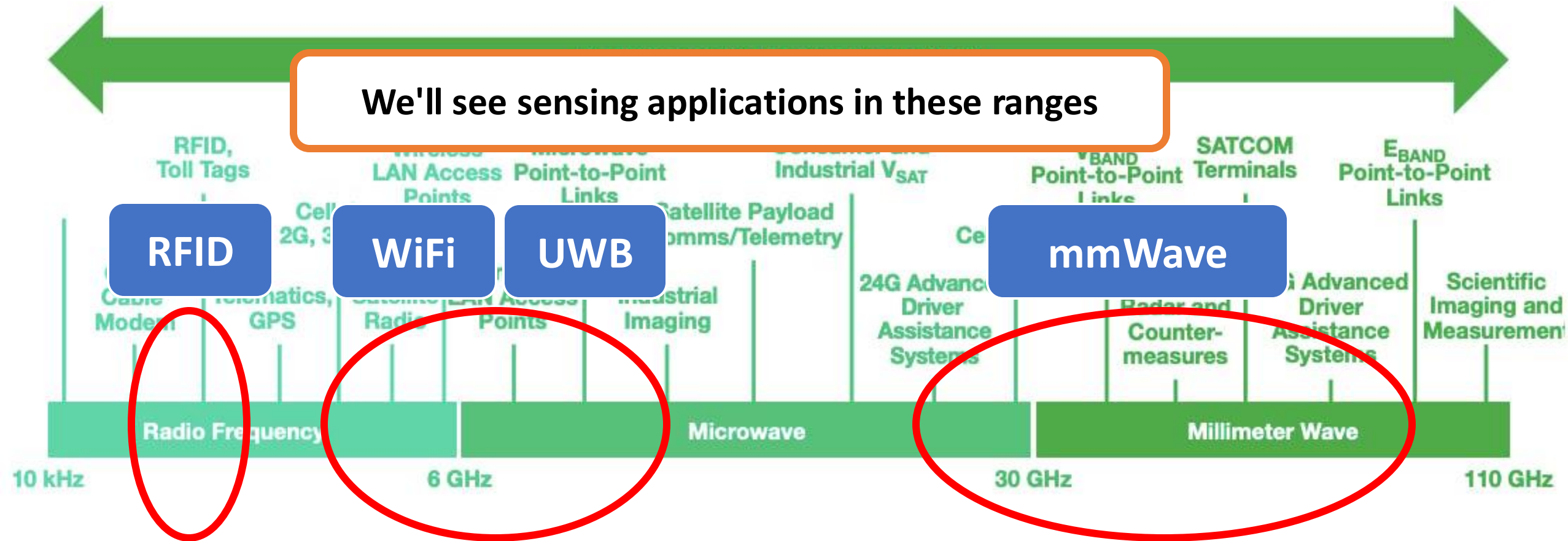
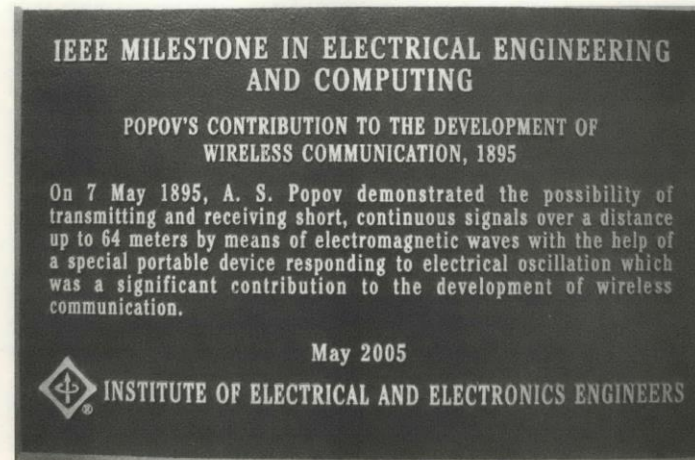
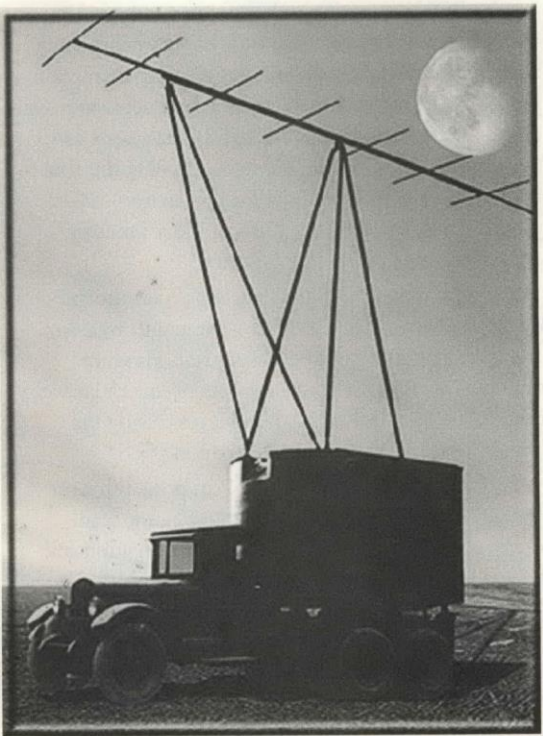


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Signal Interference to Detect Motion: The Concept of Radar

- 1897: Alexander Popov, a Russian Imperial Navy physicist, was testing an early version of wireless communication between two ships in Baltic Sea
 - He observed an interference wave pattern caused by a third ship
 - Popov proposed the idea that this might be used to detect moving objects



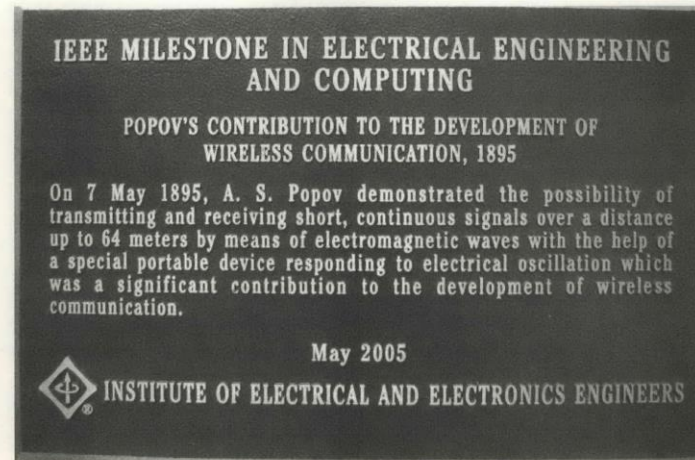
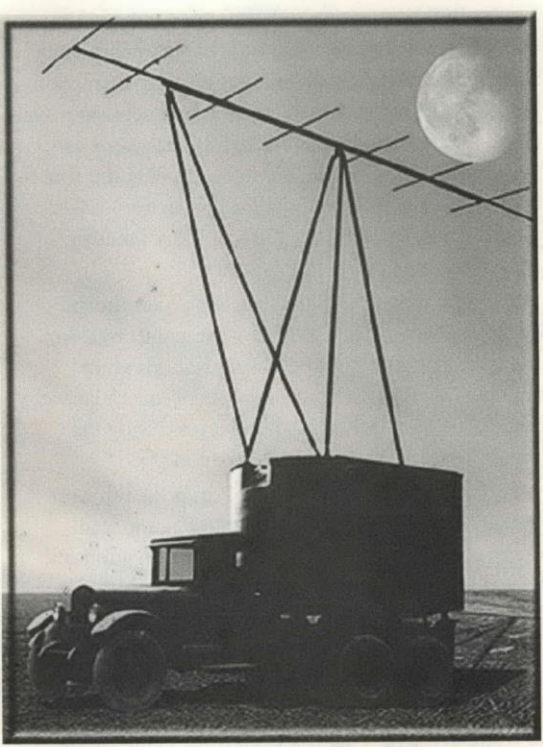
<https://ieeexplore.ieee.org/abstract/document/5282288>

**A Brief
History of Radar
in the
Soviet Union
and
Russia**

*V.S. Chernyak
I. Ya. Immoreev*

Signal Interference to Detect Motion: The Concept of Radar

- 1932: Radar as a technical equipment was proposed by a military engineer Piotr Oshchepkov
 - "RUS-1": The first industrial radar (1939): 4m wavelength, transmitter and receiver seperated by 35 km



<https://ieeexplore.ieee.org/abstract/document/5282288>

**A Brief
History of Radar
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Basic Principle: The Doppler Effect

- Change in the frequency of the wave in relation to an observer who is moving relative to the wave source
 - **Wave source moves towards the observer:** Each successive wave cycle is emitted from a position closer to the observer
 - Time between cycles is reduced; frequency is increased
 - **Wave source moves away from the observer:** Each successive wave cycle is emitted from a position farther from the observer
 - Time between cycles is increased; frequency is decreased

 © 2008 Christian Pflüger

Image Source: Wikipedia

The Doppler Effect

- Let,
 - f_0 is the emitted frequency
 - f is the observed frequency
 - c is the propagation speed of the wave in the medium
 - v_r is the speed of the receiver, v_s is the speed of the source

$$f = \left(\frac{c \pm v_r}{c \mp v_s} \right) f_0$$

- v_r is added to c if the receiver is moving towards the source, subtracted if the receiver is moving away from the source (opposite for v_s)

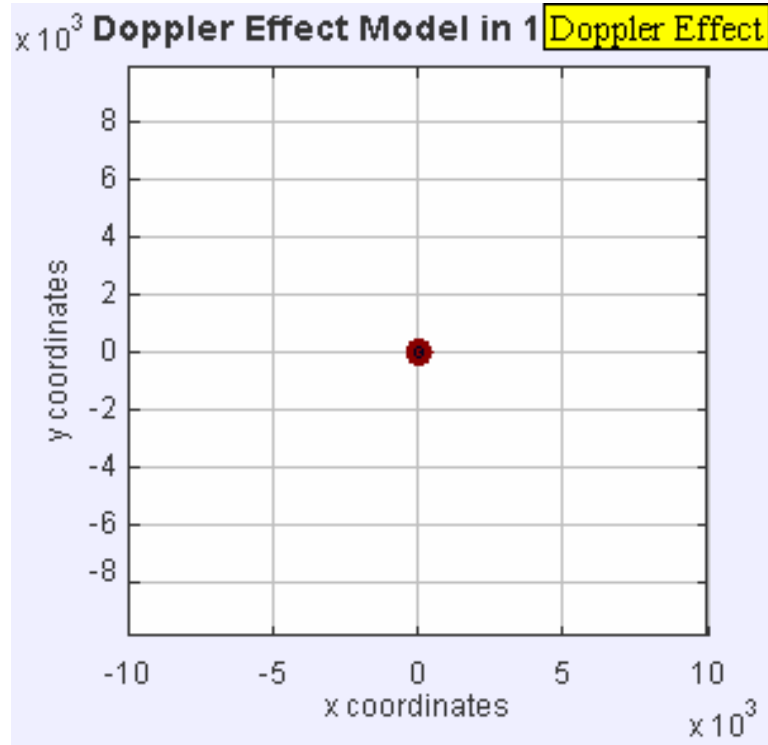
The Doppler Effect

- Equivalently, if the source is directly approaching or receding from the observer,

$$\frac{f}{v_{wr}} = \frac{f_0}{v_{ws}} = \frac{1}{\lambda}$$

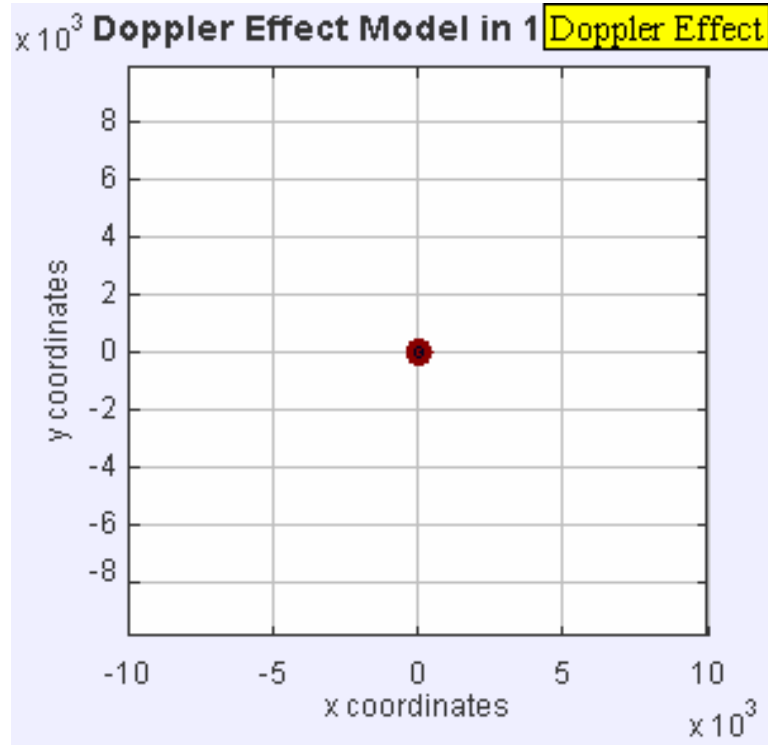
- v_{wr} is the wave speed related to the receiver, v_{ws} is the wave speed related to the source
- λ is the wavelength

Doppler Effect for the Sound Sources

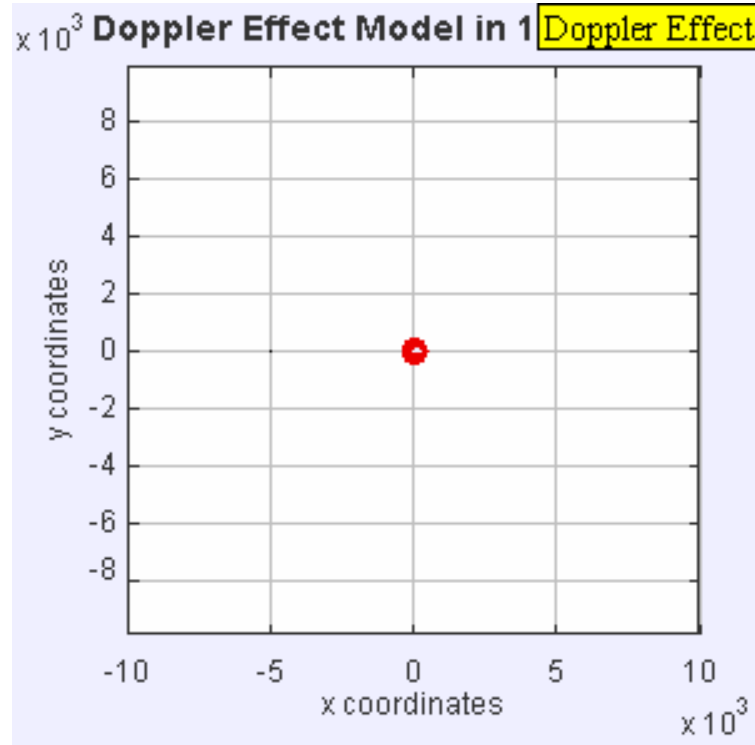


Stationary sound source

Doppler Effect for the Sound Sources

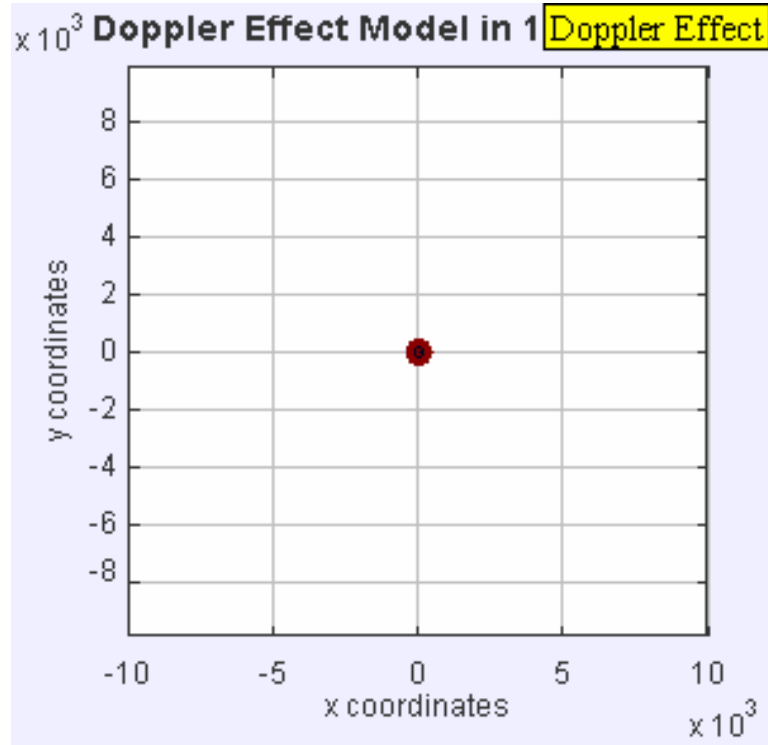


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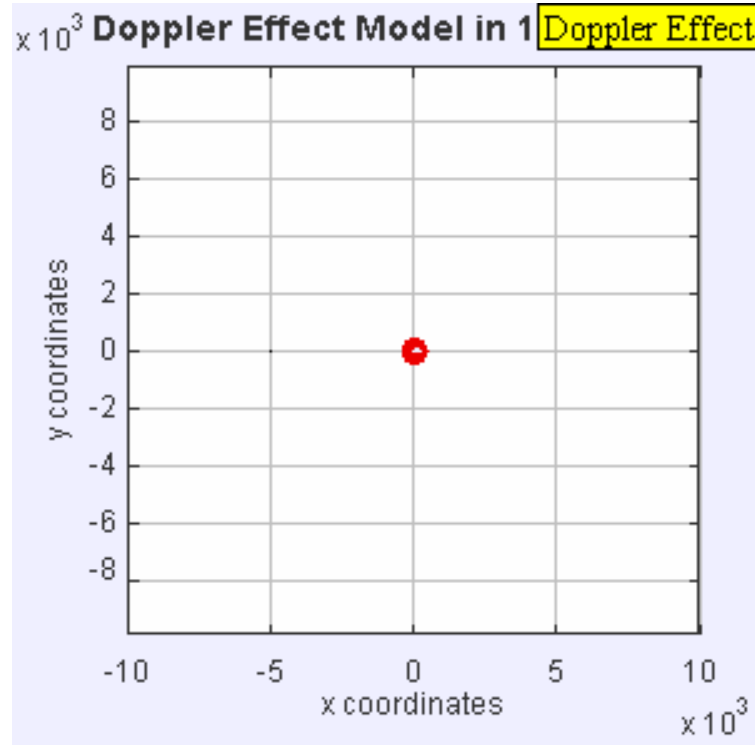


Source moves at a speed $0.7c$

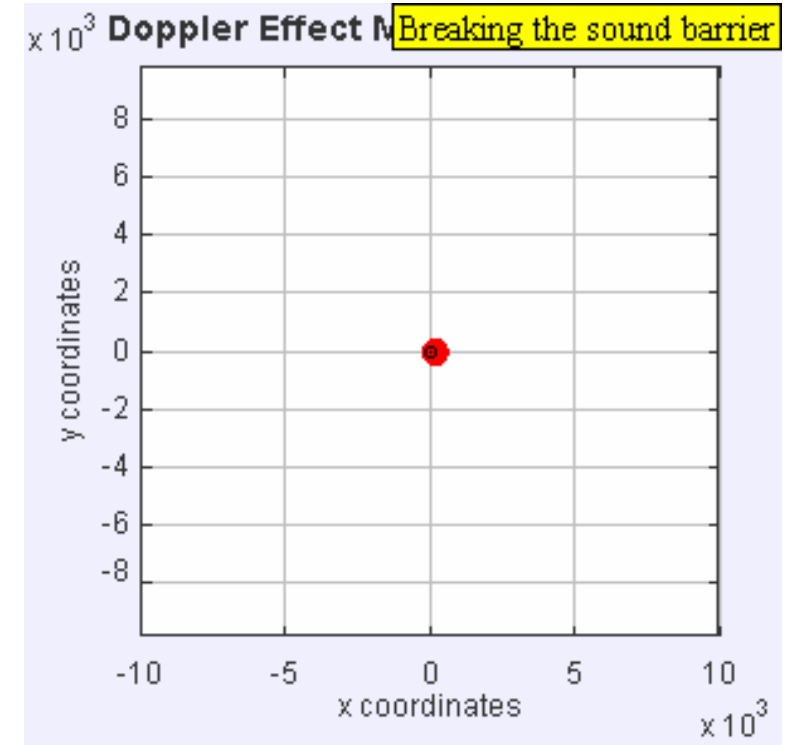
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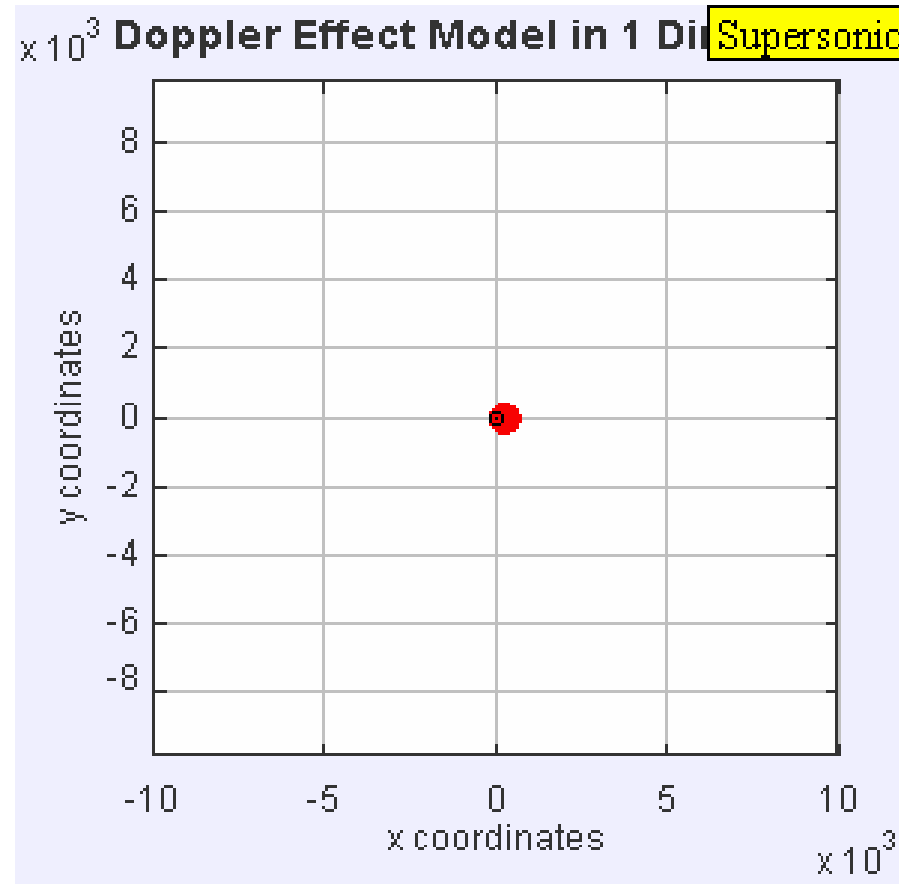


Source moves at a speed $0.7c$



Source moves at a speed c

Doppler Effect for the Sound Sources



Advancing wavefront

**Creates a shock wave
and consequently the
sonic boom**

Source moves at a speed $1.4c$

The Wave: Concept of Bandwidth

- Any arbitrary wave signal can be decomposed into a set of sinusoidal wave of multiple frequencies
 - Called the frequency components of the wave signal (or simply, the signal)

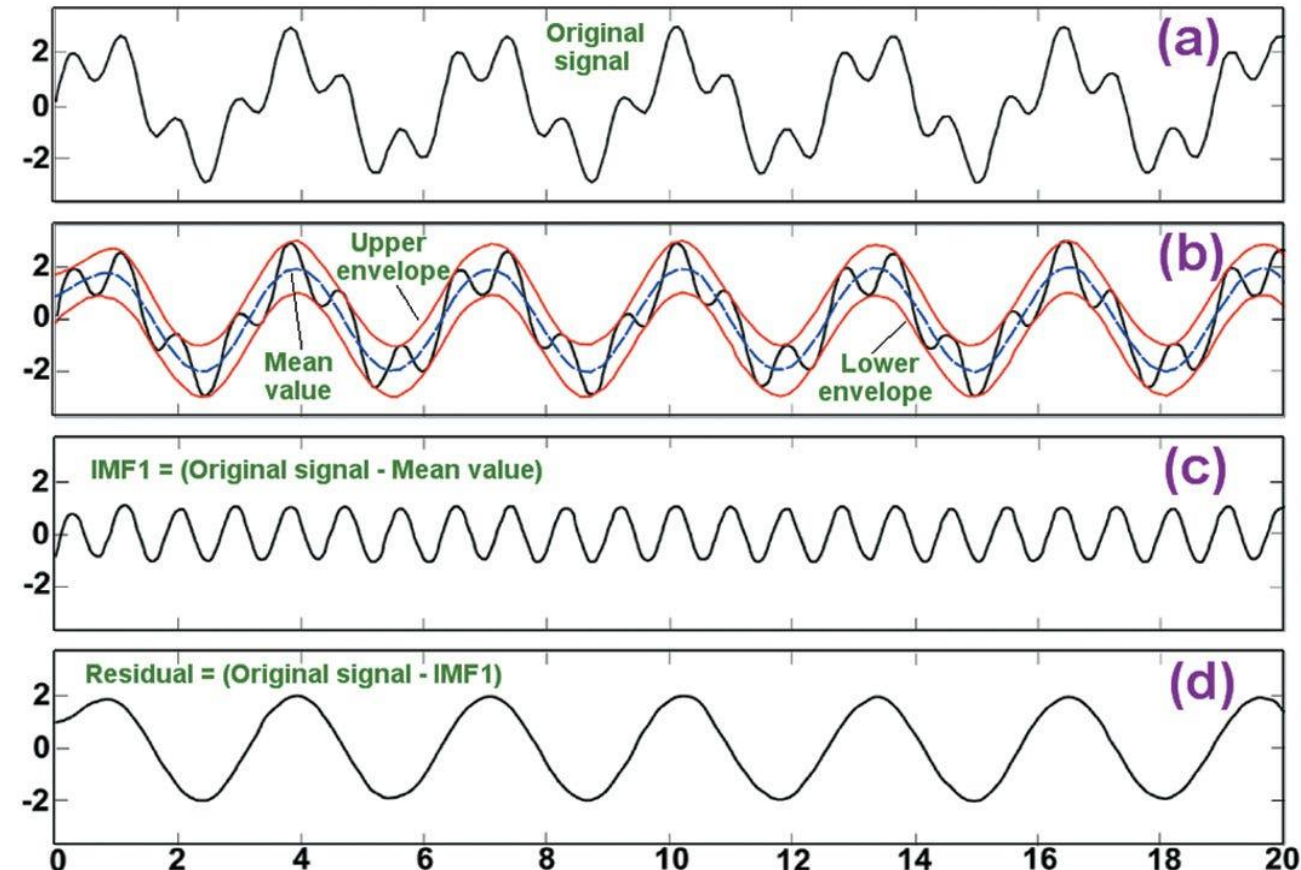
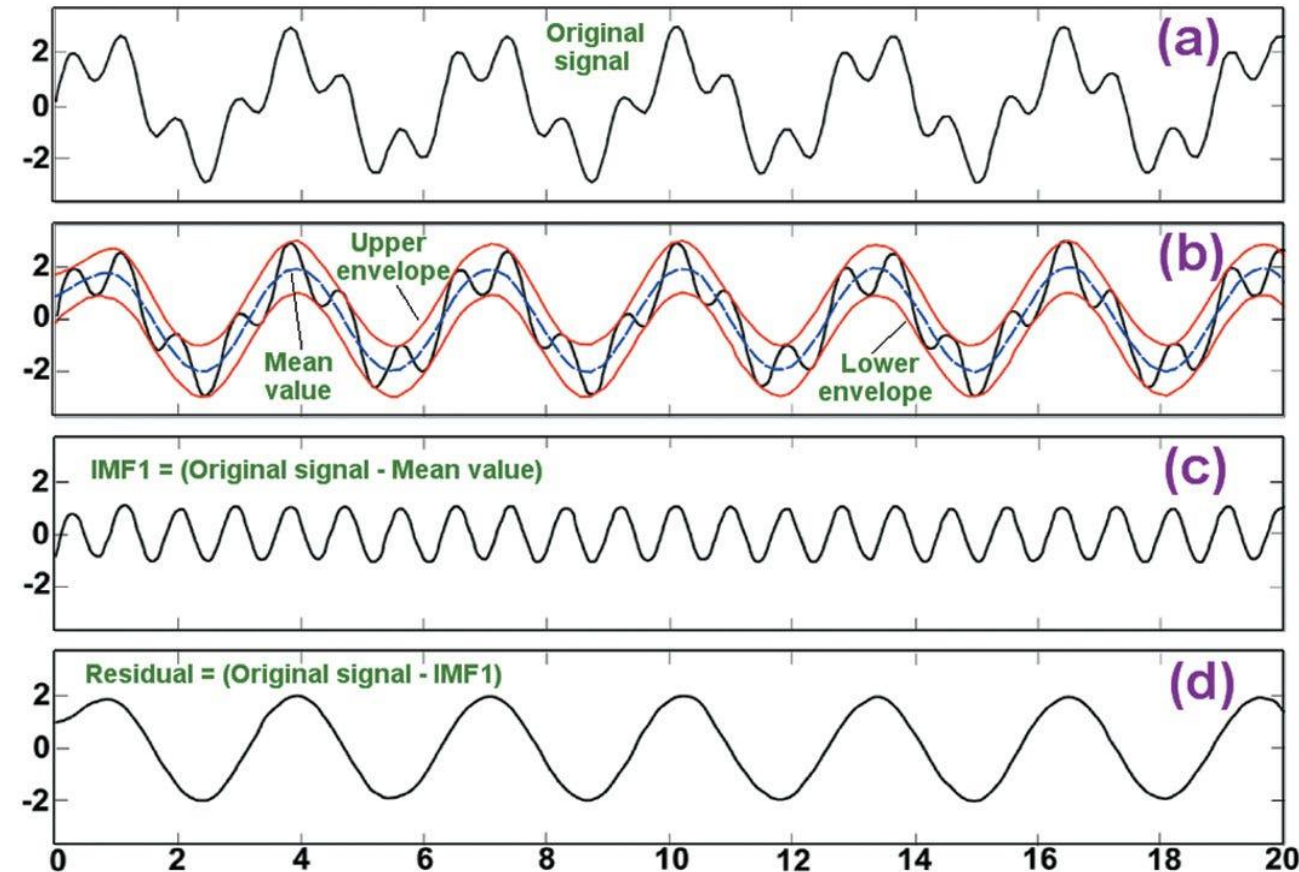


Image Source:

<https://towardsdatascience.com/decomposing-signal-using-empirical-mode-decomposition-algorithm-explanation-for-dummy-93a93304c541>

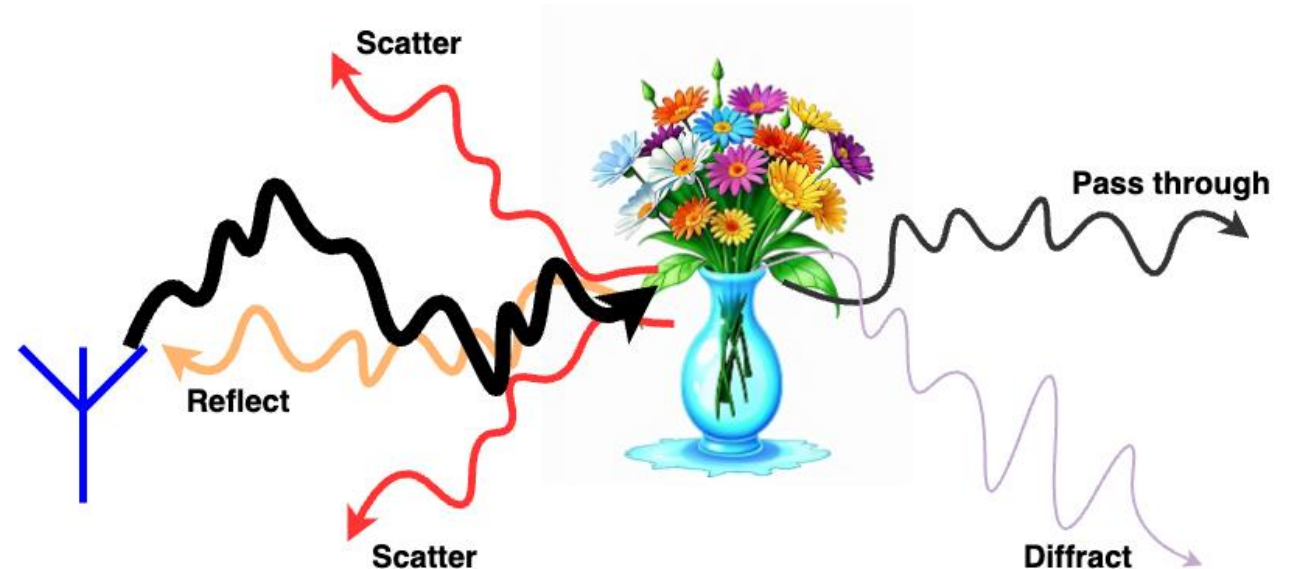
The Wave: Concept of Bandwidth

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- The bandwidth of a signal is the difference between the highest and the lowest frequency components of that signal
 - We assume that any of the frequency components between the highest and the lowest can be present in the signal



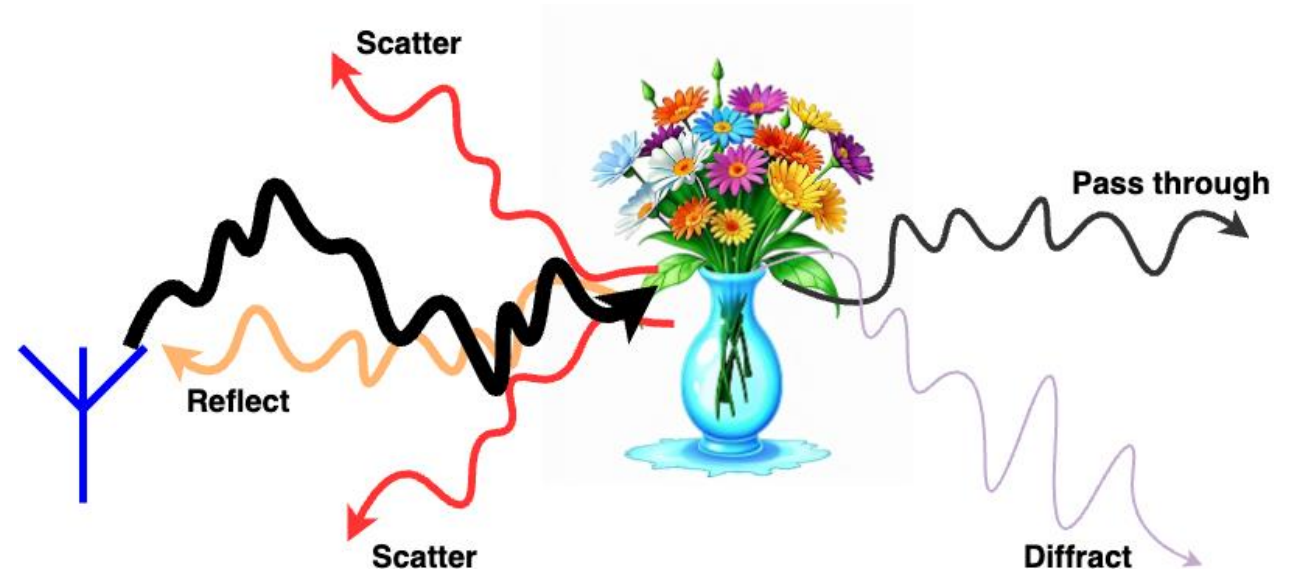
Impact of Bandwidth on Sensing

- Higher bandwidth signal means it is likely to have more number of signal components
 - Therefore, scattering, diffraction, reflection, etc., are likely to be more on high-bandwidth signals



Impact of Bandwidth on Sensing

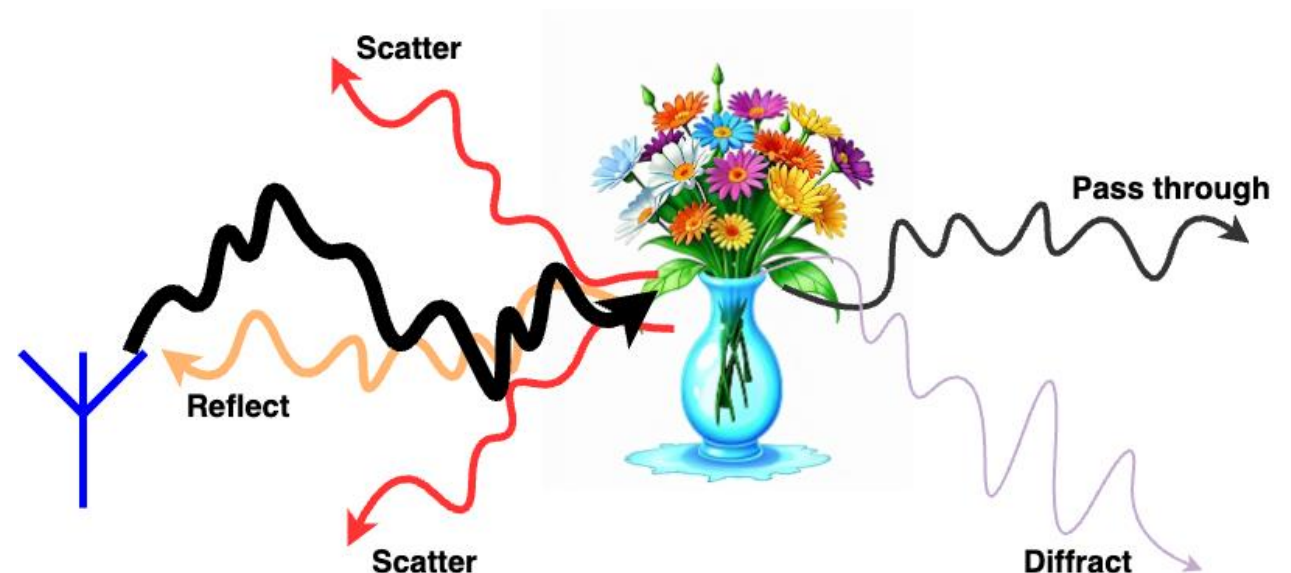
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- **Although create noises for communication, but good for sensing**



Impact of Bandwidth on Sensing

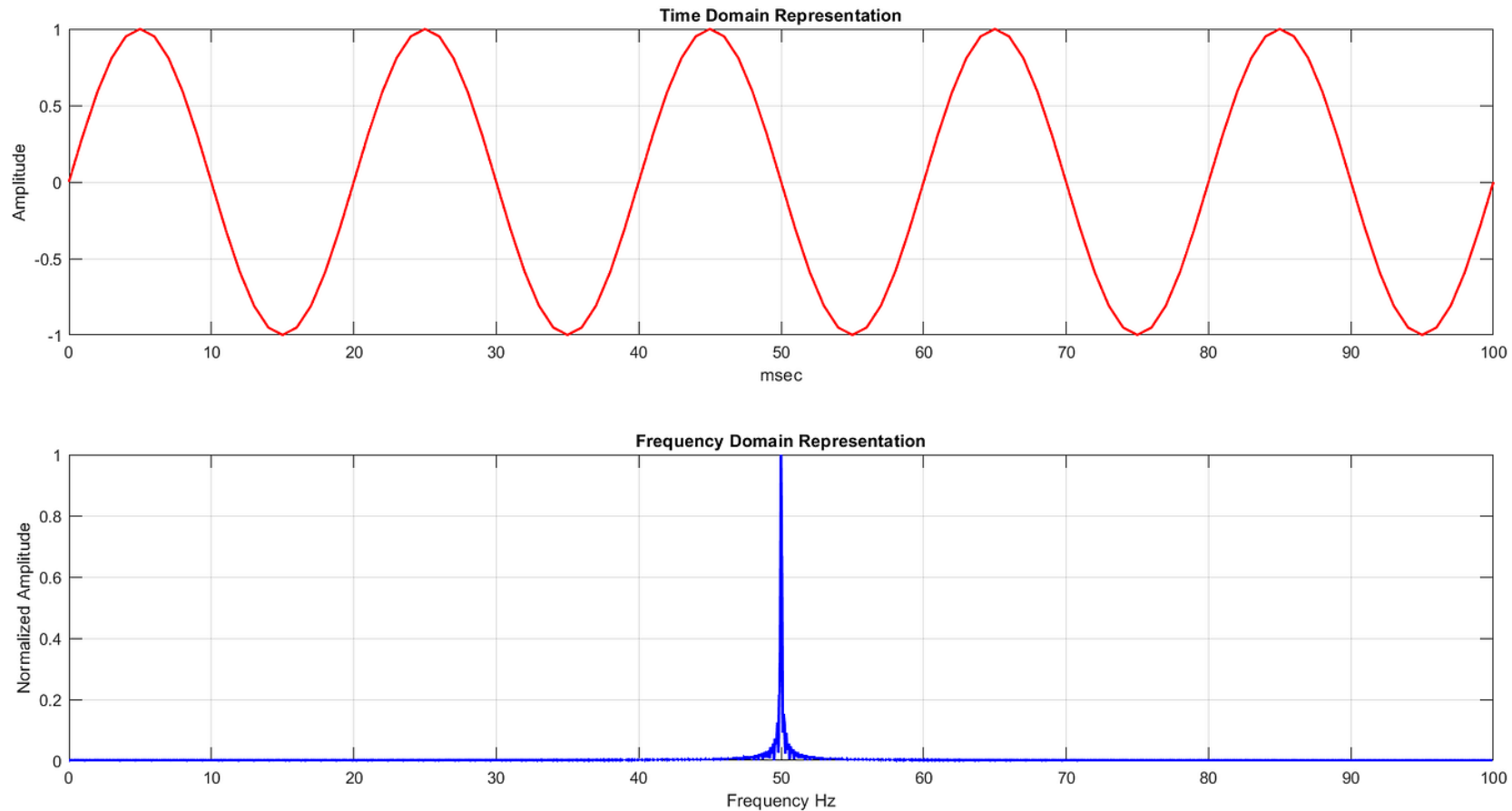
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How do we get the various frequency components of a signal?



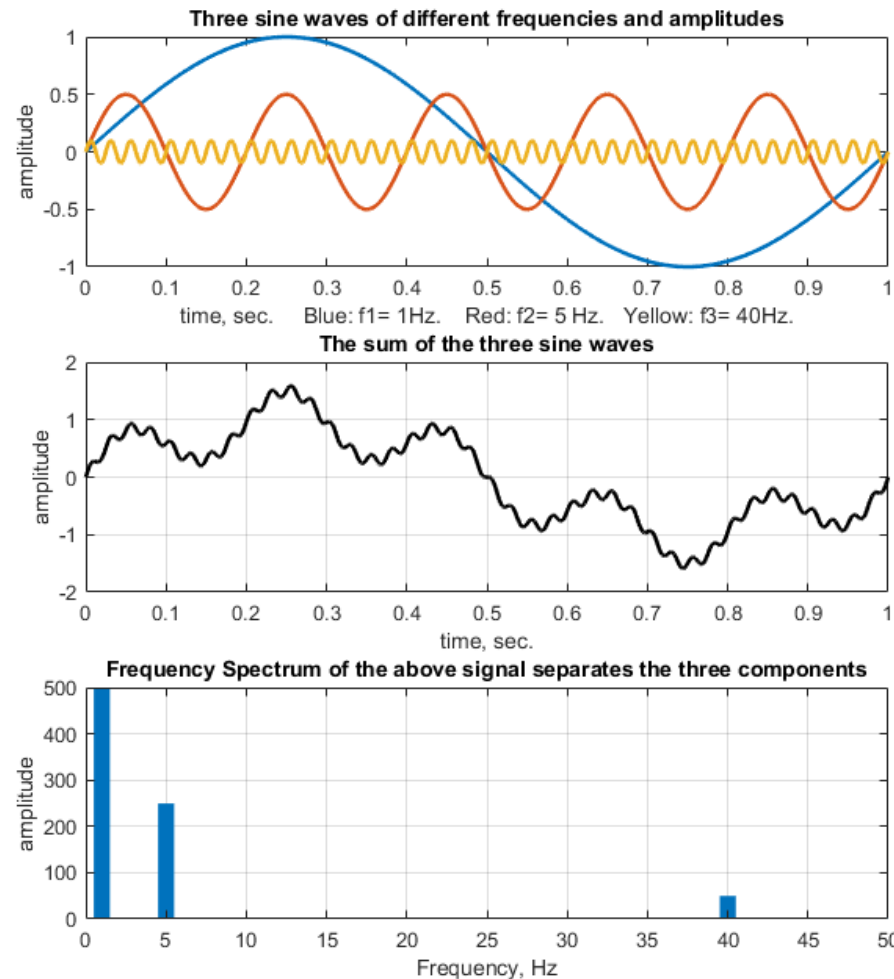
Fourier Analysis

- Extracts the frequency components of a signal



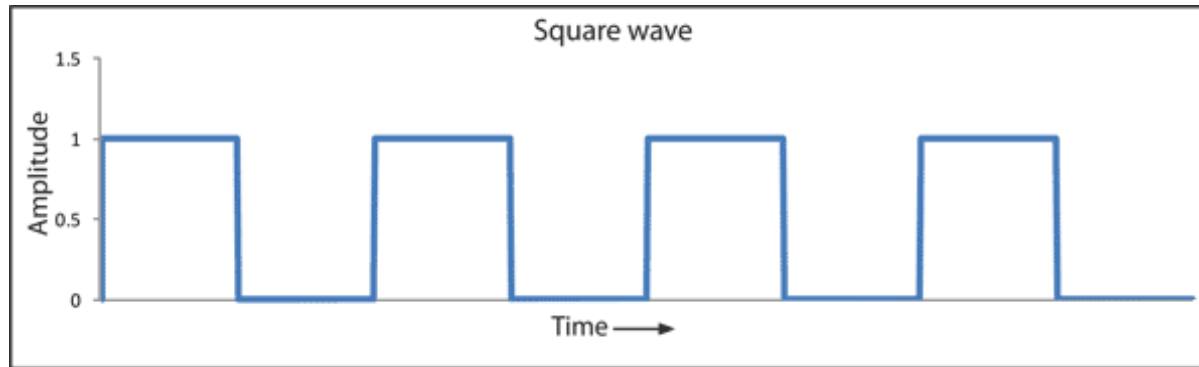
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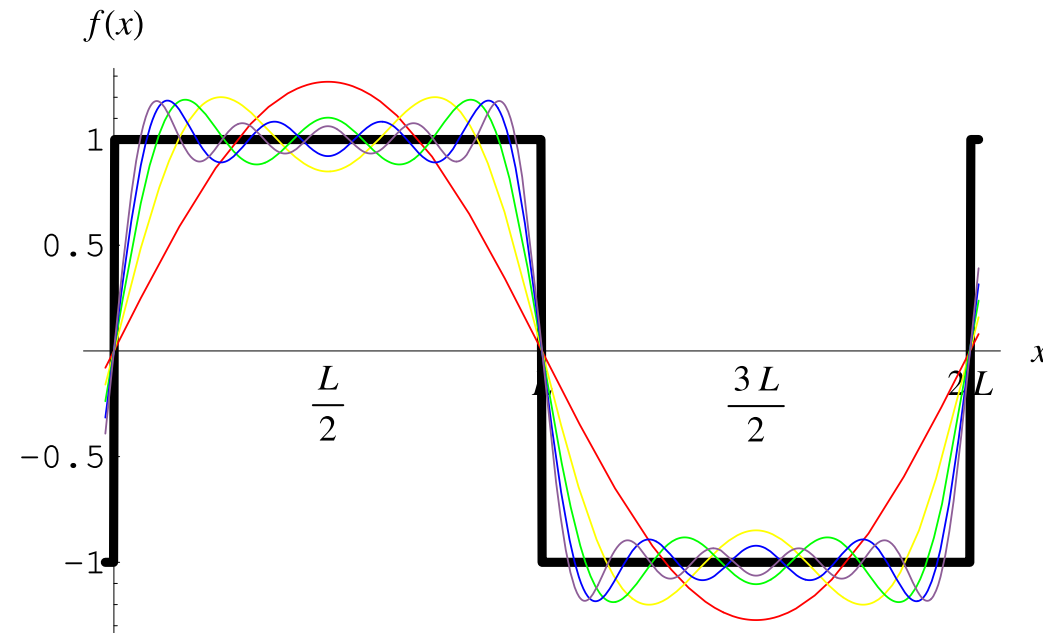
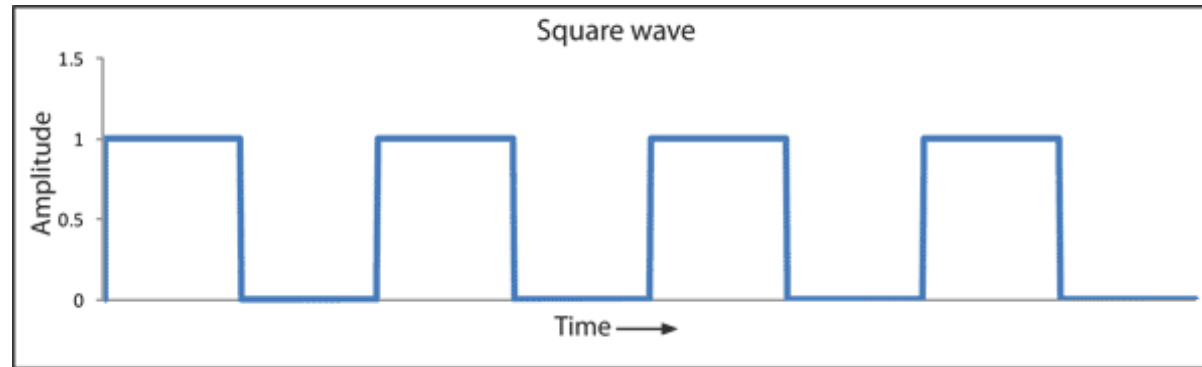
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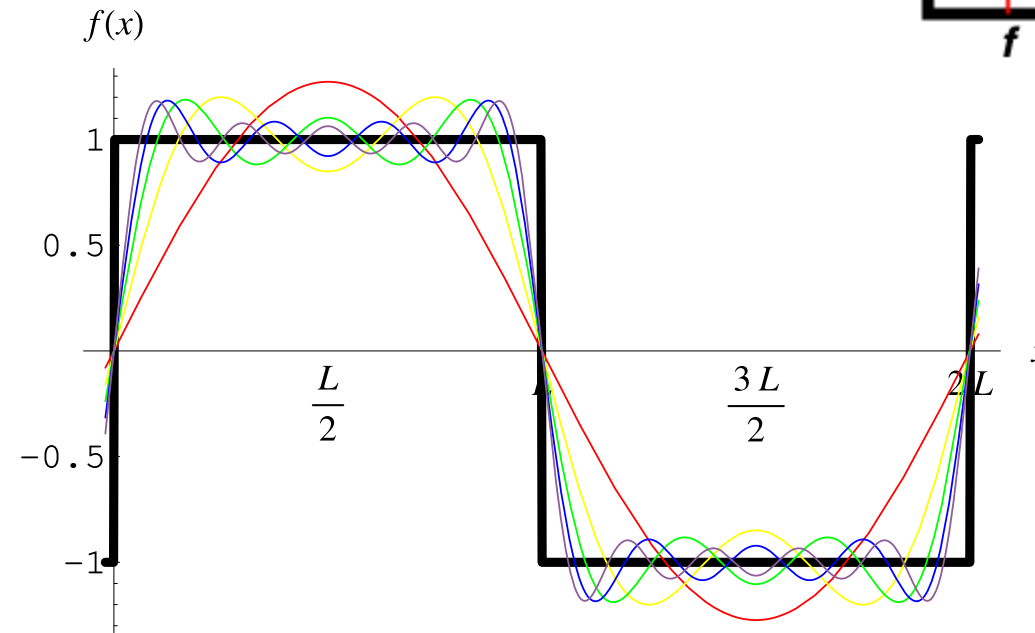
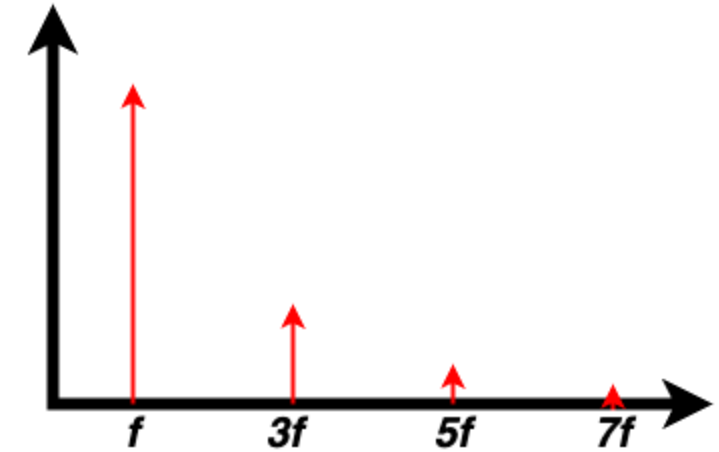
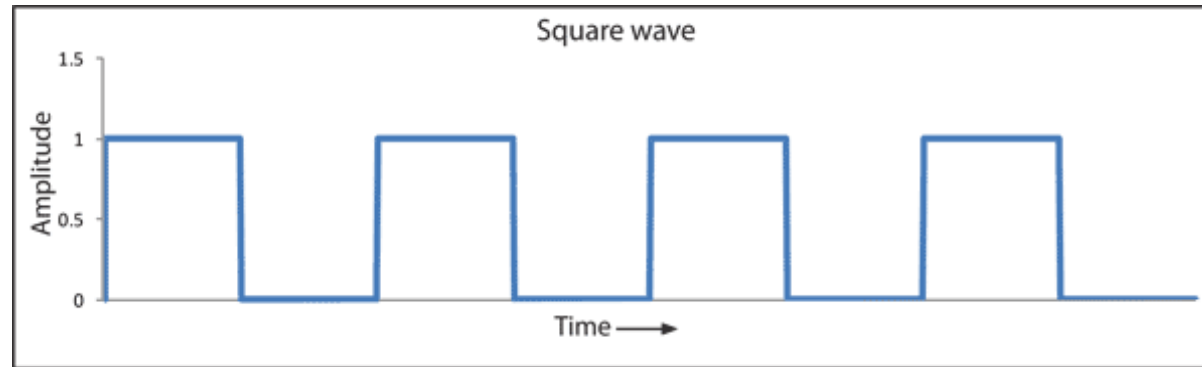
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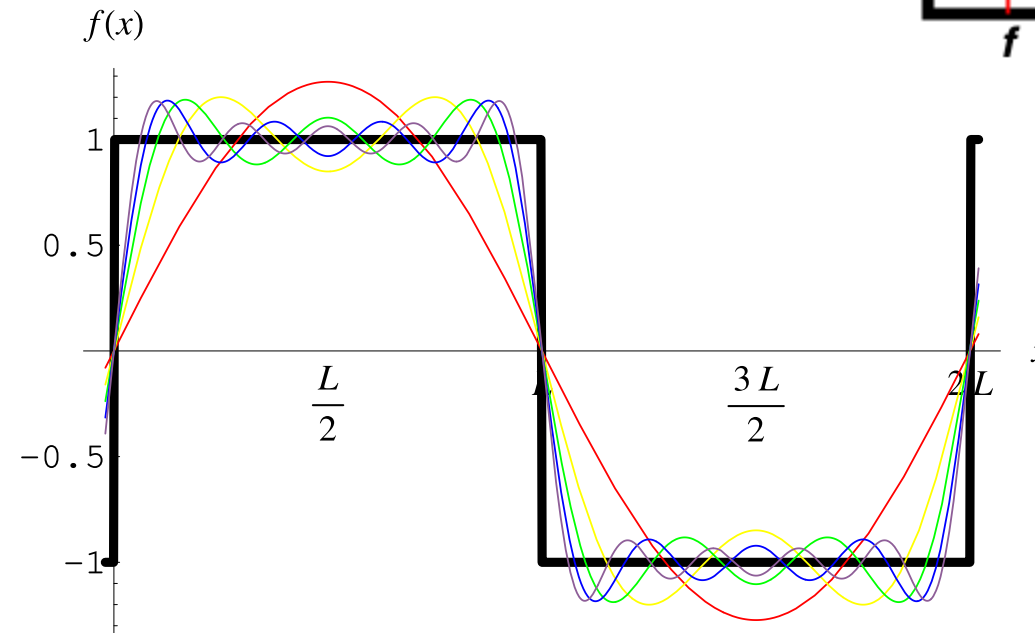
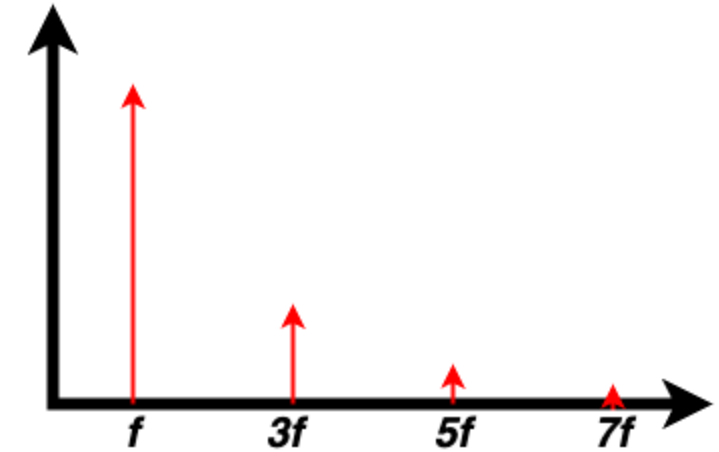
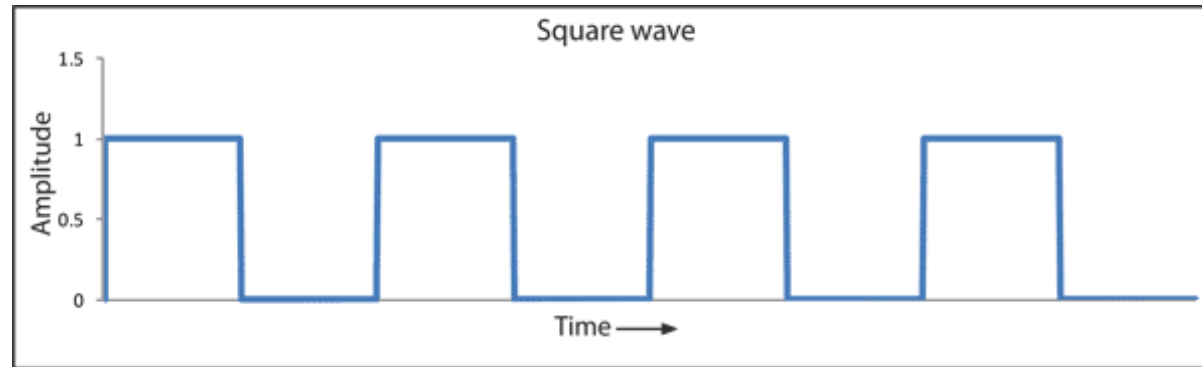
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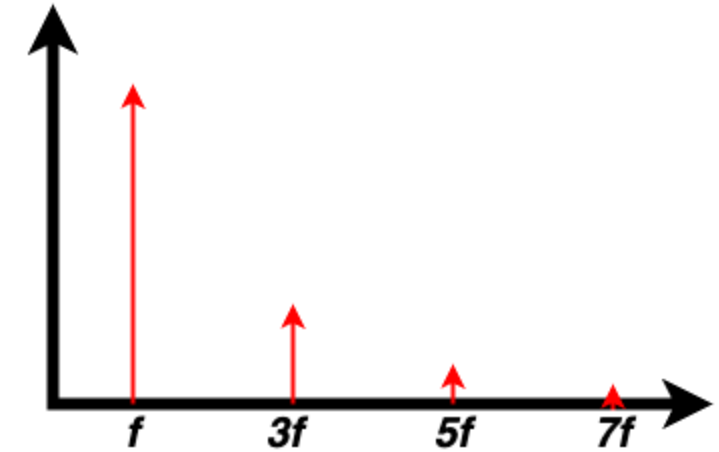
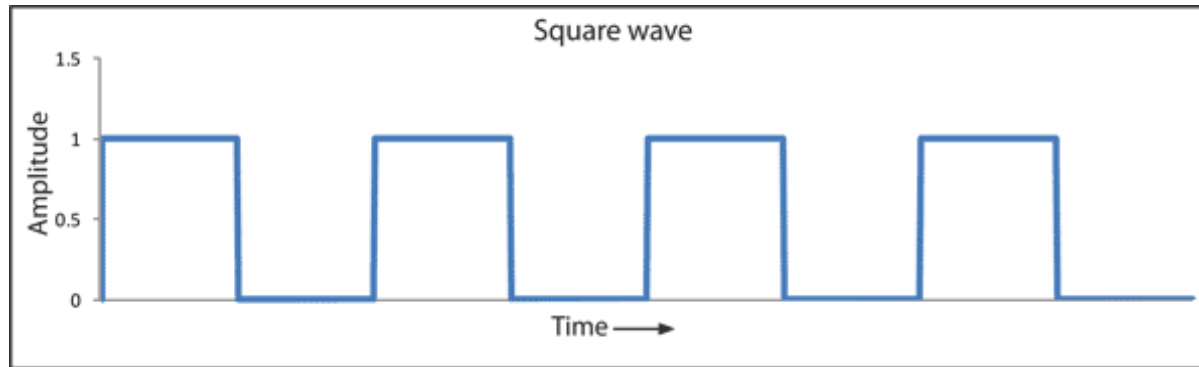
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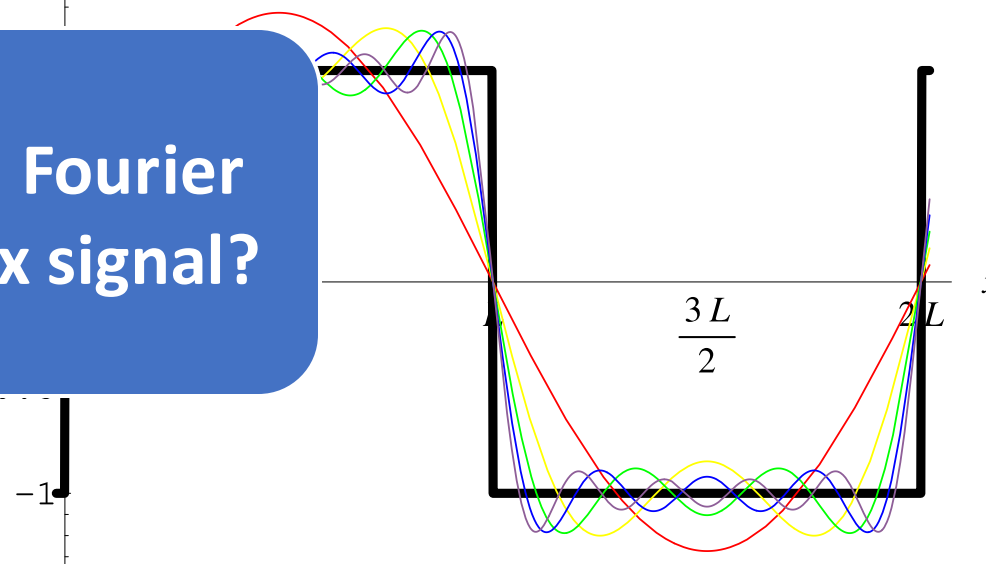


Fourier Analysis

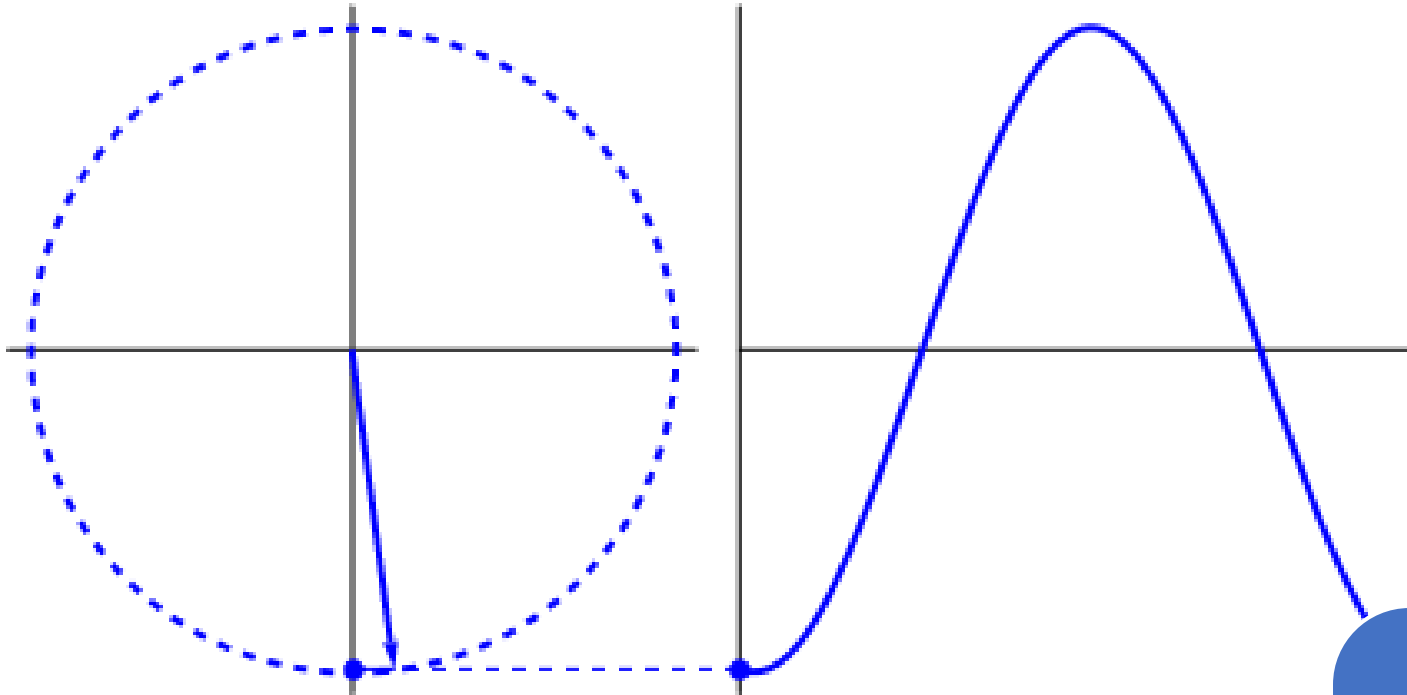
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How do we perform Fourier Analysis of a complex signal?



Representation of a Signal



$$y(t) = A \sin (\omega t + \varphi)$$

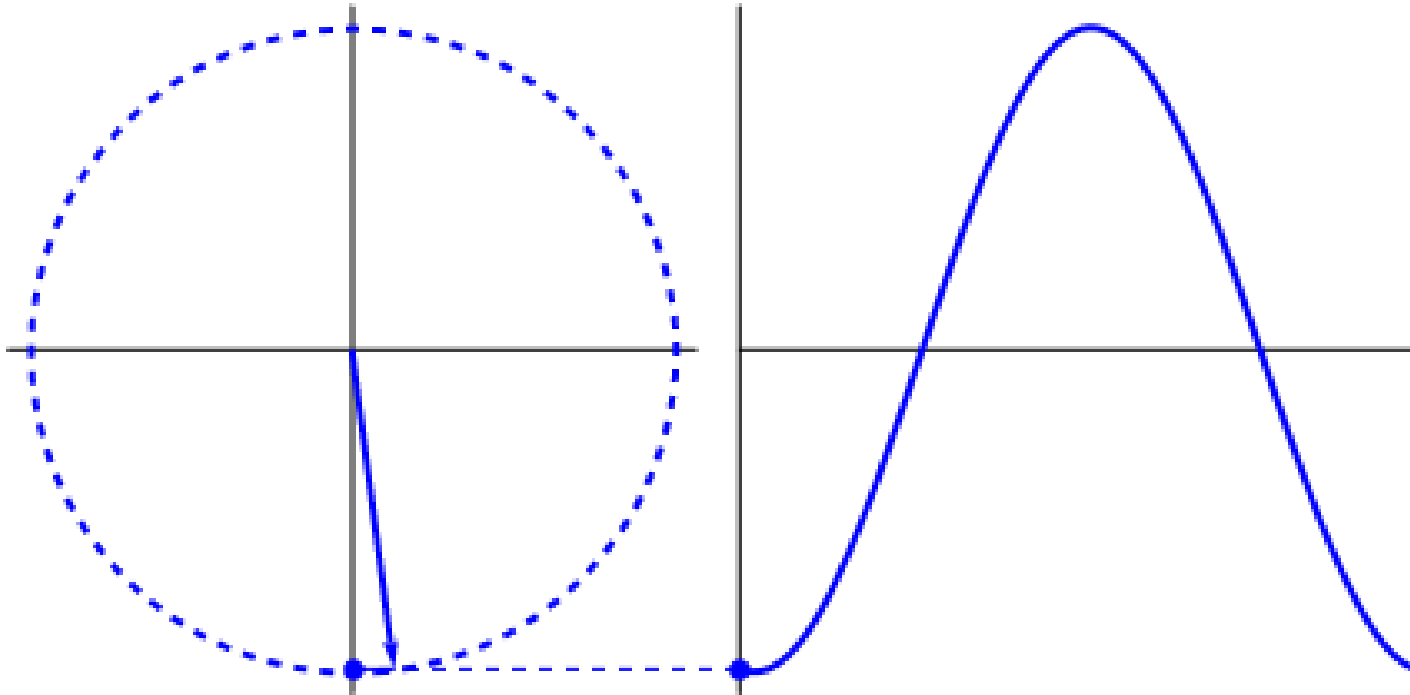
A: Amplitude

ω : Angular frequency

φ : Phase

What is the significance of frequency of this signal?

Fourier Representation in Complex Domain



$$y(t) = A \sin (2\pi f t + \varphi)$$

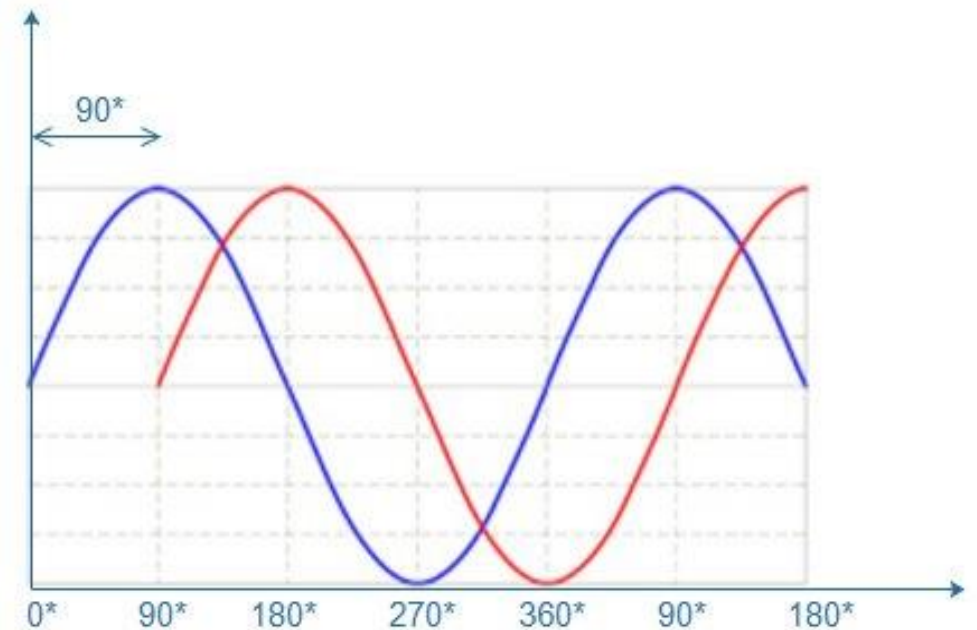
A: Amplitude

f: Ordinary frequency

φ : Phase

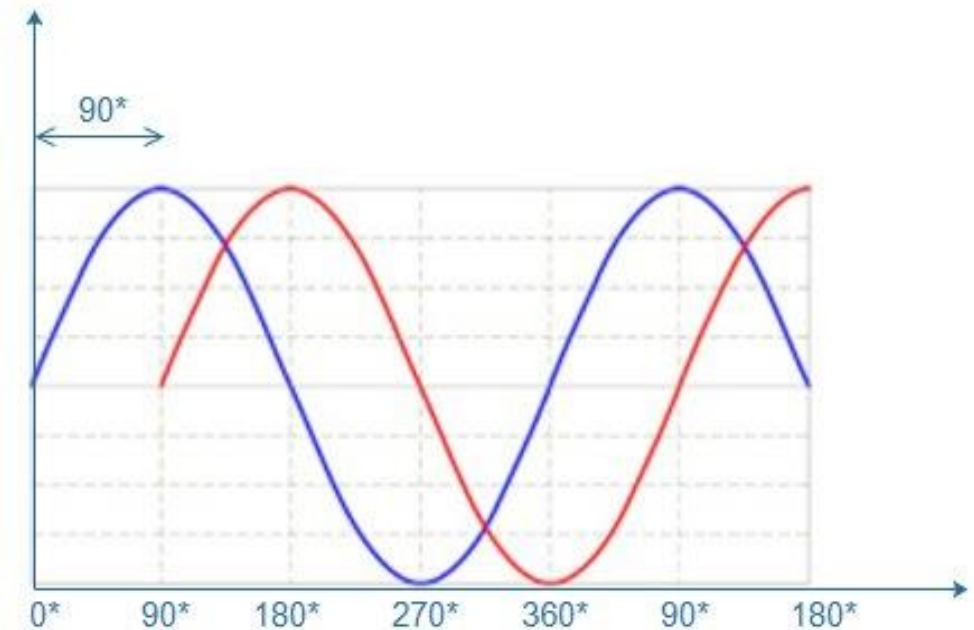
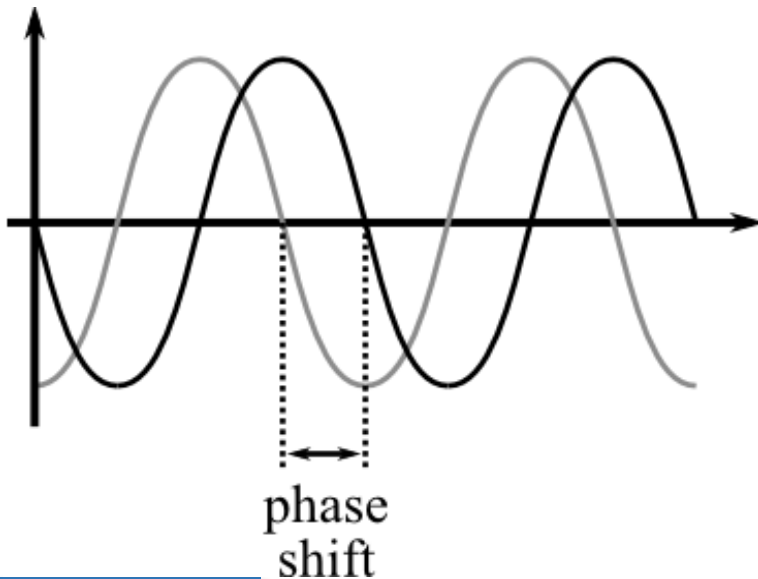
The Notion of Phase

- The fraction of the cycle covered upto some time instance t
- An important metric for sensing
 - The phase of the reflected/scattered signal depends on the type of the material where the transmitted signal hitted
 - Introduces a shift in the reflected/scattered signal

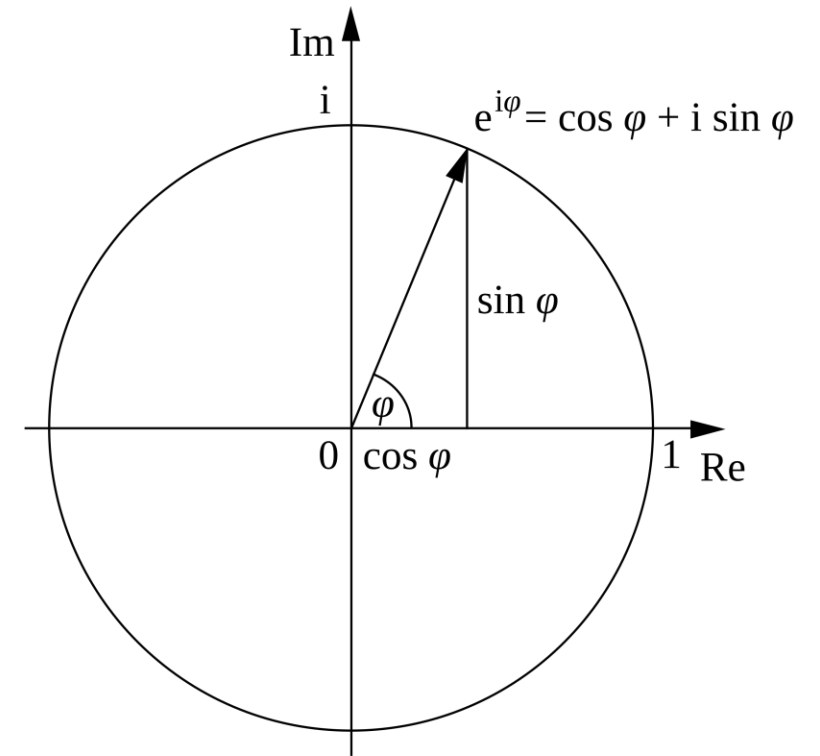
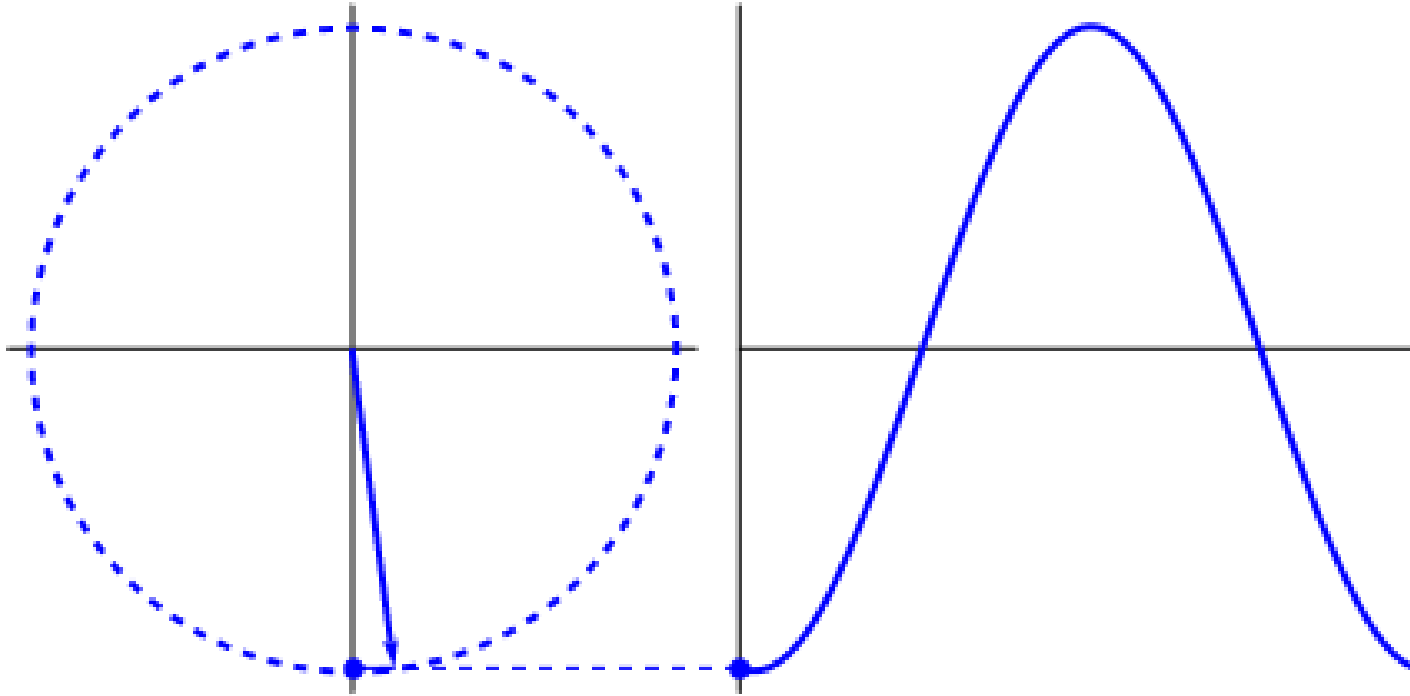


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- **Phase difference/phase shift:**

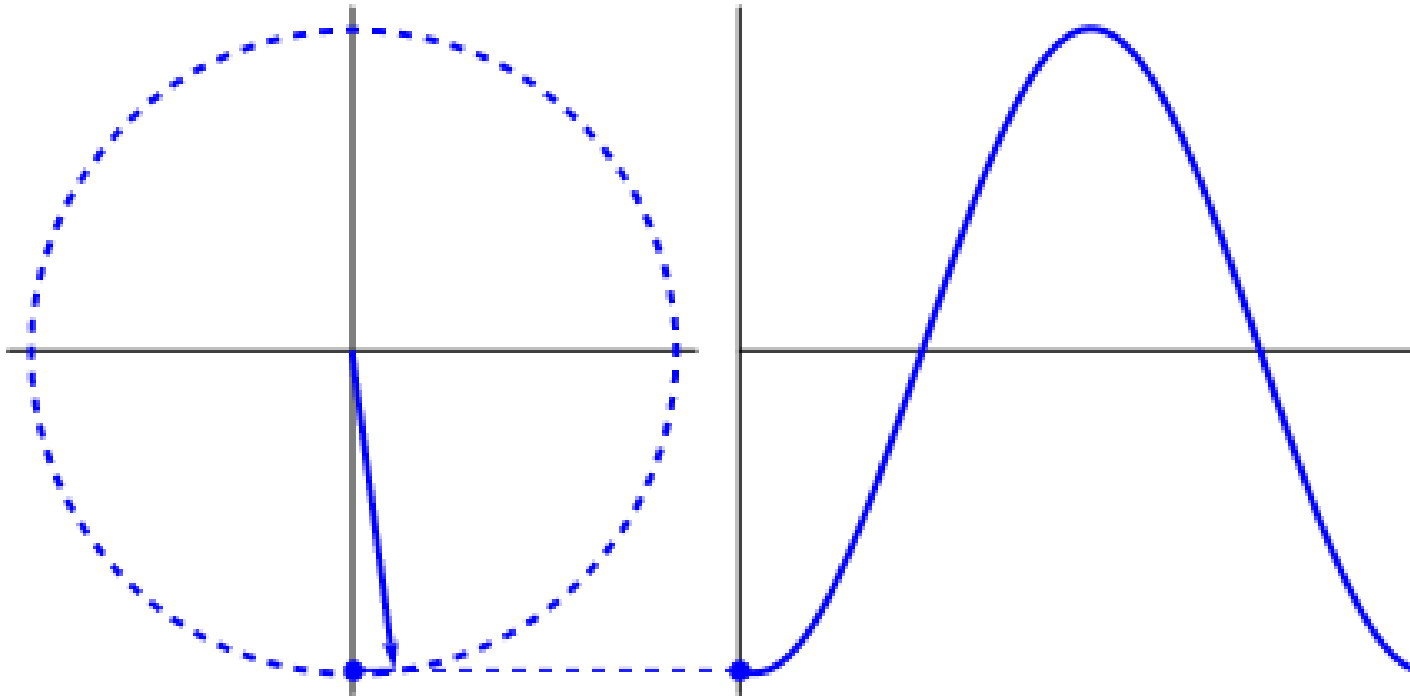


Fourier Representation in Complex Domain



$$S(f) = \int_{-\infty}^{\infty} s(t) \cdot e^{-i2\pi ft} dt.$$

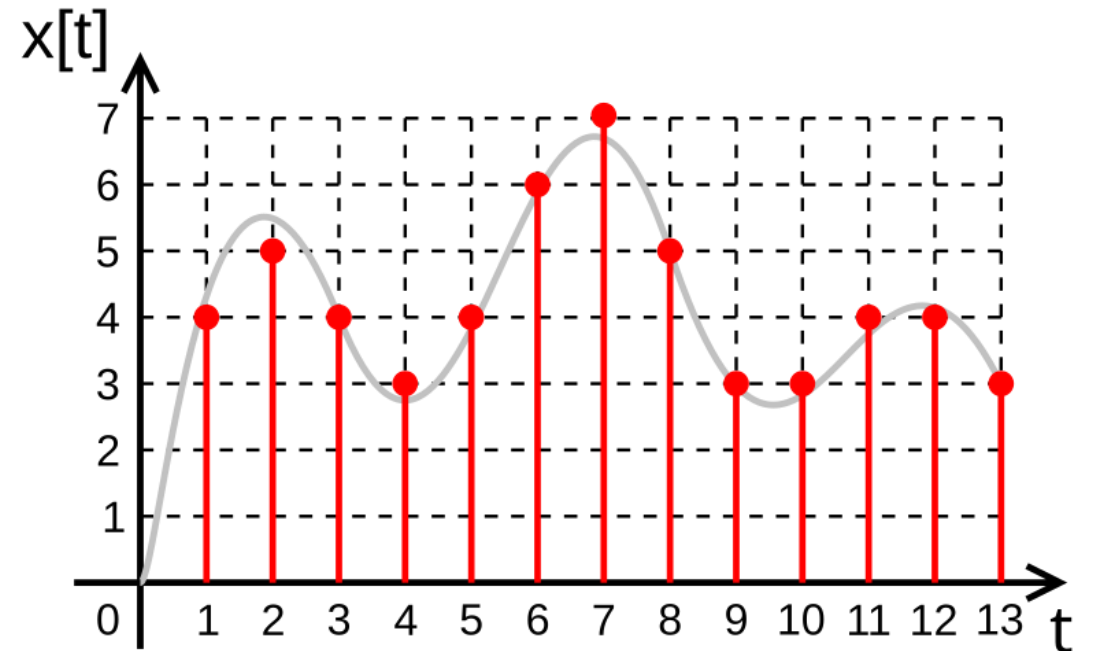
Fourier Representation in Complex Domain



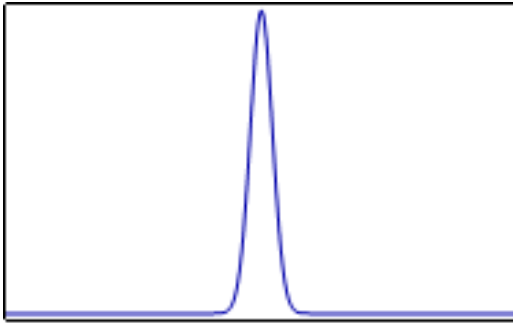
**Check this video for a
nice explanation of
Fourier transformation**

Discrete Fourier Transformation

- In practice, we record the samples of a continuous signal
 - Discrete representation of the signal
- So, we need to perform the Fourier transformation on the discrete samples of the signals
 - Discrete Time Fourier Transform (DTFT)
- However, the input and the output of the samples are also finite
 - Discrete Fourier Transform (DFT): Works on the finite time series data



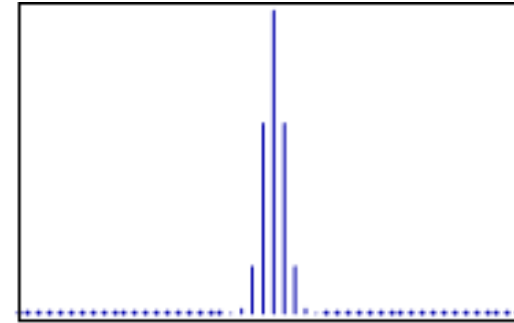
DTFT and DFT



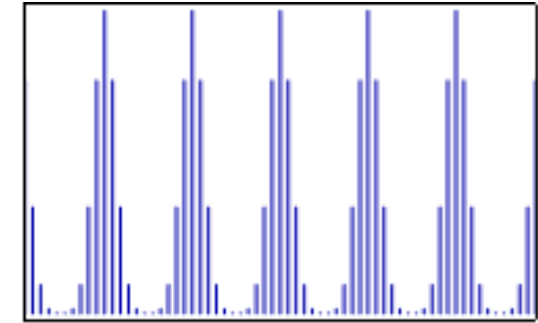
**A Continuous signal
and its Fourier
transform**



**Periodic summation of
the original signal and
its Fourier transform**



**Original signal
discretized and its
Fourier transform
(DTFT)**



**Periodic summation of
the discrete signal, DFT
computes discrete
samples of the
continuous DTFT**

Discrete Fourier Transformation (DFT)

- Let x_0, \dots, x_{n-1} be complex numbers. The DFT is defined by the formula,

$$X_k = \sum_{m=0}^{n-1} x_m e^{-i2\pi km/n} \quad k = 0, \dots, n-1,$$

o Where $e^{i2\pi/n}$ is the primitive n^{th} root of 1

- Evaluating the above equation needs $O(n^2)$ operations

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DFT Matrix

- An N-point DFT is represented as $X = Wx$

- x is the original input signal
- W is $N \times N$ DFT matrix

$$W = \left(\frac{\omega^{jk}}{\sqrt{N}} \right)_{j,k=0,\dots,N-1}$$

- ω is a primitive N^{th} root of unity

$$\omega = e^{-2\pi i/N}$$

- For different values of j , W can be represented as a matrix

$$W = \frac{1}{\sqrt{N}} \begin{bmatrix} 1 & 1 & 1 & 1 & \dots & 1 \\ 1 & \omega & \omega^2 & \omega^3 & \dots & \omega^{N-1} \\ 1 & \omega^2 & \omega^4 & \omega^6 & \dots & \omega^{2(N-1)} \\ 1 & \omega^3 & \omega^6 & \omega^9 & \dots & \omega^{3(N-1)} \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \omega^{N-1} & \omega^{2(N-1)} & \omega^{3(N-1)} & \dots & \omega^{(N-1)(N-1)} \end{bmatrix}$$

Example: 8-point DFT

$$W = \frac{1}{\sqrt{8}} \begin{bmatrix} \omega^0 & \omega^0 & \omega^0 & \omega^0 & \omega^0 & \omega^0 & \omega^0 & \omega^0 \\ \omega^0 & \omega^1 & \omega^2 & \omega^3 & \omega^4 & \omega^5 & \omega^6 & \omega^7 \\ \omega^0 & \omega^2 & \omega^4 & \omega^6 & \omega^8 & \omega^{10} & \omega^{12} & \omega^{14} \\ \omega^0 & \omega^3 & \omega^6 & \omega^9 & \omega^{12} & \omega^{15} & \omega^{18} & \omega^{21} \\ \omega^0 & \omega^4 & \omega^8 & \omega^{12} & \omega^{16} & \omega^{20} & \omega^{24} & \omega^{28} \\ \omega^0 & \omega^5 & \omega^{10} & \omega^{15} & \omega^{20} & \omega^{25} & \omega^{30} & \omega^{35} \\ \omega^0 & \omega^6 & \omega^{12} & \omega^{18} & \omega^{24} & \omega^{30} & \omega^{36} & \omega^{42} \\ \omega^0 & \omega^7 & \omega^{14} & \omega^{21} & \omega^{28} & \omega^{35} & \omega^{42} & \omega^{49} \end{bmatrix} = \frac{1}{\sqrt{8}} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & \omega & -i & -i\omega & -1 & -\omega & i & i\omega \\ 1 & -i & -1 & i & 1 & -i & -1 & i \\ 1 & -i\omega & i & \omega & -1 & i\omega & -i & -\omega \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\ 1 & -\omega & -i & i\omega & -1 & \omega & i & -i\omega \\ 1 & i & -1 & -i & 1 & i & -1 & -i \\ 1 & i\omega & i & -\omega & -1 & -i\omega & -i & \omega \end{bmatrix}$$

Example: 8-point DFT

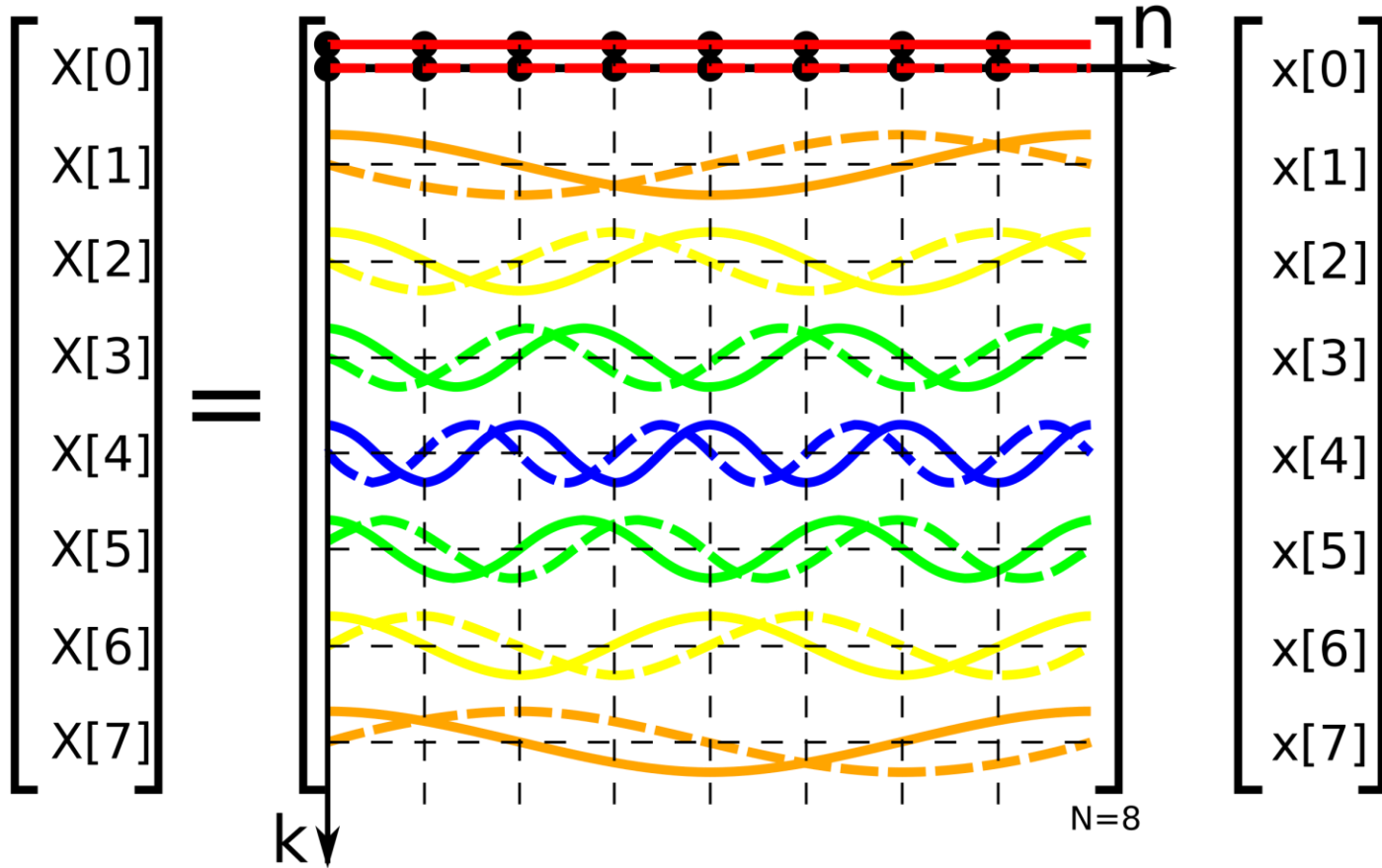
$$W = \frac{1}{\sqrt{8}} \begin{bmatrix} \omega^0 & \omega^0 & \omega^0 & \omega^0 & \omega^0 & \omega^0 & \omega^0 & \omega^0 \\ \omega^0 & \omega^1 & \omega^2 & \omega^3 & \omega^4 & \omega^5 & \omega^6 & \omega^7 \\ \omega^0 & \omega^2 & \omega^4 & \omega^6 & \omega^8 & \omega^{10} & \omega^{12} & \omega^{14} \\ \omega^0 & \omega^3 & \omega^6 & \omega^9 & \omega^{12} & \omega^{15} & \omega^{18} & \omega^{21} \\ \omega^0 & \omega^4 & \omega^8 & \omega^{12} & \omega^{16} & \omega^{20} & \omega^{24} & \omega^{28} \\ \omega^0 & \omega^5 & \omega^{10} & \omega^{15} & \omega^{20} & \omega^{25} & \omega^{30} & \omega^{35} \\ \omega^0 & \omega^6 & \omega^{12} & \omega^{18} & \omega^{24} & \omega^{30} & \omega^{36} & \omega^{42} \\ \omega^0 & \omega^7 & \omega^{14} & \omega^{21} & \omega^{28} & \omega^{35} & \omega^{42} & \omega^{49} \end{bmatrix} = \frac{1}{\sqrt{8}} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & \omega & -i & -i\omega & -1 & -\omega & i & i\omega \\ 1 & -i & -1 & i & 1 & -i & -1 & i \\ 1 & -i\omega & i & \omega & -1 & i\omega & -i & -\omega \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\ 1 & -\omega & -i & i\omega & -1 & \omega & i & -i\omega \\ 1 & i & -1 & -i & 1 & i & -1 & -i \\ 1 & i\omega & i & -\omega & -1 & -i\omega & -i & \omega \end{bmatrix}$$

$$\omega = e^{-\frac{2\pi i}{8}} = \frac{1}{\sqrt{2}} - \frac{i}{\sqrt{2}}$$

Example: 8-point DFT

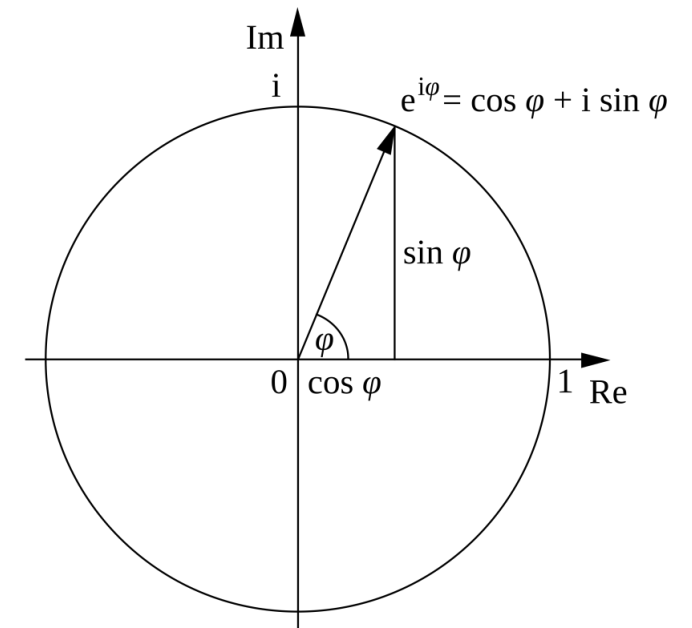
$$W = \frac{1}{\sqrt{8}} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & \frac{1-i}{\sqrt{2}} & -i & \frac{-1-i}{\sqrt{2}} & -1 & \frac{-1+i}{\sqrt{2}} & i & \frac{1+i}{\sqrt{2}} \\ 1 & -i & -1 & i & 1 & -i & -1 & i \\ 1 & \frac{-1-i}{\sqrt{2}} & i & \frac{1-i}{\sqrt{2}} & -1 & \frac{1+i}{\sqrt{2}} & -i & \frac{-1+i}{\sqrt{2}} \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\ 1 & \frac{-1+i}{\sqrt{2}} & -i & \frac{1+i}{\sqrt{2}} & -1 & \frac{1-i}{\sqrt{2}} & i & \frac{-1-i}{\sqrt{2}} \\ 1 & i & -1 & -i & 1 & i & -1 & -i \\ 1 & \frac{1+i}{\sqrt{2}} & i & \frac{-1+i}{\sqrt{2}} & -1 & \frac{-1-i}{\sqrt{2}} & -i & \frac{1-i}{\sqrt{2}} \end{bmatrix}$$

Example: 8-point DFT: Pictorial Representation

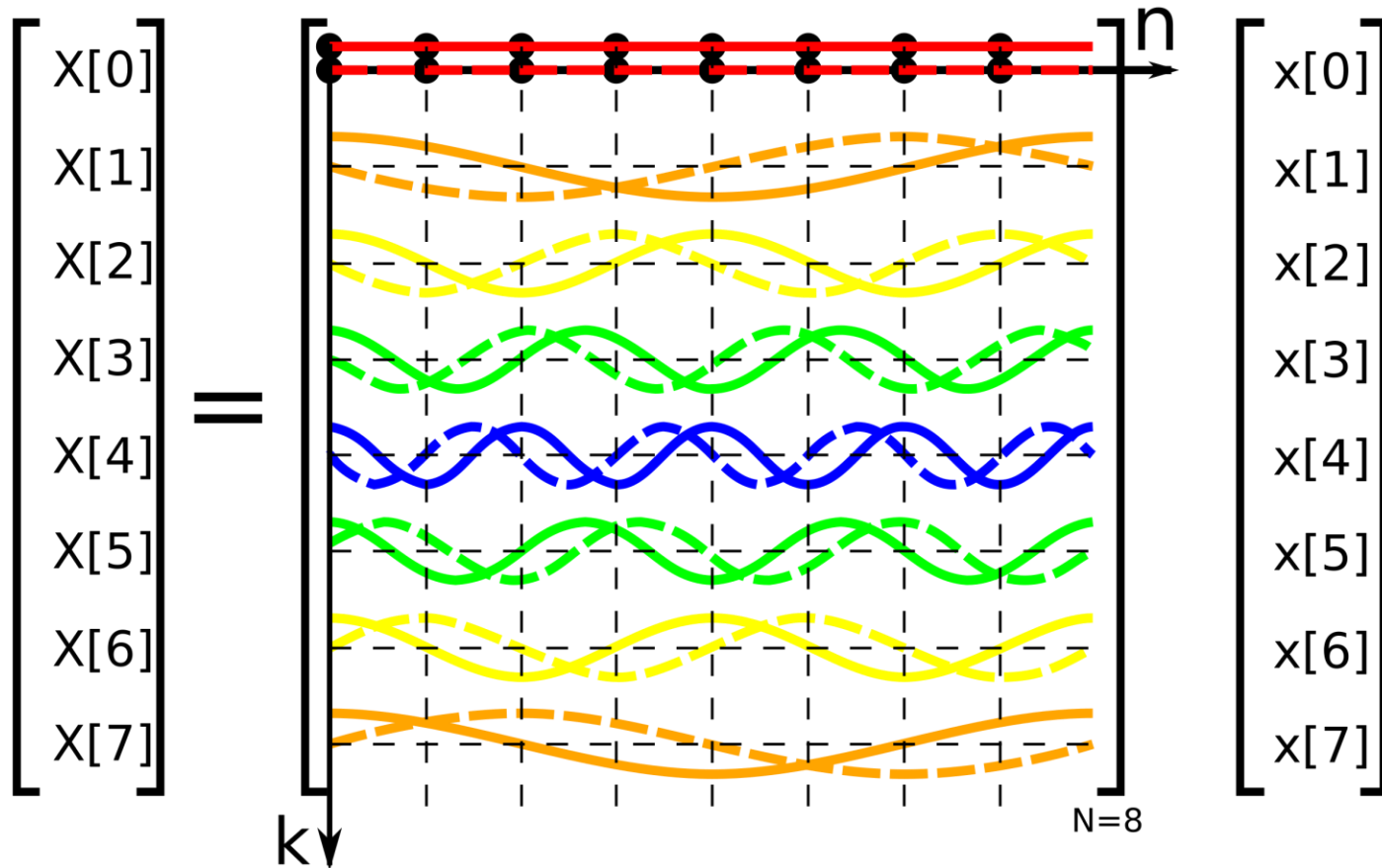


Cosine wave: Solid Line
Sine wave: Dashed line

$$W = \frac{1}{\sqrt{8}} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & \frac{1-i}{\sqrt{2}} & -i & \frac{-1-i}{\sqrt{2}} & -1 & \frac{-1+i}{\sqrt{2}} & i & \frac{1+i}{\sqrt{2}} \\ 1 & -i & -1 & i & 1 & -i & -1 & i \\ 1 & \frac{-1-i}{\sqrt{2}} & i & \frac{1-i}{\sqrt{2}} & -1 & \frac{1+i}{\sqrt{2}} & -i & \frac{-1+i}{\sqrt{2}} \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\ 1 & \frac{-1+i}{\sqrt{2}} & -i & \frac{1+i}{\sqrt{2}} & -1 & \frac{1-i}{\sqrt{2}} & i & \frac{-1-i}{\sqrt{2}} \\ 1 & i & -1 & -i & 1 & i & -1 & -i \\ 1 & \frac{1+i}{\sqrt{2}} & i & \frac{-1+i}{\sqrt{2}} & -1 & \frac{-1-i}{\sqrt{2}} & -i & \frac{1-i}{\sqrt{2}} \end{bmatrix}$$



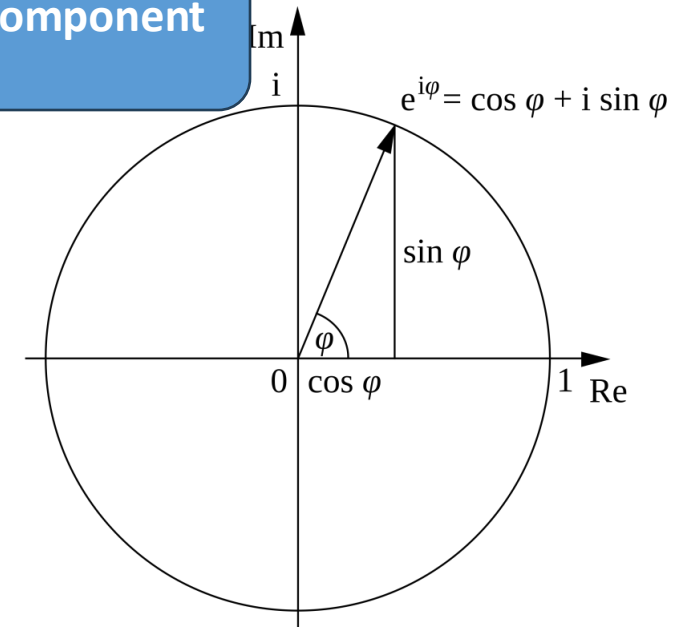
Example: 8-point DFT: Pictorial Representation



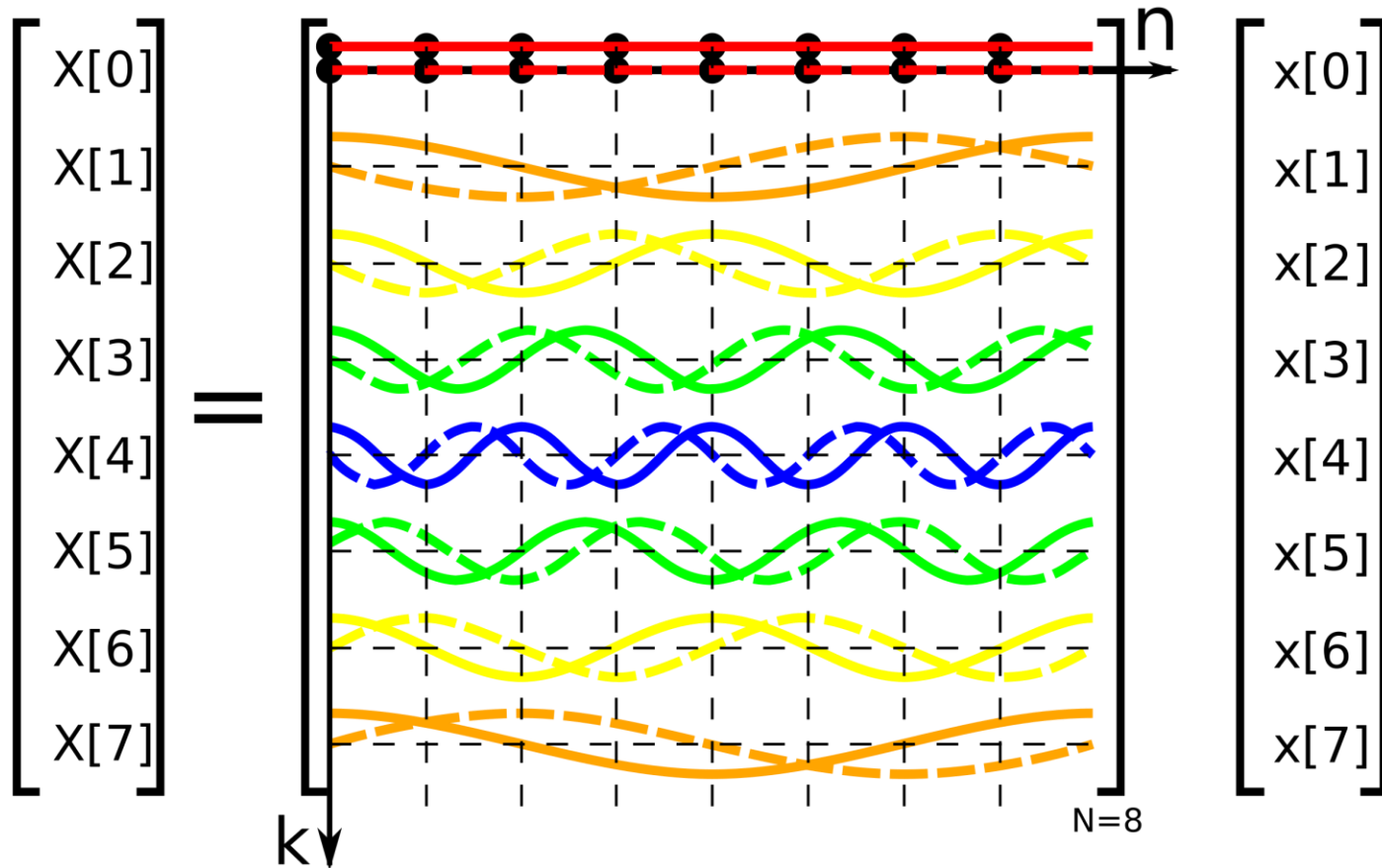
Cosine wave: Solid Line
Sine wave: Dashed line

$$W = \frac{1}{\sqrt{8}} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & \frac{1-i}{\sqrt{2}} & -i & \frac{-1-i}{\sqrt{2}} & -1 & \frac{-1+i}{\sqrt{2}} & i & \frac{1+i}{\sqrt{2}} \\ 1 & -i & -1 & i & 1 & -i & -1 & i \\ 1 & \frac{-1-i}{\sqrt{2}} & i & \frac{1-i}{\sqrt{2}} & -1 & \frac{1+i}{\sqrt{2}} & -i & \frac{-1+i}{\sqrt{2}} \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\ 1 & \frac{-1+i}{\sqrt{2}} & -i & \frac{1+i}{\sqrt{2}} & -1 & \frac{1-i}{\sqrt{2}} & i & \frac{-1-i}{\sqrt{2}} \\ 1 & i & -1 & -i & 1 & i & -1 & -i \\ 1 & \frac{1+i}{\sqrt{2}} & i & \frac{-1+i}{\sqrt{2}} & -1 & \frac{-1-i}{\sqrt{2}} & -i & \frac{1-i}{\sqrt{2}} \end{bmatrix}$$

DC Component



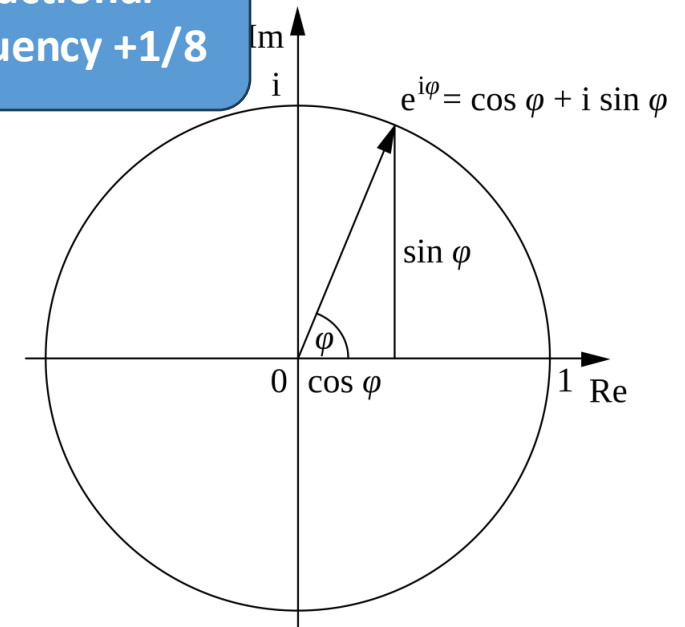
Example: 8-point DFT: Pictorial Representation



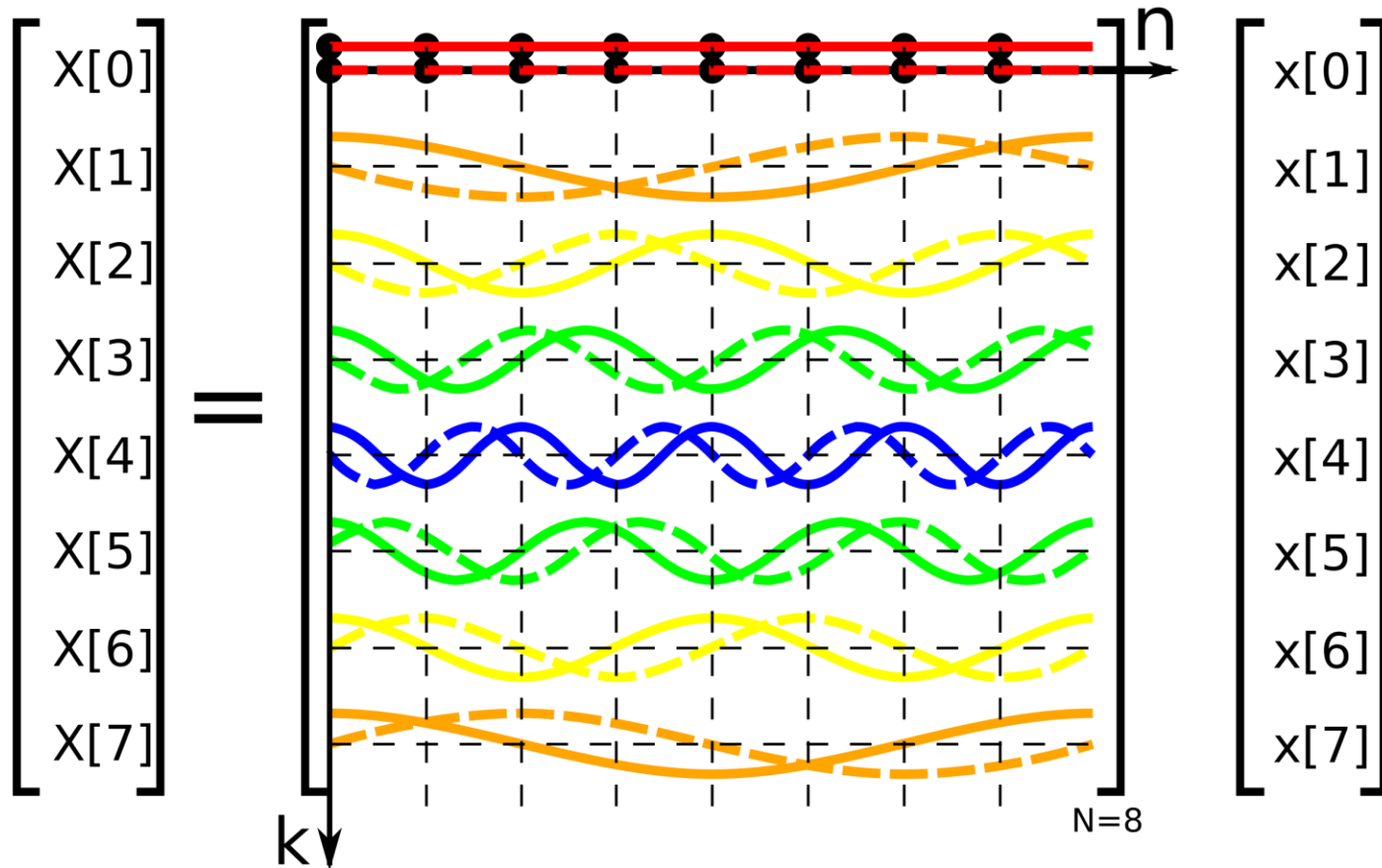
Cosine wave: Solid Line
Sine wave: Dashed line

$$W = \frac{1}{\sqrt{8}} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & \frac{1-i}{\sqrt{2}} & -i & \frac{-1-i}{\sqrt{2}} & -1 & \frac{-1+i}{\sqrt{2}} & i & \frac{1+i}{\sqrt{2}} \\ 1 & -i & -1 & i & 1 & -i & -1 & i \\ 1 & \frac{-1-i}{\sqrt{2}} & i & \frac{1-i}{\sqrt{2}} & -1 & \frac{1+i}{\sqrt{2}} & -i & \frac{-1+i}{\sqrt{2}} \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\ 1 & \frac{-1+i}{\sqrt{2}} & -i & \frac{1+i}{\sqrt{2}} & -1 & \frac{1-i}{\sqrt{2}} & i & \frac{-1-i}{\sqrt{2}} \\ 1 & i & -1 & -i & 1 & i & -1 & -i \\ 1 & \frac{1+i}{\sqrt{2}} & i & \frac{-1+i}{\sqrt{2}} & -1 & \frac{-1-i}{\sqrt{2}} & -i & \frac{1-i}{\sqrt{2}} \end{bmatrix}$$

Fractional frequency $+1/8$



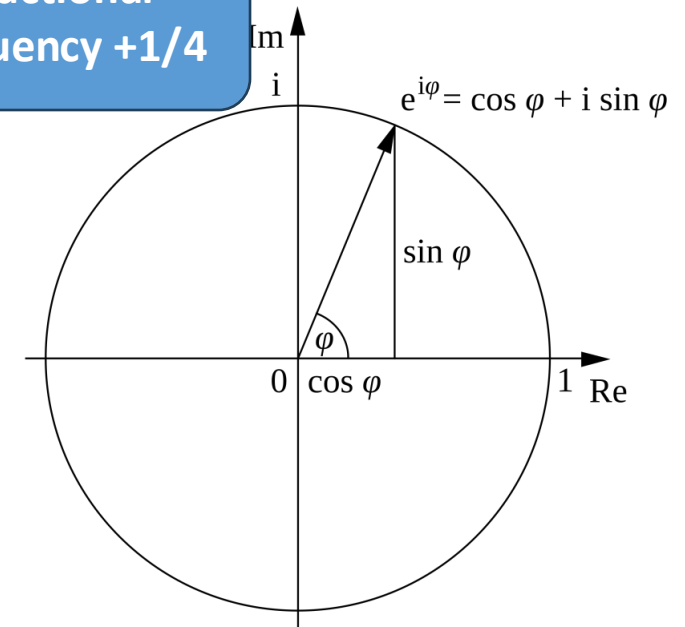
Example: 8-point DFT: Pictorial Representation



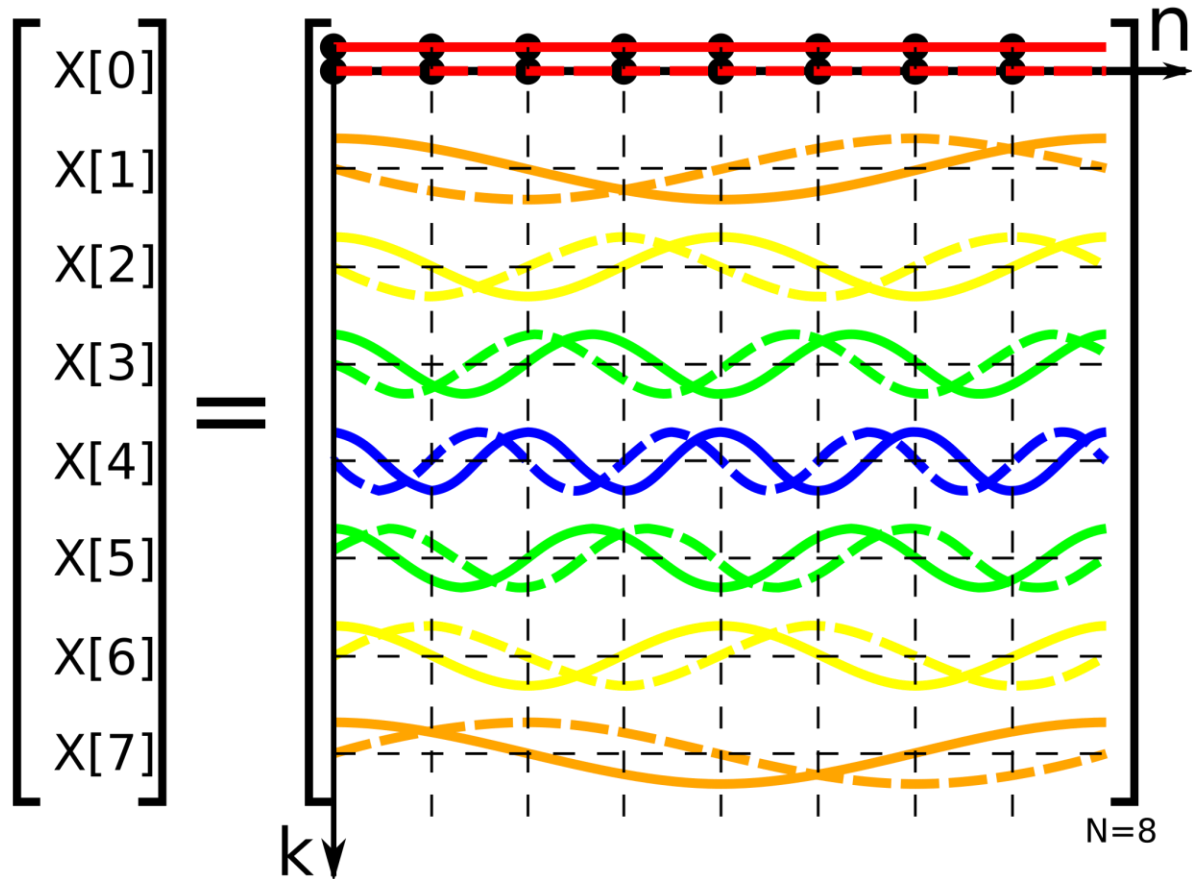
Cosine wave: Solid Line
Sine wave: Dashed line

$$W = \frac{1}{\sqrt{8}} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & \frac{1-i}{\sqrt{2}} & -i & \frac{-1-i}{\sqrt{2}} & -1 & \frac{-1+i}{\sqrt{2}} & i & \frac{1+i}{\sqrt{2}} \\ 1 & -i & -1 & i & 1 & -i & -1 & i \\ 1 & \frac{-1-i}{\sqrt{2}} & i & \frac{1-i}{\sqrt{2}} & -1 & \frac{1+i}{\sqrt{2}} & -i & \frac{-1+i}{\sqrt{2}} \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\ 1 & \frac{-1+i}{\sqrt{2}} & -i & \frac{1+i}{\sqrt{2}} & -1 & \frac{1-i}{\sqrt{2}} & i & \frac{-1-i}{\sqrt{2}} \\ 1 & i & -1 & -i & 1 & i & -1 & -i \\ 1 & \frac{1+i}{\sqrt{2}} & i & \frac{-1+i}{\sqrt{2}} & -1 & \frac{-1-i}{\sqrt{2}} & -i & \frac{1-i}{\sqrt{2}} \end{bmatrix}$$

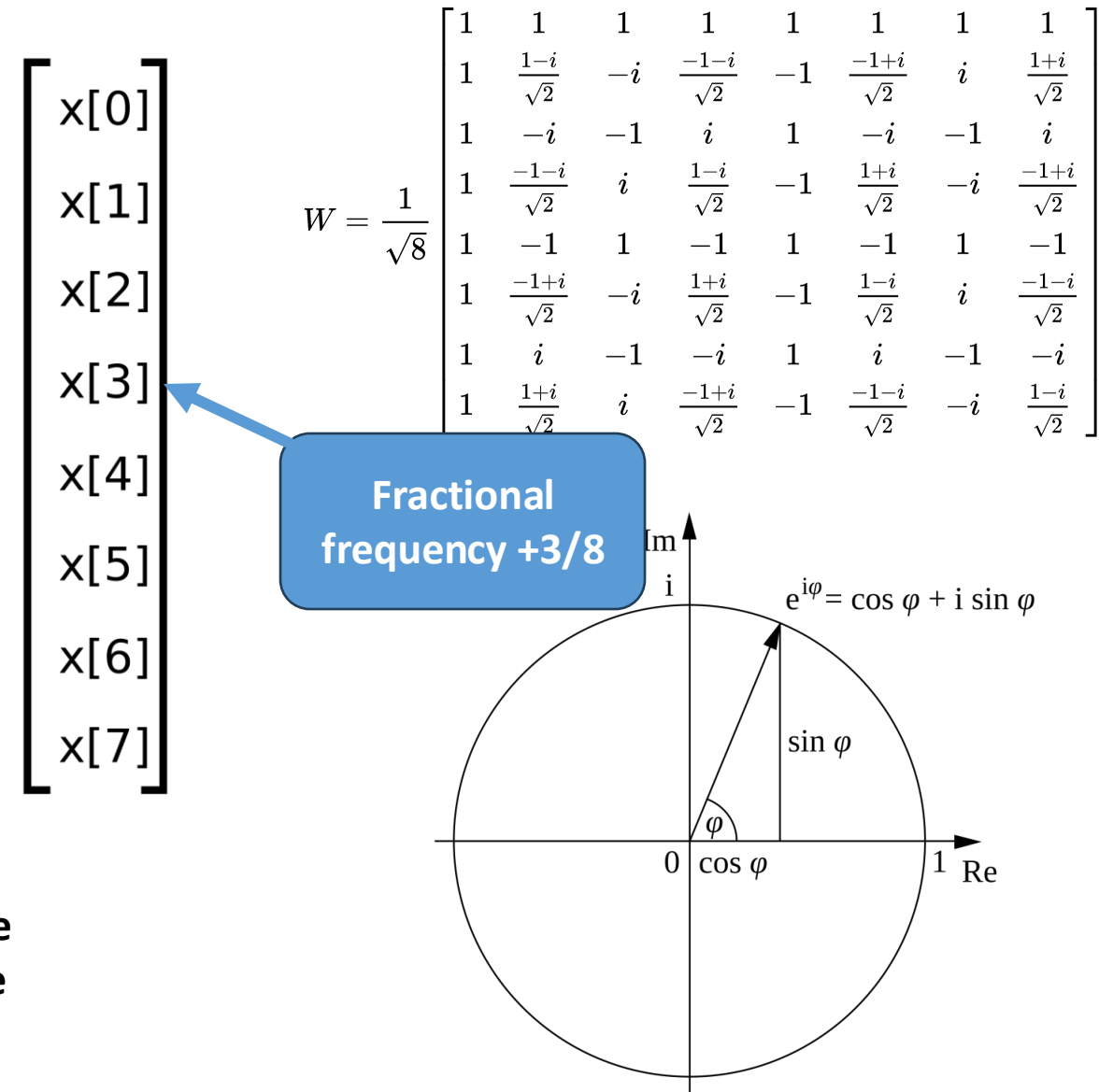
Fractional frequency +1/4



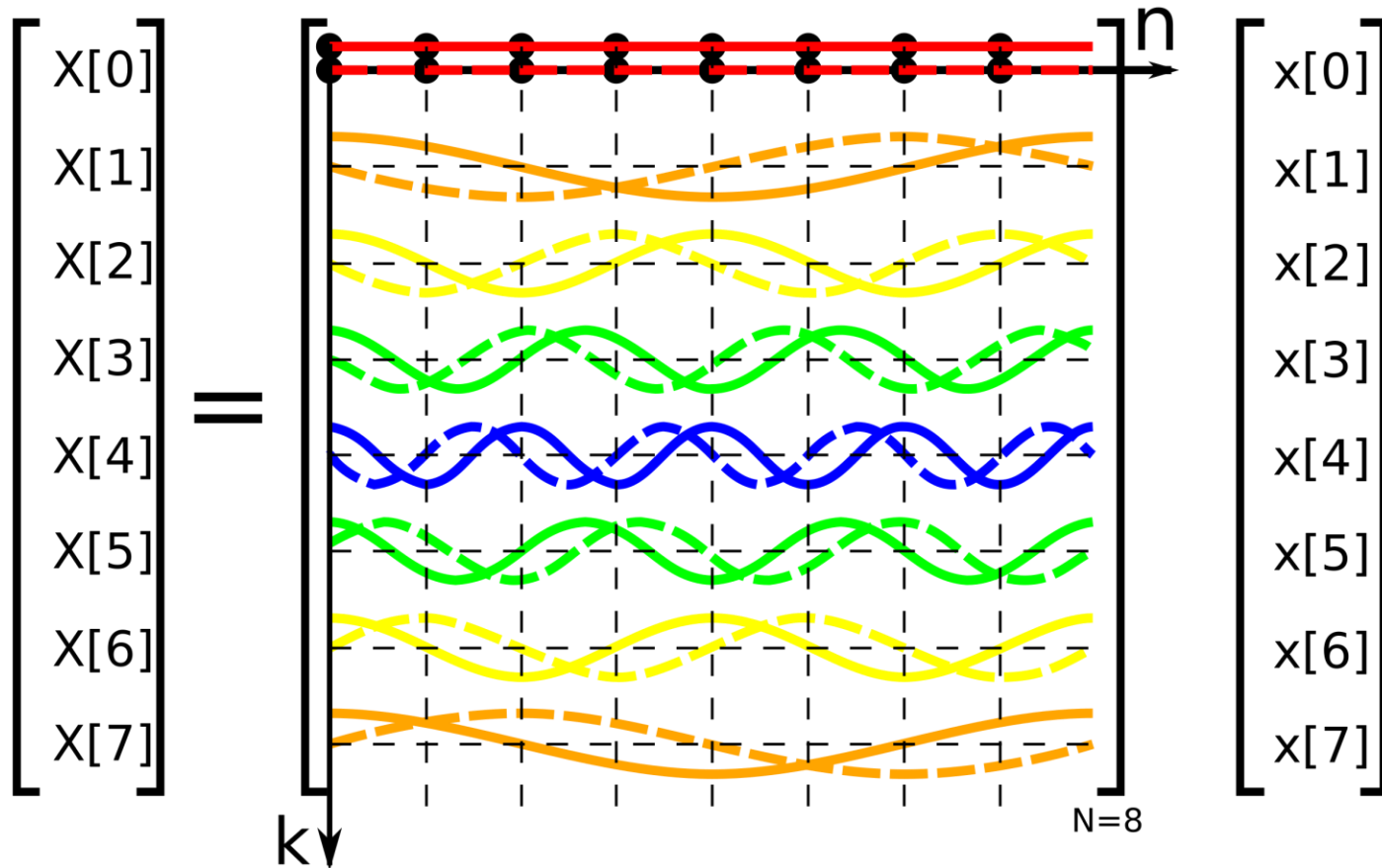
Example: 8-point DFT: Pictorial Representation



Cosine wave: Solid Line
Sine wave: Dashed line



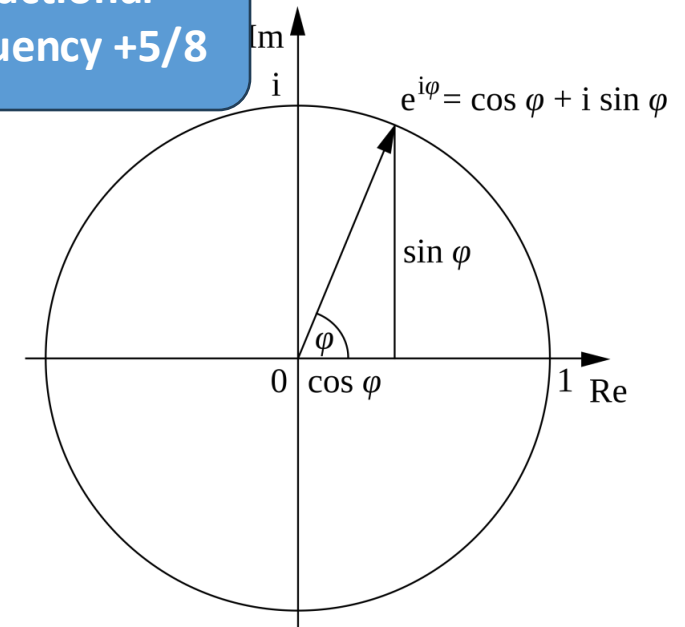
Example: 8-point DFT: Pictorial Representation



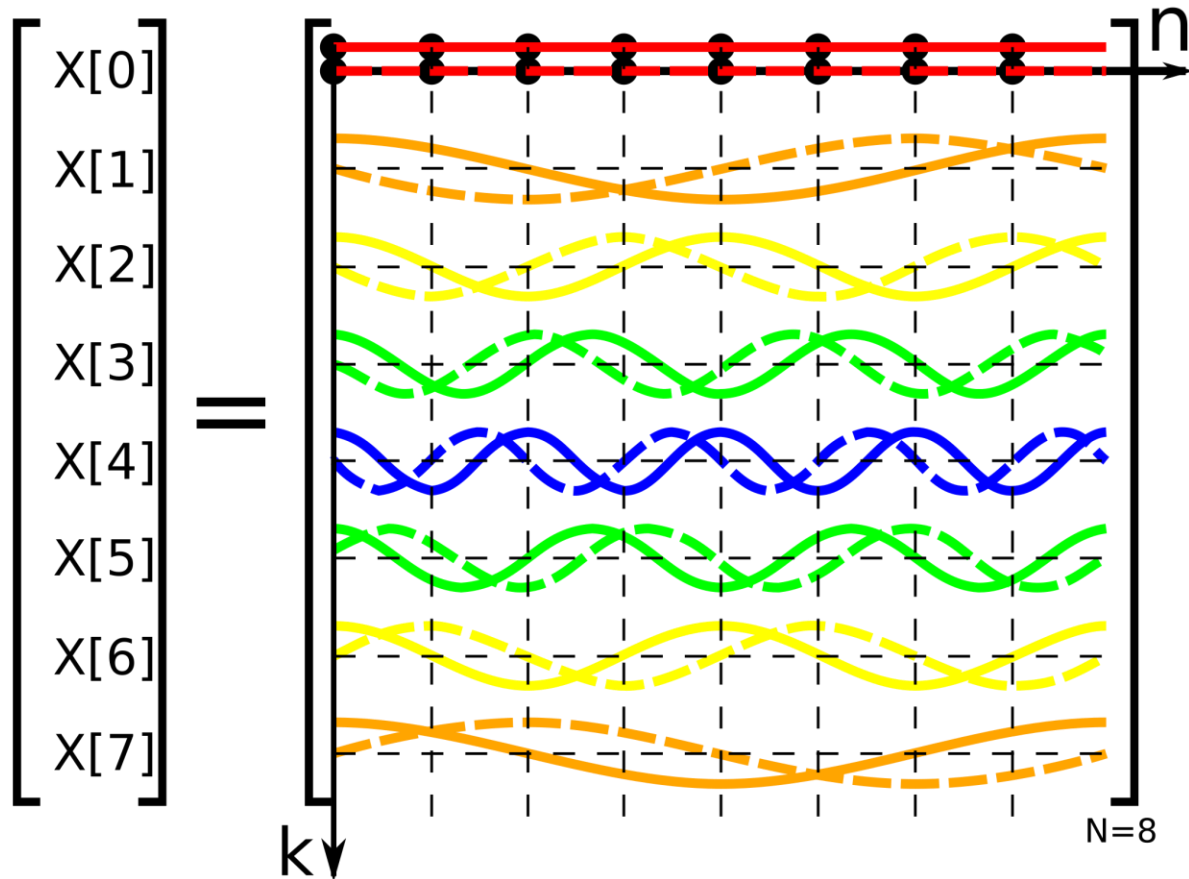
Cosine wave: Solid Line
Sine wave: Dashed line

$$W = \frac{1}{\sqrt{8}} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & \frac{1-i}{\sqrt{2}} & -i & \frac{-1-i}{\sqrt{2}} & -1 & \frac{-1+i}{\sqrt{2}} & i & \frac{1+i}{\sqrt{2}} \\ 1 & -i & -1 & i & 1 & -i & -1 & i \\ 1 & \frac{-1-i}{\sqrt{2}} & i & \frac{1-i}{\sqrt{2}} & -1 & \frac{1+i}{\sqrt{2}} & -i & \frac{-1+i}{\sqrt{2}} \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\ 1 & \frac{-1+i}{\sqrt{2}} & -i & \frac{1+i}{\sqrt{2}} & -1 & \frac{1-i}{\sqrt{2}} & i & \frac{-1-i}{\sqrt{2}} \\ 1 & i & -1 & -i & 1 & i & -1 & -i \\ 1 & \frac{1+i}{\sqrt{2}} & i & \frac{-1+i}{\sqrt{2}} & -1 & \frac{-1-i}{\sqrt{2}} & -i & \frac{1-i}{\sqrt{2}} \end{bmatrix}$$

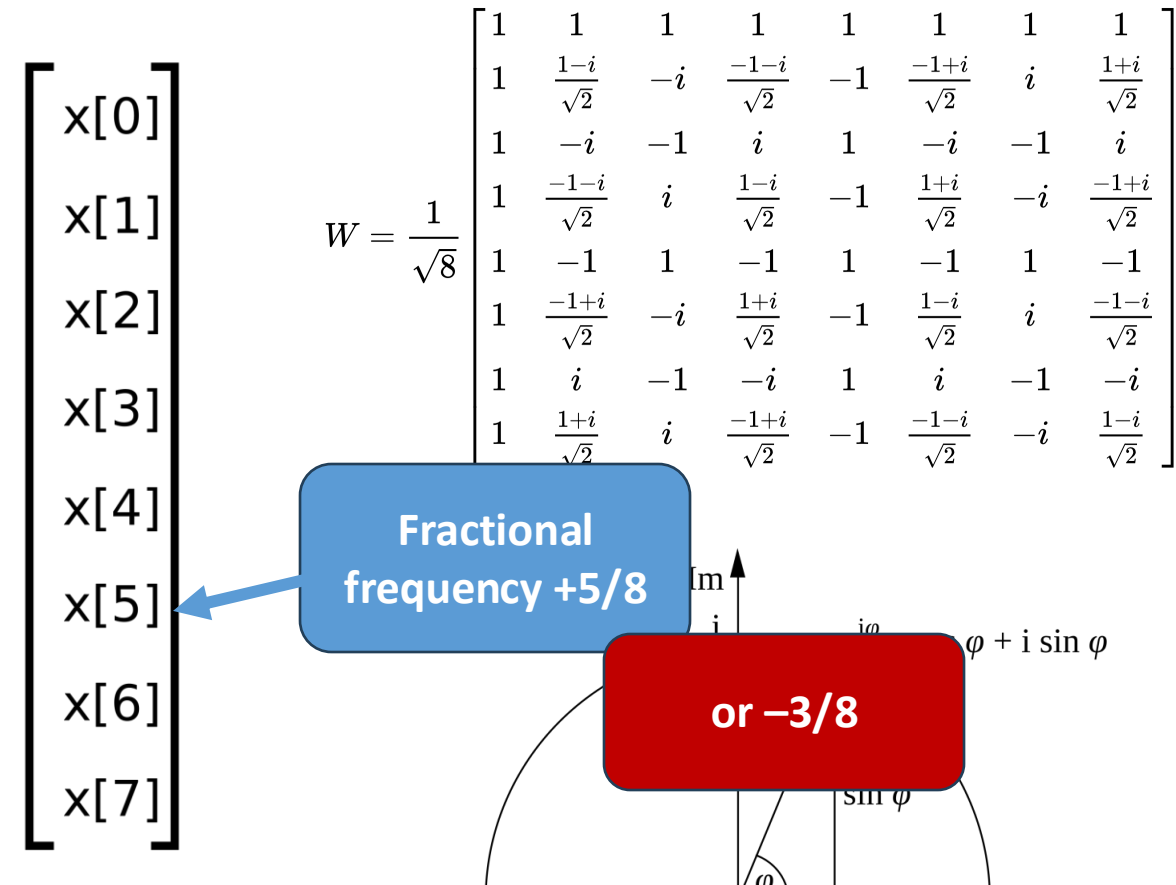
Fractional frequency +5/8



Example: 8-point DFT: Pictorial Representation



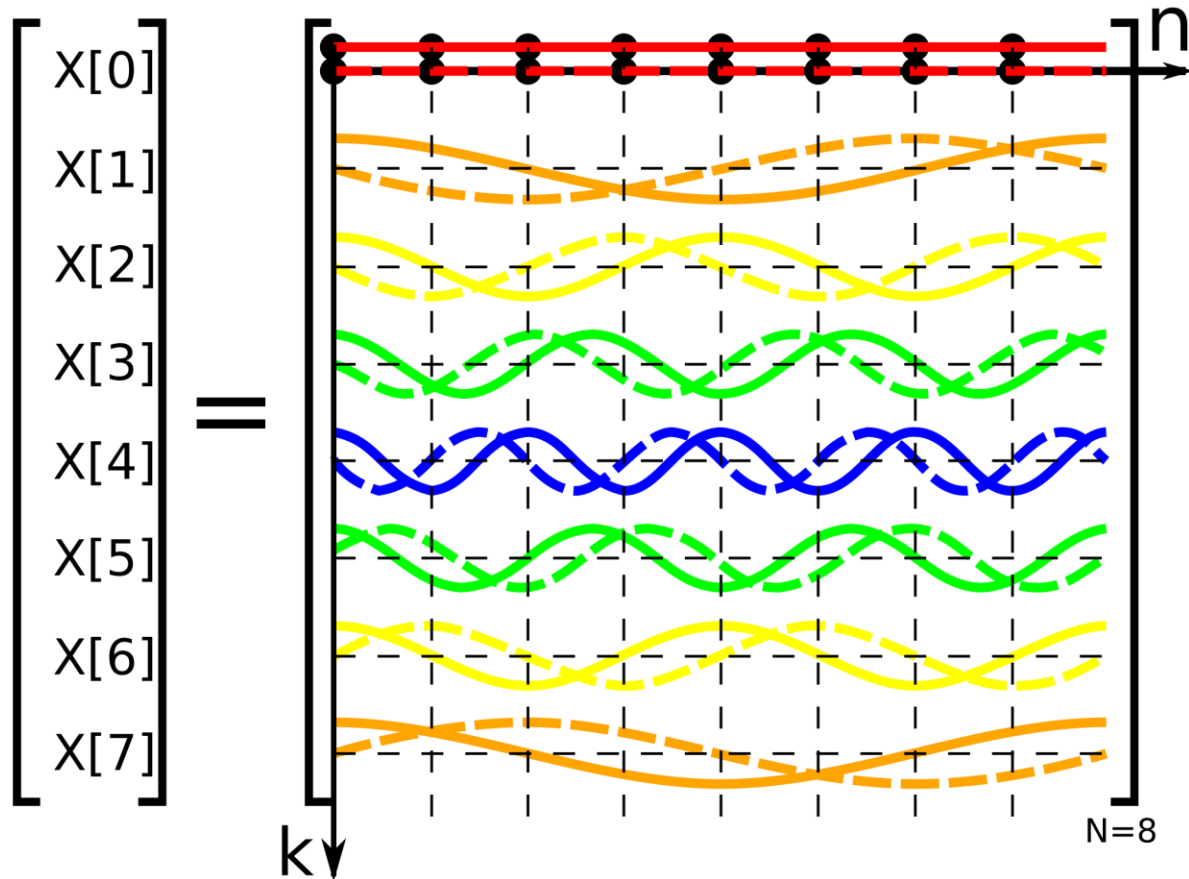
Cosine wave: Solid Line
Sine wave: Dashed line



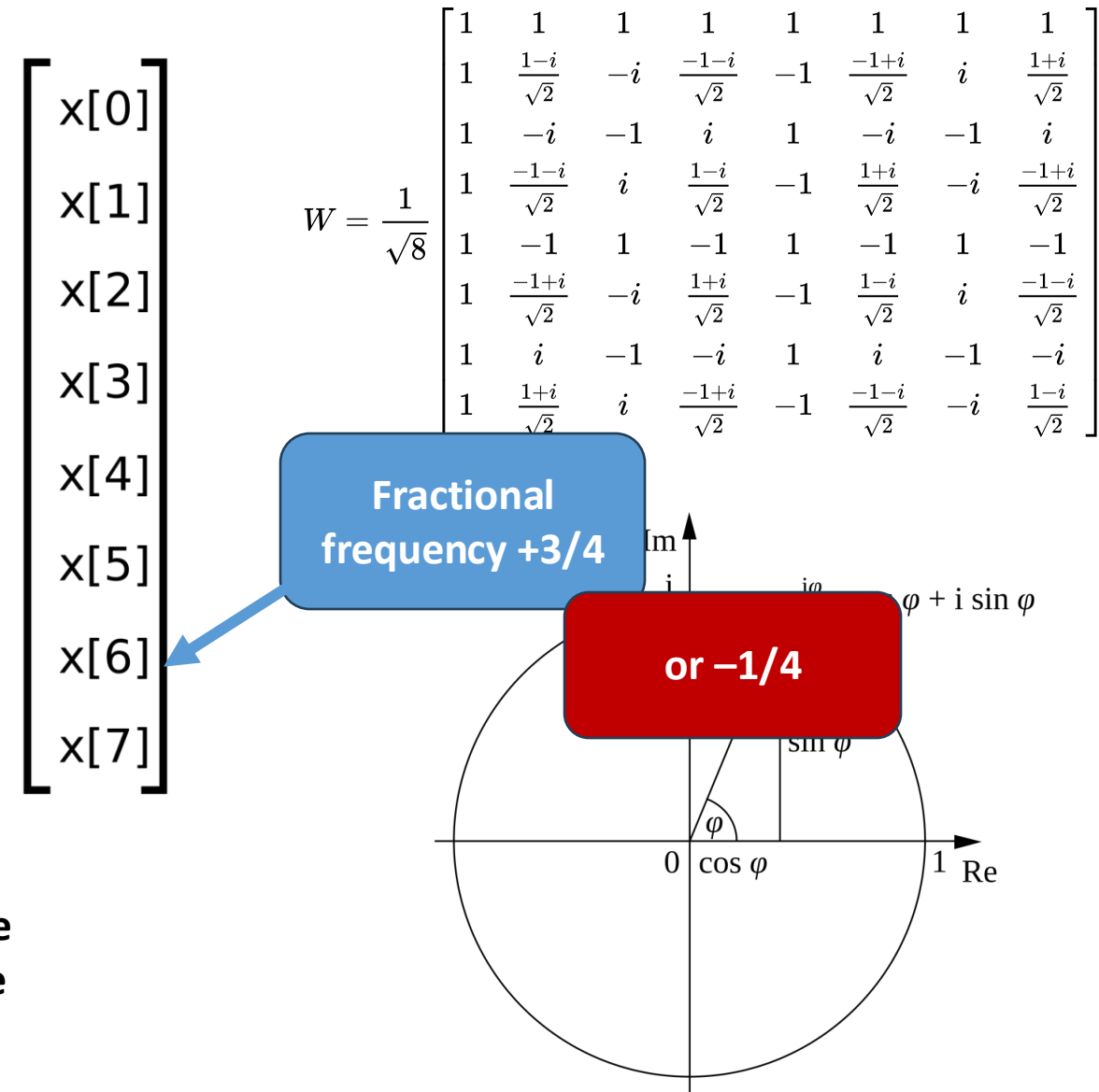
$$\cos(2\pi - \phi) = \cos \phi$$

$$\sin(2\pi - \phi) = -\sin \phi$$

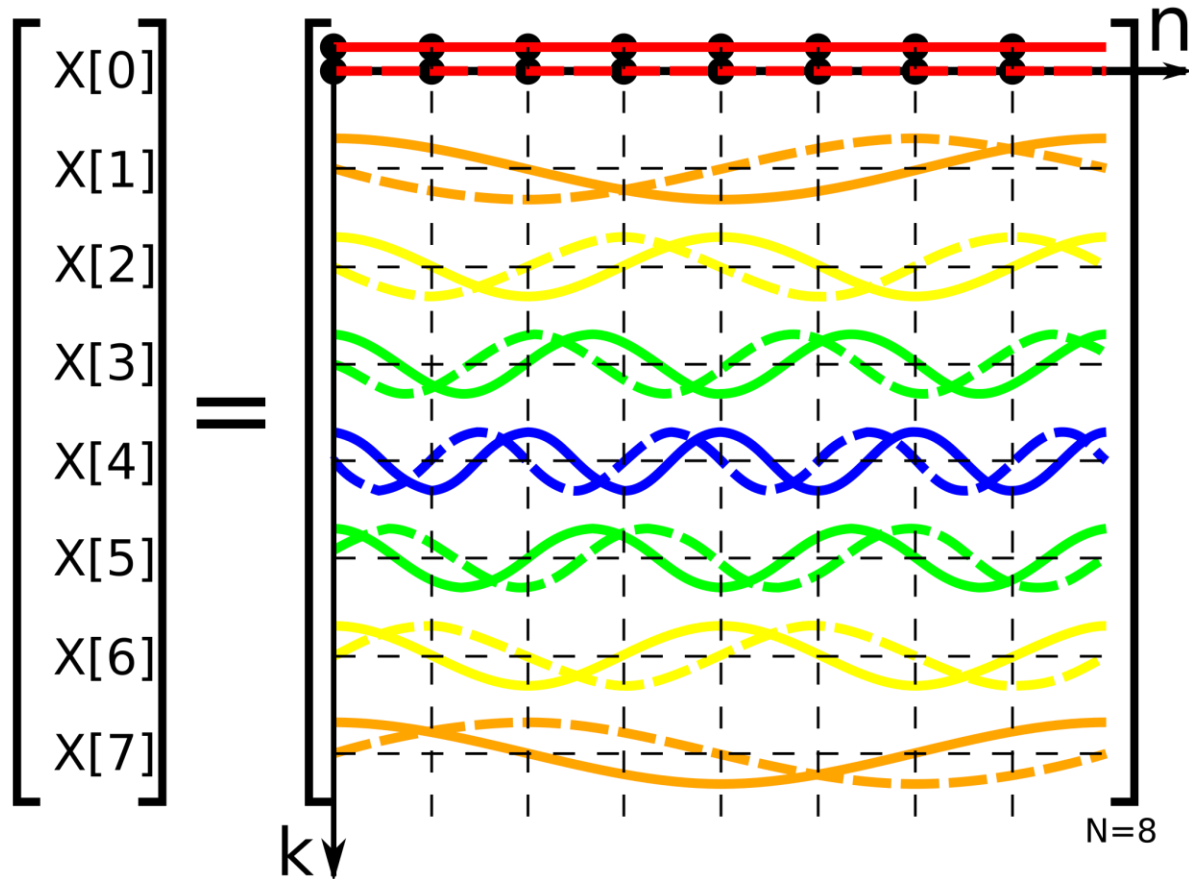
Example: 8-point DFT: Pictorial Representation



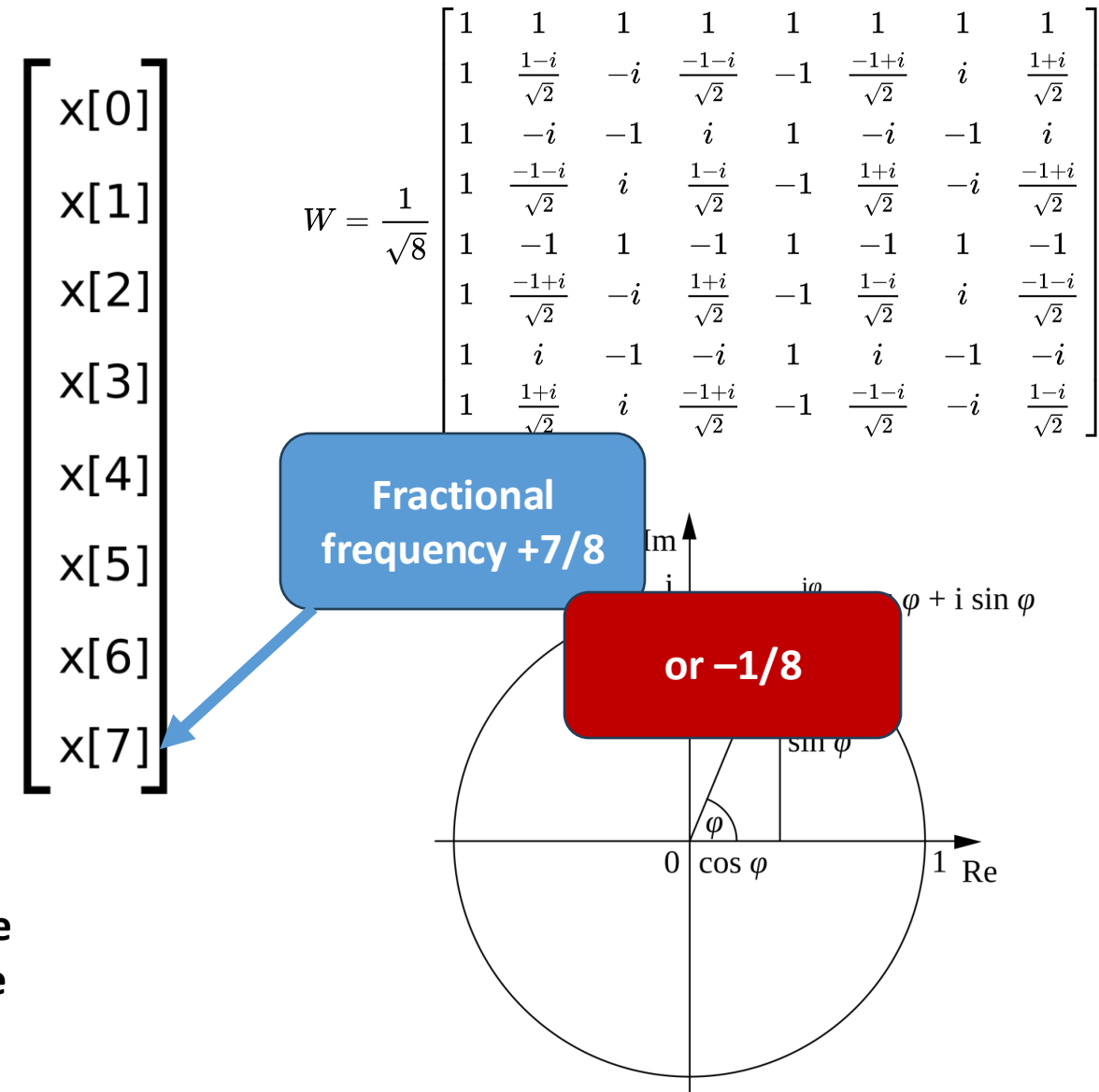
Cosine wave: Solid Line
Sine wave: Dashed line



Example: 8-point DFT: Pictorial Representation



Cosine wave: Solid Line
Sine wave: Dashed line



Discrete Fourier Transformation (DFT)

- Let x_0, \dots, x_{n-1} be complex numbers. The DFT is defined by the formula,

$$X_k = \sum_{m=0}^{n-1} x_m e^{-i2\pi km/n} \quad k = 0, \dots, n-1,$$

o Where $e^{i2\pi/n}$ is the primitive n^{th} root of 1

- Evaluating the above equation needs $O(n^2)$ operations

Can we reduce the complexity?

Fast Fourier Transform (FFT)

- Divide the DFT matrix recursively into smaller DFTs and then combine them
 - Based on the multiplications on complex root of unity
- **Radix-2 decimation-in-time (DIT) FFT**
 - Divide between odd and even inputs

$$X_k = \sum_{m=0}^{N/2-1} x_{2m} e^{-\frac{2\pi i}{N}(2m)k} + \sum_{m=0}^{N/2-1} x_{2m+1} e^{-\frac{2\pi i}{N}(2m+1)k}$$

Fast Fourier Transform (FFT)

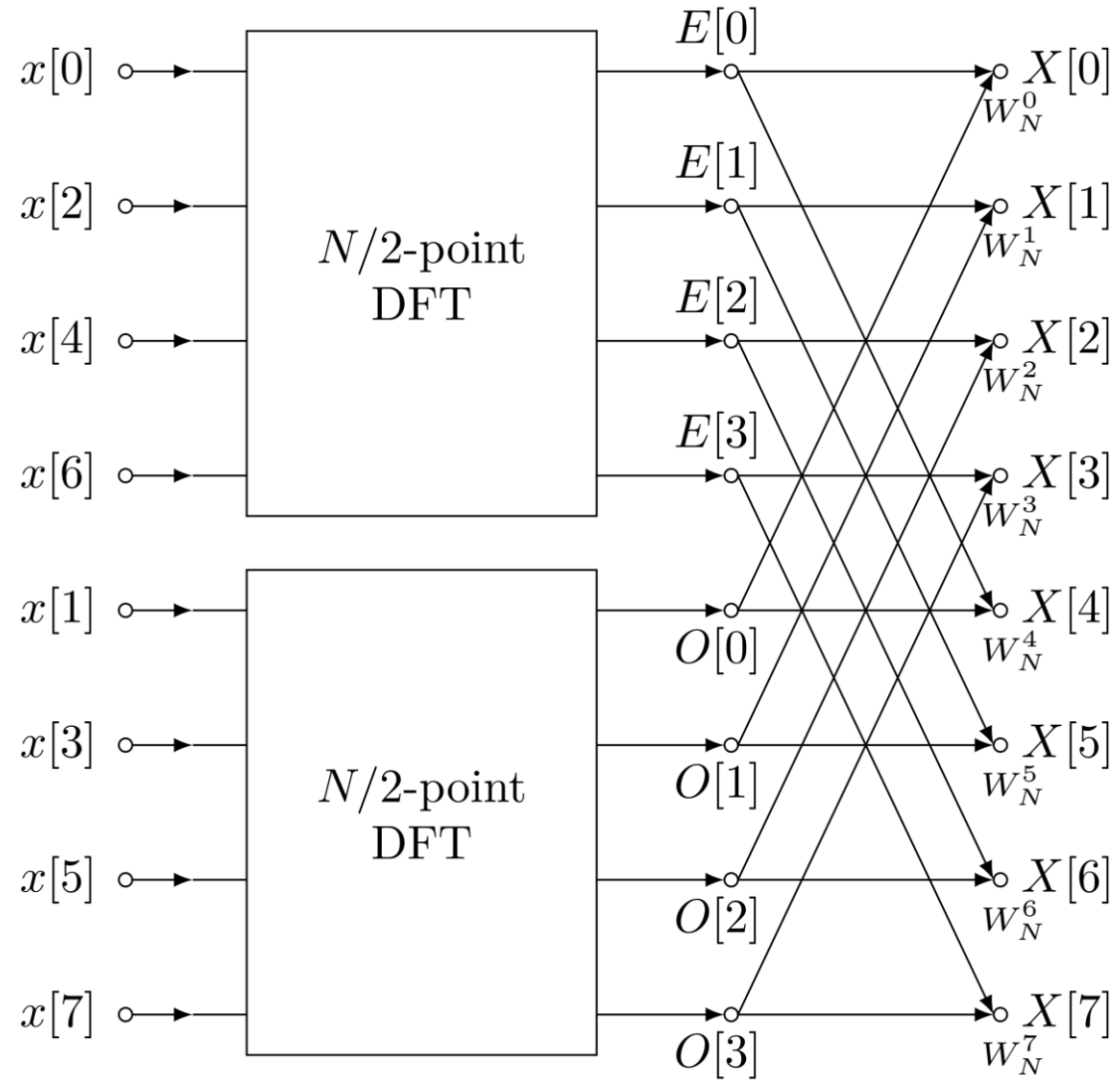
- Divide the DFT matrix recursively into smaller DFTs and then combine them
 - Based on the multiplications on complex root of unity
- **Radix-2 decimation-in-time (DIT) FFT**
 - Divide between odd and even inputs

$$X_k = \sum_{m=0}^{N/2-1} x_{2m} e^{-\frac{2\pi i}{N}(2m)k} + \sum_{m=0}^{N/2-1} x_{2m+1} e^{-\frac{2\pi i}{N}(2m+1)k}$$

- By rearranging,

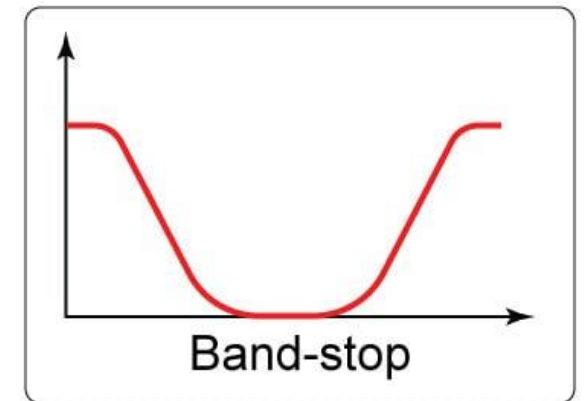
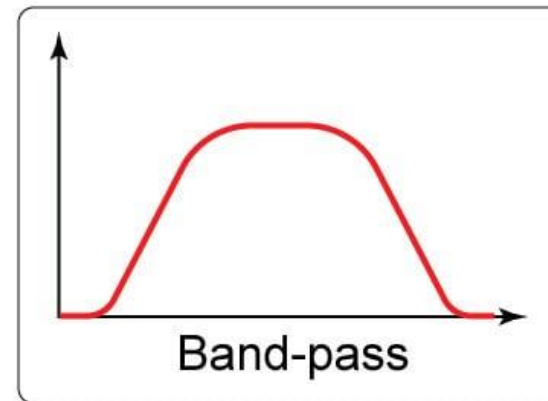
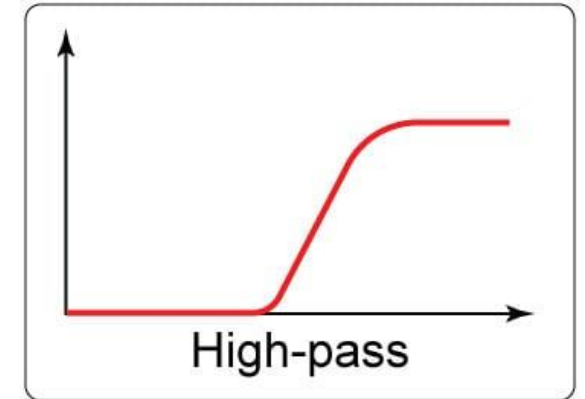
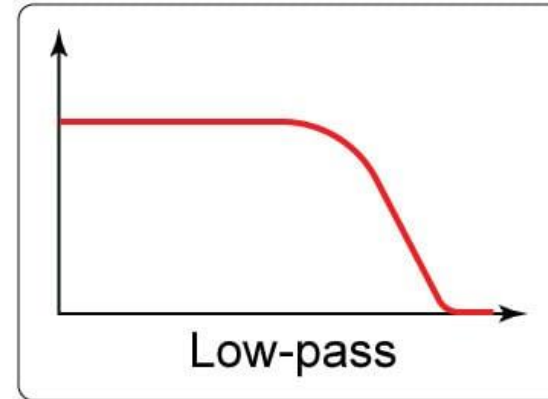
$$X_k = \underbrace{\sum_{m=0}^{N/2-1} x_{2m} e^{-\frac{2\pi i}{N/2}mk}}_{\text{DFT of even-indexed part of } x_n} + e^{-\frac{2\pi i}{N}k} \underbrace{\sum_{m=0}^{N/2-1} x_{2m+1} e^{-\frac{2\pi i}{N/2}mk}}_{\text{DFT of odd-indexed part of } x_n} = E_k + e^{-\frac{2\pi i}{N}k} O_k \quad \text{for } k = 0, \dots, \frac{N}{2} - 1.$$

FFT for $N = 8$



Application of FFT in Sensing -- Filters

- Extract a portion of the signal components
- Example: Acoustic data processing
 - You want to analyze the acoustic chirp sent from your smartphone
 - However, the sound emitted may get mixed with other environmental noises
 - Pass the received signal through a band-pass filter to extract only the components of the target frequency band



In Summary

- Signal propagation, distortion, reflection, etc., can help us sense the environment
 - Doppler analysis helps in identifying moving objects or movement patterns
 - The reflection patterns (frequency components, time of flight, etc.) can be used to compute the distance of the object from the transmitter, angle of arrival, etc.
 - Phase shift can be used to identify material properties
- Fourier transform helps us to identify the frequency components in the signal
 - Helpful for signal analysis (we'll see the details later)
 - Useful for preprocessing the signals – removing unwanted signal components – low-pass/ high-pass/ bandpass filters



Happy Learning!

Some resources
related to this topic

