

# EEE3030 Digital Signal Processing Notes

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# Introduction

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This set of notes will cover 1. The prerequisites for learning about digital signal processing methods, 2. The aforementioned methods and how we use them and where they are used.

## Vectors

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### 2.1 What is a Vector?

A **vector** is a mathematical object that has both **magnitude** (length) and **direction**. Vectors are often written in boldface, such as  $\mathbf{v}$ , or with an arrow,  $\vec{v}$ . In coordinate form, a vector in  $\mathbb{R}^n$  is written as

$$\mathbf{v} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}.$$

### 2.2 Vector Equality

Two vectors  $\mathbf{u}$  and  $\mathbf{v}$  are equal if and only if all their components are equal:

$$\mathbf{u} = \mathbf{v} \quad \Leftrightarrow \quad u_i = v_i \quad \forall i.$$

### 2.3 Vector Addition

The sum of two vectors is obtained by adding their corresponding components:

$$\mathbf{u} + \mathbf{v} = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix} + \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} = \begin{bmatrix} u_1 + v_1 \\ u_2 + v_2 \\ \vdots \\ u_n + v_n \end{bmatrix}.$$

### 2.4 Scalar Multiplication

Multiplying a vector by a scalar  $c \in \mathbb{R}$  stretches or shrinks it:

$$c\mathbf{v} = c \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} = \begin{bmatrix} cv_1 \\ cv_2 \\ \vdots \\ cv_n \end{bmatrix}.$$

### 2.5 Vector Magnitude (Norm)

The length (or magnitude) of a vector  $\mathbf{v}$  is

$$\|\mathbf{v}\| = \sqrt{v_1^2 + v_2^2 + \cdots + v_n^2}.$$

### 2.6 Unit Vectors

A **unit vector** has magnitude 1. To normalize a vector  $\mathbf{v}$ :

$$\hat{\mathbf{v}} = \frac{\mathbf{v}}{\|\mathbf{v}\|}.$$

## 2.7 Dot Product (Scalar Product)

For vectors  $\mathbf{u}, \mathbf{v} \in \mathbb{R}^n$ :

$$\mathbf{u} \cdot \mathbf{v} = \sum_{i=1}^n u_i v_i = \|\mathbf{u}\| \|\mathbf{v}\| \cos \theta,$$

where  $\theta$  is the angle between  $\mathbf{u}$  and  $\mathbf{v}$ .

- If  $\mathbf{u} \cdot \mathbf{v} = 0$ , the vectors are orthogonal.

## 2.8 Cross Product (in $\mathbb{R}^3$ )

For vectors  $\mathbf{u}, \mathbf{v} \in \mathbb{R}^3$ :

$$\mathbf{u} \times \mathbf{v} = \begin{vmatrix} \mathbf{i} & \mathbf{j} & \mathbf{k} \\ u_1 & u_2 & u_3 \\ v_1 & v_2 & v_3 \end{vmatrix} = (u_2 v_3 - u_3 v_2)\mathbf{i} - (u_1 v_3 - u_3 v_1)\mathbf{j} + (u_1 v_2 - u_2 v_1)\mathbf{k}.$$

The cross product is a vector perpendicular to both  $\mathbf{u}$  and  $\mathbf{v}$ .

## 2.9 Projection of One Vector onto Another

The projection of  $\mathbf{u}$  onto  $\mathbf{v}$  is

$$\text{proj}_{\mathbf{v}} \mathbf{u} = \left( \frac{\mathbf{u} \cdot \mathbf{v}}{\|\mathbf{v}\|^2} \right) \mathbf{v}.$$

## 2.10 Orthogonality

Vectors are **orthogonal** (perpendicular) if

$$\mathbf{u} \cdot \mathbf{v} = 0.$$

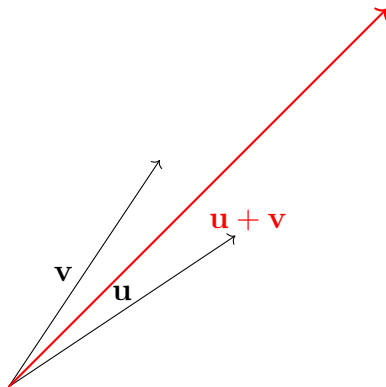
## 2.11 Linear Combination

A vector  $\mathbf{w}$  is a linear combination of  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$  if

$$\mathbf{w} = c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_k \mathbf{v}_k$$

for some scalars  $c_1, c_2, \dots, c_k$ .

## 2.12 Geometric Visualization



# Matrices

---

A **matrix** is a rectangular array of numbers arranged in rows and columns. An  $m \times n$  matrix has  $m$  rows and  $n$  columns:

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}.$$

## 3.1 Matrix Equality

Two matrices  $A$  and  $B$  are equal if they have the same size and

$$a_{ij} = b_{ij} \quad \forall i, j.$$

## 3.2 Matrix Addition

If  $A, B \in \mathbb{R}^{m \times n}$ , then

$$A + B = \begin{bmatrix} a_{11} + b_{11} & \cdots & a_{1n} + b_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} + b_{m1} & \cdots & a_{mn} + b_{mn} \end{bmatrix}.$$

## 3.3 Scalar Multiplication

For  $c \in \mathbb{R}$ ,

$$cA = \begin{bmatrix} ca_{11} & ca_{12} & \cdots & ca_{1n} \\ \vdots & \vdots & \ddots & \vdots \\ ca_{m1} & ca_{m2} & \cdots & ca_{mn} \end{bmatrix}.$$

## 3.4 Matrix Multiplication

If  $A \in \mathbb{R}^{m \times n}$  and  $B \in \mathbb{R}^{n \times p}$ , then

$$C = AB \in \mathbb{R}^{m \times p},$$

where each entry is computed as

$$c_{ij} = \sum_{k=1}^n a_{ik} b_{kj}.$$

### 3.4.1 Example

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix} = \begin{bmatrix} 1 \cdot 5 + 2 \cdot 7 & 1 \cdot 6 + 2 \cdot 8 \\ 3 \cdot 5 + 4 \cdot 7 & 3 \cdot 6 + 4 \cdot 8 \end{bmatrix} = \begin{bmatrix} 19 & 22 \\ 43 & 50 \end{bmatrix}.$$

## 3.5 Transpose

The **transpose** of  $A \in \mathbb{R}^{m \times n}$  is  $A^T \in \mathbb{R}^{n \times m}$ , defined by

$$(A^T)_{ij} = a_{ji}.$$

### 3.6 Identity Matrix

The identity matrix  $I_n$  is an  $n \times n$  matrix with ones on the diagonal and zeros elsewhere:

$$I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

It satisfies  $AI_n = I_nA = A$  for compatible  $A$ .

### 3.7 Determinant

For a square matrix  $A \in \mathbb{R}^{n \times n}$ , the **determinant**  $\det(A)$  is a scalar with geometric meaning (volume scaling).

For  $2 \times 2$  matrices:

$$\det \begin{bmatrix} a & b \\ c & d \end{bmatrix} = ad - bc.$$

For  $3 \times 3$  matrices:

$$\det \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix} = a(ei - fh) - b(di - fg) + c(dh - eg).$$

### 3.8 Inverse of a Matrix

A square matrix  $A$  is invertible if there exists  $A^{-1}$  such that

$$AA^{-1} = A^{-1}A = I.$$

For  $2 \times 2$  matrices:

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}, \quad A^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}, \quad \det(A) \neq 0.$$

### 3.9 Rank of a Matrix

The **rank** of a matrix  $A$  is the dimension of its column space (or row space). It equals the maximum number of linearly independent rows or columns.

### 3.10 Eigenvalues and Eigenvectors

For a square matrix  $A$ , a nonzero vector  $\mathbf{v}$  is an eigenvector if

$$A\mathbf{v} = \lambda\mathbf{v},$$

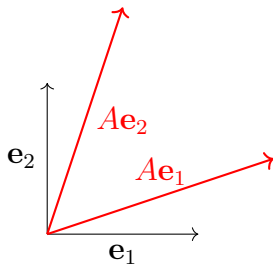
where  $\lambda$  is the corresponding eigenvalue.

To find  $\lambda$ , solve the characteristic equation:

$$\det(A - \lambda I) = 0.$$

### 3.11 Geometric Interpretation

- Matrix multiplication can be seen as a linear transformation of space.
- Determinant measures area/volume scaling and orientation.
- Eigenvectors are directions that remain unchanged under the transformation.



## Complex Numbers

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A complex number is defined as

$$z = x + iy, \quad x, y \in \mathbb{R}, \quad i^2 = -1.$$

$$\Re(z) = x, \quad \Im(z) = y, \quad \bar{z} = x - iy.$$

$$|z| = \sqrt{x^2 + y^2}, \quad \arg(z) = \theta.$$

### 4.1 Forms

$$z = x + iy \quad (\text{rectangular}), \quad z = r(\cos \theta + i \sin \theta) \quad (\text{polar}), \quad z = re^{i\theta} \quad (\text{exponential}).$$

### 4.2 Euler's Formula

$$e^{i\theta} = \cos \theta + i \sin \theta.$$

Special cases:

$$e^{i\pi} + 1 = 0, \quad e^{i\pi/2} = i.$$

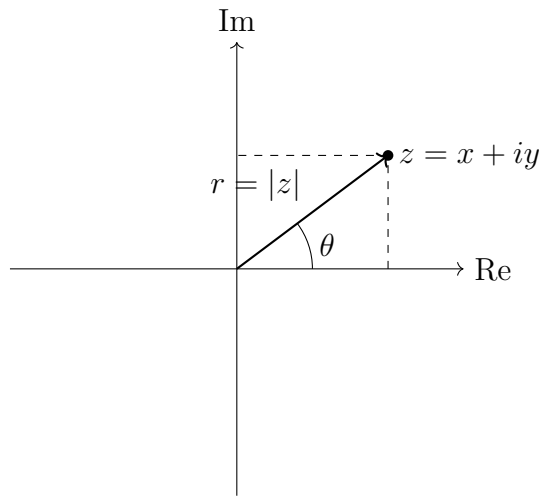
### 4.3 De Moivre's Theorem

$$(re^{i\theta})^n = r^n e^{in\theta} = r^n (\cos n\theta + i \sin n\theta).$$

### 4.4 Argand Diagram

A geometric representation of  $z = x + iy$  as a point  $(x, y)$ .



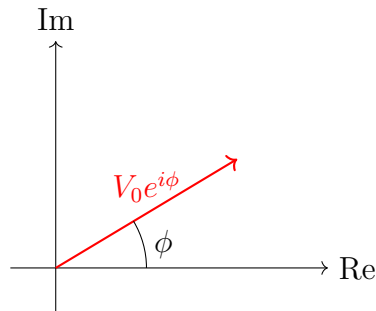


## 4.5 Phasor Diagram

A sinusoid

$$v(t) = V_0 \cos(\omega t + \phi) = \Re\{V_0 e^{i(\omega t + \phi)}\}$$

is represented by a rotating phasor vector.



## 4.6 Exponential Trigonometric Relations

$$\cos \theta = \frac{e^{i\theta} + e^{-i\theta}}{2}, \quad \sin \theta = \frac{e^{i\theta} - e^{-i\theta}}{2i}.$$

## 4.7 Hyperbolic Functions

$$\cosh x = \frac{e^x + e^{-x}}{2}, \quad \sinh x = \frac{e^x - e^{-x}}{2}.$$

Relations:

$$\cos(ix) = \cosh x, \quad \sin(ix) = i \sinh x.$$

## 4.8 Trigonometric Identities

### 4.8.1 Pythagorean

$$\sin^2 \theta + \cos^2 \theta = 1, \quad 1 + \tan^2 \theta = \sec^2 \theta, \quad 1 + \cot^2 \theta = \csc^2 \theta.$$

### 4.8.2 Sum and Difference

$$\sin(a \pm b) = \sin a \cos b \pm \cos a \sin b,$$

$$\cos(a \pm b) = \cos a \cos b \mp \sin a \sin b,$$

$$\tan(a \pm b) = \frac{\tan a \pm \tan b}{1 \mp \tan a \tan b}.$$

### 4.8.3 Double Angle

$$\sin(2\theta) = 2 \sin \theta \cos \theta, \quad \cos(2\theta) = \cos^2 \theta - \sin^2 \theta,$$
$$\tan(2\theta) = \frac{2 \tan \theta}{1 - \tan^2 \theta}.$$

### 4.8.4 Half Angle

$$\sin^2 \frac{\theta}{2} = \frac{1 - \cos \theta}{2}, \quad \cos^2 \frac{\theta}{2} = \frac{1 + \cos \theta}{2}.$$

### 4.8.5 Product-to-Sum

$$\begin{aligned}\sin a \sin b &= \frac{1}{2}[\cos(a - b) - \cos(a + b)], \\ \cos a \cos b &= \frac{1}{2}[\cos(a - b) + \cos(a + b)], \\ \sin a \cos b &= \frac{1}{2}[\sin(a + b) + \sin(a - b)].\end{aligned}$$

### 4.8.6 Sum-to-Product

$$\begin{aligned}\sin a \pm \sin b &= 2 \sin \left( \frac{a \pm b}{2} \right) \cos \left( \frac{a \mp b}{2} \right), \\ \cos a + \cos b &= 2 \cos \left( \frac{a + b}{2} \right) \cos \left( \frac{a - b}{2} \right), \\ \cos a - \cos b &= -2 \sin \left( \frac{a + b}{2} \right) \sin \left( \frac{a - b}{2} \right).\end{aligned}$$

## 4.9 Hyperbolic Identities

### 4.9.1 Fundamental

$$\cosh^2 x - \sinh^2 x = 1.$$

### 4.9.2 Sum and Difference

$$\begin{aligned}\sinh(a \pm b) &= \sinh a \cosh b \pm \cosh a \sinh b, \\ \cosh(a \pm b) &= \cosh a \cosh b \pm \sinh a \sinh b.\end{aligned}$$

### 4.9.3 Double Angle

$$\sinh(2x) = 2 \sinh x \cosh x, \quad \cosh(2x) = \cosh^2 x + \sinh^2 x.$$

### 4.9.4 Half Angle

$$\cosh^2 \frac{x}{2} = \frac{\cosh x + 1}{2}, \quad \sinh^2 \frac{x}{2} = \frac{\cosh x - 1}{2}.$$

## 4.10 Roots of Unity

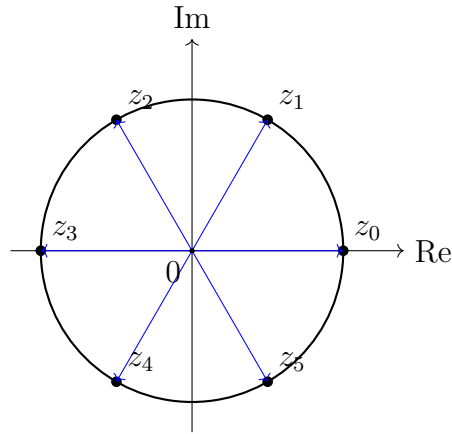
The  $n$ -th roots of unity are the solutions of

$$z^n = 1.$$

They are given by

$$z_k = e^{i \frac{2\pi k}{n}} = \cos \left( \frac{2\pi k}{n} \right) + i \sin \left( \frac{2\pi k}{n} \right), \quad k = 0, 1, 2, \dots, n-1.$$

These correspond to  $n$  equally spaced points on the unit circle in the Argand plane.



For example, for  $n = 6$  the roots are

$$z_k = e^{i\frac{2\pi k}{6}}, \quad k = 0, 1, 2, 3, 4, 5,$$

which are vertices of a regular hexagon inscribed in the unit circle.

## 4.11 Poles and Zeroes

In complex analysis and systems theory, the behaviour of a function

$$F(s) = \frac{N(s)}{D(s)}$$

is characterized by its **zeroes** and **poles**.

- A **zero** is a value  $s_0$  such that  $F(s_0) = 0$  (i.e.  $N(s_0) = 0$ ).
- A **pole** is a value  $s_p$  where  $F(s)$  tends to infinity (i.e.  $D(s_p) = 0$ ).

### 4.11.1 Example:

Consider

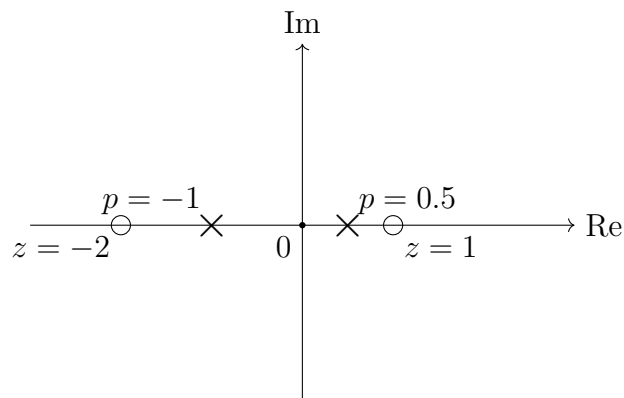
$$F(s) = \frac{(s-1)(s+2)}{(s-0.5)(s+1)}.$$

Zeroes:  $s = 1, s = -2$       Poles:  $s = 0.5, s = -1$

### 4.11.2 Pole–Zero Diagram

Poles and zeroes are plotted in the complex plane (often the  $s$ -plane or  $z$ -plane):

- Zeroes are marked with a  $\circ$  (circle).
- Poles are marked with a  $\times$  (cross).



### 4.11.3 Interpretation

- Zeroes indicate where the system response is **cancelled**.
- Poles indicate where the system response is **resonant** or tends to infinity.
- The relative locations of poles and zeroes determine the stability and frequency response of a system.

## Fourier Series

---

Any periodic function  $f(t)$  with period  $T$  can be expanded into a Fourier series:

$$f(t) = \frac{a_0}{2} + \sum_{n=1}^{\infty} \left( a_n \cos\left(\frac{2\pi n}{T}t\right) + b_n \sin\left(\frac{2\pi n}{T}t\right) \right).$$

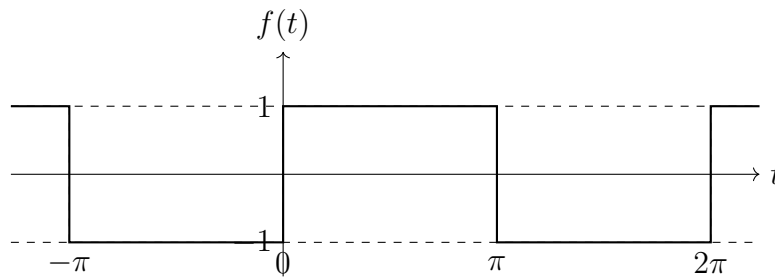
### 5.1 Fourier Coefficients

$$a_0 = \frac{2}{T} \int_0^T f(t) dt, \quad a_n = \frac{2}{T} \int_0^T f(t) \cos\left(\frac{2\pi n}{T}t\right) dt,$$
$$b_n = \frac{2}{T} \int_0^T f(t) \sin\left(\frac{2\pi n}{T}t\right) dt.$$

### 5.2 Example: Square Wave

Consider a square wave  $f(t)$  with period  $T = 2\pi$  defined as

$$f(t) = \begin{cases} 1, & 0 < t < \pi, \\ -1, & -\pi < t < 0, \end{cases} \quad \text{and extended periodically.}$$



### 5.3 Step 1: Compute $a_0$

$$a_0 = \frac{2}{2\pi} \int_{-\pi}^{\pi} f(t) dt = \frac{1}{\pi} \left( \int_{-\pi}^0 -1 dt + \int_0^{\pi} 1 dt \right) = 0.$$

### 5.4 Step 2: Compute $a_n$

$$a_n = \frac{1}{\pi} \int_{-\pi}^{\pi} f(t) \cos(nt) dt.$$

Since  $f(t)$  is odd and  $\cos(nt)$  is even, their product is odd. Thus

$$a_n = 0.$$

### 5.5 Step 3: Compute $b_n$

$$b_n = \frac{1}{\pi} \int_{-\pi}^{\pi} f(t) \sin(nt) dt.$$

Now  $f(t)$  is odd and  $\sin(nt)$  is odd  $\implies$  product is even:

$$b_n = \frac{2}{\pi} \int_0^{\pi} (1) \sin(nt) dt.$$

$$b_n = \frac{2}{\pi} \left[ \frac{1 - \cos(n\pi)}{n} \right] = \frac{2}{n\pi} (1 - (-1)^n).$$

So:

$$b_n = \begin{cases} \frac{4}{n\pi}, & n \text{ odd,} \\ 0, & n \text{ even.} \end{cases}$$

### 5.6 Final Fourier Series (Sine Form)

$$f(t) = \frac{4}{\pi} \left( \sin t + \frac{1}{3} \sin 3t + \frac{1}{5} \sin 5t + \cdots \right).$$

### 5.7 Complex Form of Fourier Series

We can also write the Fourier series using complex exponentials:

$$f(t) = \sum_{n=-\infty}^{\infty} c_n e^{int}, \quad c_n = \frac{1}{2\pi} \int_{-\pi}^{\pi} f(t) e^{-int} dt.$$

### 5.8 Step 1: Compute $c_n$

For the square wave:

$$c_n = \frac{1}{2\pi} \left( \int_0^{\pi} 1 \cdot e^{-int} dt + \int_{-\pi}^0 -1 \cdot e^{-int} dt \right).$$

$$c_n = \frac{1}{2\pi} \left( \frac{1 - e^{-in\pi}}{-in} - \frac{e^{in\pi} - 1}{-in} \right).$$

$$c_n = \frac{1}{\pi in} (1 - (-1)^n).$$

Thus:

$$c_n = \begin{cases} \frac{2}{i\pi n}, & n \text{ odd,} \\ 0, & n \text{ even.} \end{cases}$$

### 5.9 Step 2: Final Complex Series

$$f(t) = \sum_{\substack{n=-\infty \\ n \text{ odd}}}^{\infty} \frac{2}{i\pi n} e^{int}.$$

## 5.10 Interpretation

- The sine form shows explicitly the harmonic content (only odd harmonics).
- The complex form is more compact and symmetric, useful in analysis and physics.
- Both represent the same square wave in the limit of infinite harmonics.

# Fourier Transforms

---

The Fourier Transform (FT) converts a time-domain signal into its frequency-domain representation.

## 6.1 From Fourier Series to Fourier Transform (Step by Step)

A periodic signal  $x_T(t)$  of period  $T$  can be expressed as a Fourier Series:

$$x_T(t) = \sum_{n=-\infty}^{\infty} c_n e^{in\omega_0 t}, \quad \omega_0 = \frac{2\pi}{T}.$$

The coefficients are

$$c_n = \frac{1}{T} \int_{-T/2}^{T/2} x_T(t) e^{-in\omega_0 t} dt.$$

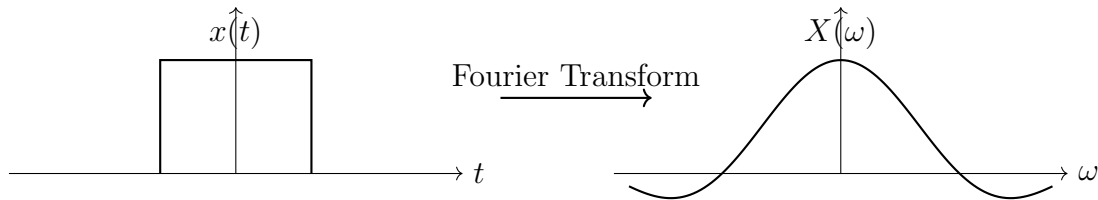
1. As  $T \rightarrow \infty$ ,  $x_T(t) \rightarrow x(t)$  (aperiodic signal).
2. The spacing between harmonics  $\omega_0 = 2\pi/T \rightarrow 0$ , giving a continuous frequency axis  $\omega$ .
3. The discrete coefficients  $c_n$  become continuous spectrum values  $X(\omega)$ .
4. The summation turns into an integral.

Thus, the Fourier Transform pair is obtained:

$$X(\omega) = \int_{-\infty}^{\infty} x(t) e^{-i\omega t} dt, \quad x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\omega) e^{i\omega t} d\omega.$$

## 6.2 Time and Frequency Domains

- **Time domain**  $x(t)$ : signal evolution with respect to time.
- **Frequency domain**  $X(\omega)$ : how much of each sinusoidal frequency is present.



A localised rectangular pulse in time spreads into a sinc in frequency. Conversely, a single sinusoid in time becomes a delta spike in frequency.

## 6.3 Fourier Transform Properties: Shifts

- **Time Shift Theorem:** If  $x(t) \leftrightarrow X(\omega)$ , then

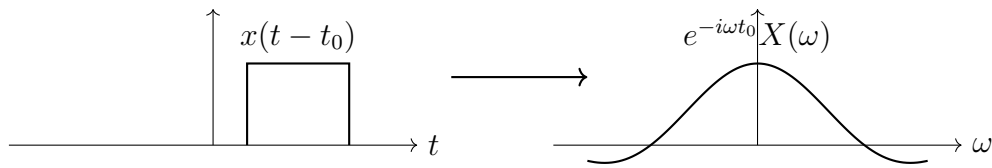
$$x(t - t_0) \iff e^{-i\omega t_0} X(\omega).$$

(Time delay  $\rightarrow$  frequency phase shift)

- **Frequency Shift Theorem:** If  $x(t) \leftrightarrow X(\omega)$ , then

$$e^{i\omega_0 t} x(t) \iff X(\omega - \omega_0).$$

(Modulation in time  $\rightarrow$  spectrum shift)



## 6.4 Worked Example: Rectangular Pulse

Let

$$x(t) = \begin{cases} 1, & |t| \leq \frac{T}{2}, \\ 0, & |t| > \frac{T}{2}. \end{cases}$$

Fourier Transform:

$$X(\omega) = \int_{-T/2}^{T/2} e^{-i\omega t} dt = \frac{2 \sin(\omega T/2)}{\omega}.$$

Thus

$$X(\omega) = T \cdot \text{sinc}\left(\frac{\omega T}{2\pi}\right).$$

# Derivation of the Discrete Fourier Transform (DFT)

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## 7.1 Step 1: Continuous-Time Fourier Transform (CTFT)

For a continuous-time signal  $x(t)$ , the Fourier Transform is

$$X(\omega) = \int_{-\infty}^{\infty} x(t) e^{-i\omega t} dt.$$

## 7.2 Step 2: Sampling the Signal

We sample  $x(t)$  every  $T_s$  seconds:

$$x[n] = x(nT_s), \quad n \in \mathbb{Z}.$$

This produces a discrete-time sequence.

## 7.3 Step 3: Discrete-Time Fourier Transform (DTFT)

The DTFT is

$$X(e^{i\omega}) = \sum_{n=-\infty}^{\infty} x[n] e^{-i\omega n}.$$

Properties:

- $X(e^{i\omega})$  is continuous in  $\omega$ .
- $X(e^{i\omega})$  is  $2\pi$ -periodic.

## 7.4 Step 4: Finite-Length Signals

In practice, we only keep  $N$  samples:

$$x[n] = 0 \quad \text{for } n < 0 \text{ or } n \geq N.$$

Equivalently, we treat  $x[n]$  as  $N$ -periodic:

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