

Fourier Transform summary

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- Periodic functions
- Fourier Transform
- Fourier Transform Properties
- Fourier Transform and LTI systems



Periodic signals



A continuous-time signal x(t) is **periodic**, when there is a positive non-zero value *T* for which:

$$x(t + T) = x(t)$$
, for all t .

• *T* is referred to as the period of the signal.

• **Frequency**: $F = \frac{1}{\tau}$. It is measured in Hertz (Hz).

• Angular frequency: $\Omega = 2\pi F = \frac{2\pi}{\tau}$.



Periodic signals



 $\frac{\sin t}{\cos t}$

Trigonometric signals (sine, cosine):

$$x(t) = A\cos(\Omega t + \varphi) = \frac{A}{2} \left[e^{i(\Omega t + \varphi)} + e^{-i(\Omega t + \varphi)} \right]$$

$$x(t) = A\sin(\Omega t + \varphi) = \frac{A}{2i} \left[e^{i(\Omega t + \varphi)} - e^{-i(\Omega t + \varphi)} \right]_{x(t)}$$

- A: amplitude,
- Ω : angular frequency, φ : phase
- The period of the sinusoid is $T = \frac{2\pi}{\Omega}$.



Periodic signals



Complex exponential signal:

$$x(t) = e^{st} = e^{(\sigma + i\Omega)t} = e^{\sigma t} (\cos\Omega t + i \sin\Omega t).$$

Special case:

 $x(t) = Ae^{i\Omega t} = A\cos\Omega t + iA\sin\Omega t.$





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Fourier transform (**FT**) of signal x(t) is a function transformation $X(\Omega)$:

$$X(\Omega) \triangleq \mathcal{F}\{x(t)\} = \int_{-\infty}^{\infty} x(t)e^{-i\Omega t}dt.$$

Inverse Fourier transform of $X(\Omega)$ is defined as: $x(t) \triangleq \mathcal{F}^{-1}\{X(\Omega)\} = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\Omega) e^{i\Omega t} d\Omega.$





- Fourier transform pair is denoted by: $x(t) \leftrightarrow X(\Omega)$.
- Fourier transform essentially decomposes an 1D signal x(t) into a sum of periodic complex exponential functions, resulting in signal *spectrum* (frequency content) X(Ω) at frequency Ω.





The Fourier transform $X(\Omega)$ is, in general, a complex one:

$$X(\Omega) = R(\Omega) + iI(\Omega) = |X(\Omega)|e^{i\varphi(\Omega)}.$$

- $X(\Omega)$ is referred as the (Fourier) **spectrum** of x(t).
- $|X(\Omega)|$ is the **spectrum magnitude**.
- $\varphi(\Omega)$ is the **spectrum phase**.





Band-limited signals (or low-pass signals) are those satisfying:

 $|X(\Omega)| = 0, \qquad |\Omega| > \Omega_{max}.$

• Ω_{max} is the band-limited signal **bandwidth**.





FT $X(\Omega)$ is given by: $X(\Omega) = \int_{-\infty}^{\infty} x(t) e^{-i\Omega t} dt = \int_{-\infty}^{\infty} x(t) (\cos(\Omega t) - i\sin(\Omega t)) dt.$

If x(t) is a real-valued signal, then: $R(\Omega) = \int_{-\infty}^{\infty} x(t) \cos(\Omega t) dt.$ $I(\Omega) = -\int_{-\infty}^{\infty} x(t) \sin(\Omega t) dt.$





Consequently, if x(t) is a real function:

- $R(\Omega), I(\Omega)$ are **even/odd functions**, respectively:
 - $R(-\Omega) = R(\Omega), \qquad I(-\Omega) = -I(\Omega).$

FT satisfies the *complex conjugation* property: $X(-\Omega) = X^*(\Omega),$

* denotes complex conjugation.

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Dirichlet conditions are sufficient for $FT X(\Omega)$ convergence:

1. x(t) is absolutely integrable, as defined by:

 $\int_{-\infty}^{\infty} |x(t)| dt < \infty.$

2. x(t) has bounded variation within any finite interval [a, b]. 3. x(t) is a continuous function or contains finite number of finite discontinuities in any given finite interval. If conditions 2,3 are met, x(t) is a **piecewise smooth** *function*.





The **square window function**:

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

satisfies Dirichlet conditions. Its FT is a real **sync function**:

$$X(\Omega) = \int_{-\infty}^{\infty} x(t)e^{-i\Omega t}dt = \int_{-T/2}^{T/2} e^{-i\Omega t}dt = \frac{2}{\Omega}\sin\left(\frac{\Omega T}{2}\right)$$
$$= T\frac{\sin\left(\frac{\Omega T}{2}\right)}{\Omega T/2}.$$

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Fourier transform:

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

is a special case of bilateral *Laplace transform* (*LT*):

$$X(s) = \int_{-\infty}^{\infty} x(t)e^{-st}dt,$$

for $s = i\Omega$.





If $s = \sigma + i\Omega$, Laplace transform becomes:

$$X(\sigma + i\Omega) = \int_{-\infty}^{\infty} x(t)e^{-(\sigma + i\Omega)t}dt = \int_{-\infty}^{\infty} |x(t)e^{-\sigma t}|e^{-i\Omega t}dt$$

and:

$$X(\sigma + i\Omega) = \mathcal{F}\{x(t)e^{-\sigma t}\}.$$

- The bilateral Laplace transform of x(t) can be interpreted as the Fourier transform of signal $x(t)e^{-\sigma t}$.
- This holds only if x(t) is absolutely integrable.

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Delta function transforms:

• LT of delta function $\delta(t)$ is given by:

 $\mathcal{L}{\delta(t)} = 1,$ for all *s*.

• By definition, the FT of delta function is: $\mathcal{F}\{\delta(t)\} = \int_{-\infty}^{\infty} \delta(t) e^{-i\Omega t} dt = 1.$







LT of right-sided exponential signal: $x(t) = e^{-at}u(t), a > 0,$ is given by:

$$\mathcal{L}\{x(t)\} = X(s) = \frac{1}{s+a}, \qquad \operatorname{Re}(s) > -a.$$

By definition, the FT of the same signal is:
$$\mathcal{F}\{x(t)\} = X(\Omega) = \int_{-\infty}^{\infty} e^{-at} u(t) e^{-i\Omega t} dt = \int_{0}^{\infty} e^{-(a+i\Omega)t} dt = \frac{1}{i\Omega + a}.$$





$$X(\Omega) = X(s) \Big|_{s=i\Omega}.$$

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• Note that x(t) is absolutely integrable.





Fourier Transform pairs

Signal $x(t)$	Fourier Transform $X(\Omega)$	
$\delta(t)$	1	
$\delta(t-t_0)$	$e^{-i\Omega t_0}$	
1	$2\pi\delta(arOmega)$	
$e^{i\Omega_0 t}$	$2\pi\delta(\Omega-\Omega_0)$	
$\cos(\Omega_0 t)$	$\pi[\delta(\varOmega-\varOmega_0)+\delta(\varOmega+\varOmega_0)]$	
$\sin(\Omega_0 t)$	$-i\pi[\delta(\Omega-\Omega_0)-\delta(\Omega+\Omega_0)]$	
u(t)	$\pi\delta(\Omega) + \frac{1}{i\Omega}$	
u(-t)	$\pi\delta(\Omega) - rac{1}{i\Omega}$	





Fourier Transform pairs

Signal $x(t)$	Fourier Transform $X(\Omega)$	
$e^{-at}u(t), \qquad a>0$	$\frac{1}{i\Omega + a}$	
$e^{-a t }$, $a > 0$	$\frac{2a}{a^2 + \Omega^2}$	
e^{-at^2} , $a > 0$	$\sqrt{\frac{\pi}{a}}e^{-\Omega^2/4a}$	
$x(t) = \begin{cases} 1, & t < a \\ 0, & t > a \end{cases}$	$2a \frac{\sin(\Omega a)}{\Omega a}$	
$\frac{\sin(at)}{\pi t}$	$X(\Omega) = \begin{cases} 1, & \Omega < a \\ 0, & \Omega > a \end{cases}$	
$\operatorname{sgn}(t)$	$\frac{2}{i\Omega}$	

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• Linearity:

$$a_1 x_1(t) + a_2 x_2(t) \leftrightarrow a_1 X_1(\Omega) + a_2 X_2 \Omega.$$

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• Time Shifting:

$$x(t-t_0) \leftrightarrow e^{-i\Omega t_0} X(\Omega).$$

 $e^{i\Omega_0 t} x(t) \leftrightarrow X(\Omega - \Omega_0).$

- A time shift results only in a FT phase change.
- Frequency Shifting:



Fourier Transform Properties x(t)x(at)• Time Scaling: 2 -1-21 -2-1 $\mathbf{2}$ t1 $x(at) \leftrightarrow \frac{1}{|a|} X\left(\frac{\Omega}{a}\right).$ • Effect of time scaling on $X(\Omega)$ $\frac{1}{|a|}X(\frac{\Omega}{a})$ a square wave signal: $x(t) = \begin{cases} 1, & |t| < 2\\ 0, & |t| > 2 \end{cases}$ 2 m for a = 2. Signal frequency range Effects of signal time scaling increases.

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- Time Reversal (signal flipping about t = 0): $x(-t) \leftrightarrow X(-\Omega)$.
- Signal flipping is used in convolution definition.

• Duality (or Symmetry): $X(t) \leftrightarrow 2\pi x(-\Omega).$



- First order differentiation in the temporal domain: $\frac{dx(t)}{dt} \leftrightarrow i\Omega X(\Omega).$
- Higher order differentiation in the temporal domain:

$$\frac{d^n x(t)}{dt^n} \leftrightarrow (i\Omega)^n X(\Omega).$$

- Multiplication by $(i\Omega)^n$ amounts to high-pass signal filtering.
- Therefore, differentiation is a high-pass system.

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- Integration in the temporal domain: $\int_{-\infty}^{t} x(-\tau) d\tau \leftrightarrow \pi X(0) \delta(\Omega) + \frac{1}{i\Omega} X(\Omega).$
- Division by *i*Ω amounts to low-pass signal filtering. Therefore, integration is a *low-pass system*.

 $(-it)x(t) \leftrightarrow \frac{dX(\Omega)}{d\Omega}.$

Differentiation in the Frequency Domain:



Signal convolution:

 $x(t) * h(t) \leftrightarrow X(\Omega)H(\Omega).$

- As FT can be computed through Discrete Fourier Transform (DFT) for discrete signals, using the Fast Fourier Transform (FFT) algorithms, it leads to fast convolution calculation.
- Signal multiplication:

$$x_1(t)x_2(t) \leftrightarrow \frac{1}{2\pi}X_1(\Omega) * X_2(\Omega).$$



• Parseval's theorem:

$$\int_{-\infty}^{\infty} x_1(t) x_2(t) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} X_1(\Omega) X_2(-\Omega) d\Omega$$

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$$\int_{-\infty}^{\infty} |x(t)|^2 dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} |X(\Omega)|^2 d\Omega$$

• Essentially, FT preserves energy after transformation.





 $X(-\Omega)=X^*(\Omega),$

• If $x(t) = x_e(t) + x_o(t)$, where $x_e(t) = x_e(-t)$ and $x_o(t) = -x_o(-t)$ are the even and odd components of x(t) and: $x_e(t) \leftrightarrow A(\Omega),$ $x_o(t) \leftrightarrow iB(\Omega).$

then:

 $x(t) \leftrightarrow X(\Omega) = A(\Omega) + iB(\Omega).$





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 The output y(t) of a continuous-time LTI system is given by the convolution:

y(t) = x(t) * h(t).

- Applying the convolution property, we get: $Y(\Omega) = X(\Omega)H(\Omega).$
- LTI system *frequency response* $H(\Omega)$: $H(\Omega) = \frac{Y(\Omega)}{X(\Omega)}$

is the FT of its impulse response h(t).

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$$\delta(t) \longrightarrow T[\delta(t)] \longrightarrow h(t)$$

$$x(t) \longrightarrow \boxed{T[x(t)]} \longrightarrow y(t) = x(t) * h(t)$$

$$x(t) \longrightarrow \boxed{\mathcal{F}\{x(t)\}} \longrightarrow Y(\Omega) = X(\Omega)H(\Omega)$$

$$h(t) \longrightarrow \boxed{\mathcal{F}\{h(t)\}}$$

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Polar representation of system frequency response $H(\Omega)$:

 $H(\Omega) = |H(\Omega)|e^{i\theta_H(\Omega)}.$

- $|H(\Omega)|$ is the system **frequency response magnitude**.
- $\theta_H(\Omega)$ is the system **frequency response phase**.





The behavior of the LTI system in the frequency domain is completely characterized by its frequency response $H(\Omega)$:

- If $|H(\Omega)| < 1$, the respective input signal frequency is attenuated.
- If $|H(\Omega)| = 0$, the respective frequency is cutoff.
- If $|H(\Omega)| = 1$, the respective frequency passes through the system.
- If $|H(\Omega)| > 1$, the respective frequency is amplified.





results in system output:

 $Y(\Omega) = 2\pi H(\Omega_0)\delta(\Omega - \Omega_0).$

• Taking the inverse FT of $Y(\Omega)$, we get: $y(t) = H(\Omega_0)e^{i\Omega_0 t}$.

• An LTI does **not** change the frequency Ω_0 .

• $e^{i\Omega t}$ is an LTI system eigenfunction with eigenvalue $H(\Omega_0)$.





$$X(\Omega) = |X(\Omega)|e^{i\theta_X(\Omega)}, \quad H(\Omega) = |H(\Omega)|e^{i\theta_H(\Omega)},$$
$$Y(\Omega) = |Y(\Omega)|e^{i\theta_Y(\Omega)}.$$

- Then from $Y(\Omega) = X(\Omega)H(\Omega)$ we get: $|Y(\Omega)| = |X(\Omega)||H(\Omega)|,$ $\theta_Y(\Omega) = \theta_X(\Omega) + \theta_H(\Omega).$
- System frequency response magnitude $|H(\Omega)|$ is referred as the gain of the system.

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- Both signal differentiation and integration can be considered as linear systems having frequency response $H(\Omega)$.
- In signal differentiation:

$$\frac{d^n x(t)}{dt^n} \leftrightarrow Y(\Omega) = (i\Omega)^n X(\Omega).$$

multiplication in the frequency domain by $(i\Omega)^n$ results in a **high-pass system**:

 $H(\Omega)=(i\Omega)^n.$





- The higher the differentiation order is, the more profound its high-pass characteristics are.
- Differentiation typically enhances high-frequency signal noise.
- Therefore, low-pass signal filtering must be performed before signal differentiation.
- Second-order differentiation and beyond are typically avoided, due to their noise sensitivity.



• In integration:

$$\int_{-\infty}^{t} x(-\tau) d\tau \leftrightarrow \frac{1}{i\Omega} X(\Omega), \qquad \text{if } X(0) = 0.$$

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division by $i\Omega$ the frequency domain amounts to low-pass signal filtering:

$$H(\Omega) = \frac{1}{i\Omega}$$

Therefore, integration is a *low-pass system*.





A LTI system can be described by a linear differential equation with constant coefficients:

$$a_{n} \frac{d^{n} y(t)}{dt^{n}} + a_{n-1} \frac{d^{n-1} y(t)}{dt^{n-1}} + \dots + a_{0} y(t)$$

= $b_{m} \frac{d^{m} x(t)}{dt^{m}} + b_{m-1} \frac{d^{m-1} x(t)}{dt^{m-1}} + \dots + b_{0} x(t),$

• $a_i, b_i \in \mathbb{R}$.





Then, by applying the LT in both parts of the previous formula, we have:

$$H(i\Omega) = \frac{Y(i\Omega)}{X(i\Omega)} = \frac{b_m (i\Omega)^m + b_{m-1}(i\Omega)^{m-1} + \dots + b_0}{a_n (i\Omega)^n + a_{n-1} (i\Omega)^{n-1} + \dots + a_0}$$

- Frequency response has rational form.
- Under certain conditions, its magnitude |H(iΩ)| can become:
 - 0, thus completely attenuating the respective frequencies;
 - Very large (towards infinity), thus greatly amplifying the respective frequencies.





Filtering is the process where the amplitude/phase of the frequency components of a signal are modified or even reduced to zero.

• A "filter" is an LTI system, whose frequency response shows this selective frequency modification behavior.





- An *ideal filter* allows some selected signal frequencies to pass, while completely attenuating the rest.
- **Pass-band** is the range of frequencies passed by the filter.
- Stop-band is the range of frequencies rejected by the filter.















_ . .



 $\hat{\Omega}$

 $X(\Omega)$

 Ω_1

 Ω_2

$$|H(\Omega)| = \begin{cases} 1, & \Omega_1 < |\Omega| < \Omega_2 \\ 0, & \text{eslewhere.} \end{cases}$$

 Ω_2

 $-\Omega_2$

 $-\Omega_1$

 Ω_1, Ω_2 : **BP cut-off frequencies**. •

- Bandpass resonator is a band-pass filter having very narrow passpand band, typically around resonation frequency $\Omega_p: \Omega_1, \simeq \Omega_2 \simeq \Omega_s$.
- Radio transmitter/receiver oscillators ulletand information Analysis Lab



 Ω



- typically around frequency Ω_s : $\Omega_1, \simeq \Omega_2 \simeq \Omega_s$.
- 50/60 Hz rejection filters.

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• To avoid phase distortion, a filter phase response $\theta_H(\Omega)$ is typically linear over the frequency range of interest:

$$\theta_H(\Omega) = a\Omega.$$

Ideal frequency-selective filters are noncausal systems.





- *Filter bandwidth* is the difference between filter cut-off frequencies, if its pass-band is not infinite:
- For an ideal low-pass filter, its bandwidth is equal to its cutoff frequency: $W_B = \Omega_c$.
- For an ideal bandpass filter, its bandwidth is the difference between its two cuttoff frequencies: $W_B = \Omega_2 \Omega_1$.
- Typically a *transition band* exists between passpands and stop bands.





- *Half-Power Bandwidth* $W_{3 dB}$ is defined by the frequency at which the frequency response amplitude $|H(\Omega)|$ drops to a value equal to $|H(\Omega_m)|/\sqrt{2}$.
- $H(\Omega_m)$ is the maximal frequency response amplitude.
- It shows the point at which the output power has dropped to half of its peak value.
- At this frequency, we have 3 dB attenuation:
- $L = 10 \log_{10}(|H(\Omega)|^2 / |H(\Omega_m)|^2) = 10 \log_{10}(1/2) \approx 3 \ dB.$





Analog *electric filters*:

- They are electric networks consisting of *resistances*, *capacitors*, *inductors*.
- The input-output relation of an *RC* filter is given by:

 $RC\frac{dy(t)}{dt} + y(t) = x(t).$













 Ω

 $|H(\Omega)|$

 Ω_0

 $\frac{\overline{1}}{\sqrt{2}}$

 $-\Omega_0$

• The frequency response amplitude $|H(\Omega)|$ is: $|H(\Omega)| = \frac{1}{1/2}$.

The *RC* filter is a low-pass one.

1+

 $\left[\left(\Omega \right)^2 \right]^{1/2}$

• High frequencies are attenuated.





 $\hat{\Omega}$

 $\mathbf{h}\theta_H(\Omega)$

 Ω_0

 $\frac{\pi}{4}$

 $\frac{\pi}{4}$

 $-\Omega_0$

1D Filters

• Its frequency response phase $\theta_H(\Omega)$ is given by:

$$\theta_H(\Omega) = -\tan^{-1}\left(\frac{\Omega}{\Omega_0}\right).$$

- It is a nonlinear function of Ω.
- It can become almost linear for small Ω:

$$\theta_H(\Omega) \approx -\frac{\Omega}{\Omega_0}$$
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Its differentiation results in a second-order differential equation:

$$L\frac{d^2x(t)}{dt^2} + R\frac{dx(t)}{dt} + \frac{1}{C}x(t) = \frac{dy(t)}{dt}$$

and frequency response:

$$H(\Omega) = \frac{Y(\Omega)}{X(\Omega)} = \frac{i\Omega}{-L\Omega^2 + iR\Omega + \frac{1}{C}}$$



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1D Filters

- y(t)• If R = 0, the frequency response becomes: $H(\Omega) = \frac{i\Omega}{-L\Omega^2 + \frac{1}{C}}.$ x(t)It is an *electric oscillator* typically resonating at frequency: $\Omega = 1/\sqrt{LC}.$ • It has been extensively used in radio
- transmitters.

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Temporal diffusion system:

$$\frac{dy(t)}{dt} = c(x(t) - y(t)).$$

• c: diffusion coefficient.



high concentration







Its frequency response $H(\Omega)$ can be found by taking the FT of both sides:

$$H(\Omega) = \frac{Y(\Omega)}{X(\Omega)} = \frac{c}{c + i\Omega}.$$

- It is a low-pass system.
- It can model many phenomena, e.g., pharmacokinetics
- It can be extended to 1D, 2D and 3D spatiotemporal diffusion and to information diffusion over graphs.





m

Mass-spring-damper mechanical system. If force f(t) is exercised on a mass m that is attached to a spring having constant k that follows Hooke's law and to a damper having damping constant β , the displacement y(t) is given by the exercised forces:

$$m\frac{d^2y(t)}{dt^2} + \beta\frac{dy(t)}{dt} + ky(t) = f(t).$$

• It models car suspension systems.



Fourier Transform of this system is: $-m\Omega^2 Y(\Omega) + \beta i\Omega Y(\Omega) + kY(\Omega) = F(\Omega).$

Therefore, its transfer function is given by:

$$H(\Omega) = \frac{Y(\Omega)}{F(\Omega)} = \frac{1}{-m\Omega^2 + \beta i\Omega + k}$$



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