Discrete Time Fourier Transform (DTFT)

Discrete Time Fourier Transform (DTFT)

- **The DTFT** is the Fourier transform of choice for analyzing infinite-length signals and systems
- Useful for conceptual, pencil-and-paper work, but not Matlab friendly (infinitely-long vectors)

Properties are very similar to the Discrete Fourier Transform (DFT) with a few caveats

■ We will derive the DTFT as the limit of the DFT as the signal length $N \to \infty$

Recall: DFT (Unnormalized)

- **Analysis (Forward DFT)**
	- Choose the DFT coefficients $X[k]$ such that the synthesis produces the signal x
	- The weight $X[k]$ measures the similarity between x and the harmonic sinusoid s_k
	- Therefore, $X[k]$ measures the "frequency content" of x at frequency k

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

- Synthesis (Inverse DFT)
	- Build up the signal x as a linear combination of harmonic sinusoids s_k weighted by the DFT coefficients $X[k]$

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

The Centered DFT

Both $x[n]$ and $X[k]$ can be interpreted as periodic with period N, so we will shift the intervals of interest in time and frequency to be centered around $n, k = 0$

$$
-\frac{N}{2} \leq n, k \leq \frac{N}{2} - 1
$$

■ The modified forward and inverse DFT formulas are

$$
X_u[k] = \sum_{n=-N/2}^{N/2-1} x[n] e^{-j\frac{2\pi}{N}kn}, \qquad -\frac{N}{2} \le k \le \frac{N}{2} - 1
$$

$$
x[n] = \frac{1}{N} \sum_{k=-N/2}^{N/2-1} X_u[k] e^{j\frac{2\pi}{N}kn} - \frac{N}{2} \le n \le \frac{N}{2} - 1
$$

Recall: DFT Frequencies

$$
X_u[k] = \sum_{n=-N/2}^{N/2-1} x[n] e^{-j\frac{2\pi}{N}kn}, \quad -\frac{N}{2} \le k \le \frac{N}{2} - 1
$$

 \blacksquare $X_u[k]$ measures the similarity between the time signal x and the harmonic sinusoid s_k

Therefore, $X_u[k]$ measures the "frequency content" of x at frequency

Take It To The Limit (1)

$$
X_u[k] = \sum_{n=-N/2}^{N/2-1} x[n] e^{-j\frac{2\pi}{N}kn}, \quad -\frac{N}{2} \le k \le \frac{N}{2} - 1
$$

- **■** Let the signal length N increase towards ∞ and study what happens to $X_u[k]$
- Key fact: No matter how large N grows, the frequencies of the DFT sinusoids remain in the interval 2π

$$
-\pi \leq \omega_k = \frac{2\pi}{N}k < \pi
$$

Discrete Time Fourier Transform (Forward)

As $N \to \infty$, the forward DFT converges to a function of the **continuous frequency variable** ω that we will call the forward discrete time Fourier transform (DTFT)

$$
\sum_{n=-N/2}^{N/2-1} x[n] e^{-j\frac{2\pi}{N}kn} \longrightarrow \sum_{n=-\infty}^{\infty} x[n] e^{-j\omega n} = X(\omega), \qquad -\pi \le \omega < \pi
$$

Recall: Inner product for infinite-length signals

$$
\langle x, y \rangle = \sum_{n=-\infty}^{\infty} x[n] y[n]^*
$$

Analysis interpretation: The value of the DTFT $X(\omega)$ at frequency ω measures the similarity of the infinite-length signal $x[n]$ to the infinite-length sinusoid $e^{j\omega n}$

Discrete Time Fourier Transform (Inverse)

Inverse unnormalized DFT

$$
x[n] = \frac{2\pi}{2\pi N} \sum_{k=-N/2}^{N/2-1} X_u[k] e^{j\frac{2\pi}{N}kn}
$$

In the limit as the signal length $N \to \infty$, the inverse DFT converges in a more subtle way:

$$
e^{j\frac{2\pi}{N}kn} \longrightarrow e^{j\omega n}, \qquad X_u[k] \longrightarrow X(\omega), \qquad \frac{2\pi}{N} \longrightarrow d\omega, \qquad \sum_{k=-N/2}^{N/2-1} \longrightarrow \int_{-\pi}^{\pi}
$$

resulting in the inverse DTFT

$$
x[n] = \int_{-\pi}^{\pi} X(\omega) e^{j\omega n} \frac{d\omega}{2\pi}, \quad \infty < n < \infty
$$

Synthesis interpretation: Build up the signal x as an infinite linear combination of sinusoids $e^{j\omega n}$ weighted by the DTFT $X(\omega)$

Summary

Discrete-time Fourier transform (DTFT)

$$
X(\omega) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\omega n}, \qquad -\pi \le \omega < \pi
$$

$$
x[n] = \int_{-\pi}^{\pi} X(\omega) e^{j\omega n} \frac{d\omega}{2\pi}, \qquad \infty < n < \infty
$$

The core "basis functions" of the DTFT are the sinusoids $e^{j\omega n}$ with arbitrary frequencies ω

- The DTFT can be derived as the limit of the DFT as the signal length $N \to \infty$
- The analysis/synthesis interpretation of the DFT holds for the DTFT, as do most of its properties

Eigenanalysis of LTI Systems (Infinite-Length Signals)

LTI Systems for Infinite-Length Signals

$$
x \longrightarrow \begin{array}{|c|c|}\n\hline\n\mathcal{H} & \longrightarrow y \\
\hline\ny & = \mathbf{H}x\n\end{array}
$$

For infinite length signals, H is an infinitely large **Toeplitz matrix** with entries

$$
[\mathbf{H}]_{n,m} = h[n-m]
$$

where h is the **impulse response**

- Goal: Calculate the eigenvectors and eigenvalues of H
- **Eigenvectors** v are input signals that emerge at the system output unchanged (except for a scaling by the eigenvalue λ) and so are somehow "fundamental" to the system

Eigenvectors of LTI Systems

Fact: The eigenvectors of a Toeplitz matrix (LTI system) are the complex sinusoids

$$
s_{\omega}[n] = e^{j\omega n} = \cos(\omega n) + j\sin(\omega n), \qquad -\pi \le \omega < \pi, \quad -\infty < n < \infty
$$

Sinusoids are Eigenvectors of LTI Systems

$$
s_{\omega} \longrightarrow \boxed{\mathcal{H}} \longrightarrow \lambda_{\omega} s_{\omega}
$$

Prove that harmonic sinusoids are the eigenvectors of LTI systems simply by computing the convolution with input s_{ω} and applying the periodicity of the sinusoids (infinite-length)

$$
s_{\omega}[n] * h[n] = \sum_{m=-\infty}^{\infty} s_{\omega}[n-m] h[m] = \sum_{m=-\infty}^{\infty} e^{j\omega(n-m)} h[m]
$$

=
$$
\sum_{m=-\infty}^{\infty} e^{j\omega n} e^{-j\omega m} h[m] = \left(\sum_{m=-\infty}^{\infty} h[m] e^{-j\omega m}\right) e^{j\omega n}
$$

=
$$
\lambda_{\omega} s_{\omega}[n] \checkmark
$$

Eigenvalues of LTI Systems

The eigenvalue $\lambda_{\omega} \in \mathbb{C}$ corresponding to the sinusoid eigenvector s_{ω} is called the **frequency response** at frequency ω since it measures how the system "responds" to s_k

$$
\lambda_\omega = \sum_{n=-\infty}^{\infty} h[n] e^{-\omega n} = \langle h, s_\omega \rangle = H(\omega) \text{ (DTFT of } h)
$$

Recall properties of the **inner product**: λ_{ω} grows/shrinks as h and s_{ω} become more/less similar

Eigendecomposition and Diagonalization of an LTI System

$$
x \longrightarrow \begin{array}{c} \mathcal{H} \\ \hline \mathcal{H} \end{array} \longrightarrow y
$$

$$
y[n] = x[n] * h[n] = \sum_{m = -\infty}^{\infty} h[n - m] x[m]
$$

- While we can't explicitly display the infinitely large matrices involved, we can use the DTFT to "diagonalize" an LTI system
- **Taking the DTFTs of x and h**

$$
X(\omega) = \sum_{m=-\infty}^{\infty} x[n] e^{-\omega n}, \quad H(\omega) = \sum_{m=-\infty}^{\infty} h[n] e^{-\omega n}
$$

we have that

$$
Y(\omega) = X(\omega)H(\omega)
$$

and then

$$
y[n] = \int_{-\pi}^{\pi} Y(\omega) e^{j\omega n} \frac{d\omega}{2\pi}
$$

Summary

Complex sinusoids are the eigenfunctions of LTI systems for infinite-length signals (Toeplitz matrices)

Therefore, the discrete time Fourier transform (DTFT) is the natural tool for studying LTI systems for infinite-length signals

Figure Frequency response $H(\omega)$ equals the DTFT of the impulse response $h[n]$

Diagonalization by eigendecomposition implies

$$
Y(\omega) = X(\omega) H(\omega)
$$

Discrete Time Fourier Transform **Examples**

Discrete Time Fourier Transform

$$
X(\omega) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\omega n}, \qquad -\pi \le \omega < \pi
$$

$$
x[n] = \int_{-\pi}^{\pi} X(\omega) e^{j\omega n} \frac{d\omega}{2\pi}, \quad \infty < n < \infty
$$

 \blacksquare The Fourier transform of choice for analyzing infinite-length signals and systems

Useful for conceptual, pencil-and-paper work, but not Matlab friendly (infinitely-long vectors)

DTFT of the Unit Pulse (1)

■ Compute the DTFT of the symmetrical **unit pulse**
$$
p[n] = \begin{cases} 1 & -M \le n \le M \\ 0 & \text{otherwise} \end{cases}
$$

- Note: Duration $D_x = 2M + 1$ samples
- Example for $M = 3$

Forward DTFT

$$
P(\omega) = \sum_{n=-\infty}^{\infty} p[n] e^{-j\omega n} = \sum_{n=-M}^{M} e^{-j\omega n} \quad \dots
$$

DTFT of the Unit Pulse (2)

Apply the finite geometric series formula

$$
P(\omega) = \sum_{n=-\infty}^{\infty} p[n] e^{-j\omega n} = \sum_{n=-M}^{M} e^{-j\omega n} = \sum_{n=-M}^{M} (e^{-j\omega})^n = \frac{e^{j\omega M} - e^{-j\omega(M+1)}}{1 - e^{-j\omega}}
$$

This is an answer but it is not simplified enough to make sense, so we continue simplifying

$$
P(\omega) = \frac{e^{j\omega M} - e^{-j\omega(M+1)}}{1 - e^{-j\omega}} = \frac{e^{-j\omega/2} \left(e^{j\omega \frac{2M+1}{2}} - e^{-j\omega \frac{2M+1}{2}} \right)}{e^{-j\omega/2} \left(e^{j\omega/2} - e^{-j\omega/2} \right)}
$$

=
$$
\frac{2j \sin \left(\omega \frac{2M+1}{2} \right)}{2j \sin \left(\frac{\omega}{2} \right)}
$$

DTFT of the Unit Pulse (3)

Simplified DTFT of the unit pulse of duration $D_x = 2M + 1$ samples

$$
P(\omega) = \frac{\sin\left(\frac{2M+1}{2}\omega\right)}{\sin\left(\frac{\omega}{2}\right)}
$$

- This is called the Dirichlet kernel or "digital sinc"
	- It has a shape reminiscent of the classical $\sin x/x$ sinc function, but it is 2π -periodic
- **If** p[n] is interpreted as the impulse response of the moving average system, then $P(\omega)$ is the frequency response (eigenvalues) (low-pass filter)

DTFT of a One-Sided Exponential

- Recall the impulse response of the recursive average system: $h[n] = \alpha^n u[n]$, $|\alpha| < 1$
- **Compute the frequency response** $H(\omega)$
- **Forward DTFT**

$$
H(\omega) \,\,=\,\, \sum_{n=-\infty}^{\infty} h[n] \, e^{-j\omega n} \,\,=\,\, \sum_{n=0}^{\infty} \alpha^n \, e^{-j\omega n} \,\,=\,\, \sum_{n=0}^{\infty} (\alpha \, e^{-j\omega})^n \,\,=\,\, \frac{1}{1-\alpha \, e^{-j\omega}}
$$

Recursive system with $\alpha = 0.8$ is a low-pass filter

Impulse Response of the Ideal Lowpass Filter (1)

The frequency response $H(\omega)$ of the ideal low-pass filter passes low frequencies (near $\omega = 0$) but blocks high frequencies (near $\omega = \pm \pi$)

$$
H(\omega) = \begin{cases} 1 & -\omega_c \leq \omega \leq \omega_c \\ 0 & \text{otherwise} \end{cases}
$$

- **Compute the impulse response** $h[n]$ given this $H(\omega)$
- **Apply the inverse DTFT**

$$
h[n] = \int_{-\pi}^{\pi} H(\omega) e^{j\omega n} \frac{d\omega}{2\pi} = \int_{-\omega_c}^{\omega_c} e^{j\omega n} \frac{d\omega}{2\pi} = \frac{e^{j\omega n}}{jn} \Big|_{-\omega_c}^{\omega_c} = \frac{e^{j\omega_c n} - e^{-j\omega_c n}}{jn} = 2\omega_c \frac{\sin(\omega_c n)}{\omega_c n}
$$

Impulse Response of the Ideal Lowpass Filter (2)

The frequency response $H(\omega)$ of the ideal low-pass filter passes low frequencies (near $\omega = 0$) but blocks high frequencies (near $\omega = \pm \pi$)

The infamous "sinc" function!

DTFT of a rectangular pulse is a Dirichlet kernel

■ DTFT of a one-sided exponential is a low-frequency bump

Inverse DTFT of the ideal lowpass filter is a sinc function

Work some examples on your own!

Discrete Time Fourier Transform of a Sinusoid

Discrete Fourier Transform (DFT) of a Harmonic Sinusoid

Thanks to the orthogonality of the length-N harmonic sinusoids, it is easy to calculate the DFT Thanks to the orthogonality of the length- N harmonic sinusoid $x[n] = s_l[n] = e^{j\frac{2\pi}{N}ln}/\sqrt{N}$ N

$$
X[k] = \sum_{n=0}^{N-1} s_l[n] \frac{e^{-j\frac{2\pi}{N}kn}}{\sqrt{N}} = \langle s_l, s_k \rangle = \delta[k-l]
$$

$$
S_4[n]
$$

$$
S_4[k]
$$

So what is the DTFT of the infinite length sinusoid $e^{j\omega_0 n}$?

DTFT of an Infinite-Length Sinusoid

The calculation for the DTFT and infinite-length signals is much more delicate than for the DFT and finite-length signals

Calculate the value $X(\omega_0)$ for the signal $x[n]=e^{j\omega_0n}$

$$
X(\omega_0) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\omega_0 n} = \sum_{n=-\infty}^{\infty} e^{j\omega_0 n} e^{-j\omega_0 n} = \sum_{n=-\infty}^{\infty} 1 = \infty
$$

Calculate the value $X(\omega)$ for the signal $x[n] = e^{j\omega_0 n}$ at a frequency $\omega \neq \omega_0$

$$
X(\omega_0) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\omega n} = \sum_{n=-\infty}^{\infty} e^{j\omega_0 n} e^{-j\omega n} = \sum_{n=-\infty}^{\infty} e^{-j(\omega - \omega_0)n} = ??"
$$

Dirac Delta Function (1)

- One semi-rigorous way to deal with this quandary is to use the **Dirac delta "function,"** which is defined in terms of the following limit process
- **Consider the following function** $d_{\epsilon}(\omega)$ of the continuous variable ω

Note that, for all values of the width ϵ , $d_{\epsilon}(\omega)$ always has unit area

$$
\int d_\epsilon(\omega)\,d\omega\ =\ 1
$$

Dirac Delta Function (2)

- What happens to $d_{\epsilon}(\omega)$ as we let $\epsilon \to 0$?
	- Clearly $d_{\epsilon}(\omega)$ is converging toward something that is infinitely tall and infinitely narrow but still with unit area
- **The safest way to handle a function like** $d_{\epsilon}(\omega)$ **is inside an integral, like so**

$$
\int X(\omega)\,d_\epsilon(\omega)\,d\omega
$$

Dirac Delta Function (3)

 \cdot

As $\epsilon \to 0$, it seems reasonable that

$$
\int X(\omega) d_{\epsilon}(\omega) d\omega \stackrel{\epsilon \to 0}{\longrightarrow} X(0)
$$

and

$$
\int X(\omega)\,d_\epsilon(\omega-\omega_0)\,d\omega\;\xrightarrow{\epsilon\to 0}\;X(\omega_0)
$$

- So we can think of $d_{\epsilon}(\omega)$ as a kind of "sampler" that picks out values of functions from inside an integral
- We describe the results of this limiting process (as $\epsilon \to 0$) as the **Dirac delta "function"** $\delta(\omega)$

Dirac Delta Function (4)

■ We write

$$
\quad \text{and} \quad
$$

$$
\int X(\omega)\,\delta(\omega-\omega_0)\,d\omega = X(\omega_0)
$$

Remarks and caveats

- Do not confuse the Dirac delta "function" with the nicely behaved discrete delta function $\delta[n]$
- The Dirac has lots of "delta," but it is not really a "function" in the normal sense (it can be made more rigorous using the theory of generalized functions)
- Colloquially, engineers will describe the Dirac delta as "infinitely tall and infinitely narrow"

Scaled Dirac Delta Function

If we scale the area of $d_{\epsilon}(\omega)$ by L, then it has the following effect in the limit

$$
\int X(\omega) \, L \, \delta(\omega) \, d\omega \ = \ L \, X(0)
$$

And Now Back to Our Regularly Scheduled Program . . .

■ Back to determining the DTFT of an infinite length sinusoid

Rather than computing the DTFT of a sinusoid using the forward DTFT, we will show that an infinite-length sinusoid is the inverse DTFT of the scaled Dirac delta function $2\pi\delta(\omega-\omega_0)$

$$
\int_{-\pi}^{\pi} 2\pi \delta(\omega - \omega_0) e^{j\omega n} \frac{d\omega}{2\pi} = e^{j\omega_0 n}
$$

■ Thus we have the (rather bizarre) DTFT pair

$$
e^{j\omega_0 n}\ \stackrel{\rm DTFT}{\longleftrightarrow}\ 2\pi\,\delta(\omega-\omega_0)
$$

DTFT of Real-Valued Sinusoids

Since

$$
\cos(\omega_0 n) = \frac{1}{2} \left(e^{j\omega_0 n} + e^{-j\omega_0 n} \right)
$$

we can calculate its DTFT as

$$
\cos(\omega_0 n) \stackrel{\text{DTFT}}{\longleftrightarrow} \pi \delta(\omega - \omega_0) + \pi \delta(\omega + \omega_0)
$$

Since

$$
\sin(\omega_0 n) = \frac{1}{2j} \left(e^{j\omega_0 n} - e^{-j\omega_0 n} \right)
$$

we can calculate its DTFT as

$$
\sin(\omega_0 n) \stackrel{\text{DTFT}}{\longleftrightarrow} \frac{\pi}{j} \delta(\omega - \omega_0) + \frac{\pi}{j} \delta(\omega + \omega_0)
$$

The DTFT would be of limited utility if we could not compute the transform of an infinite-length sinusoid

Hence, the Dirac delta "function" (or something else) is a necessary evil

The Dirac delta has infinite energy (2-norm); but then again so does an infinite-length sinusoid

Discrete Time Fourier Transform Properties

Recall: Discrete-Time Fourier Transform (DTFT)

Forward DTFT (Analysis)

$$
X(\omega) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\omega n}, \qquad -\pi \le \omega < \pi
$$

Inverse DTFT (Synthesis)

$$
x[n] = \int_{-\pi}^{\pi} X(\omega) e^{j\omega n} \frac{d\omega}{2\pi}, \qquad \infty < n < \infty
$$

DTFT pair

$$
x[n] \overset{\text{DTFT}}{\longleftrightarrow} X(\omega)
$$

The DTFT is Periodic

We defined the DTFT over an interval of ω of length 2π , but it can also be interpreted as **periodic** with period 2π

$$
X(\omega) = X(\omega + 2\pi k), \quad k \in \mathbb{Z}
$$

Proof

DTFT Frequencies

$$
X(\omega) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\omega n}, \qquad -\pi \le \omega < \pi
$$

- $X(\omega)$ measures the similarity between the time signal x and and a sinusoid $e^{j\omega n}$ of frequency ω
- **Therefore,** $X(\omega)$ measures the "frequency content" of x at frequency ω

DFT Frequencies and Periodicity

Periodicity of DFT means we can treat frequencies mod 2π

 $X(\omega)$ measures the "frequency content" of x at frequency $(\omega)_{2\pi}$

DTFT Frequency Ranges

- **Periodicity of DTFT means every length-2**π interval of ω carries the same information
- Typical interval 1: $0 \leq \omega < 2\pi$

■ Typical interval 2: $-\pi \leq \omega < \pi$ (more intuitive)

DTFT and Time Shift

If $x[n]$ and $X(\omega)$ are a DTFT pair then

$$
x[n-m] \stackrel{\text{DFT}}{\longleftrightarrow} e^{-j\omega m} X(\omega)
$$

Proof: Use the change of variables $r = n - m$

$$
\sum_{n=-\infty}^{\infty} x[n-m] e^{-j\omega n} = \sum_{r=-\infty}^{\infty} x[r] e^{-j\omega(r+m)} = \sum_{r=-\infty}^{\infty} x[r] e^{-j\omega r} e^{-j\omega m}
$$

$$
= e^{-j\omega m} \sum_{r=-\infty}^{\infty} x[r] e^{-j\omega r} = e^{-j\omega m} X(\omega) \checkmark
$$

DTFT and Modulation

If $x[n]$ and $X(\omega)$ are a DFT pair then

$$
e^{j\omega_0 n} x[n] \stackrel{\text{DFT}}{\longleftrightarrow} X(\omega - \omega_0)
$$

Remember that the DTFT is 2π -periodic, and so we can interpret the right hand side as $X((\omega - \omega_0)_{2\pi})$

Proof:

$$
\sum_{n=-\infty}^{\infty} e^{j\omega_0 n} x[n] e^{-j\omega n} = \sum_{n=-\infty}^{\infty} x[n] e^{-j(\omega - \omega_0)n} = X(\omega - \omega_0) \quad \checkmark
$$

DTFT and Convolution

$$
x \longrightarrow \boxed{h} \longrightarrow y
$$

$$
y[n] = x[n] * h[n] = \sum_{m = -\infty}^{\infty} h[n - m] x[m]
$$

 \blacksquare If

then

$$
x[n] \stackrel{\text{DTFT}}{\longleftrightarrow} X(\omega), \qquad h[n] \stackrel{\text{DTFT}}{\longleftrightarrow} H(\omega), \qquad y[n] \stackrel{\text{DTFT}}{\longleftrightarrow} Y(\omega)
$$

$$
Y(\omega) = H(\omega) X(\omega)
$$

Convolution in the time domain $=$ **multiplication in the frequency domain**

The DTFT is Linear

 \blacksquare It is trivial to show that if

$$
x_1[n] \overset{\text{DTFT}}{\longleftrightarrow} X_1(\omega) \qquad x_2[n] \overset{\text{DTFT}}{\longleftrightarrow} X_2(\omega)
$$

then

$$
\alpha_1 x_1[n] + \alpha_2 x[2] \stackrel{\text{DFT}}{\longleftrightarrow} \alpha_1 X_1(\omega) + \alpha_2 X_2(\omega)
$$

DTFT Symmetry Properties

The sinusoids $e^{j\omega n}$ of the DTFT have symmetry properties:

$$
Re (e^{j\omega n}) = cos (\omega n) \text{ (even function)}
$$

$$
Im (e^{j\omega n}) = sin (\omega n) \text{ (odd function)}
$$

These induce corresponding symmetry properties on $X(\omega)$ around the frequency $\omega = 0$

Even signal/DFT

$$
x[n] = x[-n], \qquad X(\omega) = X(-\omega)
$$

Odd signal/DFT

$$
x[n] = -x[-n], \qquad X(\omega) = -X(-\omega)
$$

Proofs of the symmetry properties are identical to the DFT case; omitted here

DFT Symmetry Properties Table

DTFT is periodic with period 2π

Convolution in time becomes multiplication in frequency

DTFT has useful symmetry properties

Acknowledgements

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