## Fourier series representation of periodic signals – Chapter 3<sup>2</sup>

We have seen that a signal can generally be represented as a linear combination of shifted impulse (or sample) functions.

We will show that a signal can also be represented as a linear combination of complex exponential functions, provided certain conditions are satisfied.

## Why is this useful? Section 3.2

Reason is due to this important fact: The output of an LTI system due to a complex exponential input is the same complex exponential multiplied by a (possibly complex) gain factor.

We say that complex exponentials are

<sup>&</sup>lt;sup>2</sup>These slides are based on lecture notes of Professors L. Lampe, C. Leung, and R. Schober

eigenfunctions of LTI systems; the gain factors are termed eigenvalues. (cf. eigenvectors and eigenvalues of a matrix)

We now prove the following.

Continuous-time: If  $x(t)=e^{st}$  is input to a LTI system with impulse response h(t), the output y(t) is

$$y(t) = H(s)e^{st}$$

where

$$H(s) = \int_{-\infty}^{\infty} h(\tau) e^{-s\tau} d\tau$$

Proof:

### Example:

Suppose 
$$y(t) = x(t-1)$$
 and  $x(t) = e^{j2\pi t}$ .

## Example:

Suppose y(t) = x(t-1) and  $x(t) = \cos 2\pi t + \cos 3\pi t$ .

Discrete-time: If  $x[n]=z^n$  is input to a LTI system with impulse response h[n], the output y[n] is

$$y[n] = H(z)z^n$$

where

$$H(z) = \sum_{k=-\infty}^{\infty} h[k] \ z^{-k}$$

**Proof**:

# Fourier series representation of CT periodic signals – Section 3.3

Let x(t) be a periodic signal with fundamental period T. Then if certain conditions (Dirichlet, pp. 197-200) are satisfied, we can represent x(t) as a Fourier series (FS) (an infinite sum of complex exponentials):

$$x(t) = \sum_{k=-\infty}^{\infty} a_k \ e^{jk\left(\frac{2\pi}{T}\right)t}$$

where the (possibly complex) Fourier coefficients  $\{a_k\}$  are given by

$$a_k = \frac{1}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} x(t) e^{-jk(\frac{2\pi}{T})t} dt,$$

$$k = 0, \pm 1, \pm 2, \dots$$

The above equations are referred to as the *synthesis* and *analysis* equations respectively.

Section 3.4 provides a good discussion of the *convergence* of the FS representation.

#### Notes:

- 1. Using Euler's relationship, we can also express the FS representation as an infinite sum of sine and cosine terms.
- 2. Given x(t), we can determine  $\{a_k\}_{k=-\infty}^{\infty}$ ; conversely, we can reconstruct x(t) from  $\{a_k\}_{k=-\infty}^{\infty}$ .

 $\{a_k\}_{k=-\infty}^{\infty}$  give the *frequency-domain* description of the signal and are called its *spectral coefficients*.

3. A periodic signal x(t) of fundamental period T has components at frequencies  $0, \pm \frac{2\pi}{T}, \pm \frac{4\pi}{T}, \ldots$ , i.e. at multiples of the fundamental frequency  $\omega_0 = \frac{2\pi}{T}$  or  $f_0 = \frac{1}{T}$ .

The component at freq  $nf_0$  is called the nth harmonic.

4. Conjugate symmetry property: If x(t) is real, then  $a_{-k} = a_k^*$ .

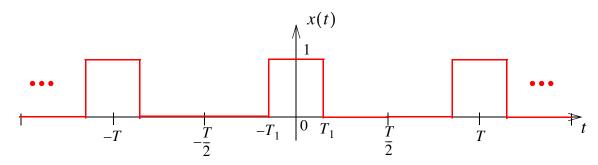
**Proof**:

As a result,

$$|a_{-k}| = |a_k^*| = |a_k|$$
  
and  $\angle a_{-k} = \angle a_k^* = -\angle a_k$ .

Thus, when x(t) is real, the amplitudes of the spectral coefficients have even symmetry whereas their phases have odd symmetry.

Example: Periodic square wave shown below.



Periodic square wave

We would like to determine its Fourier or spectral coefficients  $\{a_k\}_{k=-\infty}^{\infty}$ .

#### Recall that

$$a_k = \frac{1}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} x(t) e^{-jk(\frac{2\pi}{T})t} dt, k = 0, \pm 1, \pm 2, \dots$$

In this case, we have

$$a_{k} = \frac{1}{T} \int_{-T_{1}}^{T_{1}} e^{-jk(\frac{2\pi}{T})t} dt$$

$$= -\frac{1}{jk2\pi} e^{-jk(\frac{2\pi}{T})t} \Big]_{-T_{1}}^{T_{1}}$$

$$= \frac{2}{k2\pi} \left[ \frac{e^{jk(\frac{2\pi}{T})T_{1}} - e^{-jk(\frac{2\pi}{T})T_{1}}}{2j} \right]_{\sin k(\frac{2\pi}{T})T_{1}}^{T_{1}}$$

$$= \frac{\sin k \left(\frac{2\pi}{T}\right) T_1}{k\pi}$$

$$= \frac{\sin k \left(\frac{2\pi}{T}\right) T_1}{k\pi \frac{2T_1}{T}} \times \frac{2T_1}{T}$$

$$= \frac{2T_1}{T} \operatorname{sinc}\left(\frac{2kT_1}{T}\right)$$

As expected, the amplitudes of the FS coefficients have even symmetry; in this case, the phases are 0.

e.g. 
$$T_1 = T/4$$

Then, the average value is

$$a_0 = \frac{1}{2}$$

and

$$a_k = \frac{1}{2}\operatorname{sinc}\left(\frac{k}{2}\right)$$

$$= \frac{1}{2}\frac{\sin \pi k/2}{\pi k/2}$$

$$= \frac{\sin \pi k/2}{\pi k}.$$

k	1	2	3	4	5	6	7
$a_k$	$\frac{1}{\pi}$	0	$-\frac{1}{3\pi}$	0	$\frac{1}{5\pi}$	0	$-\frac{1}{7\pi}$

## Properties of CT Fourier series – Section 3.5

See Table 3.1, p. 206 for a list of CTFS properties. We will look at a few of the more commonly used ones here. Most of these properties can also be obtained from our future study of CT Fourier transform.

Notation: The pairing of a periodic signal x(t) and its FS coefficients  $a_k$  is represented by

$$x(t) \stackrel{\mathcal{FS}}{\longleftrightarrow} a_k$$

#### 1. Linearity – Section 3.5.1

Let x(t) and y(t) be two periodic signal with period T and  $x(t) \stackrel{\mathcal{FS}}{\longleftrightarrow} a_k$ ,  $y(t) \stackrel{\mathcal{FS}}{\longleftrightarrow} b_k$ .

Then,

$$z(t) = Ax(t) + By(t) \stackrel{\mathcal{FS}}{\longleftrightarrow} c_k = Aa_k + Bb_k$$
.

#### **Proof**:

#### 2. Time shift – Section 3.5.2

If 
$$x(t) \stackrel{\mathcal{FS}}{\longleftrightarrow} a_k$$
, then  $x(t-t_0) \stackrel{\mathcal{FS}}{\longleftrightarrow} e^{-jk\left(\frac{2\pi}{T}\right)t_0} a_k$ .

As a result, when a periodic signal is shifted in time, the magnitudes of its FS coefficients are unchanged.

#### **Proof**:

3. Time reversal – Section 3.5.3 If  $x(t) \stackrel{\mathcal{FS}}{\longleftrightarrow} a_k$ , then  $x(-t) \stackrel{\mathcal{FS}}{\longleftrightarrow} a_{-k}$ .

As a result, if x(t) is even, i.e. x(-t) = x(t), then  $a_{-k} = a_k$ , i.e. the FS coefficients are also even.

Moreover, if x(t) is odd, i.e. x(-t) = -x(t), then  $a_{-k} = -a_k$ , i.e. the FS coefficients are also odd.

**Proof**:

4. Time Scaling – Section 3.5.4

If  $x(t) = \sum_{k=-\infty}^{\infty} a_k \ e^{jk\left(\frac{2\pi}{T}\right)t}$ ,

then  $x(\alpha t) = \sum_{k=-\infty}^{\infty} a_k \ e^{jk\left(\alpha\frac{2\pi}{T}\right)t}$ .

Note that each of the "tones" in x(t) are simply compressed in time by a factor of  $\alpha$ . *Proof:* 

## 5. Multiplication – Section 3.5.5

If x(t) and y(t) are periodic with a common period T and FS coefficients  $\{a_k\}$  and  $\{b_k\}$  respectively, then

$$x(t)y(t) \stackrel{\mathcal{FS}}{\longleftrightarrow} h_k = \sum_{l=-\infty}^{\infty} a_l b_{k-l}.$$

Note that the RHS is the DT convolution of the sequences representing the FS coefficients of x(t) and y(t).

#### **Proof:**

6. Conjugation – Section 3.5.6 If  $x(t) \stackrel{\mathcal{FS}}{\longleftrightarrow} a_k$ , then  $x^*(t) \stackrel{\mathcal{FS}}{\longleftrightarrow} a_{-k}^*$ .

The conjugate symmetry property of the CTFS we saw previously follows easily from the above conjugation property.

**Proof**:

### 7. Parseval's relation – Section 3.5.7

If 
$$x(t) \stackrel{\mathcal{FS}}{\longleftrightarrow} a_k$$
, then 
$$\frac{1}{T} \int_T |x(t)|^2 dt = \sum_{k=-\infty}^{\infty} |a_k|^2.$$

The LHS is the average power of x(t). Also since

$$\frac{1}{T} \int_{T} \left| a_k \ e^{jk\left(\frac{2\pi}{T}\right)t} \right|^2 \ dt = \frac{1}{T} \int_{T} \left| a_k \right|^2 \ dt = \left| a_k \right|^2,$$

we see that  $|a_k|^2$  is the average power in the kth harmonic component of x(t).

Parseval's relation states that the average power in x(t) is equal to the sum of the average powers in all its harmonics.

Proof: Problem 3.46.

# Fourier series representation of DT periodic signals – Section 3.6

Let x[n] be a periodic signal with fundamental period N. Then, as it is shown in the

textbook, equations (3.94) and (3.95), we have the following DT Fourier series (DTFS) pair

$$x[n] = \sum_{k=\langle N \rangle} a_k \ e^{jk\left(\frac{2\pi}{N}\right)n}$$

where the (possibly complex) Fourier coefficients  $\{a_k\}$ , also known as spectral coefficients, are given by

$$a_k = \frac{1}{N} \sum_{n=\langle N \rangle} x[n] e^{-jk\left(\frac{2\pi}{N}\right)n}$$

The notation  $\sum_{k=< N>}$  is used to indicate that the summation is over N consecutive integers, starting with any value of k.

The above two equations are referred to as the *synthesis* and *analysis* equations respectively.

There are similarities to as well as differences from the CT case.

#### Remarks:

- 1. The synthesis equation in the *DT case* involves a finite sum, in sharp contrast to the *CT case which involves an infinite* sum. Hence, unlike the CT case, there are no convergence issues in the DT case.
- 2. Unlike the CT case,  $a_k = a_{k+N}$ .
- 3. An important set of DT complex exponentials is defined as

$$\phi_k[n] = e^{jk(\frac{2\pi}{N})n}, k = 0, \pm 1, \pm 2, \dots$$

Note: 
$$\phi_k[n] = \phi_{k+iN}[n]$$
.

#### 4. Fact:

$$\sum_{n=< N>} \phi_k[n] = \sum_{n=< N>} e^{jk\left(\frac{2\pi}{N}\right)n}$$

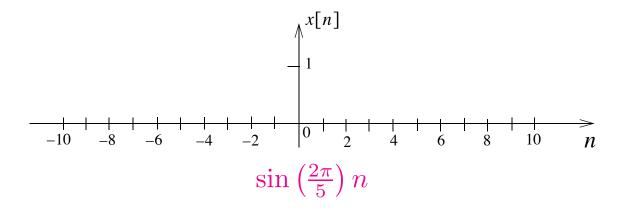
$$= \begin{cases} N, & k = 0, \pm N, \pm 2N, \dots \\ 0, & \text{otherwise.} \end{cases}$$

#### **Proof:**

#### Example:

Suppose  $x[n] = \sin\left(\frac{2\pi}{5}\right)n$ , a signal with fundamental period 5.

n	0	1	2	3	4	5	6
x[n]	0	0.95	0.59	-0.59	-0.95	0	0.95

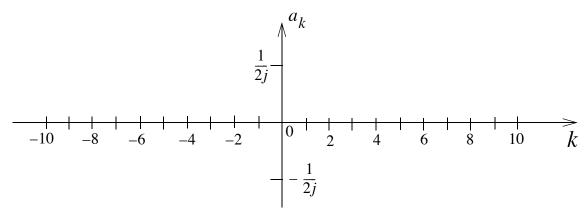


Then, we can write

$$x[n] = \frac{1}{2j} \left[ e^{j\left(\frac{2\pi}{5}\right)n} - e^{-j\left(\frac{2\pi}{5}\right)n} \right].$$

By direct comparison with the *synthesis* equation for the *DTFS*, we obtain

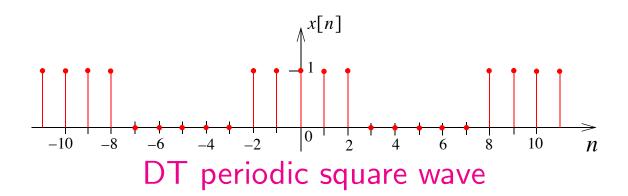
$$a_1 = \frac{1}{2j}, \quad a_{-1} = -\frac{1}{2j}.$$



Spectral coefficients of  $\sin\left(\frac{2\pi}{5}\right)n$ 

Example: Consider the following signal with fundamental period N=10: in the interval  $-5 \le n \le 5$ ,

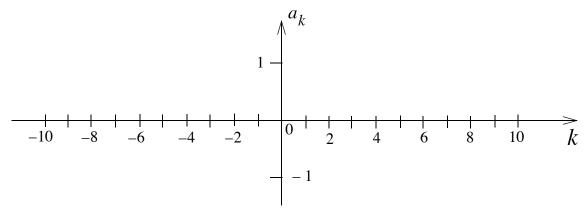
$$x[n] = \begin{cases} 1, & -2 \le n \le 2 \\ 0, & \text{otherwise.} \end{cases}$$



Then,

$$a_k = \frac{1}{10} \sum_{n=-2}^{2} e^{-jk(\frac{2\pi}{10})n}$$

n	-5	-4	-3	-2	-1	0	1	2	3	4
$a_n$	0.1	0.0	-0.12	0.0	0.32	0.5	0.32	0.0	-0.12	0.0



FS coefficients of DT periodic square wave

More generally, suppose that x[n] is periodic with fundamental period N and in the interval  $-\left\lfloor\frac{N}{2}\right\rfloor \leq n \leq \left\lfloor\frac{N-1}{2}\right\rfloor$ ,

$$x[n] = \begin{cases} 1, & -N_1 \le n \le N_1 \\ 0, & \text{otherwise.} \end{cases}$$

Then, from Example 3.12 on p. 218, we have

$$a_k = \begin{cases} \frac{\sin\left[2\pi k\left(N_1+\frac{1}{2}\right)/N\right]}{N\sin(\pi k/N)}, & k \text{ not a multiple of } N \\ \frac{2N_1+1}{N}, & k \text{ a multiple of } N. \end{cases}$$

## Properties of DT Fourier series – Section 3.7

Although there are *important differences* such as the *periodicity of the FS coefficients* and the *Gibbs phenomenon* between the CT and DT Fourier series, there are also *many similarities*.

See Table 3.2 on p. 221 for a list of DTFS properties. Most of them can be derived in similar fashion to the CTFS cases.

We next look at a couple of properties to illustrate the similarities and differences.

Notation: The pairing of a periodic signal x[n] and its FS coefficients  $a_k$  is represented by

$$x[n] \stackrel{\mathcal{FS}}{\longleftrightarrow} a_k$$

### 1. Multiplication – Section 3.7.1

If x[n] and y[n] are periodic with a common period N and FS coefficients  $\{a_k\}$  and  $\{b_k\}$  respectively, then

$$x[n]y[n] \stackrel{\mathcal{FS}}{\longleftrightarrow} d_k = \sum_{l=\langle N \rangle} a_l b_{k-l}.$$

Note that the RHS is called the DT *periodic* convolution of the sequences representing the FS coefficients of x[n] and y[n].

# 2. Parseval's relation – Section 3.7.3 If $x[n] \stackrel{\mathcal{FS}}{\longleftrightarrow} a_k$ , then

$$\frac{1}{N} \sum_{n=< N>} |x[n]|^2 = \sum_{k=< N>} |a_k|^2.$$

The LHS is the average power of x[n]. Also

since

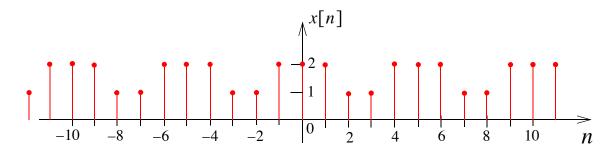
$$\frac{1}{N} \sum_{n=\langle N \rangle} \left| a_k e^{jk\left(\frac{2\pi}{N}\right)n} \right|^2 = \frac{1}{N} \sum_{n=\langle N \rangle} \left| a_k \right|^2$$
$$= \left| a_k \right|^2,$$

we see that  $|a_k|^2$  is the average power in the kth harmonic component of x[n].

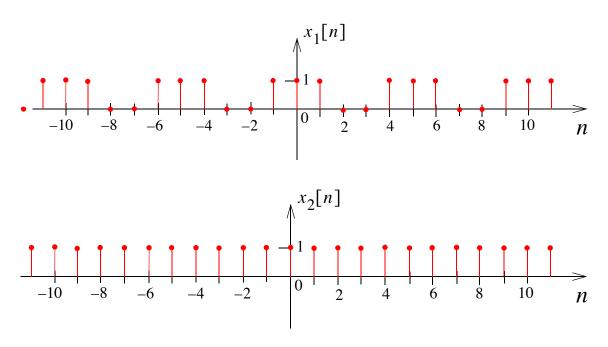
Parseval's relation states that the average power in x[n] is equal to the sum of the average powers in all its harmonics.

### Example:

Determine the DTFS coefficients of the signal x[n], with a fundamental period of 5, as shown in the following figure.



To solve this problem, we can make use of the linearity property of the DTFS by noting that  $x[n] = x_1[n] + x_2[n]$ , where  $x_1[n]$  and  $x_2[n]$  are shown in the following figures.



If

$$x[n] \stackrel{\mathcal{FS}}{\longleftrightarrow} a_k, x_1[n] \stackrel{\mathcal{FS}}{\longleftrightarrow} b_k, x_2[n] \stackrel{\mathcal{FS}}{\longleftrightarrow} c_k$$

then

$$a_k = b_k + c_k .$$

Based on the Example on p. 79 of the notes, we have

$$b_k = \begin{cases} \frac{\frac{3}{5}}{5}, & k \text{ a multiple of 5} \\ \frac{1}{5} \frac{\sin(3\pi k/5)}{\sin(\pi k/5)}, & \text{otherwise.} \end{cases}$$

We also note that

$$c_k \stackrel{\triangle}{=} \frac{1}{N} \sum_{n=< N>} x_2[n] e^{-jk(\frac{2\pi}{N})n}$$

$$= \frac{1}{5} \sum_{n=< 5>} e^{-jk(\frac{2\pi}{5})n}$$

$$= \begin{cases} 1, & k \text{ a multiple of 5} \\ 0, & \text{otherwise.} \end{cases}$$

## We thus obtain

$$a_k = \begin{cases} \frac{8}{5}, & k \text{ a multiple of 5} \\ \frac{1}{5} \frac{\sin(3\pi k/5)}{\sin(\pi k/5)}, & \text{otherwise.} \end{cases}$$

## Frequency response of LTI systems – Section 3.8

Continuous-time: Recall that if  $x(t) = e^{st}$  is input to a LTI system with impulse response h(t), the output y(t) is

$$y(t) = H(s)e^{st}$$

where

$$H(s) = \int_{-\infty}^{\infty} h(t) e^{-st} dt.$$

We will be mostly interested in the special case when  $s=j\omega$ . The input is then the complex exponential  $e^{j\omega t}$  and

$$H(j\omega) = \int_{-\infty}^{\infty} h(t) e^{-j\omega t} dt$$

is called the *frequency response* of the system.

Let x(t) be a periodic signal with a Fourier series (FS) representation

$$x(t) = \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t}, \ \omega_0 = \frac{2\pi}{T}.$$

Then it follows that the output is given by

$$y(t) = \sum_{k=-\infty}^{\infty} a_k H(jk\omega_0) e^{jk\omega_0 t}.$$

Similar results hold for DT systems.

Discrete-time: Recall that if  $x[n] = z^n$  is input to a LTI system with impulse response h[n], the output y[n] is

$$y[n] = H(z)z^n$$

where

$$H(z) = \sum_{n=-\infty}^{\infty} h[n] \ z^{-n} \ .$$

We will be mostly interested in the special case when  $z=e^{j\omega}$ . The input is then the complex exponential  $e^{j\omega n}$  and

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

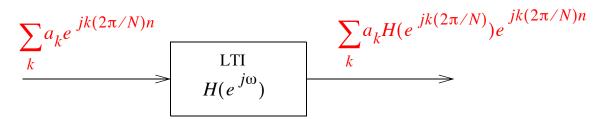
is called the *frequency response* of the system.

Let x[n] be a periodic signal with a Fourier series representation

$$x[n] = \sum_{k=\langle N \rangle} a_k \ e^{jk\left(\frac{2\pi}{N}\right)n}.$$

Then it follows that the output is given by

$$y[n] = \sum_{k=\langle N \rangle} a_k H\left(e^{jk\left(\frac{2\pi}{N}\right)}\right) e^{jk\left(\frac{2\pi}{N}\right)n}.$$



Block diagram illustration of I/O relationship of LTI system

#### Example:

Recall that  $x[n] = \cos \omega_0 n$  is periodic only if  $\frac{2\pi}{\omega_0}$  is rational.

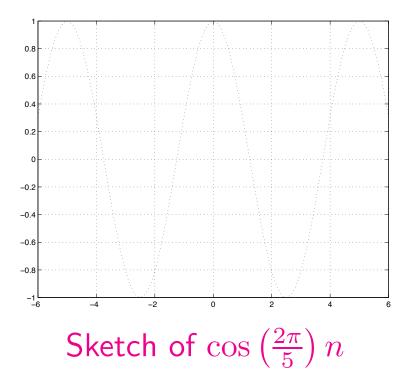
Also,  $\cos\left(\frac{2\pi}{N}\right)n$  is periodic with fundamental period N.

#### Example:

How do  $\cos\left(\frac{2\pi}{N}\right)n$  and  $\cos2\left(\frac{2\pi}{N}\right)n$  look like for N=5?

n	0	1	2	3	4
$\cos \frac{2\pi n}{5}$	1	0.31	-0.81	-0.81	0.31

n	0	1	2	3	4	
$\cos 2\left(\frac{2\pi n}{5}\right)$	1	-0.81	0.31	0.31	-0.81	

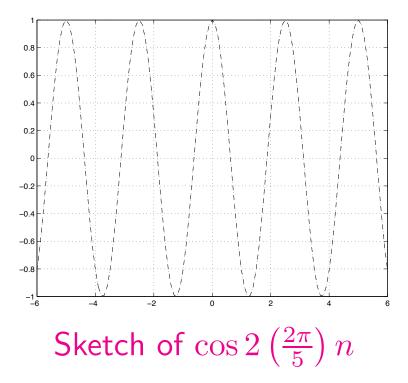


#### Note:

$$\cos\left(\frac{2\pi}{5}\right)n = \frac{1}{2} e^{j\left(\frac{2\pi}{5}\right)n} + \frac{1}{2} e^{-j\left(\frac{2\pi}{5}\right)n}.$$

A direct comparison with the DTFS synthesis equation shows that

$$a_1 = \frac{1}{2}, \ a_{-1} (= a_{5-1}) = \frac{1}{2}.$$



#### Note:

$$\cos 2\left(\frac{2\pi}{5}\right)n = \frac{1}{2}e^{j2\left(\frac{2\pi}{5}\right)n} + \frac{1}{2}e^{-j2\left(\frac{2\pi}{5}\right)n}.$$

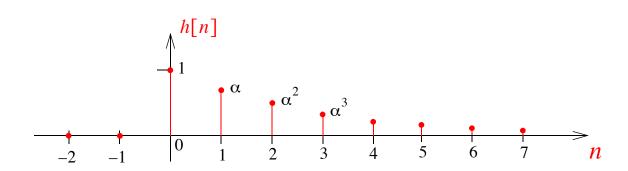
A direct comparison with the DTFS synthesis equation shows that

$$a_2 = \frac{1}{2}, \ a_{-2} (= a_{5-2}) = \frac{1}{2}.$$

Consider an LTI system with impulse response

#### shown below

$$h[n] = \alpha^n \ u[n], \ |\alpha| < 1.$$



Then,

$$H(z) \stackrel{\triangle}{=} \sum_{n=-\infty}^{\infty} h[n] z^{-n}$$

$$= \sum_{n=0}^{\infty} \alpha^n z^{-n}$$

$$= \sum_{n=0}^{\infty} (\alpha z^{-1})^n$$

$$= \frac{1}{1-\alpha z^{-1}}$$

and

$$H(e^{j\omega}) = \frac{1}{1-\alpha e^{-j\omega}}$$
.

Suppose that  $x[n] = \cos\left(\frac{2\pi}{N}\right)n$  is input into such an LTI system.

Question: How can we determine the output, y[n]?

Since

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

we can easily write

$$\begin{split} y[n] &= a_1 H\left(e^{j\frac{2\pi}{N}}\right) e^{j\frac{2\pi}{N}n} + a_{-1} H\left(e^{-j\frac{2\pi}{N}}\right) e^{-j\frac{2\pi}{N}n} \\ &= \frac{1}{2} \left[ \left(\frac{1}{1-\alpha e^{-j\frac{2\pi}{N}}}\right) e^{j\frac{2\pi}{N}n} + \left(\frac{1}{1-\alpha e^{j\frac{2\pi}{N}}}\right) e^{-j\frac{2\pi}{N}n} \right]. \end{split}$$

### Letting

$$re^{j\theta} = \frac{1}{1 - \alpha e^{-j\frac{2\pi}{N}}} ,$$

we have

$$y[n] = r \cos\left(\frac{2\pi}{N}n + \theta\right).$$

Question: How can we determine r and  $\theta$ ?

$$re^{j\theta} = \frac{1}{1 - \alpha \cos \frac{2\pi}{N} + j\alpha \sin \frac{2\pi}{N}}$$

so that

$$r = \frac{1}{\sqrt{(1 - 2\alpha \cos \frac{2\pi}{N}) + \alpha^2}}$$

$$\tan \theta = \frac{-\alpha \sin \frac{2\pi}{N}}{1 - \alpha \cos \frac{2\pi}{N}}.$$