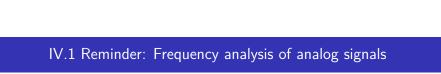
Chapter IV: Frequency Analysis of Discrete Signals



### Introduction

- Very useful to analyze signals in frequency domain
- ► The **spectrum** of a signal indicates the frequency contents
- Mathematical tools:
  - periodicac signals: Fourier series
  - non-periodical signals: Fourier transform

# Analog periodical signals

Periodical signal:

$$x(t) = x(t+T)$$

▶ The fundamental frequency is

$$F_0 = \frac{1}{T}$$

► The signal can be decomposed as a sum of complex exponential signals, with multiples of the fundamental frequency,  $kF_0$ 

$$x(t) = \sum_{k=-\infty}^{\infty} c_k e^{j2\pi k F_0 t}$$

ightharpoonup The coefficients  $c_k$  are the **spectrum** of the signal

$$c_k = \frac{1}{T} \int_{T/2}^{T/2} x(t) e^{-j2\pi k F_0 t}$$

- $\triangleright$  The coefficients  $c_k$  are complex values
  - their modulus = "amplitude spectrum"
  - ► their phase = "phase spectrum""

### Conditions for convergence

- When is the Fourier series convergent to the signal?
  - ▶ i.e. when is the relation correct,
  - i.e. when is the sum actually equal to x(t)?
- Dirichlet conditions: the sum is convergent in all continuity points if:
  - 1. x(t) is continuous or has a finite number of discontinuities in any finite interval
  - 2. x(t) has a finite number of maxima and minima in any period
  - 3. x(t) is absolutely integrable in any period, i.e.:

$$\int_{T}|x(t)|dt<\infty$$

- Weaker condition:
  - ightharpoonup if x(t) is square summable

$$\int_T x(t)^2 dt < \infty$$

- then the he difference  $d(t) = x(t) \sum_{k=-\infty}^{\infty} c_k e^{j2\pi k F_0 t}$  has zero energy
- Does not guarantee pointwise convergence

## Signal spectrum

- ▶ The coefficients  $c_k$  are complex numbers
- ▶ If the signal is **real**  $x(t) \in \mathbb{R}$ , then the  $c_k$  are **even** 
  - $|c_k| = |c_{-k}|$
  - $ightharpoonup \angle c_k = -\angle c_{-k}$
  - ▶ group the terms with  $c_k$  with  $c_{-k}$  -> cosine with amplitude  $|\mathbf{c}_{-k}|$  and phase  $\angle c_k$
- Average power of signal = energy of coefficients

$$P_T = \frac{1}{T} \int_T |x(t)|^2 = \sum_{-infty}^{\infty} |c_k|^2$$

- Interpretation of Fourier series for real signal
  - ▶ the signal is the sum of cosine signals with frequency  $0, F_0, 2F_0...$ , with amplitudes  $|c_k|$  and phase  $\angle c_k$
- No other frequencies appear in spectrum → spectrum is made of "lines"

## Time-frequency duality

- ► Time-frequency **duality**:
  - ► Real signal -> Even spectrum
  - ► Periodic signal -> Discrete spectrum

### Analog non-periodical signals

▶ The signal is composed of all frequencies (inverse Fourier transform)

$$x(t) = \int_{infty}^{\infty} X(F) e^{j2\pi ft} dF$$

The frequency content is found by the Fourier transform

$$X(F) = \int_{infty}^{\infty} x(t)e^{-j2\pi ft}dt$$

- ( Can use instead  $\Omega = 2\pi F$ )
- X(F) is a complex function
  - $\blacktriangleright |X(F)|$  is the amplitude spectrum
  - $\triangleright$   $\angle X(F)$  is the phase spectrum

## Conditions for convergence

- When is the Fourier series convergent to the signal?
  - ▶ i.e. when is the relation correct,
  - i.e. when is the sum actually equal to x(t)?
- ▶ Dirichlet conditions: the sum is convergent in all continuity points if:
  - 1. x(t) is continuous or has a finite number of discontinuities
  - 2. x(t) has a finite number of maxima and minima
  - 3. x(t) is absolutely integrable:

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

- Weaker condition:
  - if x(t) is square summable

$$\int_{-\infty}^{\infty} x(t)^2 dt < \infty$$

- then the he difference  $d(t) = x(t) \sum_{k=-\infty}^{\infty} c_k e^{j2\pi k F_0 t}$  has zero energy
- ▶ Does not guarantee *pointwise* convergence

## Signal spectrum

- X(F) is a complex function
- ▶ If the signal is **real**  $x(t) \in \mathbb{R}$ , then the X(F) is **even** 
  - ▶ |X(F)| = |X(-F)|
  - $ightharpoonup \angle X(F) = -\angle X(-F)$
  - ▶ group the terms with  $c_k$  with  $c_{-k}$  -> cosine with amplitude |X(F)| and phase  $\angle X(F)$
- Signal energy is the same in time and frequency domains

$$E = \int_{\infty}^{\infty} |x(t)|^2 dt = \int_{\infty}^{\infty} |X(F)|^2 dF$$

▶ The power spectral density of x(t) is

$$S_{xx}(F) = |X(F)|^2$$



## Fourier series of discrete periodical signals

- ▶ A discrete signal of period N: x[n] = x[n + N]
- Decomposed as a sum of complex exponentials:

$$x[n] = \sum_{k=0}^{N-1} c_k e^{j2\pi kn/N}, n = 0, 1, ...N - 1$$

Finding the coefficients:

$$c_k = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-j2\pi kn/N}$$

# Comparison with analog Fourier series

- Compared to analog signals:
  - consider fundamental frequency  $f_0 = 1/N$
  - only *N* terms, with frequencies  $k \cdot f_0$ :
    - $\triangleright$  0,  $f_0$ ,  $2f_0$ , ... $(N-1)f_0$
  - ▶ only N distinct coefficients c<sub>k</sub>
  - ▶ the *N* coefficients  $c_k$  can be chosen like  $-\frac{N}{2} < k \le \frac{N}{2} =>$  the frequencies span the range -1/2...1/2

$$-\frac{1}{2} < f_k \le \frac{1}{2}$$

$$-\pi < \omega_k \le \pi$$

### Basic properties of Fourier coefficients

1. Signal is **discrete** –> coefficients are **periodic** with period N

$$c_{k+N} = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-j2\pi(k+N)n/N} = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-j2\pi kn/N}$$

- 2. If signal is real  $x[n] \in \mathbb{R}$ , the coefficients are **even**:
  - $c_k^* = c_{-k}$
  - $|c_k| = |c_{-k}|$
  - $ightharpoonup \angle c_k = \angle c_{-k}$
- Together with periodicity:
  - $|c_k| = |c_{-k}| = |c_{N-k}|$

### Expressing as sum of sinusoids

▶ Grouping terms with  $c_k$  and  $c_{-k}$  we get

$$x[n] = c_0 + 2\sum_{k=1}^{L} |c_k| cos(2\pi \frac{k}{N} + \angle c_k)$$

where L = N/2 or L = (N-1)/2 depending if N is even or odd

- ightharpoonup Signal = DC value + a finite sum of sinusoids with frequencies  $kf_0$ 
  - $ightharpoonup |c_k|$  give the amplitudes (x 2)
  - $ightharpoonup \angle c_k$  give the phases

# Power spectral density

▶ The average power of a discrete periodic signal

$$P = \frac{1}{N} \sum_{n=0}^{N-1} |x[n]|^2$$

Is the same in the frequency domain (with proof):

$$P = \sum_{k=0}^{N-1} |c_k|^2$$

Power spectral density of the signal is

$$S_{xx}[k] = |c_k|^2$$

► Energy over one period is

$$E = \sum_{n=0}^{N-1} |x[n]|^2 = N \sum_{k=0}^{N-1} |c_k|^2$$

## Examples

Examples:

$$x_1[n] = cos(\sqrt{5}\pi n)$$
$$x_2[n] = 2sin(\frac{\pi}{3}n)$$
$$x_3[n] = \{1, 1, 0, 0\}$$

### Example in Python

```
>>> import numpy as np
>>> from scipy.fftpack import fft, ifft
>>> x = np.array([1.0, 1.0, 0.0, 0.0])
>>> y = 1.0/4.0 * fft(x)
>>> y
array([ 0.50+0.j , 0.25-0.25j, 0.00+0.j , 0.25+0.25j])
```

### 1. Linearity

If the signal  $x_1[n]$  has the Fourier series coefficients  $\{c_k^{(1)}\}$ , and  $x_2[n]$  has  $\{c_k^{(2)}\}$ , then their sum has

$$a \cdot x_1[n] + b \cdot x_2[n] \leftrightarrow \{a \cdot c_k^{(1)} + b \cdot c_k^{(2)}\}$$

Proof: via definition

### 2. Shifting in time

If  $x[n] \leftrightarrow \{c_k\}$ , then

$$x[n-n_0] \leftrightarrow \{e^{(-j2\pi k n_0/N)}c_k\}$$

Proof: via definition

▶ The amplitudes  $|c_k|$  are not affected, shifting in time affects only the phase

### 3. Modulation in time

$$e^{j2\pi k_0 n/N} \leftrightarrow \{c_{k-k_0}\}$$

### 4. Complex conjugation

$$x^*[n] \leftrightarrow \{c_{-k}^*\}$$

#### 5. Circular convolution

Circular convolution of two signals  $\leftrightarrow$  product of coefficients

$$x_1[n] \otimes x_2[n] \leftrightarrow \{N \cdot c_k^{(1)} \cdot c_k^{(2)}\}$$

Circular convolution:

$$x_1[n] \otimes x_2[n] = \sum_{k=0}^{N-1} x_1[k] x_2[(n-k)_N]$$

- ▶ takes two periodic signals of period N, result is the same
- Example at the whiteboard: how it is computed

#### 6. Product in time

Product in time  $\leftrightarrow$  circular convolution of Fourier series coefficients

$$x_1[n] \cdot x_2[n] \leftrightarrow \sum_{k=1}^{N-1} c_m^{(1)} c_{(k-m)_N}^{(2)} = c_k^{(1)} \otimes c_k^{(2)}$$

## Fourier transform of discrete non-periodical signals

- Non-periodical signals contain all frequencies, not only the multiples of  $f_0$
- ▶ The Fourier transform of a discrete signal:

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

▶ The inverse Fourier transform:

$$x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(\omega) e^{j\omega n} d\omega$$

### Comparison

- ▶ Compared with the Fourier transform of analog signals
  - sum instead of integral in Fourier transform
  - spectrum is only in range:

$$\omega \in [-\pi, \pi]$$

$$f\in\left[-\frac{1}{2},\frac{1}{2}\right]$$

- ▶ Compared with the Fourier series of discrete periodical signals
  - general  $\omega$  instead of  $2\pi k f_0$
  - spectrum is continuous, not discrete
  - ▶ integral, not sum in inverse Fourier transform

### Parseval theorem

▶ Parseval theorem: energy of the signal is the same in time and frequency domains

$$E = \sum_{-\infty}^{\infty} |x[n]|^2 = \frac{1}{2\pi} \int_{-\pi}^{\pi} |X(\omega)|^2$$

- Proof: on whiteboard
- ► The **energy spectral density** gives the energy contained for each frequency

$$S_{xx}(\omega) = |X(\omega)|^2$$

## Basic properties of Fourier transform

• It is **periodical** with period  $2\pi$ 

$$X(\omega + 2\pi) = X(\omega)$$

▶ If the signal x[n] is real, the Fourier transform is **even** 

$$x[n] \in \mathbb{R} \to X^*(\omega) = X(-\omega)$$

- ▶ This means
  - modulus is even:  $|X(\omega)| = |X(-\omega)|$
  - phase is odd:  $X(\omega) = -X(-\omega)$

## Convergence of the Fourier transform

- When are the relations valid?
- Assume we compute the Fourier transform with only 2M + 1 samples:

$$X_M(\omega) = \sum_{-M}^M x[n] e^{-j\omega n}$$

▶ If a signal x[n] is **absolutely summable**:

$$\sum_{\infty}^{\infty} |x[n]| < \infty$$

▶ then the Fourier series is **uniform convergent** for every  $\omega$  (OK):

$$\lim_{M\to\infty}X(\omega)-X_M(\omega)=0$$

# Convergence for square-summable signals

► Signals that are only **square summable** 

$$\sum_{\infty}^{\infty} |x[n]|^2 < \infty$$

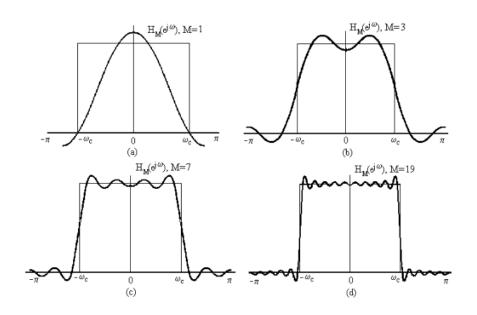
have a weaker convergence:

$$\lim_{M\to\infty}\int_{-\pi}^{\pi}|X(\omega)-X_M(\omega)|^2d\omega=0$$

### The Gibbs phenomenon

- ▶ When  $H(\omega)$  has discontinuities, then h[n] is not absolutely summable, only square summable
- ▶ Problem: if we only use *M* samples, even if *M* is very large, we will obtain **small oscillations around the discontinuity**
- ▶ As  $M \to \infty$ , the oscillations do not become smaller, but thinner –> they don't go away!
- ► The Fourier transform will always *overshoot* with about 9% below and above
- ► Known as the Gibbs phenomenon

# Gibbs phenomenon



### Relation between Fourier series and Fourier transform

- ▶ If apply Fourier transform to periodical discrete signals,  $X(\omega)$  contains Diracs
- ▶ The Diracs are at frequencies  $kf_0$ , just like the Fourier series
- ▶ The value of an impulse = the coefficient  $c_k$  of the Fourier series
- ▶ The Fourier series  $\approx$  the Fourier transform of periodic signals
  - lacktriangle we directly compute the coefficients  $c_k$  of the impulses in the spectrum

### Fourier transform and Z transform

▶ Definition of Fourier transform = Z transform with:

$$z = e^{j\omega}$$

- $e^{j\omega}$  = points on the unit circle
- ► Fourier transform = Z transform evaluated **on the unit circle** 
  - ▶ if the unit circle is in the convergence region of Z transform
  - otherwise, equivalence does not hold
- ▶ This is true for most usual signals we work with
  - there are exceptions, but they are outside the scope of this class

### Properties of Fourier transform

### 1. Linearity

$$a \cdot x_1[n] + b \cdot x_2[n] \leftrightarrow a \cdot X_1(\omega) + b \cdot X_2(\omega)$$

Proof: via definition

### Properties of Fourier transform

### 2. Shifting in time

$$x[n-n_0] \leftrightarrow e^{-j\omega n_0}X(\omega)$$

Proof: via definition

▶ The amplitudes  $|X(\omega)|$  is not affected, shifting in time affects only the phase

## Properties of Fourier transform

### 3. Modulation in time

$$e^{j\omega_0 n} x[n] \leftrightarrow X(\omega - \omega_0)$$

### 4. Complex conjugation

$$x^*[n] \leftrightarrow X^*(-\omega)$$

#### 5. Convolution

$$x_1[n] * x_2[n] \leftrightarrow X_1(\omega) \cdot X_2(\omega)$$

▶ Not circular convolution, this is the normal convolution

#### 6. Product in time

Product in time  $\leftrightarrow$  convolution of Fourier transforms

$$x_1[n] \cdot x_2[n] \leftrightarrow \frac{1}{2\pi} \int_{-\pi}^{\pi} X_1(\lambda) X_2(\omega - \lambda) d\lambda$$

#### Correlation theorem

$$r_{\mathsf{x}_1\mathsf{x}_2}[I] \leftrightarrow \mathsf{X}_1(\omega)\mathsf{X}_2(-\omega)$$

#### Wiener Khinchin theorem

Autocorrelation of a signal  $\leftrightarrow$  Power spectral density

$$r_{xx}[I] \leftrightarrow S_{xx}(\omega) = |X(\omega)|^2$$

#### Parseval theorem

Energy is the same when computed in the time or frequency domain

$$\sum |x[n]|^2 = \frac{1}{2\pi} \int_{-\pi}^{\pi} |X(\omega)|^2 d\omega$$

### Relationship of Fourier transform and Fourier series

- How are they related?
  - ▶ Fourier transform: for non-periodical signals
  - ▶ Fourier series: for periodical series
- ▶ Duality: periodic in time ↔ discrete in frequency
- ▶ If we **periodize** a signal x[n] by repeating with period N:

$$x_N[n] = \sum_{k=-\infty}^{\infty} x[n-kN]$$

then the Fourier transform w is discrete (made of Diracs):

$$X_N(\omega) = 2\pi c_k \delta(\omega - k \frac{2\pi}{N})$$

▶ The coefficients of the Diracs = exactly the Fourier series coefficients

### Relationship of Fourier transform and Fourier series

- So, Fourier transform can be considered for both periodic and non-periodic signals
- ► Fourier transform for periodic signals = discrete (sum of Diracs with some coefficients)
  - ▶ Diracs at frequencies  $f_0 = 1/N$  and its multiplies
- ► Fourier series for periodic signals = gives the coefficients of the Diracs directly
  - ▶ it just omits to write the Diracs explicitly in the equation

#### Relation of Fourier transform and Z transform

- Fourier transform:  $X(\omega) = \sum_{-\infty}^{\infty} x[n]e^{-j\omega n}$
- ▶ Z transform:  $X(z) = \sum_{-\infty}^{\infty} x[n]z^{-n}$
- **Fourier tranform = Z transform for**  $z = e^{j\omega}$
- $z = e^{j\omega}$  means evaluated on the unit circle:
  - $|z| = |e^{j\omega}| = 1 (modulus)$
  - $ightharpoonup \angle z = \angle e^{j\omega} = \omega(phase)$
- Conditions:
  - ▶ unit circle must be in the Convergence Region of Z transform
  - some signals can have Fourier transform even though unit circle not in CR
- ▶ If signal has pole on unit circle → Dirac (infinite) in Fourier transform
  - ► e.g. *u*[*n*]
  - some signals are non-convergent on unit circle, but have Fourier transform (e.g. u[n])

$$X(z) = C \cdot \frac{(z - z_1) \cdots (z - z_M)}{(z - p_1) \cdots (z - p_N)}$$

$$X(\omega) = C \cdot \frac{(e^{j\omega} - z_1) \cdots (e^{j\omega} - z_M)}{(e^{j\omega} - p_1) \cdots (e^{j\omega} - p_N)}$$

Modulus:

$$|X(\omega)| = |C| \cdot \frac{|e^{j\omega} - z_1| \cdots |e^{j\omega} - z_M|}{|e^{j\omega} - p_1| \cdots |e^{j\omega} - p_N|}$$

Phase:

$$\angle X = \angle C + \angle (e^{j\omega} - z_1) + \cdots + \angle (e^{j\omega} - z_M) - \angle (e^{j\omega} - p_1) - \cdots - \angle (e^{j\omega} - p_N)$$

- For complex numbers:
  - ightharpoonup modulus of |a-b|= the length of the segment between a and b
  - ▶ phase of |a b| = the angle of the segment from b to a (direction is important)
- ▶ So, for a point on the unit circle  $z = e^{j\omega}$ 
  - ▶ modulus  $|X(\omega)|$  is given by the distances to the zeros and to the poles
  - ▶ phase  $\angle X(\omega)$  is given by the angles from the zeros and poles to z

- Consequences:
  - when a pole is very close to unit circle -> Fourier transform is large at this point
  - when a zero is very close to unit circle -> Fourier transform is small at this point
- Examples:...
- ▶ Simple interpretation for modulus  $|X(\omega)|$ :
  - $\triangleright$  Z transform X(z) is a "landscape"
    - poles = mountains of infinite height
    - zeros = valleys of zero height
  - ▶ Fourier transform  $X(\omega) =$  "Walking over this landscape along the unit circle" -> the heights give the Fourier transform
  - ► When close to a mountain -> road is high -> Fourier transform has large amplitude
  - ► When close to a valley -> road is low -> Fourier transform has small amplitude
- Enough to sketch the Fourier transform for signals with few poles/zeros

- Note: X(z) might also have a constant C in front!
  - ▶ It does not appear in pole-zero plot
  - ▶ The value of |C| and  $\angle C$  must be determined separately
- ▶ This "geometric method" can be applied for both modulus and phase

### Time-frequency duality

- ▶ **Duality** properties related to Fourier transform/series
- ▶ Discrete ↔ Periodic
  - discrete in time -> periodic in frequency
  - periodic in time -> discrete in frequency
- ► Continuous ↔ Non-periodic
  - continous in time -> non-periodic in frequency
  - ▶ non-periodic in time → continuous in frequency

# Frequency-based classification of signals

- Based on frequency content:
  - low-frequency signals
  - mid-frequency signals (band-pass)
  - high-frequency signals
- **Band-limited** signals: spectrum is 0 over some frequency  $f_{max}$
- ▶ **Time-limited** signals: signal value is 0 outside some time interval
- ▶ **Bandwitdh** B: frequency interval  $[F_1, F_2]$  which contains 95% of energy
  - ▶  $B = F_2 F_1$
- Based on bandwidth B:
  - ▶ Narrow-band signals: \$B << \$ central frequency  $\frac{F_1+F_2}{2}$
  - ► Wide-band signals: not narrow-band