Signals, Instruments, and Systems – W4

An Introduction to Signal Processing – Signals, Series, Transforms
Signal – Definition

“A signal is any time-varying or spatial-varying quantity”
Signal – Definition

“A signal is any **time**-varying or **spatial**-varying quantity”

\[ T = f(t) \]
Signal – Definition

“A signal is any **time**-varying or **spatial**-varying quantity”

\[ h = f(x) \]
Signal – Definition

“A signal is any **time**-varying or **spatial**-varying quantity”
Signal – Definition

“A signal is any time-varying or spatial-varying quantity”

\[ h = f(x, y) \]
Signal – Definition

“A signal is any time-varying or spatial-varying quantity”

\[ I = f(x, y) \]

Photo: Maurizio Polese

\[ y \text{ [Pixel]} \]

\[ x \text{ [Pixel]} \]
Continuous versus Discrete
Continuous

Discrete

$2.478956983748...$
Analog versus Digital
Analog versus Digital

• An analog signal …
  – is **continuous** in time
  – has **continuous** values

• A digital signal …
  – is **discrete** in time
  – has **discrete** values
Analog – Digital

- Analog
  - Continuous
  - Real world

- Digital
  - Discrete
  - Digital world
Sensors Classification
Continuous – Discrete

<table>
<thead>
<tr>
<th></th>
<th>Continuous</th>
<th>Discrete</th>
</tr>
</thead>
<tbody>
<tr>
<td>Time</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Space</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Amplitude</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
Continuous: Amplitude and Time

Barometer

Thermometer
Continuous: Time and Amplitude

Barograph

Gramophone record

Thermograph
Continuous: Space and Amplitude
Discrete: Time

Camera

Reel camera
Discrete: Time and Amplitude

Weather station

Digital Watch
Discrete: Time, Space and Amplitude

Digital camera

Digital video camera
Analog-Digital-Analog Conversion

Analog to Digital Conversion (ADC)

Digital to Analog Conversion (DAC)

\[ y = x(t) \] \[ y = x[n] \]
Analog-Digital Conversion
Analog-Digital Converter (ADC)

- Transforms continuous analog signal into series of values
- Two key elements
  - Sampling (in time)
  - Quantization (of values)
- Two types of converters
  - Linear ADCs
  - Non-linear ADCs

\[ y[n] = 0 \ 0 \ -2 \ -4 \ -2 \ 0 \ 4 \ 8 \ 10 \ 10 \]
Examples of ADC Applications

- Input line of a soundcard
- Sensor of digital camera
- Mobile phone
- Computer mouse
- …
An Example from your Course Tools

- From datasheet dsPIC on e-puck: ADC 16 channels, 12 bits, 200 kspss total capacity
- $2^{12}$ -1 discrete levels of resolution per channel
- Max 200k Hz sampling frequency on 1 channel -> e.g., 10 channels: 20k Hz
Digital-Analog Conversion
Digital-Analog Converter (DAC)

- Transforms a series of values into a continuous signal
- DAC outputs piecewise constant signal
- Additional filter stage to smooth signal (often carried out by the properties of the transducer/actuator itself)

\[ y[n]=0 \ 0 \ -2 \ -4 \ -2 \ 0 \ 4 \ 8 \ 10 \ 10 \]
Examples of DAC Applications

• Mobile phone
• MP3 player
• Graphics card (with older monitors)
• Monitors (newer systems)
Typical Waves and Signals
Waves

“A wave is a disturbance that propagates through space and time, usually with transference of energy.”

Wave function:

\[ f(x, t) = \gamma \sin(kx - \omega t) \]

\[ k = \frac{2\pi}{\lambda} \] “wave number”

\[ \omega = 2\pi f \] “angular frequency”

\[ \lambda = \frac{v}{f} \] “wave length”

\[ v = \text{speed in x direction} \]
Electromagnetic Spectrum

Increasing Frequency (ν) →

- γ rays
- X rays
- UV
- IR
- Microwave
- FM Radio waves
- AM Radio waves
- Long radio waves

Increasing Wavelength (λ) ↓

Visible spectrum

Increasing Wavelength (λ) in nm ↓

400 500 600 700
Common Frequencies

- Car motor: ~50 Hz (3000 rpm)
- Power lines: 50 Hz
- Ear: 20 Hz – 20 kHz
- Ultrasound: 20 kHz – 200 MHz
- Watch quartz: 32 kHz
- Medical Sonograph: ~2-18 MHz
- CPU: 2-3 GHz
- GSM: 0.9/1.8 GHz
Standard waveforms

- Sine
- Square
- Triangle
- Sawtooth
Sinusoid (Sine Wave) in Time

\[ y(t) = A \cdot \sin(\omega t + \theta) \]

\( A \) : amplitude
\( \omega \) : angular frequency
\( \theta \) : phase
Dirac Delta Function

\[ \delta(t) = \begin{cases} + \infty, & t = 0 \\ 0, & t \neq 0 \end{cases} \]

\[ \int_{-\infty}^{+\infty} \delta(t) \, dt = 1 \]

**Note:** The height of the arrow (or small circle) is usually used to specify the value of any multiplicative constant, which will give the area under the function; the other convention is to write the area next to the arrowhead. [Wikipedia]
Fourier Series
Fourier Series

- Joseph Fourier (1768-1830) proposed that any periodic function can be decomposed into a sum of simple oscillating functions, namely sines and cosines.
“Any periodic function …”

from: Robert W. Stewart, Daniel García-Alís, “Concise DSP Tutorial”
“...can be decomposed into a sum of sines and cosines”

from: Robert W. Stewart, Daniel García-Alís, “Concise DSP Tutorial”
Fourier Series

\[ f(t) = \sum_{n=0}^{\infty} A_n \cos\left(\frac{2\pi nt}{T}\right) + \sum_{n=1}^{\infty} B_n \sin\left(\frac{2\pi nt}{T}\right) \]

• Suppose a $2\pi$ periodic function $f(t)$ integrable on $[-\pi, \pi]

• The Fourier coefficients of $x$ are:

\[
A_n = \frac{1}{\pi} \int_{-\pi}^{\pi} f(t) \cos(nt) dt, \quad n \geq 0
\]

\[
B_n = \frac{1}{\pi} \int_{-\pi}^{\pi} f(t) \sin(nt) dt, \quad n \geq 1
\]
Complex Numbers Review

Rectangular form: \( C_n = \text{Re}(C_n) + i \text{Im}(C_n) \)

Euler's formula: \( e^{i\varphi} = \cos \varphi + i \sin \varphi \)

Polar form: \( C_n = |C_n|e^{i\varphi} \)

Magnitude: \( |C_n| \)

Phase: \( \varphi \)

\( \text{Re}(C_n) = |C_n| \cos \varphi \)

\( \text{Im}(C_n) = |C_n| \sin \varphi \)
From Real to Complex Coefficients

Real Fourier coefficients:

\[ f(t) = A_0 + \sum_{n=1}^{\infty} \left[ A_n \cos\left(\frac{2\pi nt}{T}\right) + B_n \sin\left(\frac{2\pi nt}{T}\right) \right]; A_n, B_n \in \mathbb{R} \]

Using the Euler relationships:

\[ e^{i\omega} = \cos \omega + i \sin \omega \]
\[ e^{-i\omega} = \cos(-\omega) + i \sin(-\omega) = \cos \omega - i \sin \omega \]

we can express \( \sin \) and \( \cos \) using exponential functions

\[ \Rightarrow \cos \omega = \frac{e^{i\omega} + e^{-i\omega}}{2}; \sin \omega = \frac{e^{i\omega} - e^{-i\omega}}{2i} \]

And substituting \( \sin \) and \( \cos \)

we get \((\omega_0 = 2\pi/T)\):

\[ f(t) = A_0 + \sum_{n=1}^{\infty} \left[ A_n \left(\frac{e^{i\omega_0 t} + e^{-i\omega_0 t}}{2}\right) + B_n \left(\frac{e^{i\omega_0 t} - e^{-i\omega_0 t}}{2i}\right) \right] \]

and then

\[ f(t) = A_0 + \sum_{n=1}^{\infty} \left(\frac{A_n - iB_n}{2}\right) e^{i\omega_0 t} + \sum_{n=-\infty}^{-1} \left(\frac{A_n + iB_n}{2}\right) e^{i\omega_0 t} \]

We can now express our Fourier series with complex coefficients

\[ f(t) = \sum_{n=-\infty}^{\infty} C_n e^{i\omega_0 t}; C_n \in \mathbb{C} \]
Complex Fourier Coefficients

Fourier series

\[ f(t) = \sum_{n=-\infty}^{\infty} C_n e^{in\omega_0 t} \]

Fourier coefficients

\[ C_n = \frac{1}{2\pi} \int_{-\pi}^{\pi} f(t) e^{-in\omega_0 t} \, dt \]
Examples

[Matlab demo]

http://users.ece.gatech.edu/mcclella/matlabGUIs/
Fourier Transform
From Fourier Series to Transform

- $f(t)$
  - an aperiodic signal
  - view it as the limit of a periodic signal as $T \rightarrow \infty$
  - is a piecewise continuous, integrable (function space $L^1$) or even a square-integrable (function space $L^2$) function

\[ \int_{-\infty}^{\infty} |f(x)| \, dx < \infty \quad \int_{-\infty}^{\infty} |f(x)|^2 \, dx < \infty \]

- For a periodic signal, the harmonic components are spaced $\omega_0 = 2\pi/T$ apart

- As $T \rightarrow \infty$, $\omega_0 \rightarrow 0$, and harmonic components are spaced closer and closer in frequency forming a continuous spectrum
Fourier Transform

Unitary, ordinary frequency notation

\[ \hat{f}(\xi) = \int_{-\infty}^{\infty} f(t) \cdot e^{-i2\pi \xi t} \, dt \]

\[ f(t) = \int_{-\infty}^{\infty} \hat{f}(\xi) \cdot e^{i2\pi t \xi} \, d\xi \]

Notes:
• \( \hat{f} \) is often replaced with \( F \)
• the direct transform is also called “analysis equation”
• the inverse transform is also called “synthesis equation”
Fourier Transform

Non-unitary, angular frequency notation

\[ F(\omega) = \int_{-\infty}^{+\infty} f(t) e^{-i\omega t} \, dt \]

\[ f(t) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} F(\omega) e^{i\omega t} \, d\omega \]

Notes:
- F can be replaced with \( f^\wedge \)
- In electrical engineering, \( i \) is substituted by \( j \) ("i" booked for current)
- Often, in order to emphasize the frequency response aspect, the imaginary aspect of the transform is emphasized: \( F(\omega) \), \( F(i\omega) \), or \( F(j\omega) \) are all equivalent notations
Common Fourier Transform

$$f(t) = \delta(t)$$

**FT** (analysis equation)

**IFT** (synthesis equation)

$$f(\xi) = 1$$
FT of a Dirac Impulse

\[ f(t) = \delta(t) \]

\[ \hat{f}(\xi) = \int_{-\infty}^{+\infty} f(t)e^{-i2\pi\xi t} \, dt = \int_{-\infty}^{+\infty} \delta(t)e^{-i2\pi\xi t} \, dt = 1 \]  
Analysis equation

\[ f(t) = \int_{-\infty}^{+\infty} \hat{f}(\xi)e^{i2\pi t\xi} \, d\xi = \int_{-\infty}^{+\infty} e^{i2\pi t\xi} \, d\xi = \delta(t) \]  
Synthesis equation
Common Fourier Transform

\[ f(t) = 1 \]

\[ \hat{f}(\xi) = \delta(\xi) \]
Common Fourier Transform

\[ f(t) = \cos(2\pi at) \]

\[ \hat{f}(\xi) = \frac{\delta(\xi - a) + \delta(\xi + a)}{2} \]
Common Fourier Transform

\[ f(t) = \text{rect}(t) \]

\[ \hat{f}(\xi) = \text{sinc}(\xi) \]
Properties

Linearity \( h(x) = af(x) + bg(x) \implies \hat{h}(\xi) = a \hat{f}(\xi) + b \hat{g}(\xi) \)

Translation \( h(x) = f(x - x_0) \implies \hat{h}(\xi) = e^{-2\pi i x_0 \xi} \hat{f}(\xi) \)

Modulation \( h(x) = e^{-2\pi i x_\xi} f(x) \implies \hat{h}(\xi) = \hat{f}(\xi - \xi_0) \)

Scaling \( h(x) = f(ax) \implies \hat{h}(\xi) = \frac{1}{|a|} \hat{f}(\frac{\xi}{a}) \)

Convolution \( h(x) = (f * g)(x) \implies \hat{h}(\xi) = \hat{f}(\xi) \cdot \hat{g}(\xi) \)

See next lecture
Discrete Fourier Transform

\[ \hat{f}(\xi) = F(\xi) = \int_{-\infty}^{\infty} f(t) \cdot e^{-2\pi i \xi t} \, dt \]

\[ X[k] = \sum_{n=0}^{N-1} x[n] \cdot e^{-\frac{2\pi i kn}{N}}, \quad k = 0, \ldots, N - 1 \]

- Input and output sequences are both finite (the integral becomes a finite sum)
- Matlab, operating on a computer (i.e. a digital device) can only emulate continuity/infinity and therefore use this discrete version with an adjustable discretization level (in time, frequency, and amplitude) and finite bounds
- Also, since Matlab index start from 1, formulation is typically \( x[n+1] \), \( X[k+1] \)
Discrete Fourier Transform

\[ X[k] = \sum_{n=0}^{N-1} x[n] \cdot e^{-\frac{2\pi i}{N} kn} , k = 0, \cdots, N - 1 \quad \text{DFT} \]

\[ x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] \cdot e^{\frac{2\pi i}{N} kn} , k = 0, \cdots, N - 1 \quad \text{Inverse DFT} \]

- Different from the Discrete-Time Fourier Transform (DTFT), see Week 6
- Efficient implementation, for instance in Matlab, is done using a Fast Fourier Transform (FFT) algorithm (e.g., Cooley and Tukey, 1965)
- Clear correspondence with the Fourier series (see slide 41)
Fourier Tables
Fourier Transforms Table

<table>
<thead>
<tr>
<th>Function</th>
<th>Fourier transform unitary, ordinary frequency</th>
<th>Fourier transform unitary, angular frequency</th>
<th>Fourier transform non-unitary, angular frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>$f(x)$</td>
<td>$\hat{f}(\xi) = \int_{-\infty}^{\infty} f(x)e^{-2\pi i x \xi} , dx$</td>
<td>$\hat{f}(\omega) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(x)e^{-i\omega x} , dx$</td>
<td>$\hat{f}(\nu) = \int_{-\infty}^{\infty} f(x)e^{-i\nu x} , dx$</td>
</tr>
<tr>
<td>$\text{rect}(ax)$</td>
<td>$\frac{1}{</td>
<td>a</td>
<td>} \cdot \text{sinc} \left( \frac{\xi}{a} \right)$</td>
</tr>
<tr>
<td>$\text{sinc}(ax)$</td>
<td>$\frac{1}{</td>
<td>a</td>
<td>} \cdot \text{rect} \left( \frac{\xi}{a} \right)$</td>
</tr>
<tr>
<td>$\text{sinc}^2(ax)$</td>
<td>$\frac{1}{</td>
<td>a</td>
<td>} \cdot \text{tri} \left( \frac{\xi}{a} \right)$</td>
</tr>
<tr>
<td>$\text{tri}(ax)$</td>
<td>$\frac{1}{</td>
<td>a</td>
<td>} \cdot \text{sinc}^2 \left( \frac{\xi}{a} \right)$</td>
</tr>
<tr>
<td>$e^{-ax}u(x)$</td>
<td>$\frac{1}{a + 2\pi i \xi}$</td>
<td>$\frac{1}{2\pi (a + i\omega)}$</td>
<td>$\frac{1}{a + i\nu}$</td>
</tr>
<tr>
<td>$e^{-\alpha x^2}$</td>
<td>$\frac{\pi}{\sqrt{\alpha}} \cdot e^{-\left(\frac{x^2}{\alpha}\right)}$</td>
<td>$\frac{1}{\sqrt{2\alpha}} \cdot e^{-\frac{x^2}{4\alpha}}$</td>
<td>$\frac{\pi}{\sqrt{\alpha}} \cdot e^{-\frac{i^2 \nu^2}{4\alpha}}$</td>
</tr>
<tr>
<td>$e^{-a</td>
<td>x</td>
<td>}$</td>
<td>$\frac{2a}{a^2 + 4\pi^2 \xi^2}$</td>
</tr>
</tbody>
</table>

Simple example s. 58
# Fourier Properties Table

<table>
<thead>
<tr>
<th>Function</th>
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<td>$\hat{f}(\nu) = \int_{-\infty}^{\infty} f(x) e^{-i\nu x} , dx$</td>
</tr>
<tr>
<td>$a \cdot f(x) + b \cdot g(x)$</td>
<td>$a \cdot \hat{f}(\xi) + b \cdot \hat{g}(\xi)$</td>
<td>$a \cdot \hat{f}(\omega) + b \cdot \hat{g}(\omega)$</td>
<td>$a \cdot \hat{f}(\nu) + b \cdot \hat{g}(\nu)$</td>
</tr>
<tr>
<td>$f(x - a)$</td>
<td>$e^{-2\pi i x \xi} \hat{f}(\xi)$</td>
<td>$e^{-i\omega(x - a)} \hat{f}(\omega)$</td>
<td>$e^{-i\nu(x - a)} \hat{f}(\nu)$</td>
</tr>
<tr>
<td>$e^{2\pi i a x} f(x)$</td>
<td>$\hat{f}(\xi - a)$</td>
<td>$\hat{f}(\omega - 2\pi a)$</td>
<td>$\hat{f}(\nu - 2\pi a)$</td>
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<td>$f(ax)$</td>
<td>$\frac{1}{</td>
<td>a</td>
<td>} \hat{f}\left(\frac{\xi}{a}\right)$</td>
</tr>
<tr>
<td>$f(x)$</td>
<td>$f(-\xi)$</td>
<td>$f(-\omega)$</td>
<td>$2\pi f(-\nu)$</td>
</tr>
<tr>
<td>$\frac{d^n f(x)}{dx^n}$</td>
<td>$(2\pi i \xi)^n \hat{f}(\xi)$</td>
<td>$(i\omega)^n \hat{f}(\omega)$</td>
<td>$(i\nu)^n \hat{f}(\nu)$</td>
</tr>
<tr>
<td>$x^n f(x)$</td>
<td>$\left(\frac{i}{2\pi}\right)^n \frac{d^n \hat{f}(\xi)}{d\xi^n}$</td>
<td>$\frac{2\pi}{i^n} \frac{d^n \hat{f}(\omega)}{d\omega^n}$</td>
<td>$\frac{2\pi}{i^n} \frac{d^n \hat{f}(\nu)}{d\nu^n}$</td>
</tr>
<tr>
<td>$(f * g)(x)$</td>
<td>$\hat{f}(\xi) \hat{g}(\xi)$</td>
<td>$\sqrt{2\pi} \hat{f}(\omega) \hat{g}(\omega)$</td>
<td>$\hat{f}(\nu) \hat{g}(\nu)$</td>
</tr>
<tr>
<td>$f(x) g(x)$</td>
<td>$(\hat{f} \star \hat{g})(\xi)$</td>
<td>$\frac{\hat{f} \star \hat{g}}{\sqrt{2\pi}}(\omega)$</td>
<td>$\frac{1}{2\pi} (\hat{f} \star \hat{g})(\nu)$</td>
</tr>
</tbody>
</table>

Combined example s. 59
Simple Example

\[ f(t) = \text{trian}(t) \]

\[ f(\xi) = \text{sinc}^2(\xi) \]
Combined Example

\[ f(t) = \cos(2\pi(3t))e^{-\pi t^2} \]

The Fourier transform of \( f(t) \)
Conclusion
Take Home Messages

• A signal can be a varying quantity in time and/or space
• Signal classification: continuous vs. discrete in time, space and amplitude
• Fourier decomposition: every periodic signal can be decomposed into a sum of sines and cosines
• Fourier coefficients are the weights in this sum
• Fourier Transform (FT) for aperiodic signals
• Implementation on computers of Fourier transform is discretized and bound: Discrete Fourier Transform (DFT)
• The Fast FT (FFT) is a fast algorithm for solving the DFT
• Properties of FT and tables help to solve analytically analysis and synthesis equations
Additional Literature – Week 4

Related course material
- Information, Calcul, Communication (ICC)
- Analysis IV
- Quantitative Methods II

Book