

Lecture VIII: Fourier series

Maxim Raginsky

BME 171: Signals and Systems
Duke University

September 19, 2008

This lecture

Plan for the lecture:

- 1 Review of vectors and vector spaces
- 2 Vector space of continuous-time signals
- 3 Vector space of T -periodic signals
- 4 Complete orthonormal systems of functions
- 5 Trigonometric Fourier series
- 6 Complex exponential Fourier series

Review: vectors

Recall vectors in n -space \mathbb{R}^n . Each such vector \mathbf{u} can be uniquely represented as a linear combination of n unit vectors $\mathbf{e}_1, \dots, \mathbf{e}_n$:

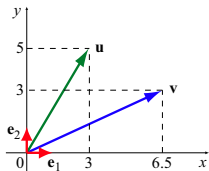
$$\mathbf{u} = \alpha_1 \mathbf{e}_1 + \alpha_2 \mathbf{e}_2 + \dots + \alpha_n \mathbf{e}_n,$$

where $\alpha_1, \dots, \alpha_n$ are real numbers (coefficients). These can be computed using the scalar (or dot) product as follows:

$$\alpha_k = \mathbf{u} \cdot \mathbf{e}_k, \quad k = 1, \dots, n$$

Note that the vectors $\mathbf{e}_1, \dots, \mathbf{e}_n$ are:

- **normalized** — $\mathbf{e}_k \cdot \mathbf{e}_k = 1$ for all k
- **orthogonal** — $\mathbf{e}_k \cdot \mathbf{e}_m = 0$ for $k \neq m$



Example: \mathbf{u}, \mathbf{v} in \mathbb{R}^2

$$\mathbf{u} = 3\mathbf{e}_1 + 5\mathbf{e}_2$$

$$\mathbf{v} = 6.5\mathbf{e}_1 + 3\mathbf{e}_2$$

By definition, vectors are objects that can be added together and multiplied by scalars:

- if $\mathbf{u} = \sum_{k=1}^n \alpha_k \mathbf{e}_k$ and $\mathbf{v} = \sum_{k=1}^n \beta_k \mathbf{e}_k$, then we can form their sum

$$\mathbf{u} + \mathbf{v} = \sum_{k=1}^n (\alpha_k + \beta_k) \mathbf{e}_k$$

- if $\mathbf{u} = \sum_{k=1}^n \alpha_k \mathbf{e}_k$ and β is a scalar, then we can form the vector

$$\beta \mathbf{u} = \sum_{k=1}^n \beta \alpha_k \mathbf{e}_k$$

More generally, a collection of objects that can be added and/or multiplied by scalars is called a **vector space**.

We have already seen one example of a vector space — the space of continuous-time signals. We can easily verify:

- we can form the sum of any two signals $x_1(t)$ and $x_2(t)$ to get another signal

$$x(t) = x_1(t) + x_2(t)$$

- we can multiply any signal $x(t)$ by a constant c to get another signal

$$z(t) = cx(t)$$

Unlike the n -space \mathbb{R}^n , the vector space of all continuous-time signals is **huge**. In fact, it is infinite-dimensional.

The space of periodic signals

Let us consider all T -periodic signals. Any such signal $x(t)$ satisfies

$$x(t + T) = x(t) \quad \text{for all } t$$

for some given $T > 0$.

It is easy to see that T -periodic signals form a vector space:

- if $x_1(t)$ and $x_2(t)$ are T -periodic, then

$$x_1(t + T) + x_2(t + T) = x_1(t) + x_2(t),$$

so their sum is T -periodic

- if $x(t)$ is T -periodic, then

$$cx(t + T) = cx(t),$$

so any scaled version of $x(t)$ is also T -periodic

If we impose even more conditions on our T -periodic signals (the so-called **Dirichlet conditions**, which hold for all signals encountered in practice), then we can represent signals as infinite linear combinations of orthogonal and normalized (**orthonormal**) vectors.

Complete orthonormal sets of functions

First of all, we can define a scalar (or dot) product of two T -periodic signals $x_1(t)$ and $x_2(t)$ as

$$\langle x_1, x_2 \rangle = \int_0^T x_1(t)x_2(t)dt$$

(note that we can integrate over any whole period, not necessarily from $t = 0$ to $t = T$).

Then we can take any sequence of T -periodic functions (signals) $\phi_0(t), \phi_1(t), \phi_2(t), \dots$ that are

❶ **normalized:** $\langle \phi_k, \phi_k \rangle = \int_0^T \phi_k^2(t)dt = 1$

❷ **orthogonal:** $\langle \phi_k, \phi_m \rangle = \int_0^T \phi_k(t)\phi_m(t)dt = 0$ if $k \neq m$

❸ **complete:** if a signal $x(t)$ is such that

$$\langle x, \phi_k \rangle = \int_0^T x(t)\phi_k(t)dt = 0 \text{ for all } k, \text{ then } x(t) \equiv 0$$

Fourier series

Let $\{\phi_k(t)\}_{k=1}^{\infty}$ be a complete, orthonormal set of functions. Then any “well-behaved” T -periodic signal $x(t)$ can be uniquely represented as an infinite series

$$x(t) = \sum_{k=0}^{\infty} \alpha_k \phi_k(t).$$

This is called the **Fourier series** representation of $x(t)$. The scalars (numbers) α_k are called the **Fourier coefficients** of $x(t)$ (with respect to $\{\phi_k(t)\}$) and are computed as follows:

$$\alpha_k = \langle x, \phi_k \rangle = \int_0^T x(t) \phi_k(t) dt$$

In analogy to vectors in n -space, you can think of α_k as the **projection** (or component) of $x(t)$ in the direction of $\phi_k(t)$.

Fourier coefficients

To derive the formula for α_k , write

$$x(t)\phi_k(t) = \sum_{m=0}^{\infty} \alpha_m \phi_m(t)\phi_k(t),$$

and integrate over one period:

$$\begin{aligned} \underbrace{\int_0^T x(t)\phi_k(t)dt}_{\langle x, \phi_k \rangle} &= \int_0^T \left(\sum_{m=0}^{\infty} \alpha_m \phi_m(t)\phi_k(t) \right) dt \\ &= \sum_{m=0}^{\infty} \alpha_m \underbrace{\int_0^T \phi_m(t)\phi_k(t)dt}_{\langle \phi_m, \phi_k \rangle} \\ &= \alpha_k, \end{aligned}$$

where in the last line we use the fact that $\{\phi_k\}$ form an orthonormal system of functions.

Convergence of Fourier series

It can be proved that, because the functions $\{\phi_k\}$ form a complete orthonormal system, the partial sums of the Fourier series

$$x(t) = \sum_{k=0}^{\infty} \alpha_k \phi_k(t)$$

converge to $x(t)$ in the following sense:

$$\lim_{N \rightarrow \infty} \int_0^T \left(x(t) - \sum_{k=1}^N \alpha_k \phi_k(t) \right)^2 dt = 0$$

So, we can use the partial sums

$$x_N(t) = \sum_{k=1}^N \alpha_k \phi_k(t)$$

to approximate $x(t)$.

Trigonometric Fourier series

Fact: the sequence of T -periodic functions $\{\phi_k(t)\}_{k=0}^{\infty}$ defined by

$$\phi_0(t) = \frac{1}{\sqrt{T}} \quad \text{and} \quad \phi_k(t) = \begin{cases} \sqrt{\frac{2}{T}} \sin(k\omega_0 t), & \text{if } k \geq 1 \text{ is odd} \\ \sqrt{\frac{2}{T}} \cos(k\omega_0 t), & \text{if } k > 1 \text{ is even} \end{cases}$$

is complete and orthonormal. Here,

$$\omega_0 = \frac{2\pi}{T}$$

is called the **fundamental frequency**. Orthonormality is quite easy to show, completeness — not so much.

Thus, any well-behaved T -periodic signal $x(t)$ can be represented as an infinite sum of sinusoids (plus a constant term $\alpha_0\phi_0$):

$$x(t) = \sum_{k=0}^{\infty} \alpha_k \phi_k(t)$$

Trigonometric Fourier series

A more common way of writing down the trigonometric Fourier series of $x(t)$ is this:

$$x(t) = a_0 + \sum_{k=1}^{\infty} a_k \cos(k\omega_0 t) + \sum_{k=1}^{\infty} b_k \sin(k\omega_0 t)$$

Then the Fourier coefficients can be computed as follows:

$$a_0 = \frac{1}{T} \int_0^T x(t) dt$$

$$a_k = \frac{2}{T} \int_0^T x(t) \cos(k\omega_0 t) dt$$

$$b_k = \frac{2}{T} \int_0^T x(t) \sin(k\omega_0 t) dt$$

Recall that $\omega_0 = 2\pi/T$.

Trigonometric Fourier series

To relate this to the orthonormal representation in terms of the ϕ_k , we note that we can write

$$\begin{aligned}a_0 &= \frac{1}{\sqrt{T}} \int_0^T x(t) \phi_0(t) dt = \frac{1}{\sqrt{T}} \alpha_0 \\a_k &= \sqrt{\frac{2}{T}} \int_0^T x(t) \phi_{2k}(t) dt = \sqrt{\frac{2}{T}} \alpha_{2k} \\b_k &= \sqrt{\frac{2}{T}} \int_0^T x(t) \phi_{2k-1}(t) dt = \sqrt{\frac{2}{T}} \alpha_{2k-1}\end{aligned}$$

Thus, we have

$$\begin{aligned}x(t) &= a_0 + \sum_{k=1}^{\infty} a_k \cos(k\omega_0 t) + \sum_{k=1}^{\infty} b_k \sin(k\omega_0 t) \\&= \frac{1}{\sqrt{T}} \alpha_0 \cdot \sqrt{T} \phi_0(t) + \sqrt{\frac{2}{T}} \left(\sum_{k=1}^{\infty} \alpha_{2k} \cdot \sqrt{\frac{T}{2}} \phi_{2k}(t) + \sum_{k=1}^{\infty} \alpha_{2k-1} \cdot \sqrt{\frac{T}{2}} \phi_{2k-1}(t) \right) \\&= \sum_{k=1}^{\infty} \alpha_k \phi_k(t)\end{aligned}$$

Symmetry properties

Things to watch out for when computing the Fourier coefficients:

- if $x(t)$ is an **even** function, i.e., $x(t) = x(-t)$ for all t , then all its sine Fourier coefficients are zero:

$$b_k = \frac{2}{T} \int_{-T/2}^{T/2} x(t) \sin(k\omega_0 t) dt = 0$$

- if $x(t)$ is an **odd** function, i.e., $x(t) = -x(-t)$, then all its cosine Fourier coefficients are zero:

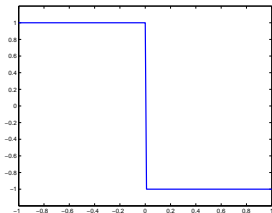
$$a_k = \frac{2}{T} \int_{-T/2}^{T/2} x(t) \cos(k\omega_0 t) dt = 0,$$

and

$$a_0 = \frac{1}{T} \int_{-T/2}^{T/2} x(t) dt = 0$$

Trigonometric Fourier series: example

Consider a 2-periodic signal $x(t)$ given by repeating the square wave:



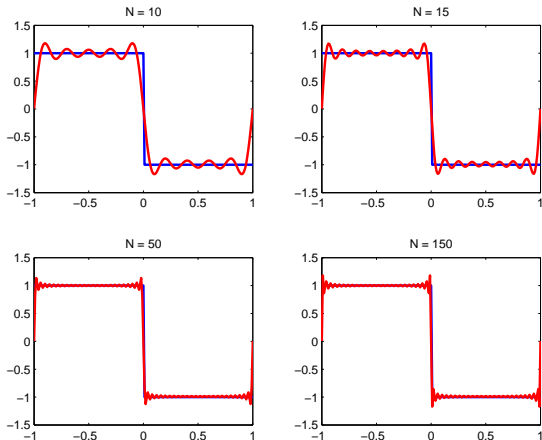
$a_0 = 0$ and $a_k = 0$ for all k
(the signal has odd
symmetry)

$$\begin{aligned} b_k &= \int_{-1}^0 \sin(k\pi t) dt - \int_0^1 \sin(k\pi t) dt \\ &= -\frac{1}{k\pi} [\cos(k\pi t)]_{-1}^0 + \frac{1}{k\pi} [\cos(k\pi t)]_0^1 \\ &= \frac{1}{k\pi} (\cos(-k\pi) - 1 + \cos(k\pi) - 1) \\ &= \frac{1}{k\pi} (2 \cos(k\pi) - 2) \\ &= -\frac{4 \sin^2(k\pi/2)}{k\pi} \end{aligned}$$

$$x(t) = -\sum_{k=1}^{\infty} \frac{4 \sin^2(k\pi/2)}{k\pi} \sin(k\pi t) = -\sum_{k \text{ odd}}^{\infty} \frac{4}{k\pi} \sin(k\pi t)$$

Trigonometric Fourier series: example

Approximating $x(t)$ by partial sums of its Fourier series:



Note the **Gibbs phenomenon**: the Fourier series (over/under)shoots the actual value of $x(t)$ at points of discontinuity. In signal processing, this effect is also called **ringing**.

Complex exponential Fourier series

Another useful complete orthonormal set is furnished by the complex exponentials:

$$\phi_k(t) = \frac{1}{\sqrt{T}} e^{jk\omega_0 t}, \quad k = \dots, -2, -1, 0, 1, 2, \dots$$

where $\omega_0 = 2\pi/T$, as before.

Note that these functions are complex-valued, and we need to redefine the dot product as

$$\langle x_1, x_2 \rangle = \int_0^T x_1(t) x_2^*(t) dt,$$

where $x_1^*(t)$ denotes the complex conjugate of $x_2(t)$. Then it is straightforward to show that

$$\langle \phi_k, \phi_m \rangle = \frac{1}{T} \int_0^T e^{jk\omega_0 t} e^{-jm\omega_0 t} dt = \begin{cases} 1, & k = m \\ 0, & k \neq m \end{cases}$$

Complex exponential Fourier series

Thus, we can expand any T -periodic $x(t)$ as

$$x(t) = \sum_{k=-\infty}^{\infty} c_k e^{jk\omega_0 t}$$

The Fourier coefficients are given by

$$c_k = \frac{1}{T} \int_0^T x(t) e^{-jk\omega_0 t} dt$$

To derive this, multiply the series representation of $x(t)$ on the right by $e^{-jk\omega_0 t}$ and integrate from 0 to T .

Symmetry properties for real signals

Consider a **real-valued** T -periodic signal $x(t)$. Then

$$c_k = \frac{1}{T} \int_0^T x(t) e^{-jk\omega_0 t} dt \quad \text{and} \quad c_{-k} = \frac{1}{T} \int_0^T x(t) e^{jk\omega_0 t} dt = c_k^*$$

Write c_k in polar form:

$$c_k = A_k e^{j\theta_k}, \quad A_k = |c_k|, \theta_k = \angle c_k$$

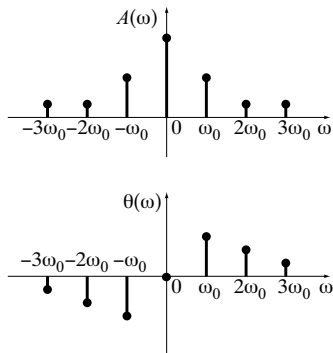
Then

$$c_k = c_{-k}^* \Rightarrow A_k = A_{-k} \text{ and } \angle c_k = -\angle c_{-k}$$

Thus, for real signals the **amplitude spectrum** A_k has even symmetry, while the **phase spectrum** θ_k has odd symmetry.

Amplitude and phase spectra

We can use the amplitudes A_k and the phases θ_k to represent the **spectrum** (or **frequency content**) of $x(t)$ graphically as follows:



Observe that the **amplitude spectrum** $A(\omega)$ has even symmetry, while the **phase spectrum** $\theta(\omega)$ has odd symmetry.