

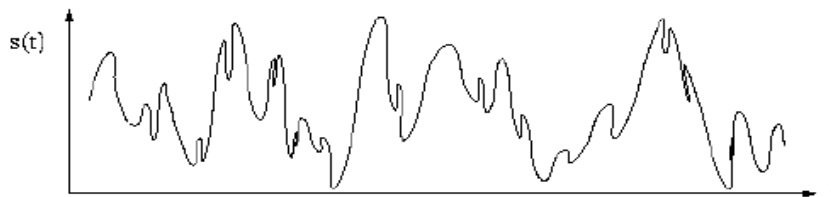
FUNDAMENTALS OF INFORMATION TRANSMISSION AND CODING (A.K.A. COMMUNICATION THEORY)

Signals and functions

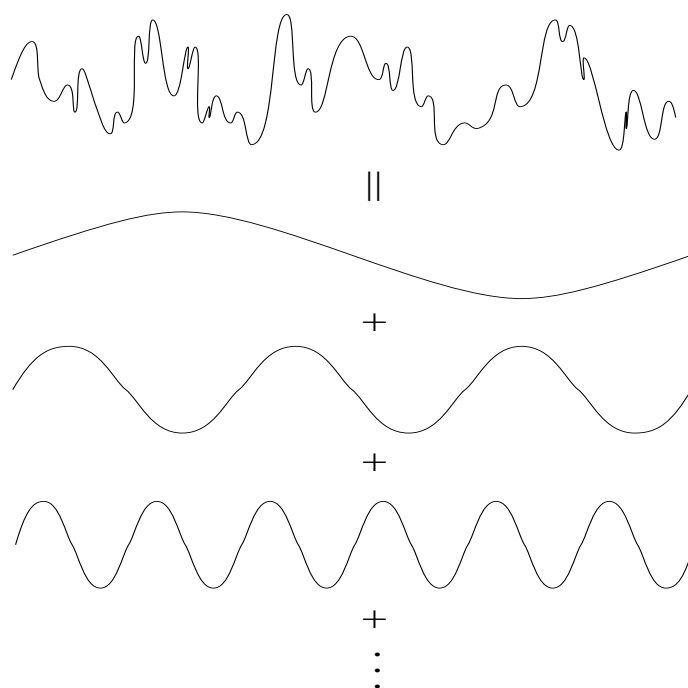
Elementary operation of communication: send *signal* on medium from A to B .

- media—copper wire, optical fiber, air/space, . . .
- signals—voltage and currents, light pulses, radio waves, microwaves, . . .
→ electromagnetic wave (let there be light!)

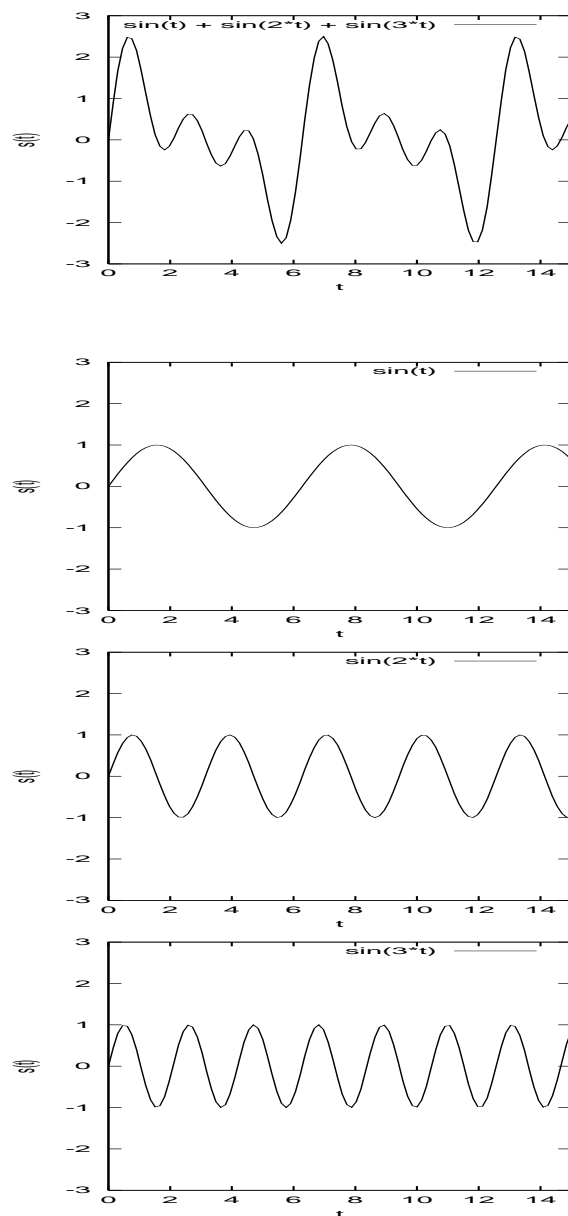
Signal can be viewed as a time-varying function $s(t)$.



If $s(t)$ is “sufficiently nice” (Dirichlet conditions) then $s(t)$ can be represented as a linear combination of complex sinusoids:



Simple example:



→ sinusoids form basis for other signals

Analogous to basis in linear algebra:

other elements can be expressed as linear combinations of “elementary” elements in the basis set

→ like atoms

Ex.: in 3-D, $\{(1, 0, 0), (0, 1, 0), (0, 0, 1)\}$ form a basis.

→ $(7, 2, 4) = 7 \cdot (1, 0, 0) + 2 \cdot (0, 1, 0) + 4 \cdot (0, 0, 1)$

→ coefficients: 7, 2, 4

→ spectrum

How many elements are there in a basis?

Vector spaces:

- finite dimensional
- infinite dimensional: signals

→ infinite number of bases

→ subject of functional analysis

Given an arbitrary element in the vector space, how to find the coefficient of basis elements?

→ e.g., given $(7, 2, 4)$, coefficient of $(0, 1, 0)$?

In linear algebra, matrix inversion:

$$A\mathbf{x} = \mathbf{y} \quad \Leftrightarrow \quad \mathbf{x} = A^{-1}\mathbf{y}$$

where A is $(n \times n)$ matrix, \mathbf{x} and \mathbf{y} are $(n \times 1)$ vectors.

→ solution techniques: e.g., Gaussian elimination

Note: the arbitrary vector \mathbf{y} (our “signal”) is represented as a linear combination

$$\mathbf{y} = A\mathbf{x} = x_1A_1 + x_2A_2 + \cdots + x_nA_n$$

where $\mathbf{x} = (x_1, x_2, \dots, x_n)$ and A_i is the i th column vector of A .

→ the A_i 's are the bases!

→ correct viewpoint of the world (for us)

For continuous (i.e., infinite dimensional) signals ...

Fourier expansion and transform:

$$s(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S(\omega) e^{i\omega t} d\omega,$$

$$S(\omega) = \int_{-\infty}^{\infty} s(t) e^{-i\omega t} dt.$$

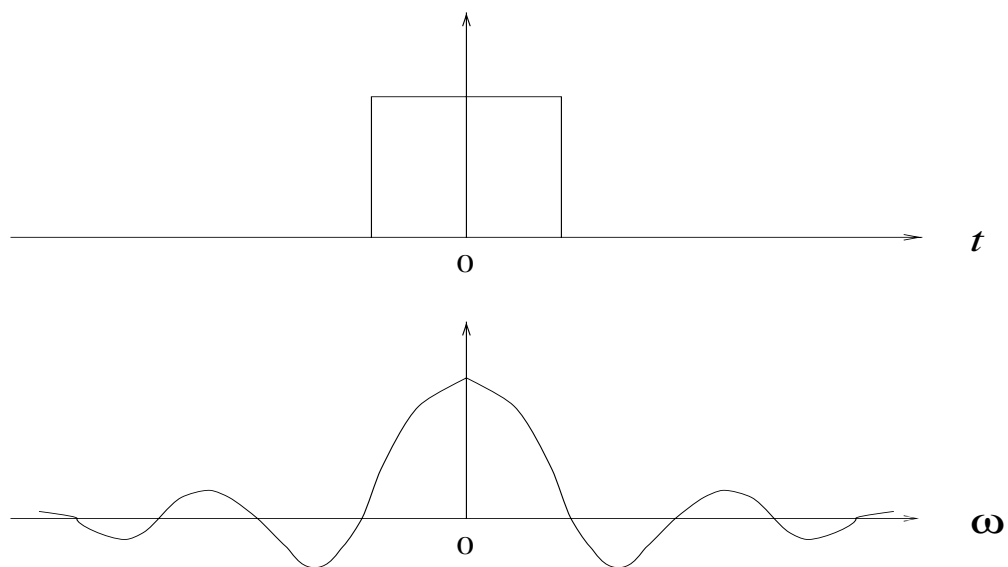
- recall: $e^{i\omega t} = \cos \omega t + i \sin \omega t$
- signal $s(t)$ is a linear combination of the $e^{i\omega t}$'s
- $S(\omega)$: coefficient of basis elements
- time domain vs. frequency domain

Frequency ω : cycles per second (Hz)

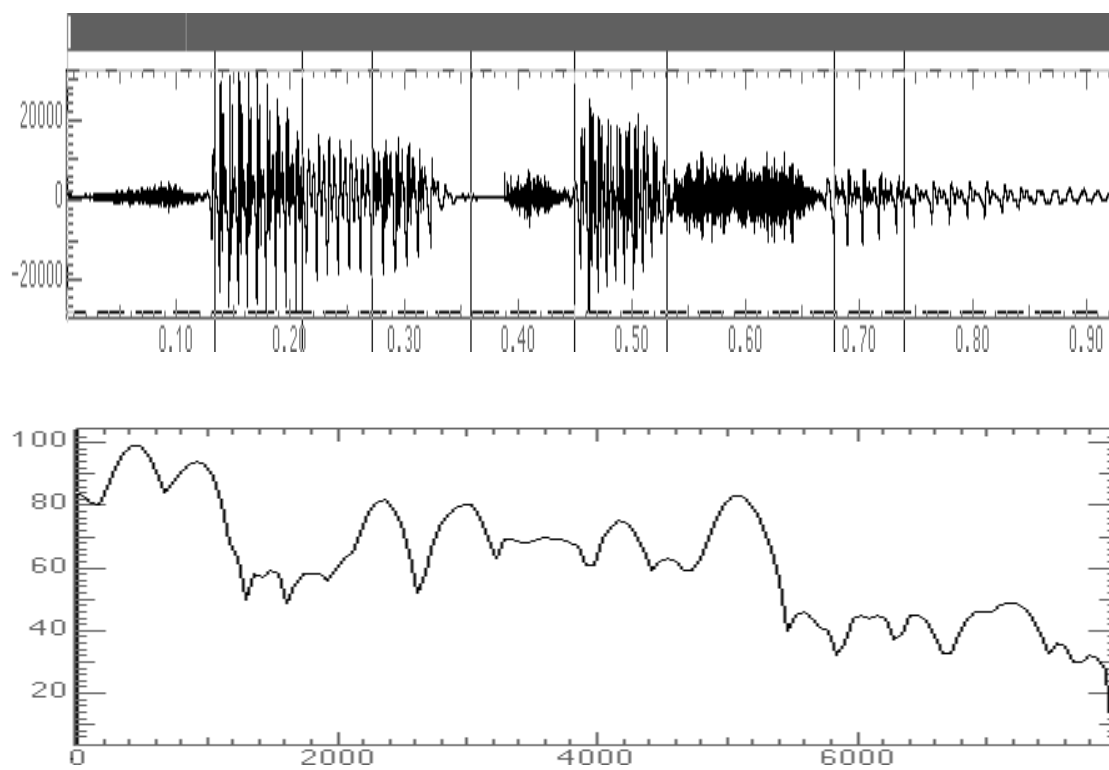
$$\longrightarrow \omega = 1/T$$

T : period of sinusoid

Example: square wave



Example: audio (e.g., speech) signal



Source: Dept. of Linguistics and Phonetics, Lund University

Random function (i.e., white noise) has “flat-looking” spectrum.

→ unbounded bandwidth

Why bother with frequency domain representation?

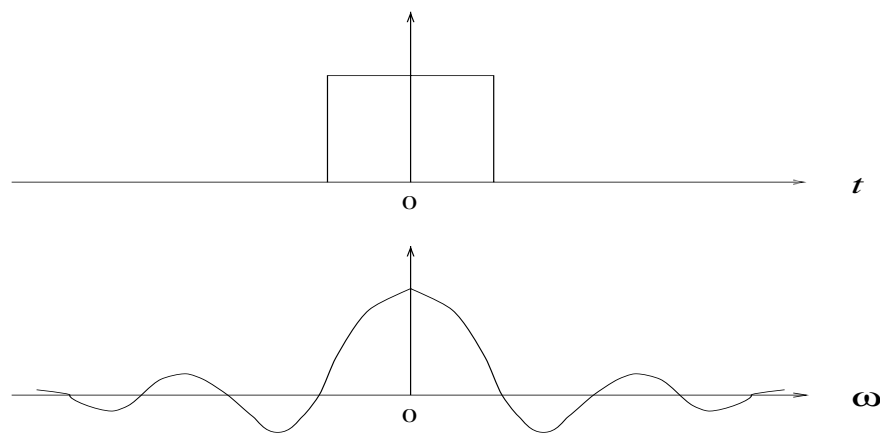
→ contains same information (invertible) ...

→ convenience

→ brings out “relevant” information

Luckily, most “interesting” functions arising in practice are “special”:

- bandlimited
- i.e., $S(\omega) = 0$ for $|\omega|$ sufficiently large
- when $S(\omega) \approx 0$, can treat as $S(\omega) = 0$
- let’s approximate!
- e.g., square wave



Ex.: human auditory system

- 20 Hz–20 kHz
- speech is intelligible at 300 Hz–3300 Hz
- broadcast quality audio; CD quality audio

Telephone systems: engineered to exploit this property

- bandwidth 3000 Hz
- copper medium: various grades
- no problem transmitting 3000 Hz signals

For communication:

Both absolute frequency and band**width** are relevant.

- baseband vs. broadband
- high-speed \Leftrightarrow broadband

Manipulate shape of different frequency sinusoids to **simultaneously** carry information (i.e., bits).

- multi-lane highway analogy
- different lane \Leftrightarrow different frequency

Manipulation of different frequencies can create complicated looking $s(t)$.

- side effect of encoding
- decoding: use Fourier transform