

Digital vs. Analog Data

Digital data: bits.

- discrete signal
- both in time and amplitude

Analog data: audio/voice, video/image

- continuous signal
- both in time and amplitude

Both forms used in today's network environment.

- burning CDs
- audio/video playback

In broadband networks:

- use analog data to carry digital data

Important task: analog data is often digitized

→ useful: why?

→ basics: digital signal processing

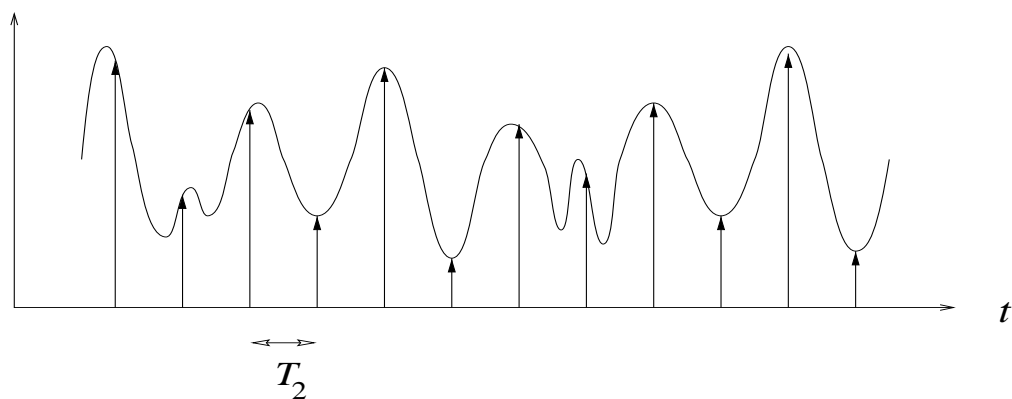
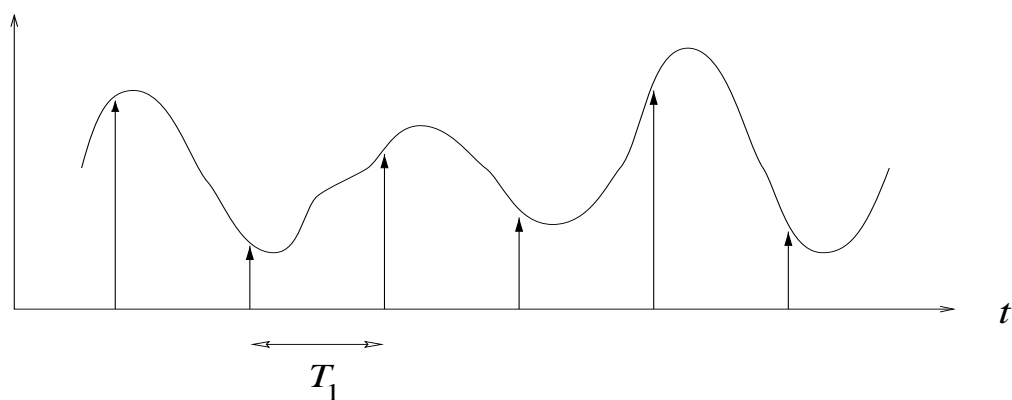
How to digitize such that digital representation is faithful?

→ sampling

→ interface between analog & digital world

Intuition behind sampling:

→ slowly vs. rapidly varying signal



If a signal varies quickly, need more samples to not miss details/changes.

$$\nu_1 = 1/T_1 < \nu_2 = 1/T_2$$

Are regularly spaced samples of fixed interval the best?

- perhaps “savings” possible
- the more samples, the more bits

Application: network probing

- goal: ascertain network state
- send sequence of probe packets
- from arriving probes infer/estimate congestion

But, do not want to disturb Schrödinger’s cat ...

- networking: minimize overhead
- irregular probing
- assurance: probabilistic

Sampling criterion for guaranteed faithfulness:

Sampling Theorem (Nyquist): Given continuous bandlimited signal $s(t)$ with $S(\omega) = 0$ for $|\omega| > W$, $s(t)$ can be reconstructed from its samples if

$$\nu > 2W$$

where ν is the sampling rate.

→ ν : samples per second

Issue of digitizing amplitude/magnitude ignored

→ problem of quantization

→ possible source of information loss

→ exploit limitations of human perception

→ logarithmic scale

Compression

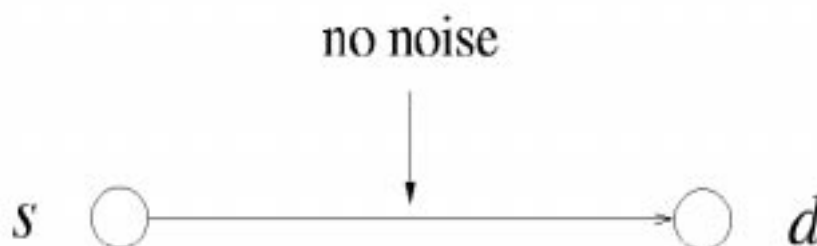
Information transmission over noiseless medium

→ medium or “channel”

Sender wants to communicate information to receiver over noiseless channel.

→ receive exactly what is sent

→ idealized scenario



Set-up:

- take a system perspective
- e.g., modem manufacturer

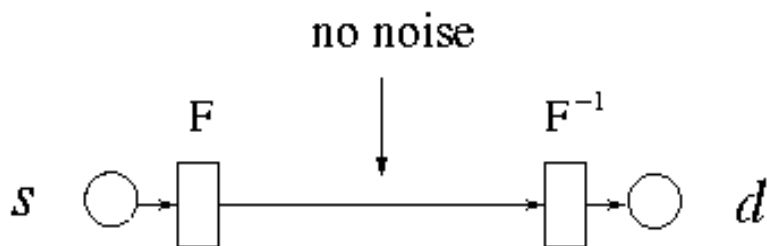
Need to specify two parts: property of source and how compression is done.

Part I. What does the source look like:

- source s emits symbols from finite alphabet set Σ
 - e.g., $\Sigma = \{0, 1\}$; $\Sigma =$ ASCII character set
- symbol $a \in \Sigma$ is generated with probability $p_a > 0$
 - e.g., books have known distribution for ‘e’, ‘x’ ...
 - let’s play “Wheel of Fortune”

Part II. Compression machinery:

- code book F assigns code word $w_a = F(a)$ for each symbol $a \in \Sigma$
 - w_a is a binary string of length $|w_a|$
 - F could be just a table
- F is invertible
 - receiver d can recover a from w_a
 - F^{-1} is the same table, different look-up



Ex.: $\Sigma = \{A, C, G, T\}$; need at least two bits

- F^1 : $w_A = 00$, $w_C = 01$, $w_G = 10$, $w_T = 11$
- F^2 : $w_A = 0$, $w_C = 10$, $w_G = 110$, $w_T = 1110$

→ pros & cons?

Note: code book F is not unique

→ find a “good” code book

→ when is a code book good?

Performance (i.e., “goodness”) measure: average code length L

$$L = \sum_{a \in \Sigma} p_a |w_a|$$

→ average number of bits consumed by given F

Ex.: If DNA sequence is 10000 letters long, then require on average $10000 \cdot L$ bits to be transmitted.

→ good to have code book with small L

Optimization problem: Given source $\langle \Sigma, \mathbf{p} \rangle$ where \mathbf{p} is a probability vector, find a code book F with least L .

A fundamental result on what is achievable to attain small L .

Entropy H of source $\langle \Sigma, \mathbf{p} \rangle$ is defined as

$$H = \sum_{a \in \Sigma} p_a \log \frac{1}{p_a}$$

Ex.: $\Sigma = \{A, C, G, T\}$; H is maximum if $p_A = p_C = p_G = p_T = 1/4$.

→ when is it minimum?

Source Coding Theorem (Shannon): For all F ,

$$H \leq L.$$

Moreover, L can be made to approach H .

Remark:

- To approach minimum H use blocks of k symbols
→ extension code
- entropy is innate property of source s
- Ensemble limitation
→ e.g., sending number $\pi = 3.1415927\dots$
→ better way?