



Analog to digital conversion

Sampling and quantization

take physics quantity (analog) ~~sample~~ have it signal sampled in T_c
(in instance n 's time)

Analog to digital conversion

Largely used in modern digital devices, music, videos, ... are all converted in digital form

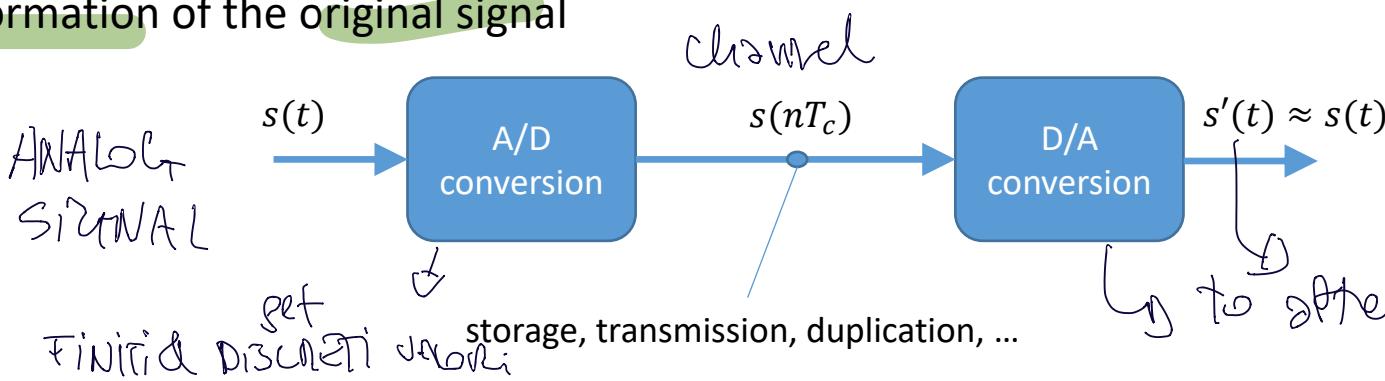
In digital form it is possible to process, manage and transmit information directly via software
without special purpose HW

- without special purpose hardware
- without altering the signal !!! \rightarrow sampling is robust

A digital signal can be converted again into an analog signal (D/A conversion)

- It is important that the A/D process does not lose too much information of the original signal

- Analog to digital conversion transforms an analog signal $s(t)$ into a sequence of integer numbers
- This numeric sequence can be stored, processed and transmitted by a digital device
 - To this purpose it is, in turn, transformed into a binary stream.
- The elements of A/D conversion are **sampling** and **quantization**



100
to other, reconstruct the signal

Sampling and quantization

Sampling

↳ consider continuous signal in interval of time

- Sampling a signal $s(t)$ means extracting from the signal a sequence of numeric values that the signal assumes at discrete time intervals
- Usually, sampling is performed at uniformly spaced time intervals, spaced of T_c i.e. the **sampling period**
- Sampling returns a stream of real numbers...

se il periodo di campionamento è T_c
→ prendi i valori del segnale a multipli
di $T_c \Rightarrow nT_c$

↳ some errors,
can't represent
all symbols

From cont. dom.
represent subset of
value of

Quantization

↳ non rappresentabile
REASON
MAPPING

- Transforms the stream of real numbers produced by the sampling into a stream of numbers:
 - either floating point or integers
 - that can be managed by a digital device
- While sampling may preserve the original signal (under some hypothesis), quantization implies a loss of information
 - i.e. after quantization it is not possible to «perfectly» reconstruct the original signal

how to represent

floats point representation
or int → want integer

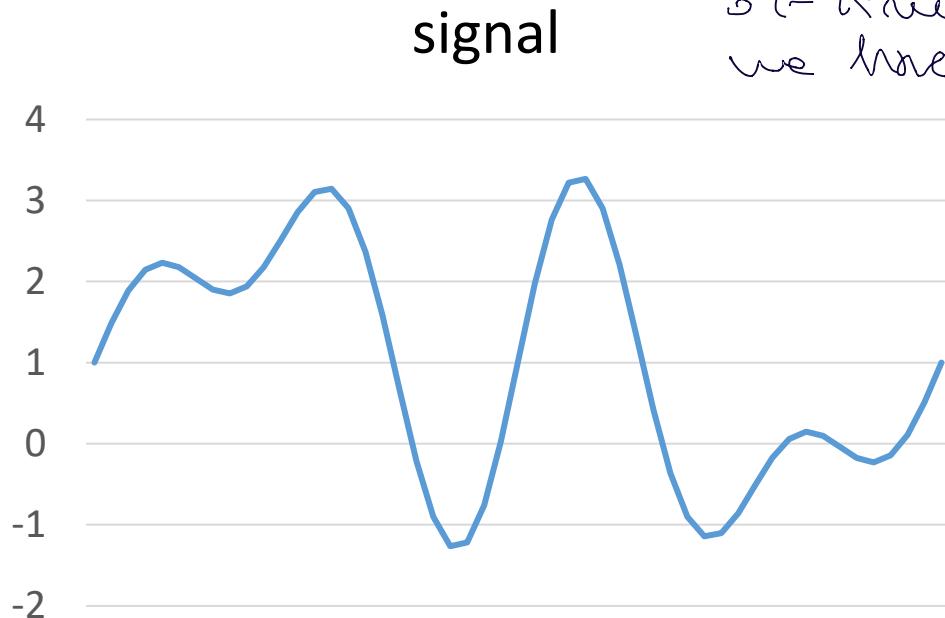
finite
bits

103

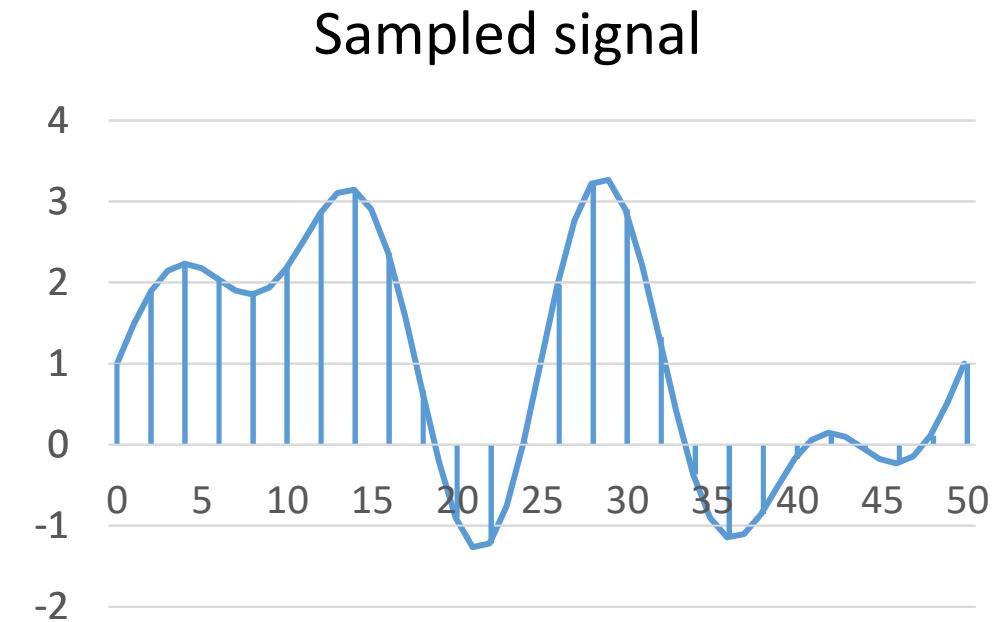
Sampling continuous signals

INPUT
A discrete signal $g(nT)$, $n \in \mathbb{Z}$ can be obtained with a uniform sampling of a continuous signal $s(t)$, where:

$$g(nT) = s(nT)$$



T is the sampling period
 $\frac{1}{T}$ is the sampling frequency

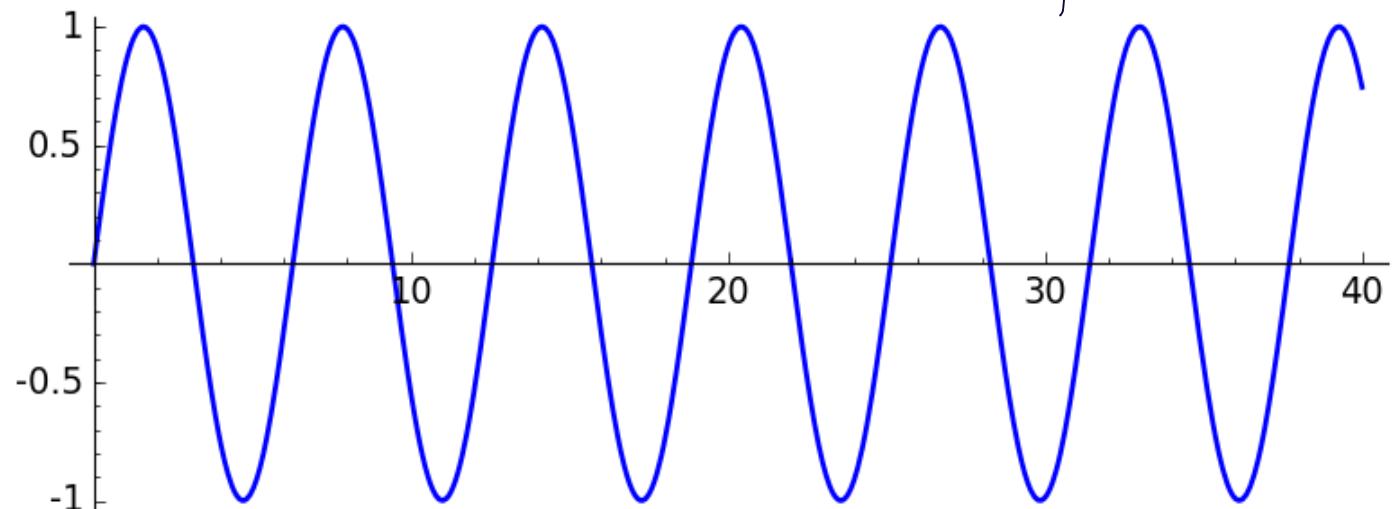


Example

Represent a signal $s(t)$ by its samples, taken at discrete times

- a sample is a value of the signal at a precise time

$$s(t) = \sin(t), \text{ frequency } f = \frac{1}{2\pi} \approx 0.159$$



$$s(nT_c) \text{ with } T_c = 3.09$$

The sampling is a signal itself, $s(nT_c)$, which takes a value for every multiple of the sampling period T_c

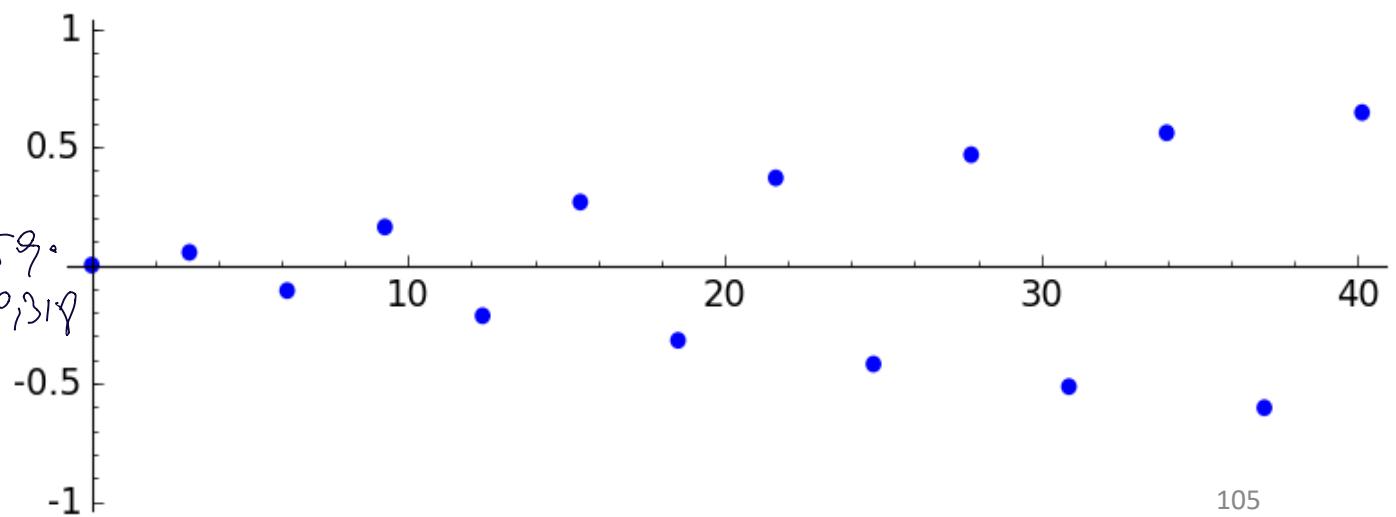
$$T_c = 3.09 \text{ implies: } f_c \approx 0.324 > 2f \approx 0.318$$

freq. of signal
2 = 0.318

focus on sample

on in sampling period

$$f_c = \frac{1}{T_c} = \frac{1}{3.09} \approx 0.324 \text{ or } 3.09$$

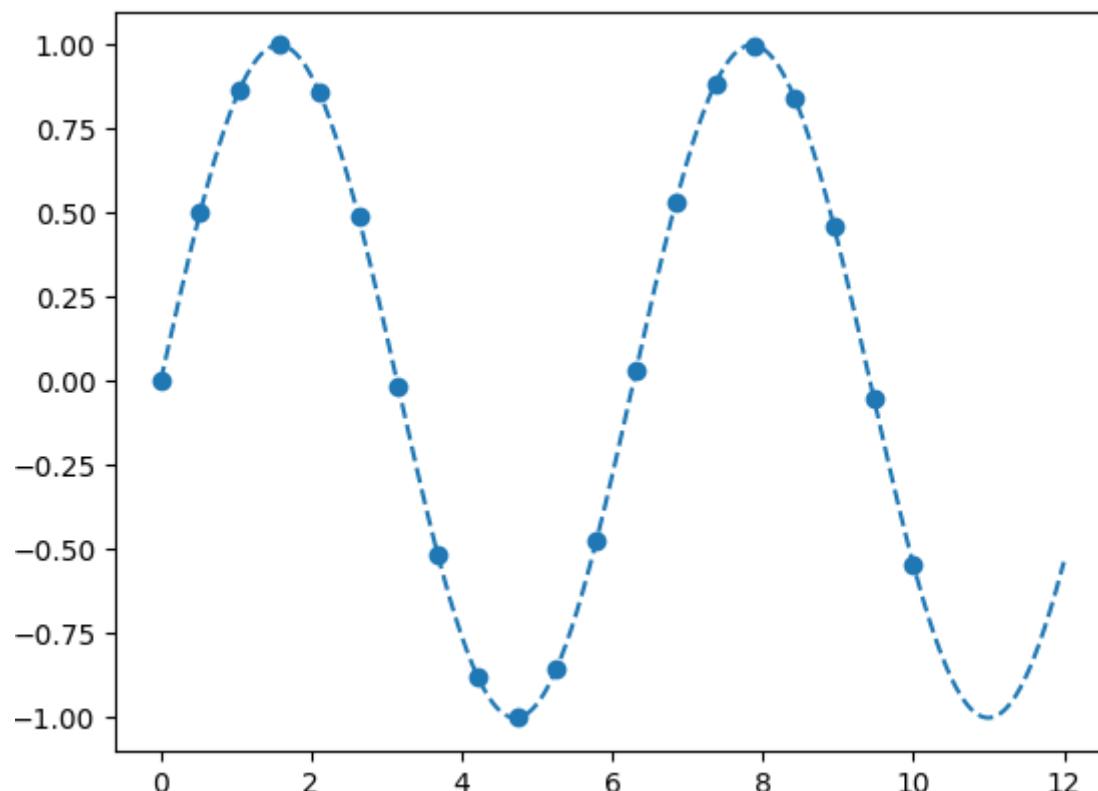


Sampling and reconstruction

Sampling Theory responds ↗

Is it possible to reconstruct a continuos signal $s(t)$ from a sequence of samples?

⇒ Ricomporre un segnale continuo
da un insieme
di samples



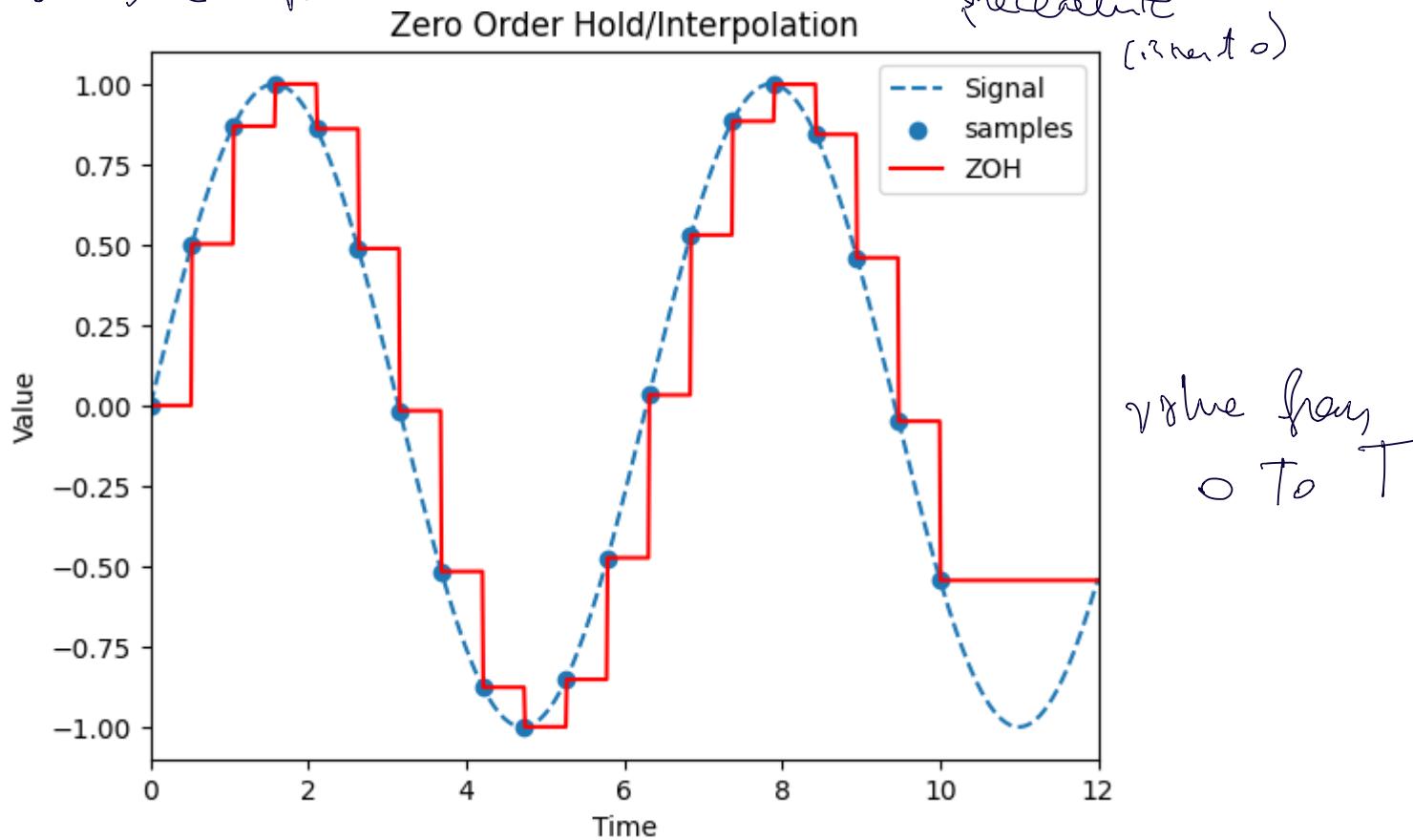
- QUANDO faccio SAMPLING
ho solo i PUNTI, devo
Ricomporre
- ⇒ IN INTERPOLATION

To reconstruct signal \rightarrow we have infos about just the points

\Rightarrow apply interp. to reconstruct the signal continuous

Interpolation - examples

allo stesso tempo ϕ : decidi valori da dare i valori
delle curve del segnale da $t = T$ \Rightarrow prendo il valore del sample
precedente
(. next o)



- Several methods for interpolation
- Several ways for reconstructing the signal:
 - Piecewise-constant signal, e.g.:
 - Zero-order hold
 - Nearest neighbour
 - Piece-wise linear, e.g.:
 - First-order hold reconstruction
 - Etc..

\Rightarrow I'm reconstructing signal approx

from a linear behaviour
2 consecutive samples

The Sampling Theorem

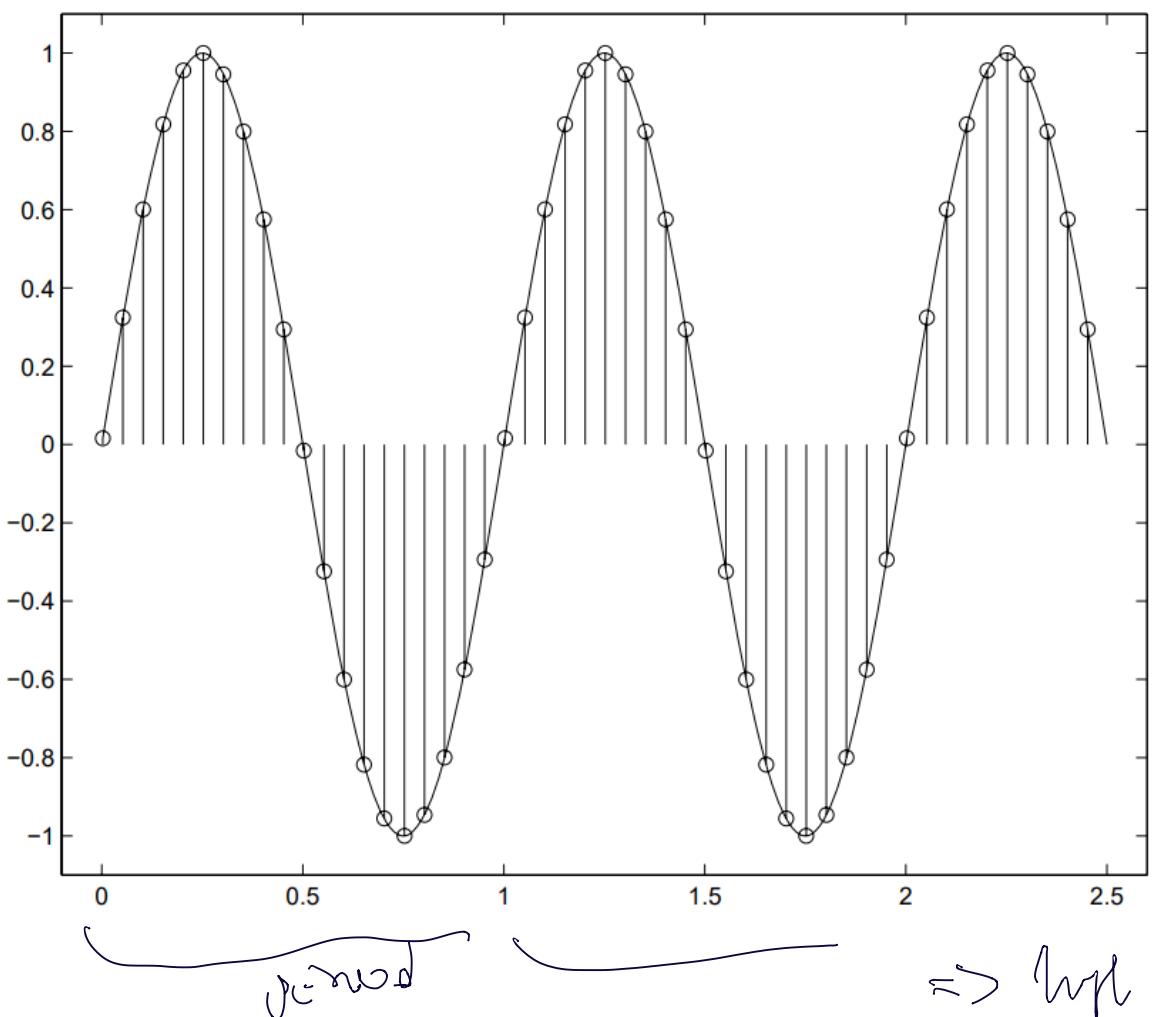
Se lo un segnale che cambia rapidamente
=> mi servono tanti samples,
Se cambia lentamente
=> pochi sample

Basic idea: a signal that changes rapidly will need to be sampled much faster than a signal that changes slowly

The sampling Theorem tells us the minimum sampling frequency required to reconstruct the signal... in ideal conditions...

sample so more often in
to reconstruct the signal

Example: Sampling a sinusoid at a high rate

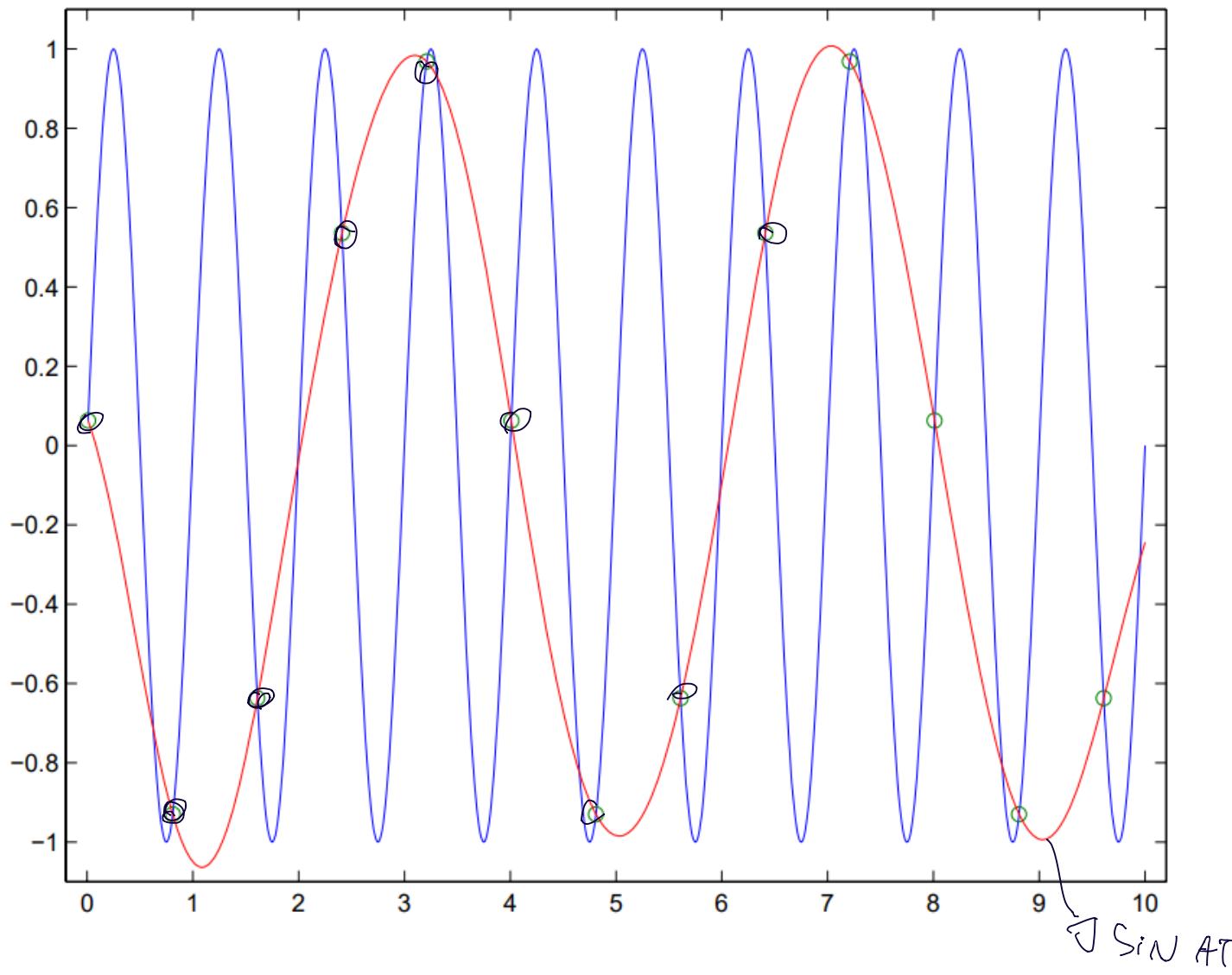


$$x(t) = \sin \frac{2\pi}{T} t$$

Samples in the example are taken
at a frequency $f_c \gg \frac{1}{T}$

$s(t) \Rightarrow$ blue ; Sampling at low rate \Rightarrow frequenza di campionamento $f_s < 2$ freq. del segnale
 \Leftrightarrow Aliasing

Example: Sampling a sinusoid at a low rate



If we sample (the blue signal) at a frequency which is too low

- we may not be able to reconstruct the signal
- We might reconstruct the lower frequency signal (red sinusoid)
- Aliasing:
 - different signals indistinguishable when sampled.
 - It depends on the sampling rate and frequency content of the signal

, L'unico \Rightarrow can sample it at low rate

Freq. anal. low than freq. d'input
 \Rightarrow don't represent just the signal

The Sampling Theorem

value
(recont)
of
Cont signal

to a discrete signal (composed by samples)

low freq. \Rightarrow effect of sample at low freq
 \Rightarrow ALIASING

Consider a signal $s(t)$ with spectrum null at frequencies above f_M

- ... hence the band of frequencies of $s(t)$ is limited \Rightarrow LIMITED in BAND (important)
- f_M is the maximum frequency of the signal

$s(t)$ is completely represented by its samples taken:

- at regular intervals $t_n = nT_c$ (with $n \in \mathbb{Z}$)
- with a sampling period of $T_c \leq \frac{1}{2f_M}$ "Nyquist"

i.e. from these samples it is possible to reconstruct $s(t)$ for any t

$\text{if freq. period} \leq \frac{1}{2f_M} \Rightarrow$ sampling the signal at freq $> \frac{1}{2}$ max freq of the signal

$$f_{c_{\min}} = \frac{1}{T_{c_{\max}}} = 2f_M$$

is called the **Nyquist frequency**

The Sampling Theorem

f reconstruct \Rightarrow interpolation; start from e sample

The Nyquist frequency $f_{c_{min}}$ is the **minimum** frequency at which to sample a signal $s(t)$ limited in band

$f_{c_{min}}$ equals the double of the maximum frequency in $s(t)$

Under this condition $s(t)$ can be reconstructed using interpolation, for example (based on cardinal sin):

multiple sampling period \rightarrow

$$s(t) = \sum_{n=-\infty}^{\infty} s(nT_c) \cdot \frac{\sin(\pi \cdot f_c \cdot (t - nT_c))}{\pi \cdot f_c \cdot (t - nT_c)}$$

Frequency of samples

RICHTWISCHER REGLA CONVOLUTIO SUMATORIA INFINITA

A sketch of the proof can be found in [1] DEI MIEI SOEPPEL

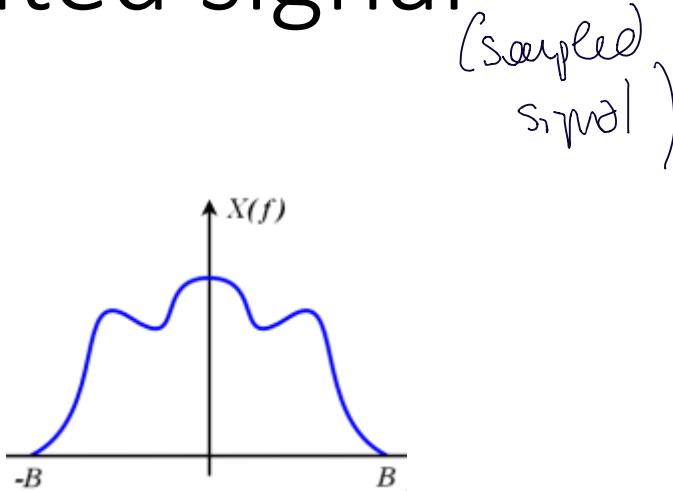
multiple

Sampling period

$s(nT_c) \cdot$ CARDINAL SIN
 $(t - nT_c)$ di

FT of a band limited signal

- Il segnale nel tempo, il suo spettro è limitato nella banda

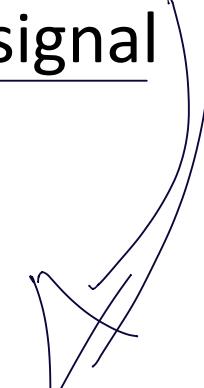


Fourier Transform of a band limited signal

come

- when a signal $x(t)$ is sampled, the Fourier transform of the sampled signal consists of shifted and scaled copies of $X(f)$

• Ho un segnale continuo
=> campioni => segnale discreto
=> Fanno la FT del
segnale campionato
=> ottego (per quanto
riguarda la trasformata
in frequenze)
=> il segnale componenti
ha uno spettro
formato da un
SET di COPIE dello
spettro del segnale continuo
(discrete)



FT of a band limited signal - sampled

$\text{FT} \text{ is a } \sum \Rightarrow \text{repliche simile}$
 $\Rightarrow \text{se copiano le frequenze}$

- The spacing between the replicas of $X(f)$ depends on the sampling rate
 - the faster we sample, the further apart are the replicas of $X(f)$.**
 - If we sample every T seconds then the spacing between the replicas in frequency domain is equal to the sampling frequency $f_s = 1/T$

$f_s \geq \text{Max freq of signal}$

\downarrow
means

le distanze fra le repliche (copie delle spettri) sono

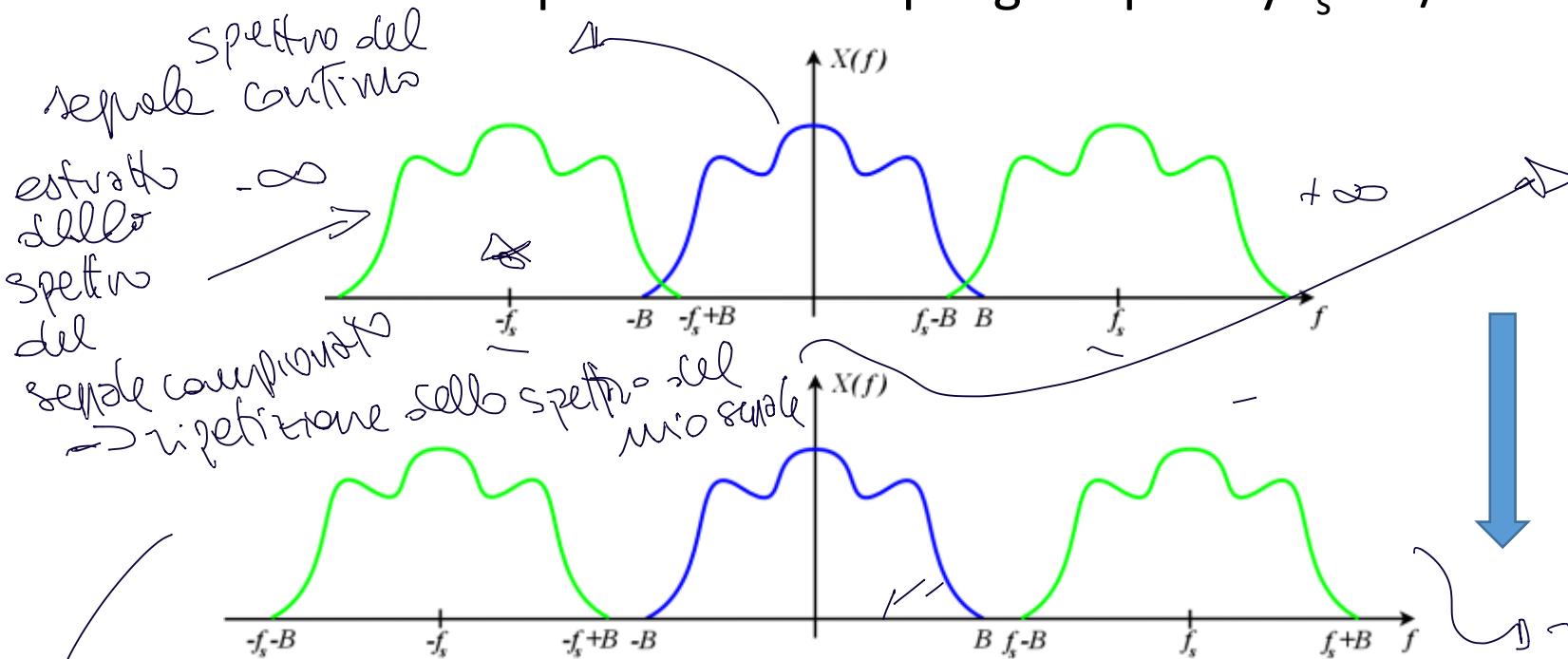
$m f_s$

NO OVERLAPPING

○ centra i multipli delle frequenze di campionamento

Increasing the sampling frequency

represents il segnale continuo nel tempo



FT

time

freq.
Densit

f d_{ig} operator

represent the

Surf contained

Aliasing

By ignoring what is between samples, the sampling process throws away lots of information

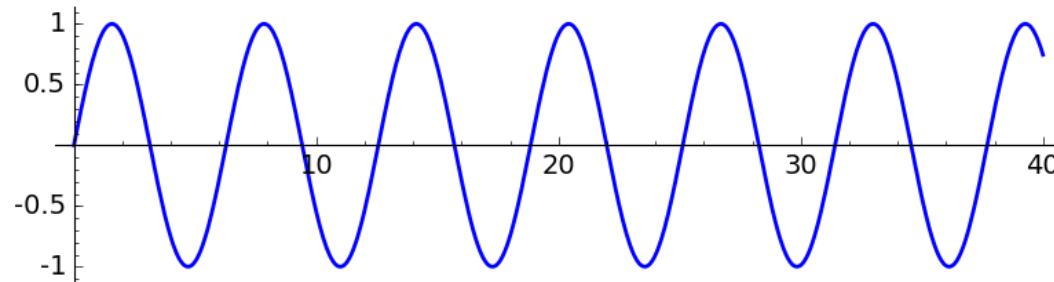
- ... however, if the signal is perfectly band-limited then no information is lost!
- ... you should consider this in your system design...

Se compriso a serv
frequency

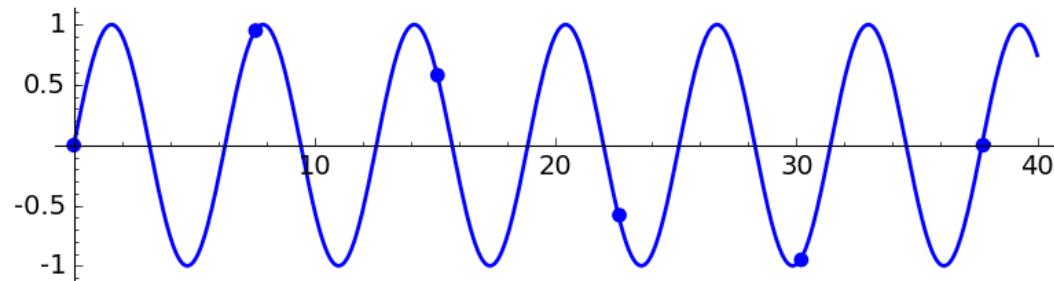
No overlapping

⇒ ALIASING

Consider for example a sinusoidal signal of frequency $f = 1/2\pi$:



If you know the frequency you can predict precisely the sampled signal:



- Mo vu segnale
- Campiona il segnale
e ricostruisce il
segnale SE SO
la frequenza del
nuovo segnale

Aliasing

- ⇒ Se campiona a bordo
DADF ⇒ DAD BW
- ⇒ Tengo il verde che
sono diversi non
si ricomposta
- ⇒ si ricomposta

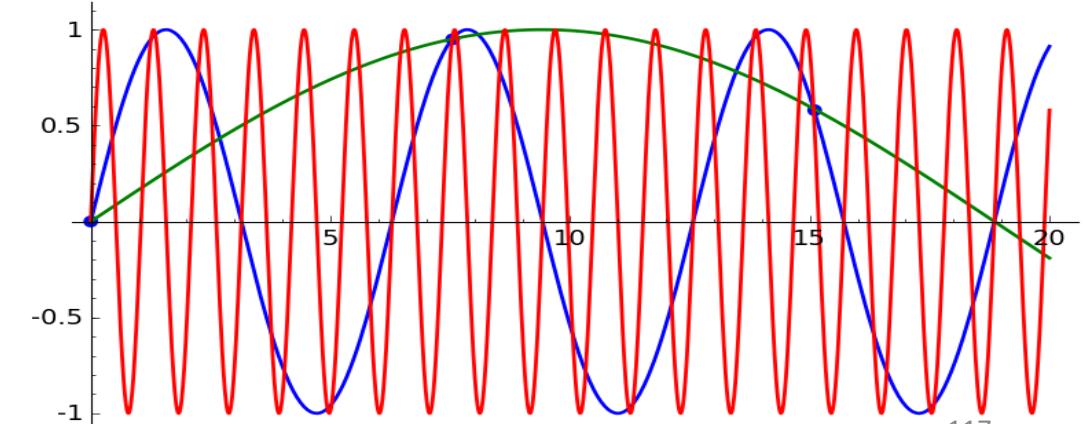
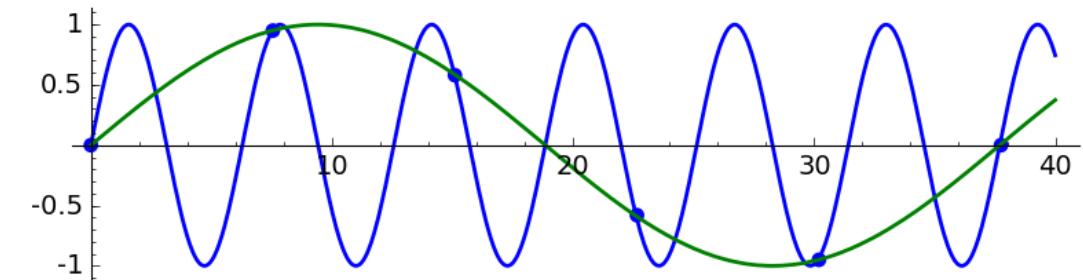
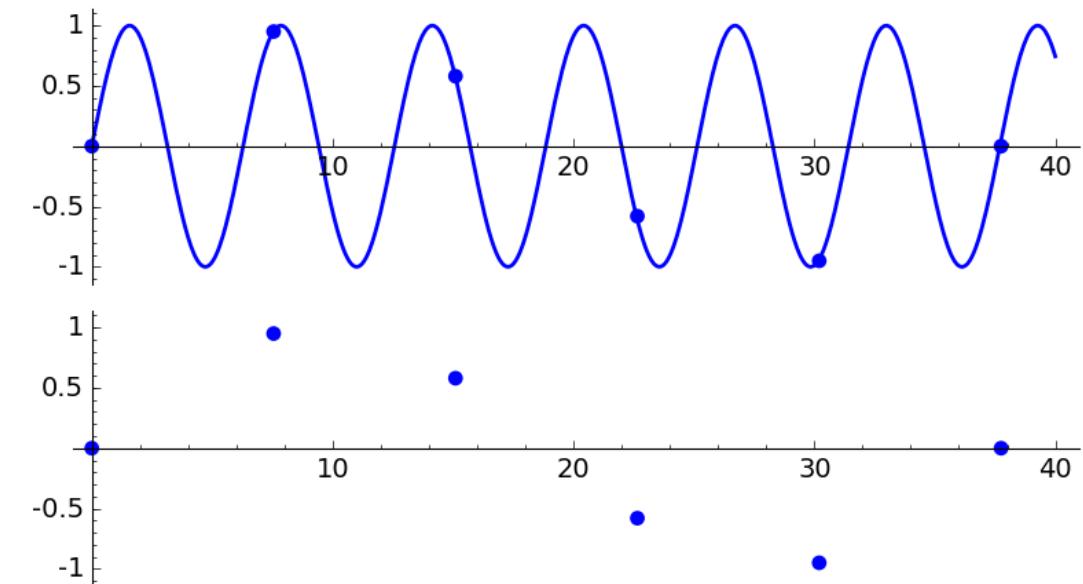
... but from the sampled signal alone you cannot necessarily find the frequency...

If you take another sinusoid at frequency $f' = f/6$ that's what you get:

This is known as the **aliasing** effect ...

- ⇒ possono ricomporre
solo il verde
- ⇒ no rosso o BW

classificazione d'interferometri → ripetere dello spettro centrale

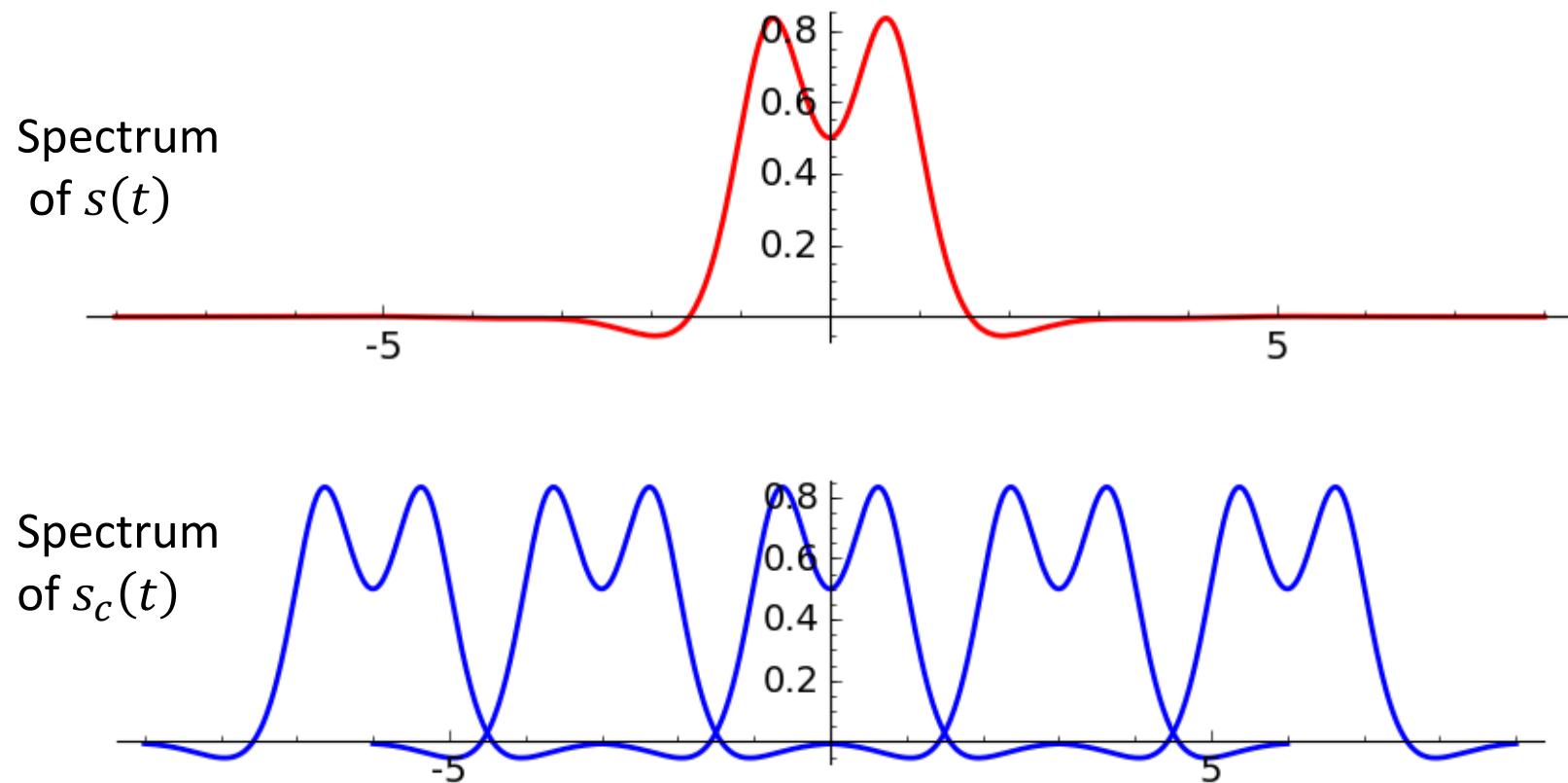


Aliasing

in multipli dello Spettro

The example seen before when sampling a sinusoid holds also for more complex signals.

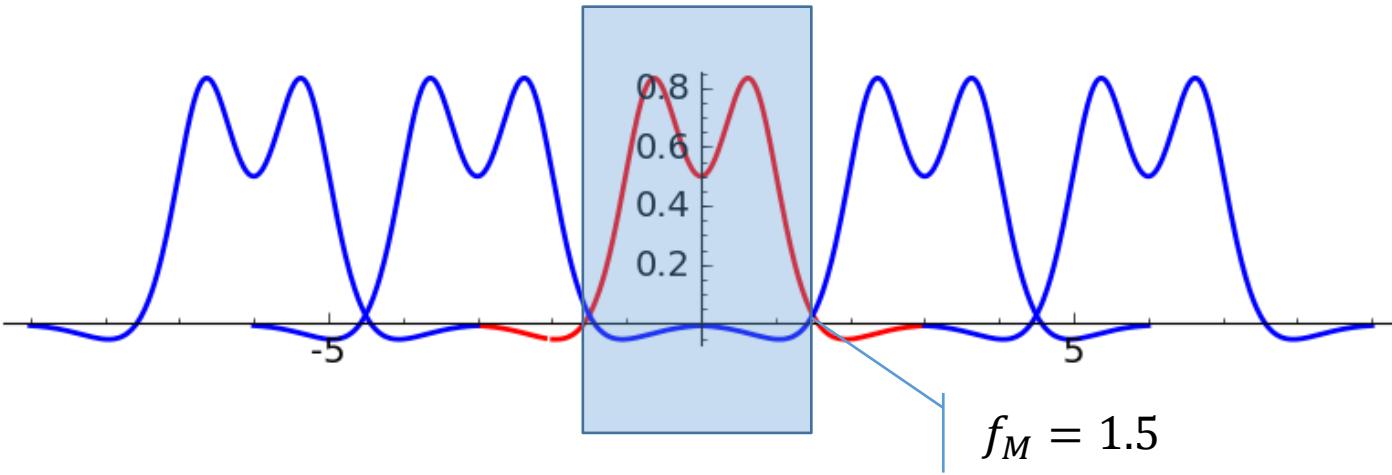
In general, the spectrum of the sampled signal $s_c(t)$ consists of infinite replicas of the spectrum of $s(t)$ centered at frequencies multiple of $f_c = 3$



Aliasing

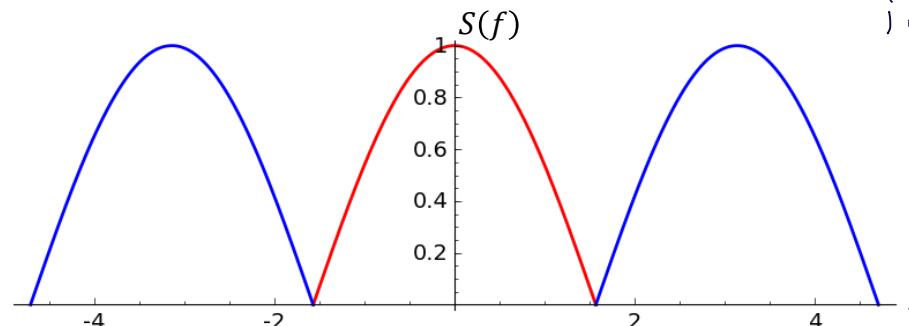
That's why Nyquist requires that $s(t)$ is limited in band!

If we limit the signal in the example at frequency $f_M = 1.5$, we can disregard all the replicas (which have higher frequency):



Hence, if $s(t)$ is strictly limited in band (i.e. it has a spectrum null at frequencies above f_M), we can choose f_c to ensure that the replicas of the spectrum of $s(t)$ do not overlap.

IDEAL
=> next difficult
even regular
limit to well
border



IDEAL : BUT HARD
more signal
limited in
band
with very limited

$$s(t) = \sin(t), \text{ with frequency } f = 1/2\pi \sim 0.159$$

Example

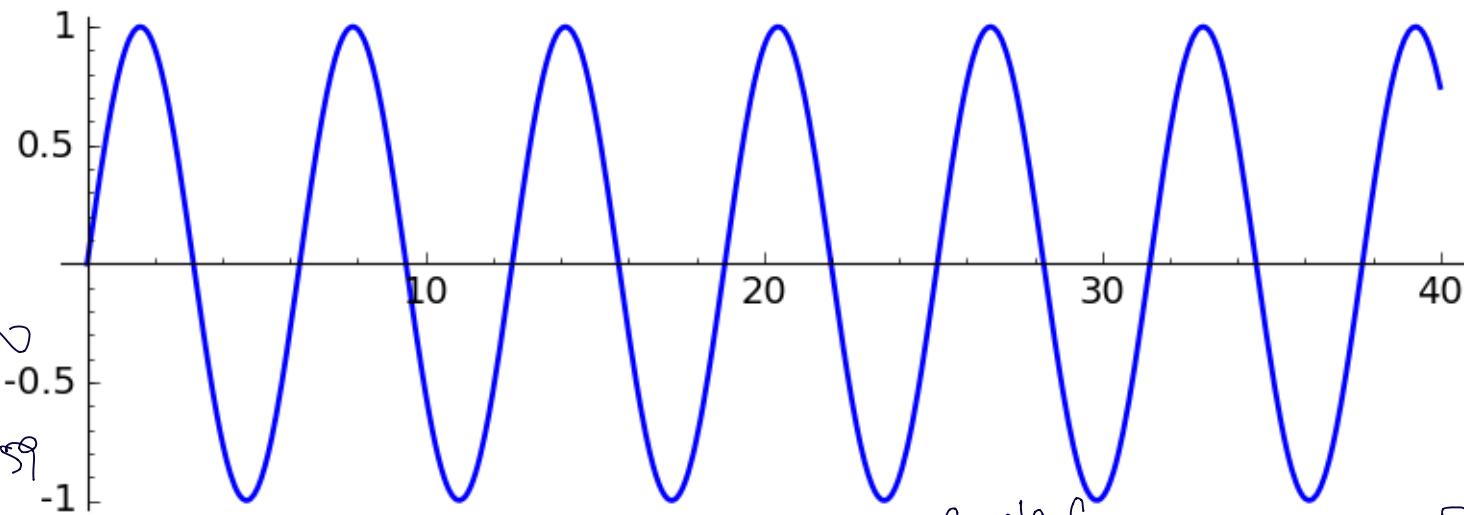
- Sampling frequency ci dà

un punto per segnale

La frequenza di campionamento

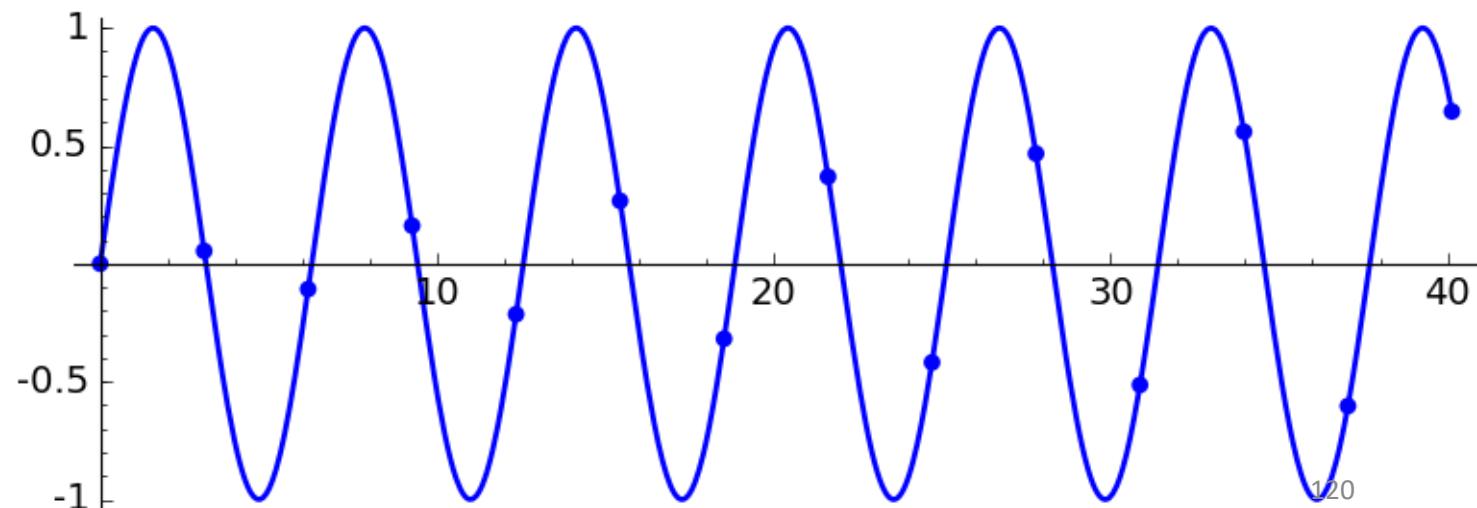
- Se segnale $s(t)$ $f = \frac{1}{2\pi} = 0.159$

ad una frequenza $f_c = 0.324 > 2f$



$s(nT_c)$ with $T_c = 3.09$; hence $f_c \sim 0.324 > 2f$

→ Sample freq
→ freq of
the signal
 $s(t)$



- X riconosci il segnale considerando
un finito numero di campioni

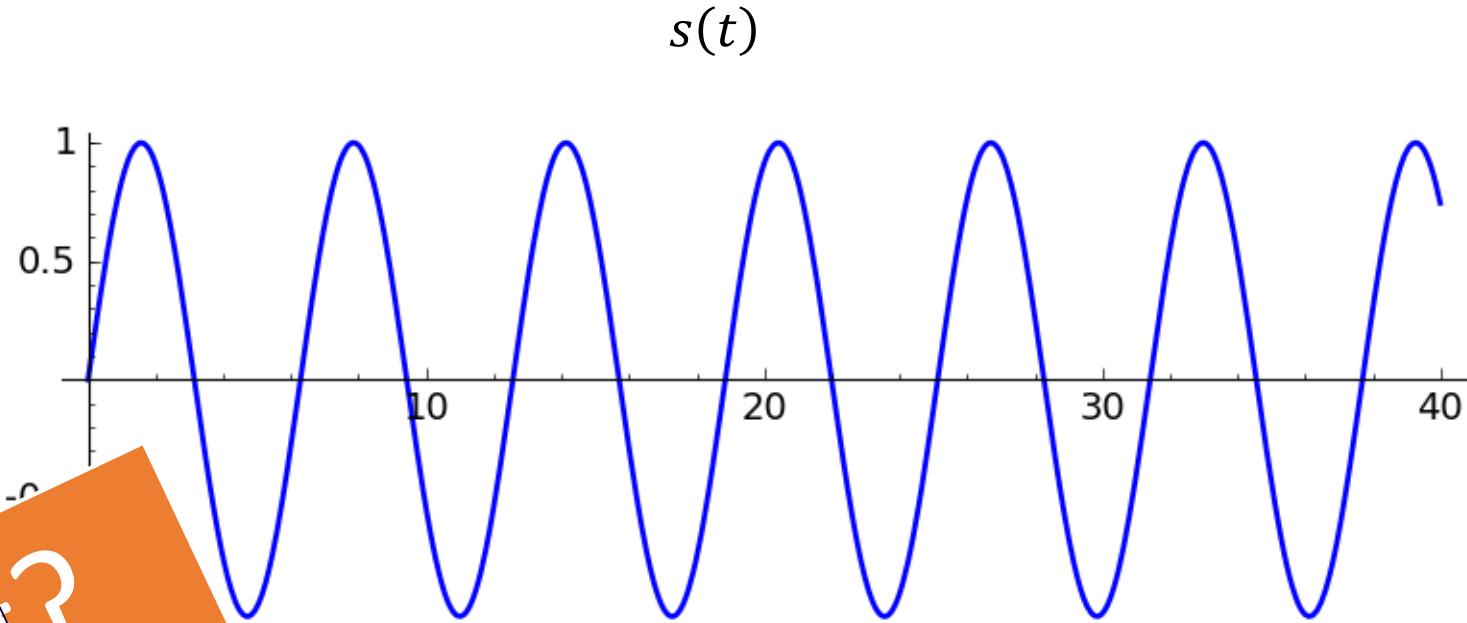
Example

($-15 \leq t \leq 15$)

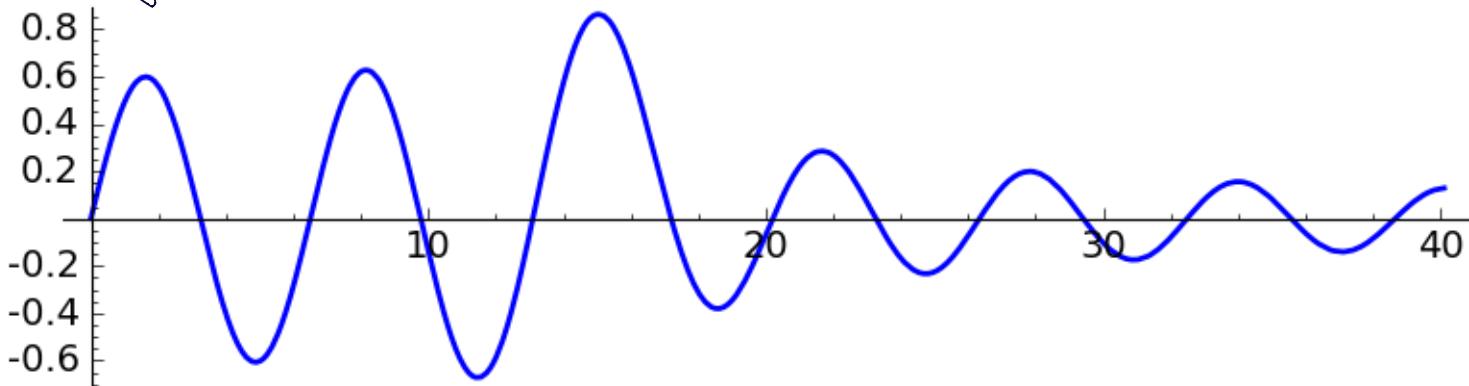
ORINGO

What's wrong?

• NUMERO DI CAMPIONI
 \Rightarrow perché il segnale
non è zero per
t > 15



$s(t)$ from interpolation (from -15 to $+15\dots$)



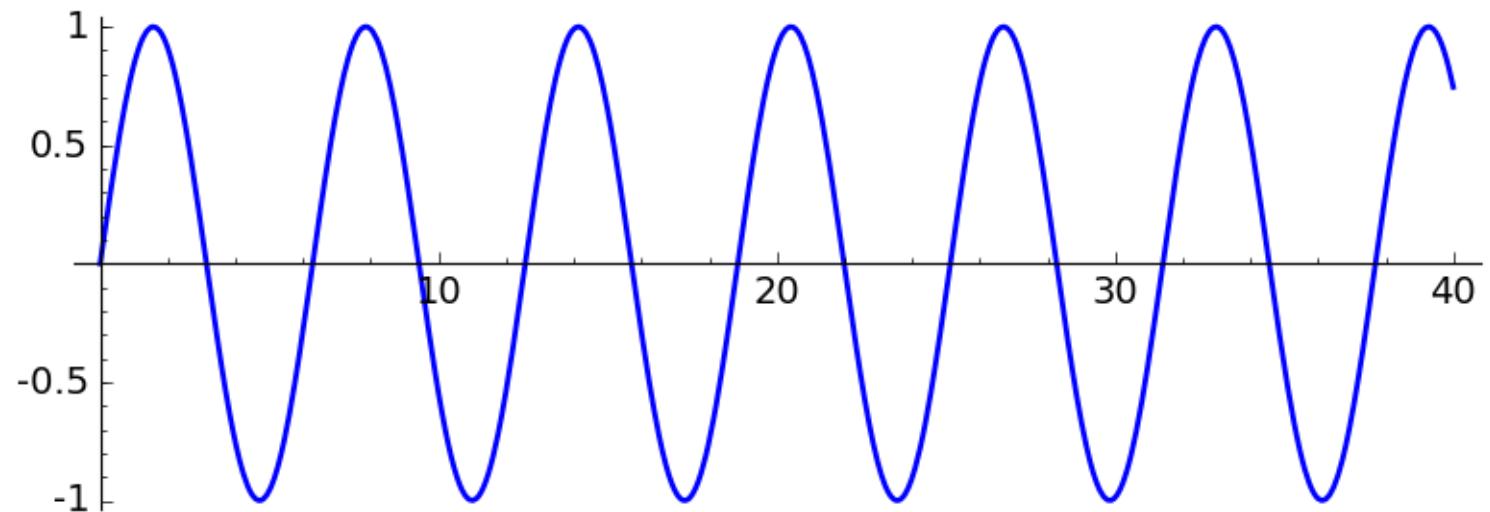
Example reloaded

Let's start again
increasing the sampling
rate...

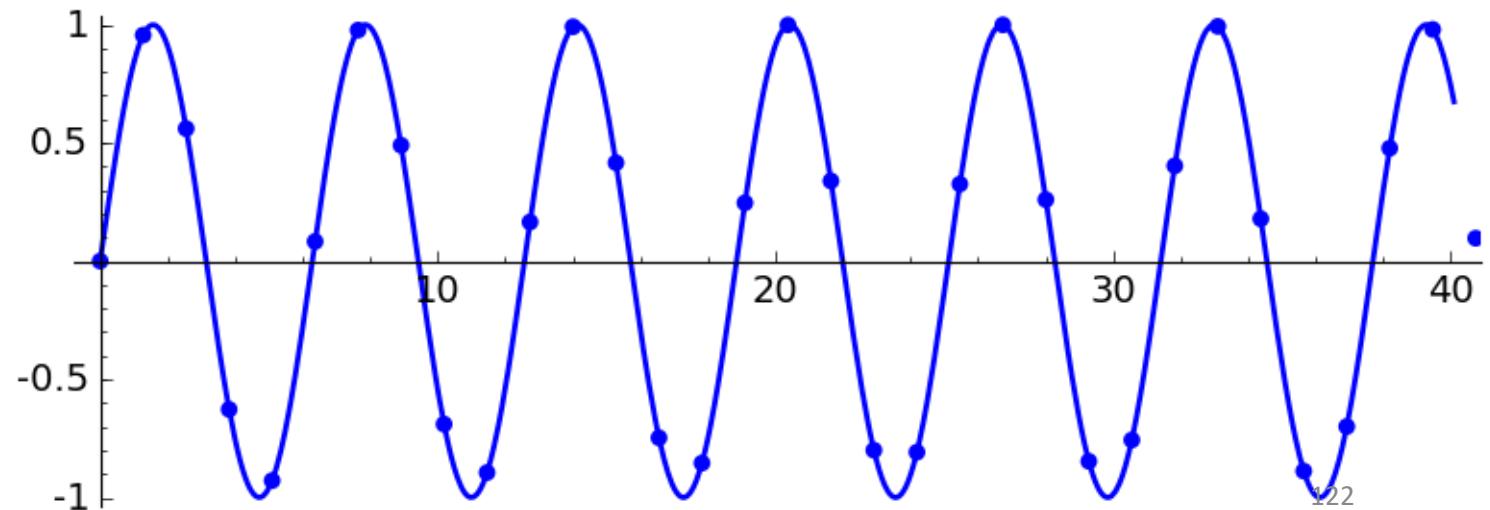
Now:

$$f_c \sim 0.787$$
$$f_c >> 2f$$

$$s(t) = \sin(t)$$

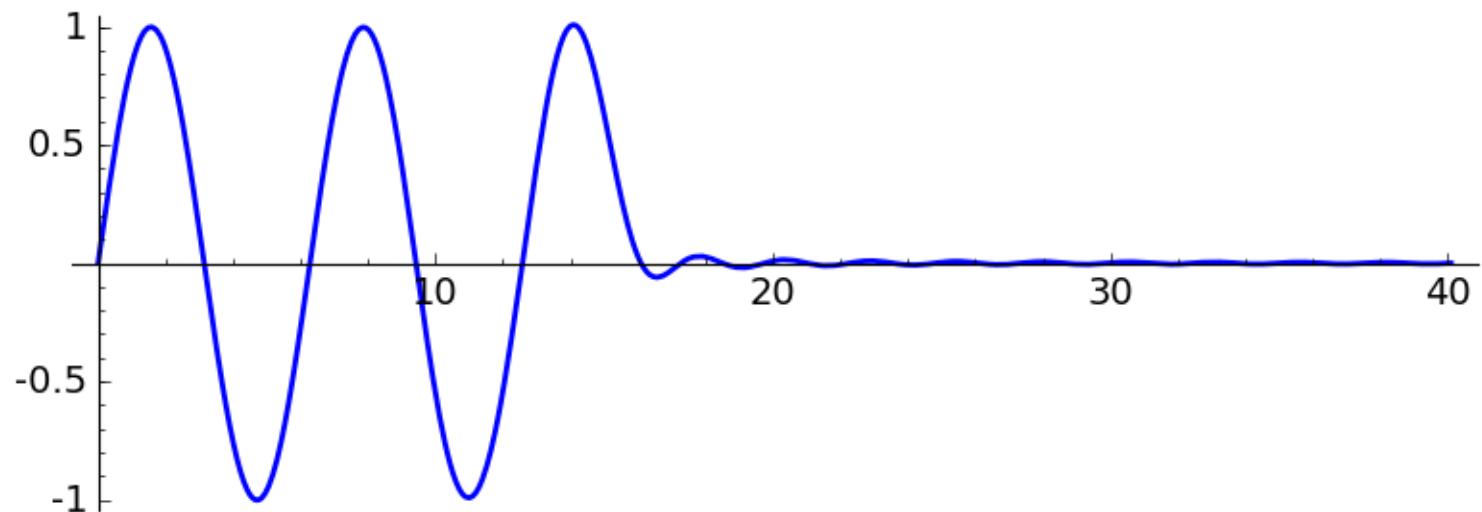


$$s(nT_c) \quad (T_c = 1.27, n \in [0, 40])$$

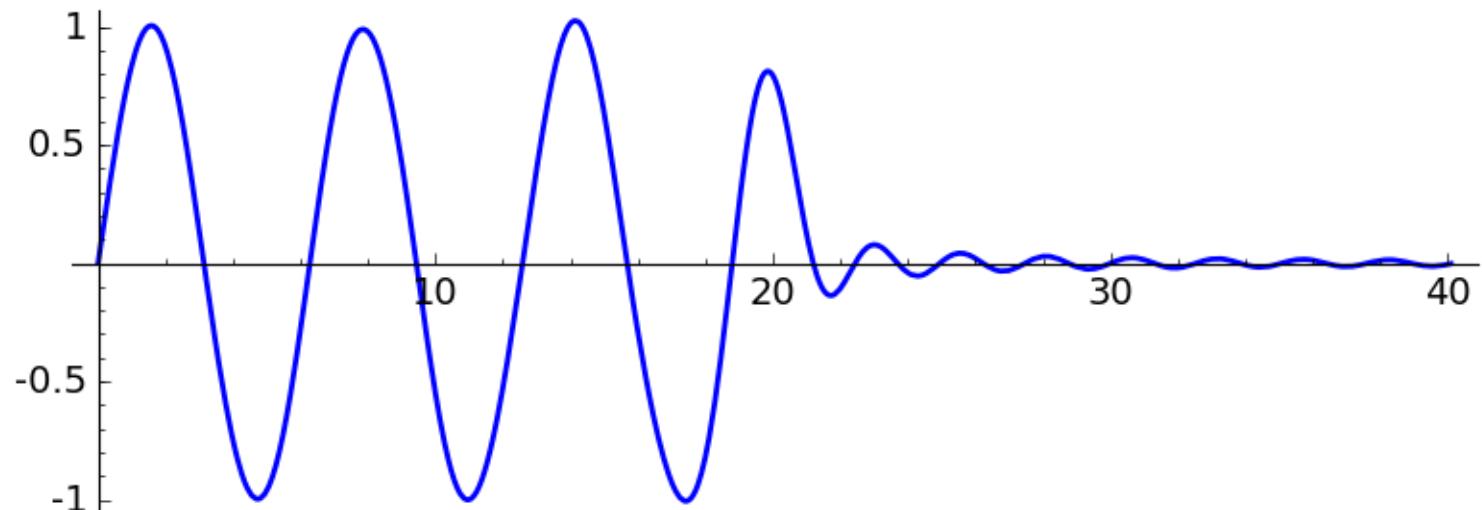


Example

now, interpolating from -15 to +15:



... interpolating from -20 to +20:



Now it's getting better...

But what's wrong with
Nyquist?

What Nyquist did say

Se ho un segnale
con durata finita
avrà uno spettro che
copre tutte le frequenze

Given a signal $s(t)$ with spectrum null at frequencies above f
(LIMITED IN BAND)

However, no real-world signal is perfectly limited in bandwidth

- such a signal should not have energy beyond a frequency band
- but for this it must extend infinitely in time

However, even if the signal is bounded in frequency there is no mean to sample a signal from the real-world and to rebuild the signal perfectly...

... unless we take an infinite number of samples

- The bound of Nyquist is on the frequency of sampling, not on the number of samples required... *(theory works with a ∞ # of samples)*

In any case, if we don't have an infinite time, we can sample a signal well enough to reasonably reconstruct it...

Sampling

• possibile appurare i fenomeni

a rendere quasi unitaria nella banda, così da focalizzarci
su una parte dello spettro dove è concentrata la maggior parte
della energia

- In reality, no signal is strictly limited in band
 - this would mean that the signal is unlimited in time
 - and this fact has no physical meaningfulness
- In practice, many signals are almost limited in a band
 - ... meaning that their spectrum beyond their band is small
 - i.e. their energy beyond the band is small
- In any sampling you can expect a little distortion due to this fact
 - but with a «good» sampling this distortion is negligible
 - furthermore, it is also possible to use a «cutoff» filter that cuts out frequencies above a threshold from a signal in order so to meet the Nyquist requirements

To cut out components of higher freq.

What Nyquist did not say

low-pass filter
(0,3 KHz)
• e campionando
si deve superare
di 8 KHz

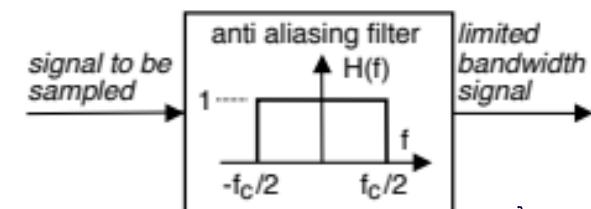
«I am going to sample at 8kHz, so I need to use a filter with a 4kHz cutoff»

- Nyquist didn't say that if you are sampling at rate N samples per second you can use an anti-aliasing filter with a cutoff at frequency $f = N/2$.
- No (analog) filter is perfect, and there is a significant amount of frequencies above 4KHz that pass, although attenuated.
- The result is that the aliased signal is not negligible and seriously disturbs your sampling

... and so what shall we do? *apply!*

- either you use more complex anti-alias filtering...

or



- ... or you simply oversample.
- In this situation the (old, analog) telephone companies used a stronger cutoff (say at 3Khz), and they oversampled at 8 KHz

Le distinte delle spettri delle varie compagnie

• APPLICANDO IL SAMPLIFY THEOREM, è difficile dare regole generali per tutti i casi

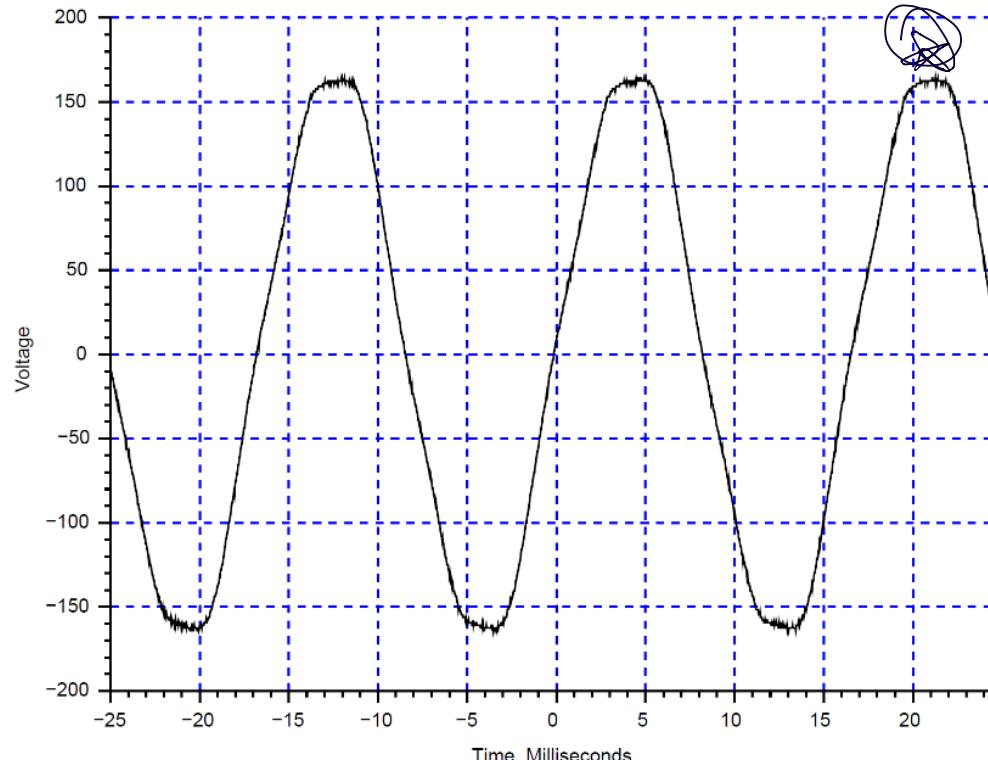
\Rightarrow ogni segnale ha
SINDACATO

What Nyquist did not say

«I need to monitor the 60Hz power line, so I need to sample at 120Hz»
NU VA
 \times
Y

«We have a signal at 1kHz we need to detect, so I need to sample at 2kHz»

- Nyquist didn't say that a signal that repeats N times a second has a bandwidth of N Hertz.
SIGNAL
- In fact the powerline for example is nominally at 60 Hz, but in fact you can see a lot of distortion, meaning that there are many other frequencies involved...
frequenze
POWERLINE



What Nyquist did not say

«I need to monitor the 60Hz power line, so I need to sample at 120Hz»

«We have a signal at 1kHz we need to detect, so I need to sample at 2kHz»

- Nyquist didn't say that a signal that repeats N times a second has a bandwidth of N Hertz.
- In fact the powerline for example is nominally at 60 Hz, but in fact you can see a lot of distortion, meaning that there many other frequencies involved...
- If you don't use anti-aliasing filters you need to figure out the maximum frequency of the signal,
- If it comes out that, as in this case, up to the fifth harmonic (300 Hz) is meaningful, you should sample at least twice that frequency (600 Hz)
- ... a bit more actually because you don't know the phase of the signal...
- ...and if you wish an accurate sampling in a short time you should better oversample seriously

What Nyquist did not say

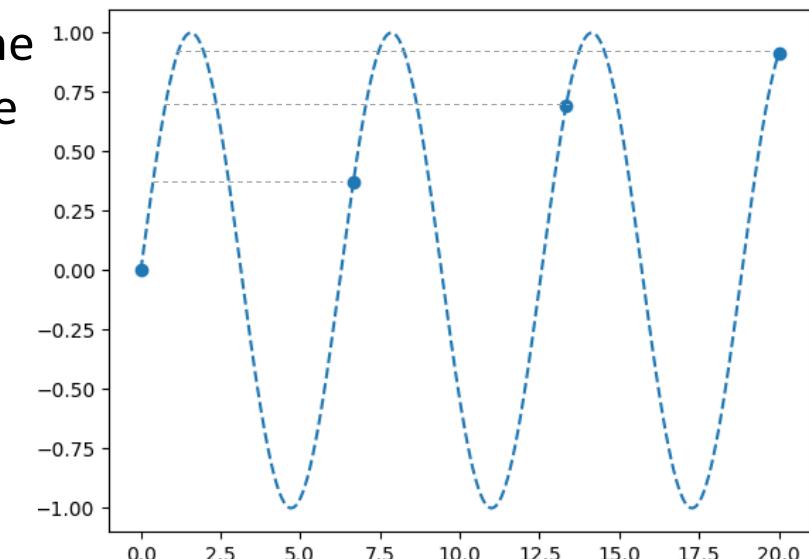
no Nyquist in here

So the quest should be rephrased: «I need to monitor the 60Hz power line, so I guess need to sample at well over 120Hz» ...

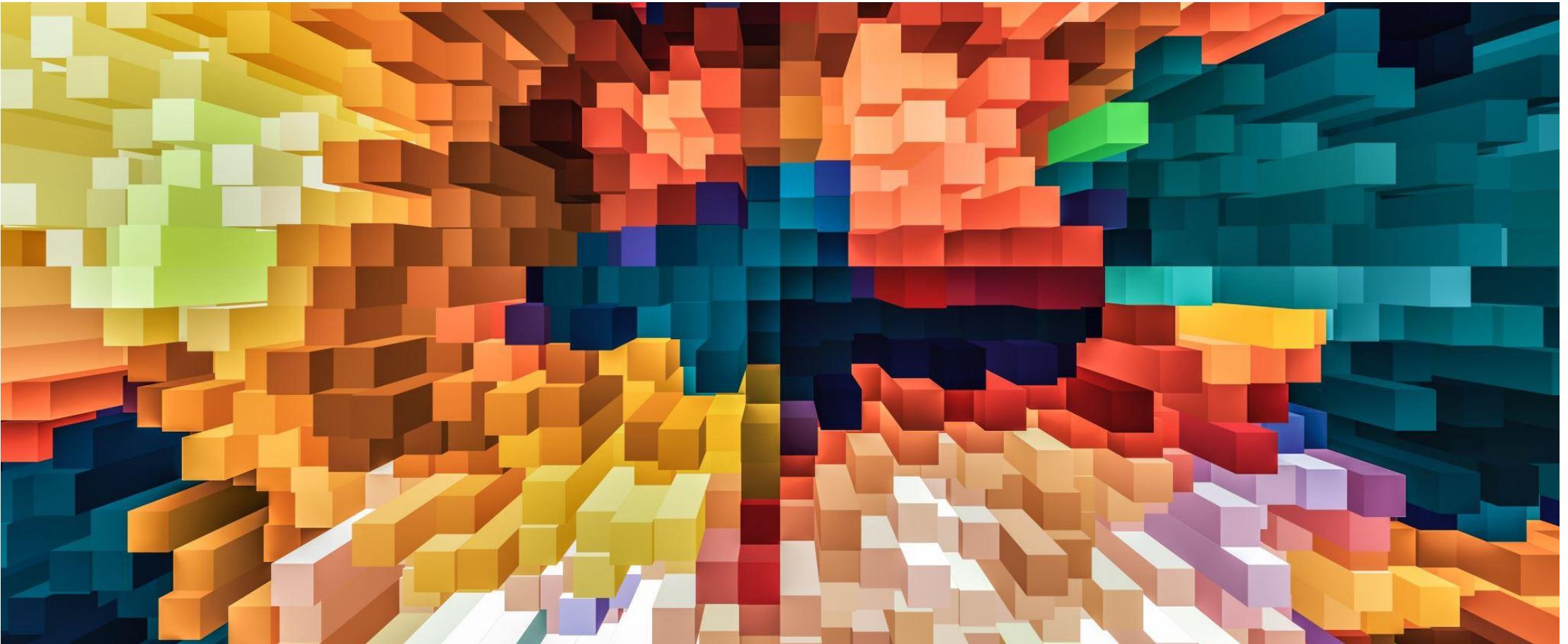
- Well... no. In some cases we could even downsample (if we have the patience to wait)
- In this specific case, **because the signal repeats continuously, it is steady and with a known and steady frequency**, if we downsample (say at a frequency slightly lower than 20Hz) we may be able to reconstruct the signal well enough

You can see how in [2]. A hint is that the signal phase within a cycle will advance a little bit with each new sample. If we sample long enough, we can have enough information to reconstruct the original signal

...but we need to know that it is a steady, repetitive signal with a steady frequency!



Quantization



SAMPLING: abbiamo un set
di campioni finiti
da un sequenze continuo
ad esempio, abbiamo
rappresentarli con una
insieme finito di bit

Quantization

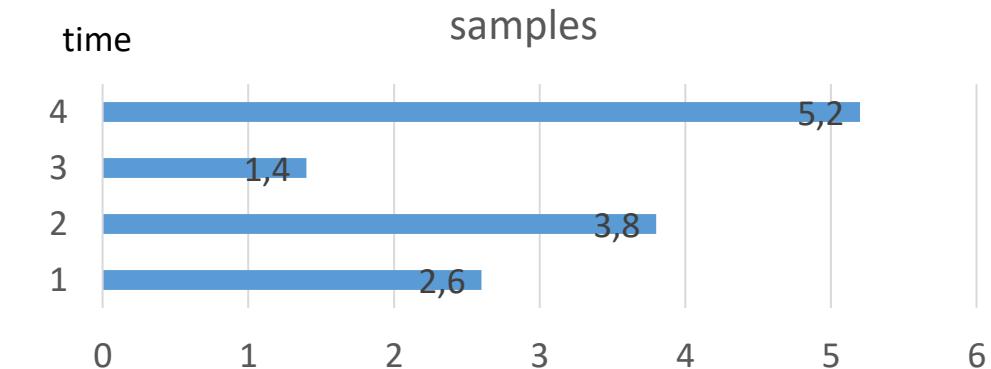
representare le quantità
con finite
tutte di valori

In general, the values obtained with sampling are Real numbers

However, in some applications even a representation in floating point (32/64 bits) is considered prohibitive (e.g. in IoT)

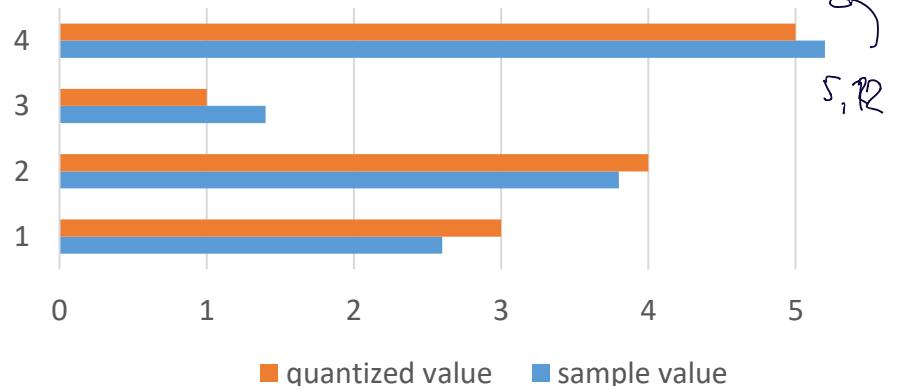
Usually the samples are represented in fixed points

- for example, in Arduino uno uses 10 bits per sample
(output of ADC)



represent integer closest = 5

samples and quantization



Quantization means to approximate the value of a sample so that it can be represented with an integer in range $[0, 2^R - 1]$

1023

losso rappresentare l'intero più vicino

Quantization

Let us consider:

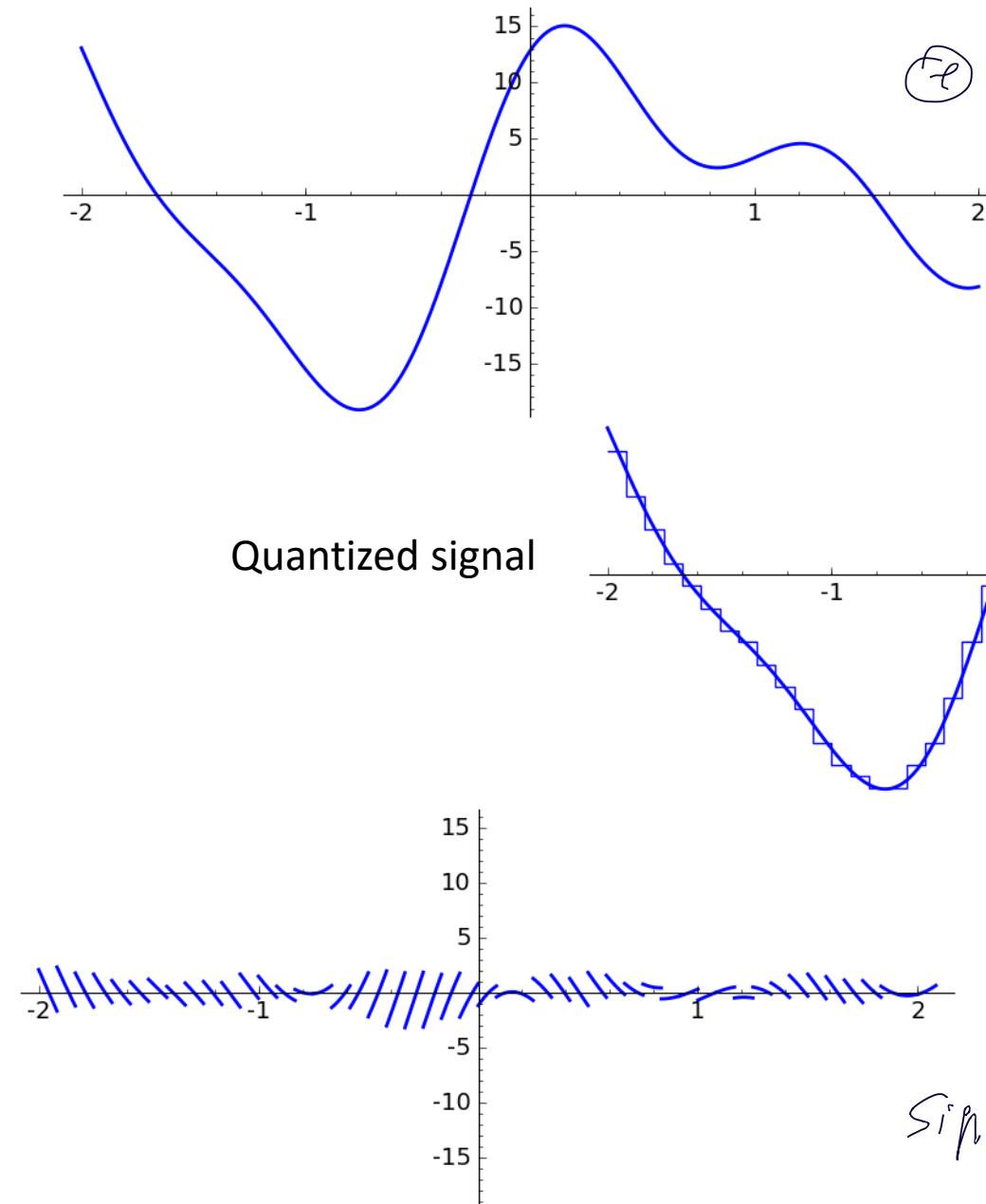
- a signal $s(t)$ sampled at frequency f_c
 - hence samples are taken at intervals $T_c = 1/f_c$
- let $x_k \in \mathbb{R}$ be the sample read at the time interval I_k
- the quantization encodes x_k in an integer value $y_k \in [0, 2^R - 1]$
 - we thus use a R -bit scalar quantizer
 - ... it is called «scalar» because the samples are quantized individually
- this requires R bits per sample (quantization rate)
- the bit rate for a signal sampled at frequency f_c is:

$$R_b = R \cdot f_c \text{ bit/sec}$$

bits \rightarrow
encode the signal
frequency can change
frequency can change

• QUANTIZANDO PRODUCE UN ENONCE
(aproximacion de intervalos pvl
vecinos)

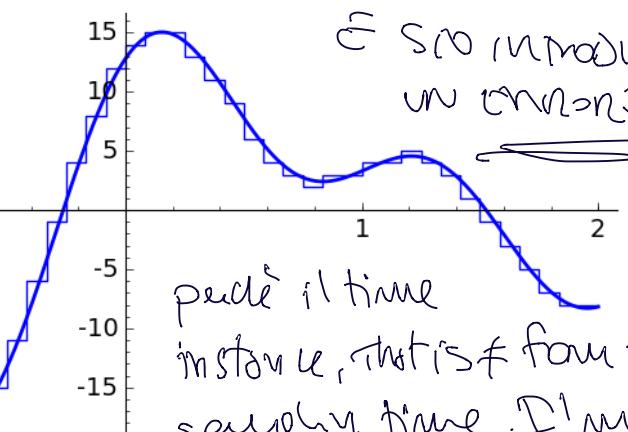
Quantization



$s(t)$ regular
signal $s(t)$

② Components represent
the separate quantized
(D-DQ) interpolation

↪ SIN (MODULANDO
UN ENONCE)



pede il time
instance, that is from the
sampling time. I'm repeating
the quantized value
un il separe
un

Quantization error
 $\rightarrow s(t) - y_k$

Signal

QUANTIZED VALUE
(VERSIONE ONE
DPCM EN REPRESENTATION)

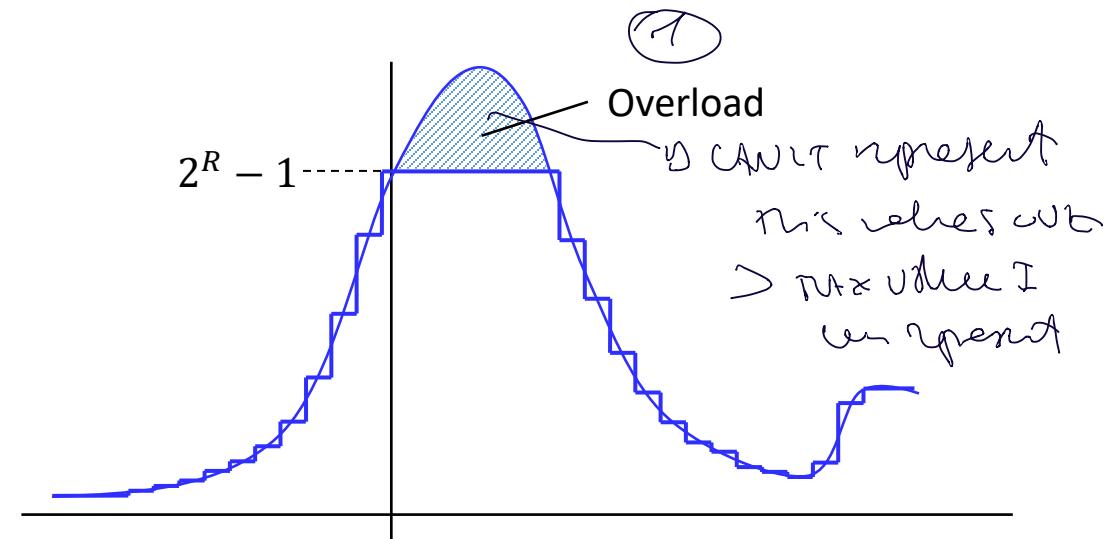
Succede quando focus
sceglie
QUANTIZAZIONE UNIFORME

altri effetti

Quantization noise

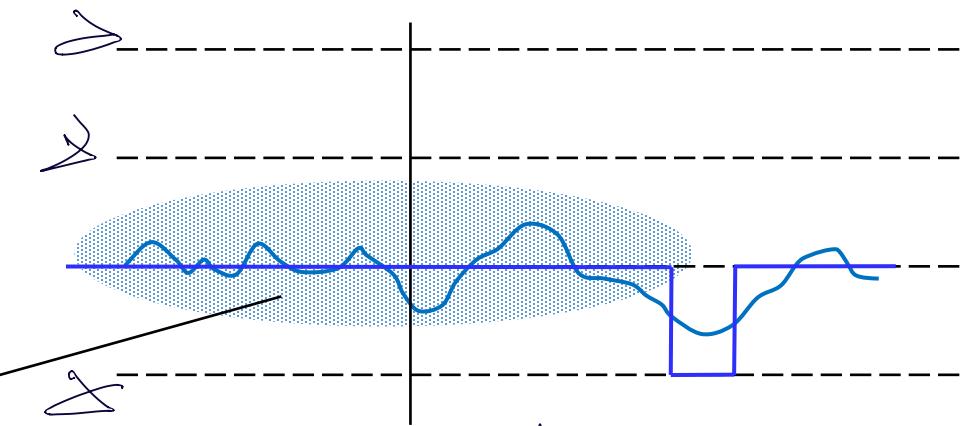
Quando
Notti i valori
di un
intervallo
venendo
rappresentati
da un
unico
valore

(1) **Overload:** occurs when
 $s(t) > 2^R - 1$ for some t
↓
R bits
↓
non uniform quantization



(2) **Granular noise:** occurs because all values in an interval I_k are represented with a unique y_k

all these values are represented with the same y_k



Example

- The human ear can hear between 20 and 20,000 Hz (20 kHz)
- With an acceptable degradation in quality, it can be cut at around 4kHz
 - as we have seen before even less, at 3KHz...
 - this is the level usually accepted at the phone
- Hence:
 $\text{Nyquist Frequency } f_c \geq 2 \cdot 3\text{kHz}$
 - sampling rate: $f_c = 8\text{KHz}$
 - each sample quantized on 8 bits ($R = 8$)
 - the bit rate is then $R_b = 64\text{Kbps}$

Cut off
Af
Ag

Exercise

min free. supply to apply the Sampling Theorem

What is the Nyquist sampling frequency of the following signals:

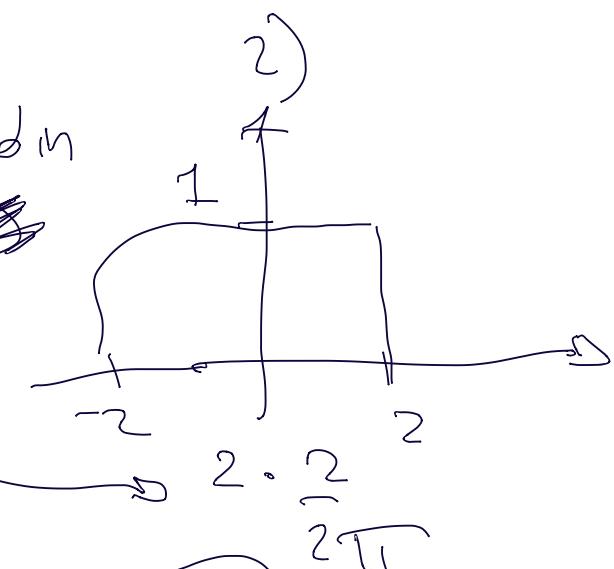
1) $s(t) = \sin(3 \cdot t)$ ~~no ripples~~ \Rightarrow ~~separable continu~~

2) $s(t) = \begin{cases} 1 & -2 < t < 2 \\ 0 & \text{otherwise} \end{cases} \Rightarrow 1.25 \frac{2}{\pi}$ limited in

3) $s(t) = \sin(2 \cdot t + 5) + \sin(8 \cdot t - 1) \Rightarrow \text{not BANDLIM}$

4) $s(t) = 7 + \sin(2 \cdot t - 1)$ ~~not limited~~ ~~separable~~

$$2 \cdot \frac{\pi}{2\pi} = \frac{1}{2}$$



Exercise

half of resolution

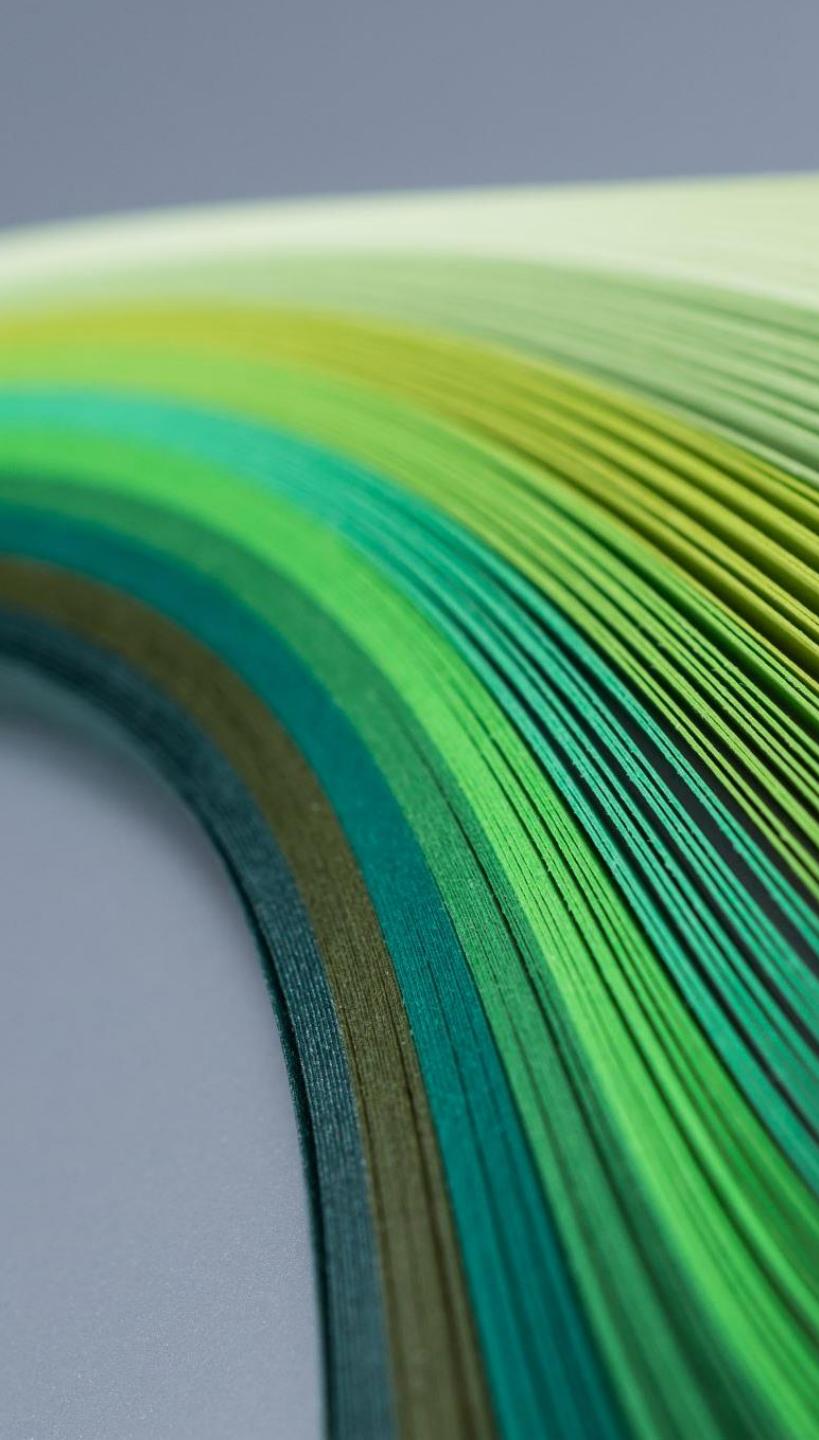
Considering ADCs that perform Q -bits quantization of signals with values ranging in $[0, M]$, compute the resolution of the quantization and the maximum absolute quantization error.

$[0, \infty]$

~~0, ∞~~

$\frac{M}{2^Q}$

Bits of quantization Q	Max.value of the input signal M	Quantization resolution	Maximum quantization error
10	5	$1024 \times \frac{1}{1024} = 1$	$1024 \times 0,5 = 512$
12	5	$4096 \times \frac{1}{4096} = 1$	$4096 \times 0,5 = 2048$
8	3,5	$256 \times \frac{1}{256} = 1$	$256 \times 0,5 = 128$
10	3,5	$1024 \times \frac{1}{1024} = 1$	$1024 \times 0,5 = 512$



Summary

- Analog, discrete and digital signals
- Decomposition of signals into harmonics by means of the Fourier series
- Switching from the domain of time to the domain of frequencies with the Fourier Transform
- Band of a signal
- From Analog to digital signals: sampling and quantization and the Nyquist theorem