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Module 2: Discrete-time signals

Video Introduction

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- ▶ **Module 2.1:** discrete-time signals and operators
- ▶ **Module 2.2:** the discrete-time complex exponential
- ▶ **Module 2.3:** the Karplus-Strong algorithm

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Module 2.1: Discrete-time signals

- ▶ discrete-time signals
- ▶ signal classes
- ▶ elementary operators
- ▶ shifts
- ▶ energy and power

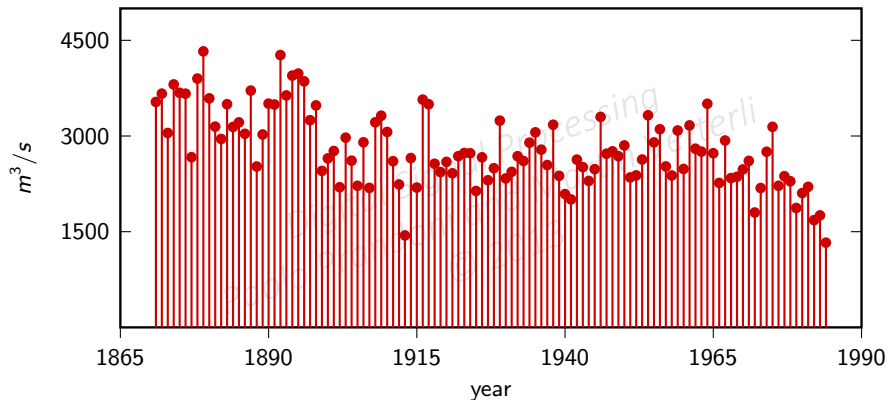
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Meteorology (limnology): the floods of the Nile



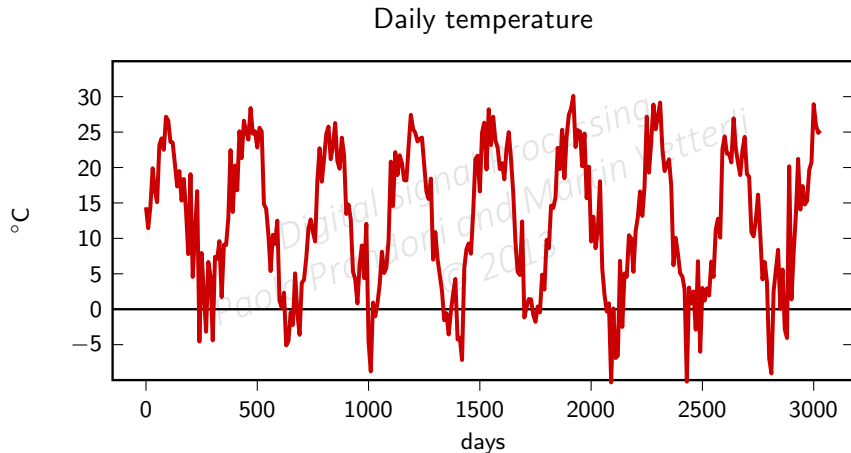
Representations of flood data: circa 2500 BC

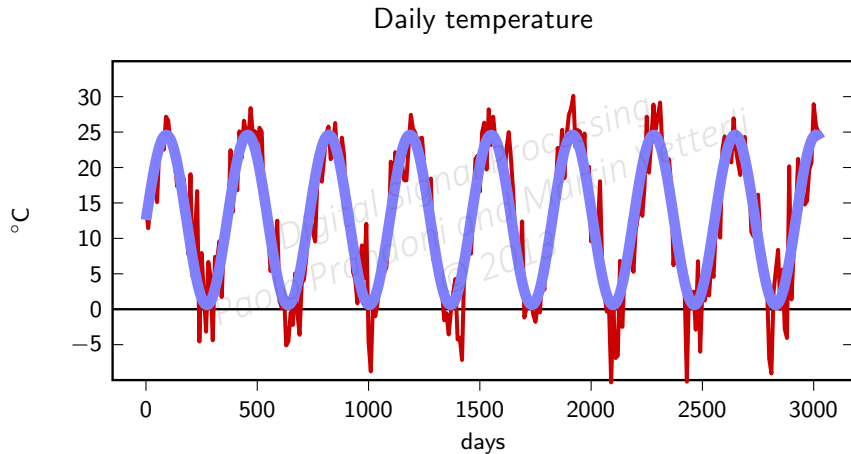
Discrete-time signals have a long tradition...



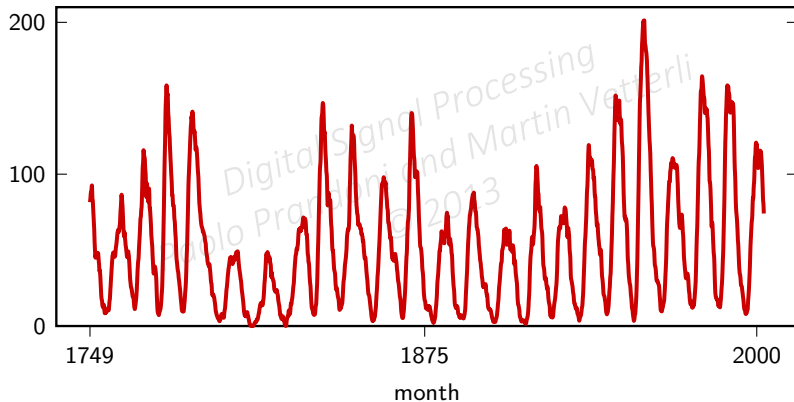
Representations of flood data: circa AD 2000

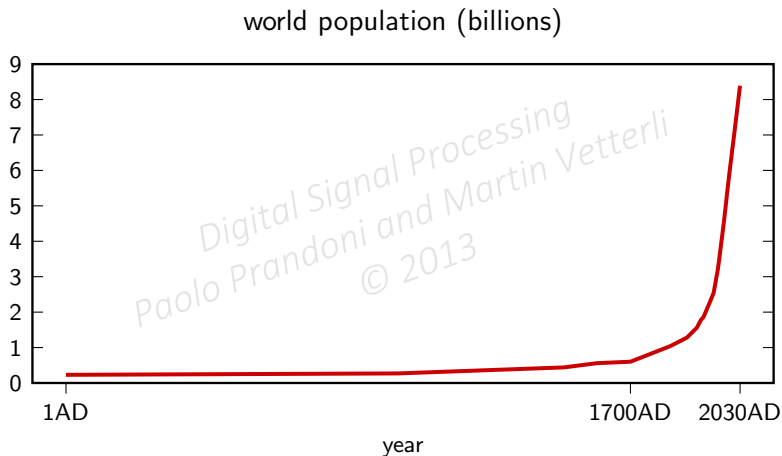
Probably your first scientific experiment...



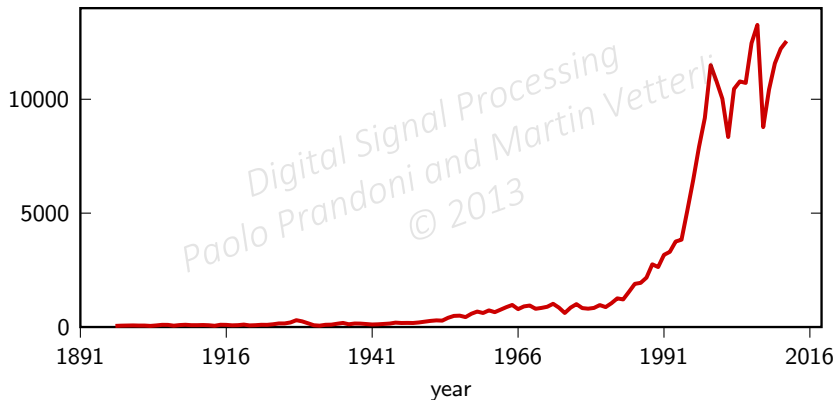


monthly solar spot activity, 1749 to 2003





a purely man-made signal: the Dow Jones industrial average



Discrete-time signal: a sequence of **complex** numbers

- ▶ one dimension (for now)
- ▶ notation: $x[n]$
- ▶ two-sided sequences: $x : \mathbb{Z} \rightarrow \mathbb{C}$
- ▶ n is dimension-less “time”
- ▶ analysis: periodic measurement
- ▶ synthesis: stream of generated samples

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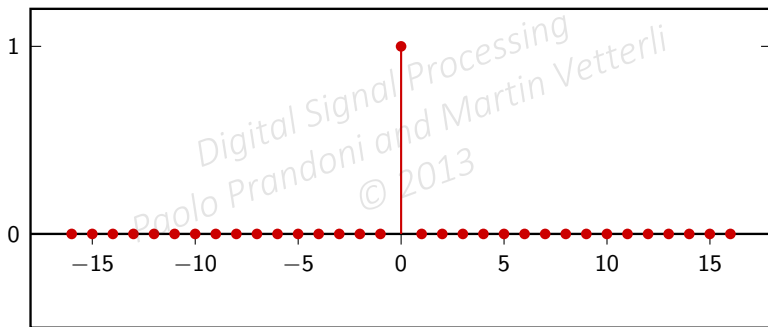
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$$x[n] = \delta[n]$$



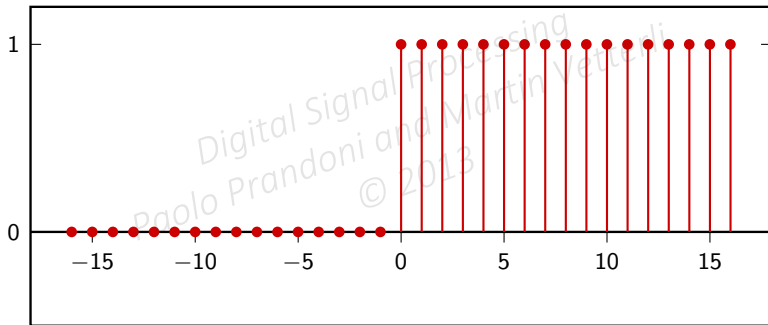
How do you synchronize audio and video...



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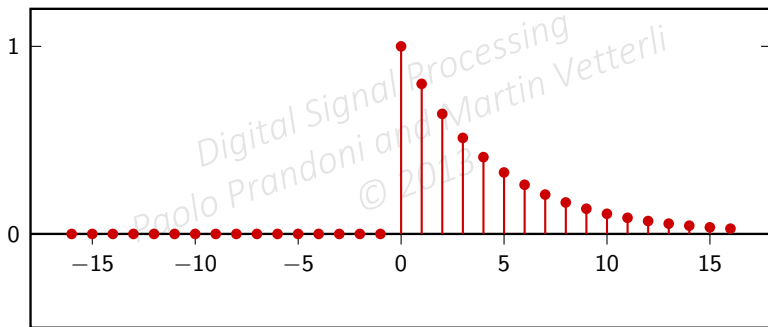
$$x[n] = u[n]$$



The Frankenstein switch...



$$x[n] = |a|^n u[n], \quad |a| < 1$$



How fast does your coffee get cold...



How fast does your coffee get cold...



Newton's law of cooling:

$$\frac{dT}{dt} = -c(T - T_{\text{env}})$$

$$T(t) = T_{\text{env}} + (T_0 - T_{\text{env}})e^{-ct}$$

In practice:

- ▶ must have convection only
- ▶ must have large conductivity

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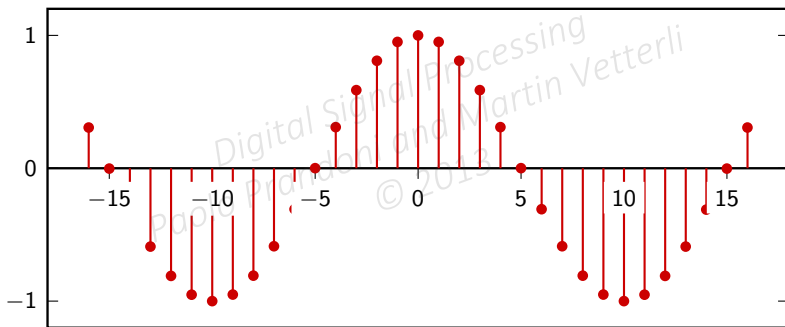
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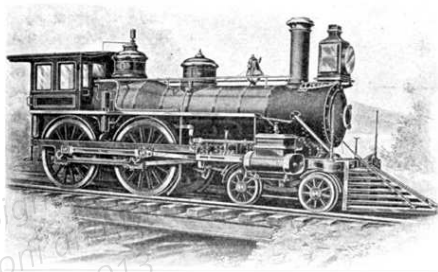
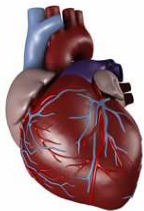
- ▶ must have convection only
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$$x[n] = \sin(\omega_0 n + \theta)$$



Oscillations are everywhere!



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- ▶ finite-length
- ▶ infinite-length
- ▶ periodic
- ▶ finite-support

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- ▶ sequence notation: $x[n]$, $n = 0, 1, \dots, N - 1$
- ▶ vector notation: $\mathbf{x} = [x_0 \ x_1 \ \dots \ x_{N-1}]^T$
- ▶ practical entities, good for numerical packages (Matlab and the like)

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- ▶ sequence notation: $x[n]$, $n \in \mathbb{Z}$

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► N -periodic sequence: $\tilde{x}[n] = \tilde{x}[n + kN]$, $n, k, N \in \mathbb{Z}$

► same information as finite-length of length N

► “natural” bridge between finite and infinite lengths

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- ▶ Finite-support sequence:

$$\bar{x}[n] = \begin{cases} x[n] & \text{if } 0 \leq n < N \\ 0 & \text{otherwise} \end{cases} \quad n \in \mathbb{Z}$$

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- ▶ another bridge between finite and infinite lengths

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- ▶ scaling:

$$y[n] = \alpha x[n]$$

- ▶ sum:

$$y[n] = x[n] + z[n]$$

- ▶ product:

$$y[n] = x[n] \cdot z[n]$$

- ▶ shift by k (delay):

$$y[n] = x[n - k]$$

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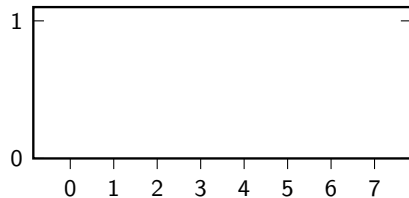
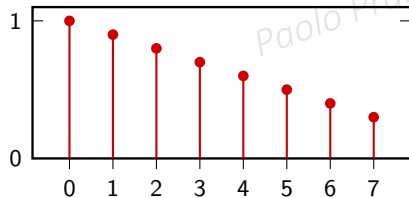
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Shift of a finite-length: finite-support

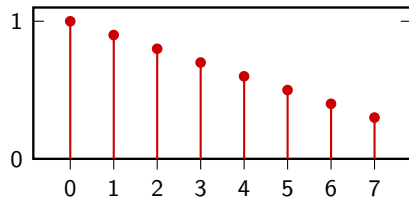
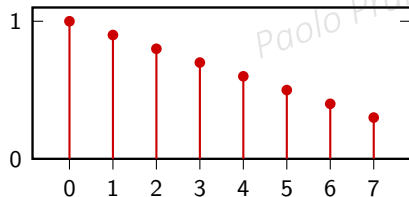
$[x_0 \ x_1 \ x_2 \ x_3 \ x_4 \ x_5 \ x_6 \ x_7]$



Shift of a finite-length: finite-support

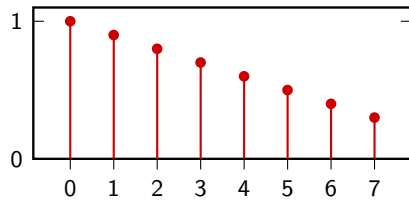
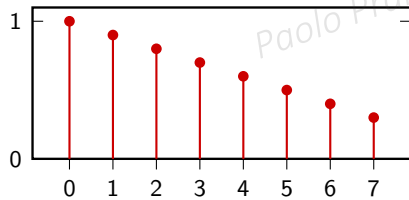
$x[n]$

... x_0 x_1 x_2 x_3 x_4 x_5 x_6 x_7 ...



$\bar{x}[n]$

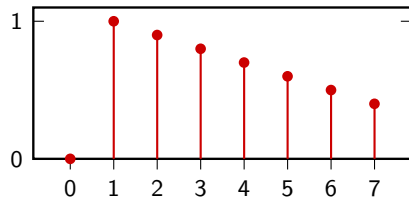
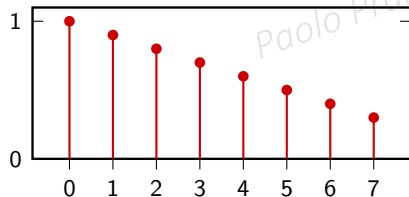
... 0 0 0 x_0 x_1 x_2 x_3 x_4 x_5 x_6 x_7 0 0 0 ...



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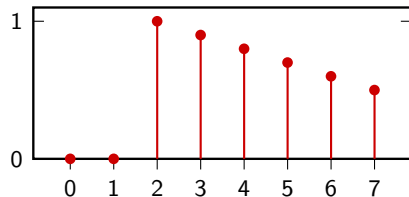
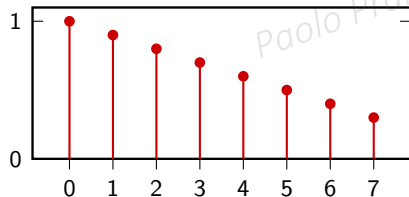
$$\bar{x}[n-1]$$

... 0 0 0 0 x_0 x_1 x_2 x_3 x_4 x_5 x_6 x_7 0 0 ...



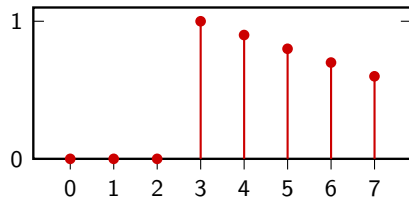
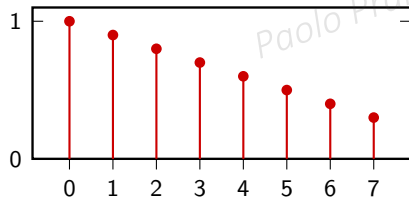
$$\bar{x}[n-2]$$

... 0 0 0 0 0 x_0 x_1 x_2 x_3 x_4 x_5 x_6 x_7 0 ...



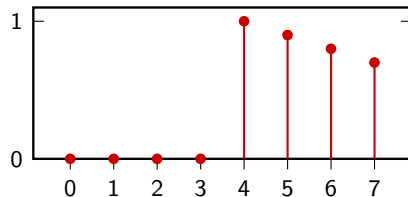
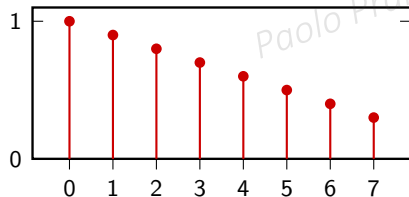
$$\bar{x}[n-3]$$

... 0 0 0 0 0 0 0 x_0 x_1 x_2 x_3 x_4 x_5 x_6 x_7 ...

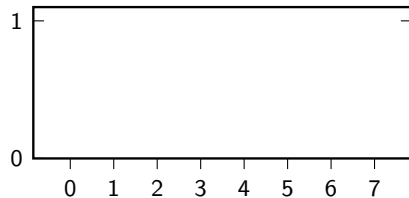
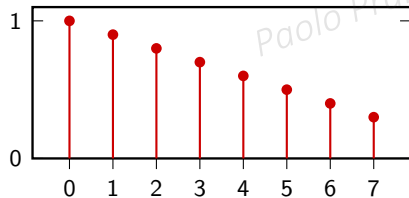


$$\bar{x}[n - 4]$$

... 0 0 0 0 0 0 0 x_0 x_1 x_2 x_3 x_4 x_5 x_6 ...

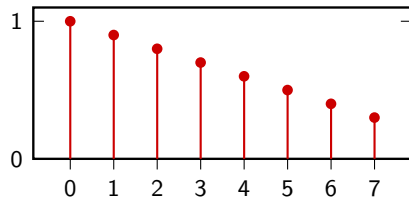
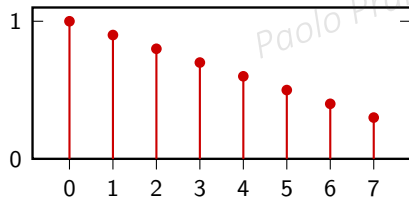


$[x_0 \ x_1 \ x_2 \ x_3 \ x_4 \ x_5 \ x_6 \ x_7]$



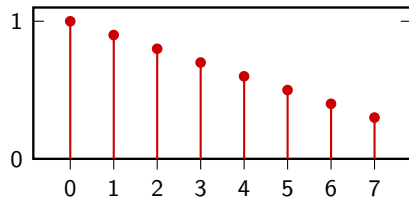
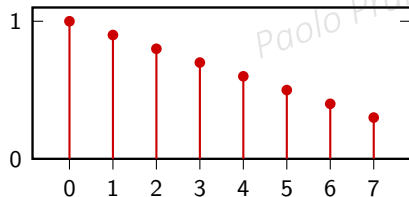
$x[n]$

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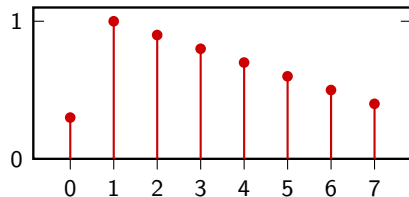
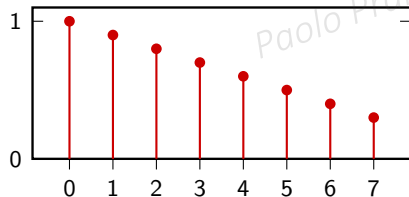
$$\tilde{x}[n]$$

... x_5 x_6 x_7 x_0 x_1 x_2 x_3 x_4 x_5 x_6 x_7 x_0 x_1 x_2 ...



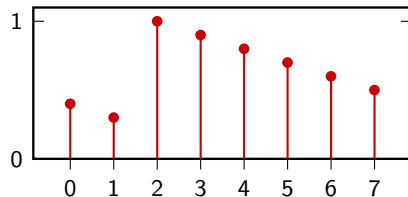
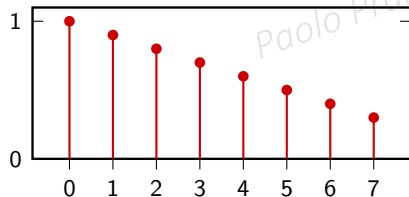
$$\tilde{x}[n-1]$$

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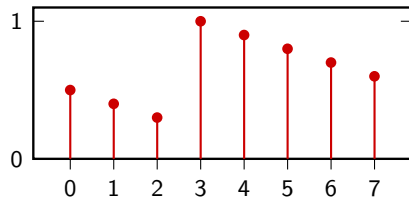
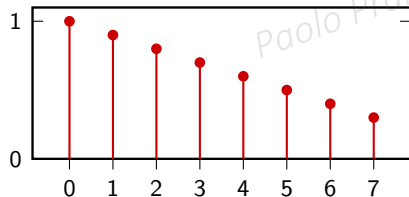
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... x_3 x_4 x_5 x_6 x_7 x_0 x_1 x_2 x_3 x_4 x_5 x_6 x_7 x_0 ...



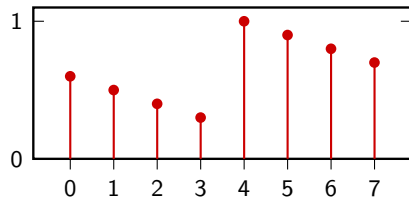
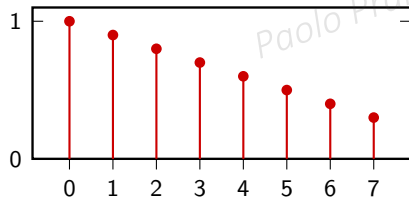
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... x_2 x_3 x_4 x_5 x_6 x_7 x_0 x_1 x_2 x_3 x_4 x_5 x_6 x_7 ...



$$\tilde{x}[n - 4]$$

... x_1 x_2 x_3 x_4 x_5 x_6 x_7 x_0 x_1 x_2 x_3 x_4 x_5 x_6 ...



$$E_x = \sum_{n=-\infty}^{\infty} |x[n]|^2$$

$$P_x = \lim_{N \rightarrow \infty} \frac{1}{2N+1} \sum_{n=-N}^N |x[n]|^2$$

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$$E_{\tilde{x}} = \infty$$

$$P_{\tilde{x}} \equiv \frac{1}{N} \sum_{n=0}^{N-1} |\tilde{x}[n]|^2$$

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END OF MODULE 2.1

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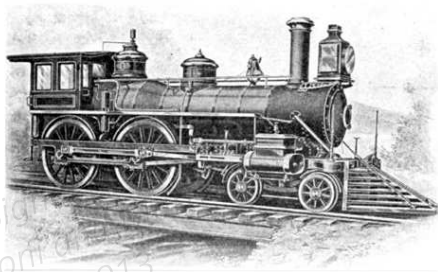
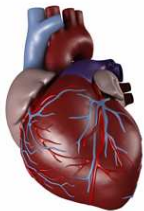
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Module 2.2: the complex exponential

- ▶ the complex exponential
- ▶ periodicity
- ▶ wagonwheel effect and maximum “speed”
- ▶ digital and real-world frequency

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Oscillations are everywhere



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Ingredients:

- ▶ a frequency ω (units: radians)
- ▶ an initial phase ϕ (units: radians)
- ▶ an amplitude A (units depending on underlying measurement)
- ▶ a trigonometric function

e.g. $x[n] = A \cos(\omega n + \phi)$

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e.g. $x[n] = A \cos(\omega n + \phi)$

the trigonometric function of choice in DSP is the complex exponential:

$$\begin{aligned}x[n] &= Ae^{j(\omega n + \phi)} \\ &= A[\cos(\omega n + \phi) + j \sin(\omega n + \phi)]\end{aligned}$$

- ▶ makes sense: sines and cosines always go together

- ▶ simpler math: trigonometry becomes algebra

- ▶ we can use complex numbers in digital systems

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- ▶ makes sense: sines and cosines always go together
- ▶ simpler math: trigonometry becomes algebra
- ▶ we can use complex numbers in digital systems

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Example: change the phase of a pure cosine

$$\cos(\omega n + \phi) = a \cos(\omega n) + b \sin(\omega n), \quad a = \cos \phi, \quad b = \sin \phi$$

- ▶ each sinusoid is always a sum of sine and cosine
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- ▶ we have to carry more terms in our equations

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► sine and cosine “live” together

► phase shift is simple multiplication

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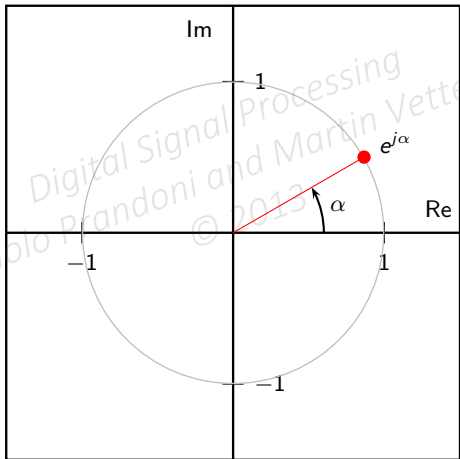
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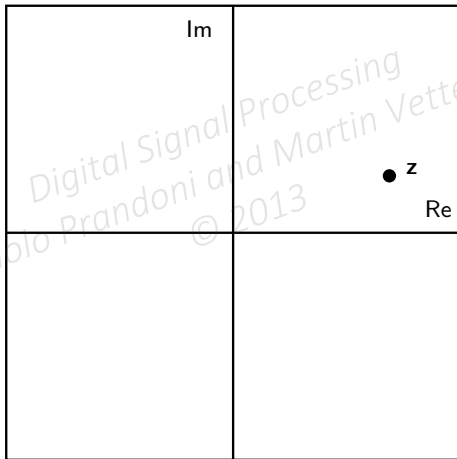
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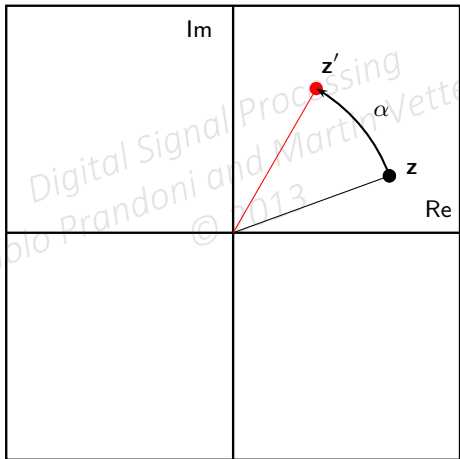
$$e^{j\alpha} = \cos \alpha + j \sin \alpha$$



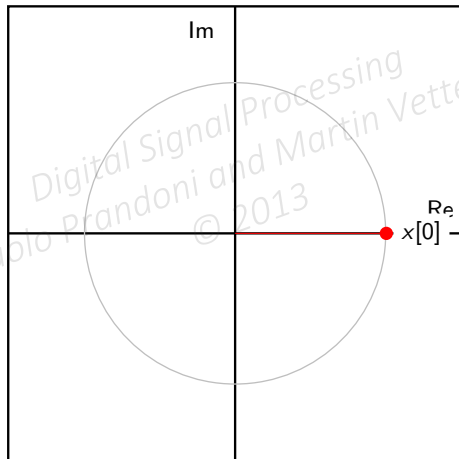
z : point on the complex plane



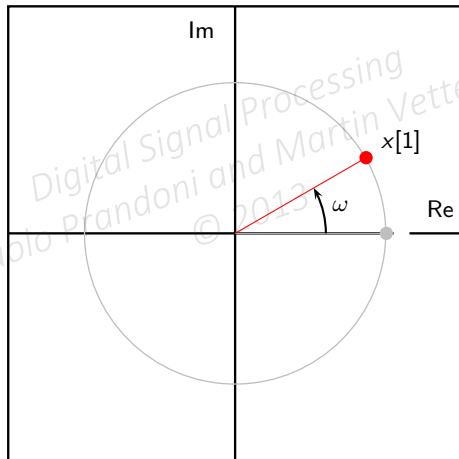
rotation: $z' = z e^{j\alpha}$



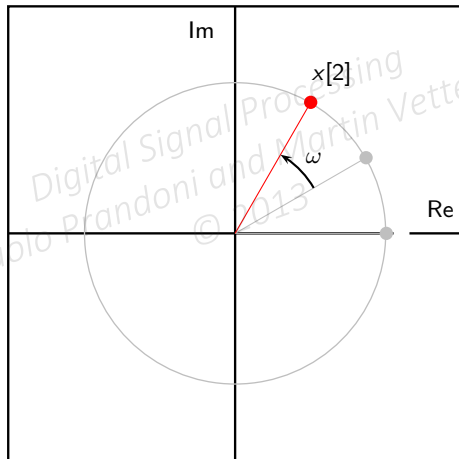
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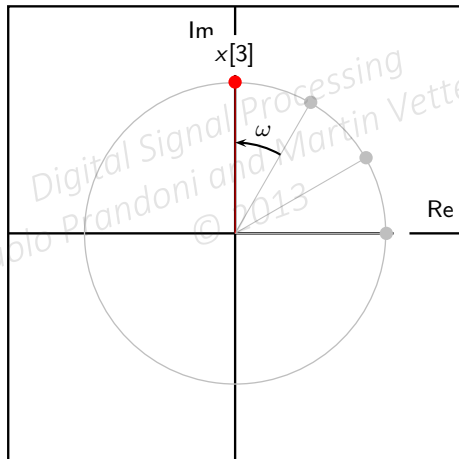
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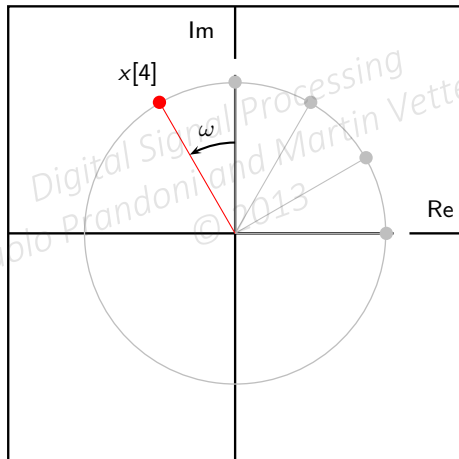
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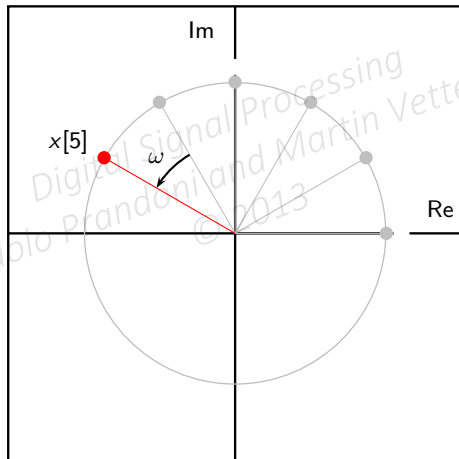
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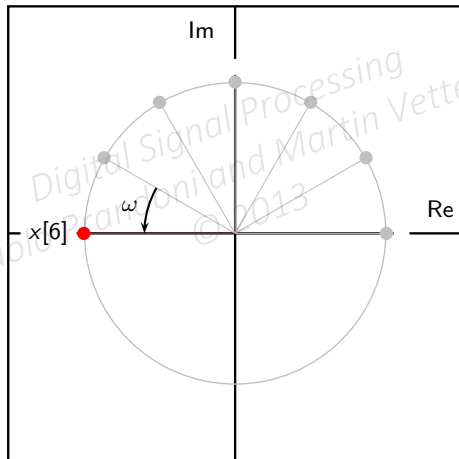
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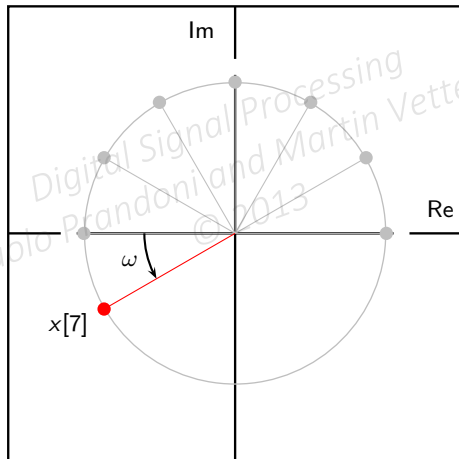
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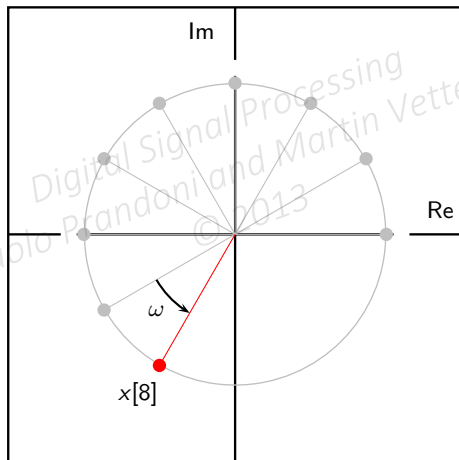
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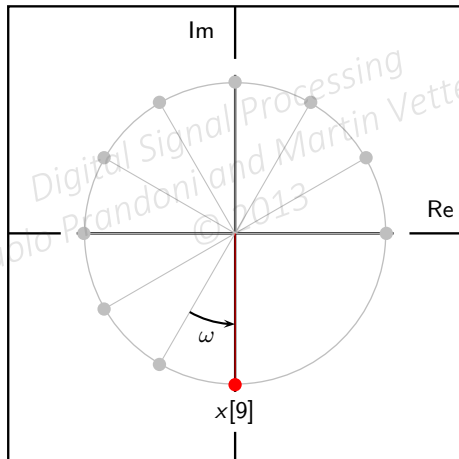
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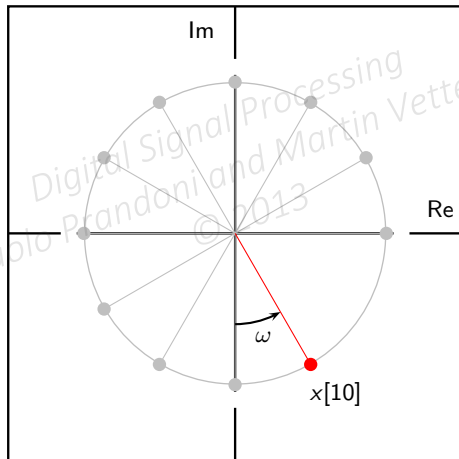
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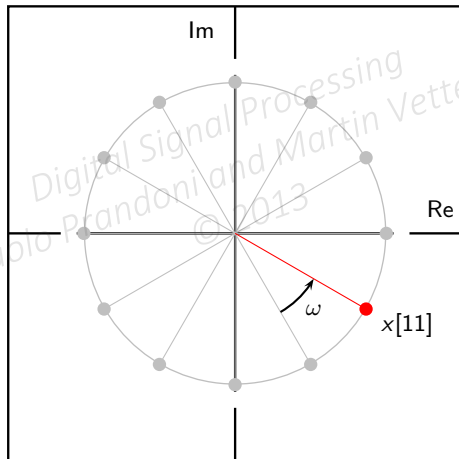
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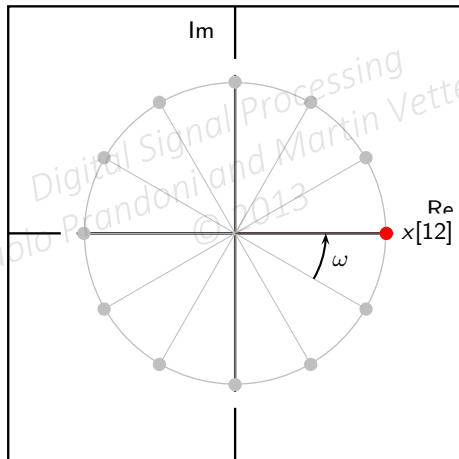
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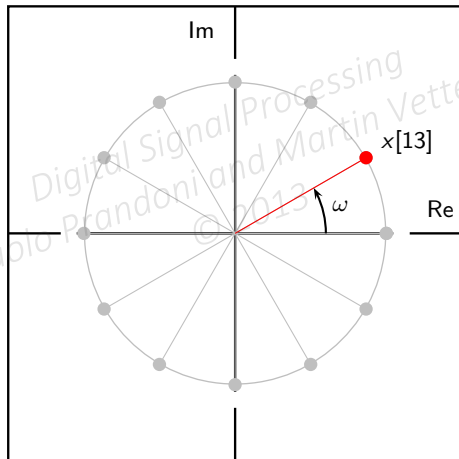
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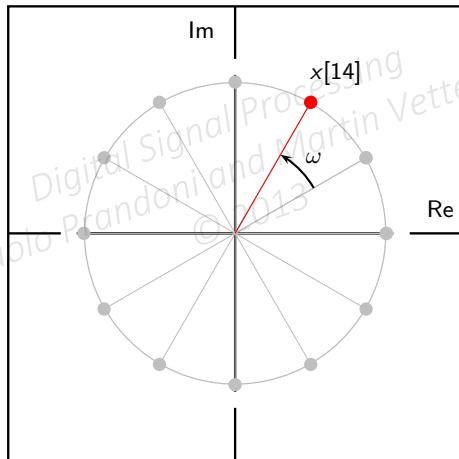
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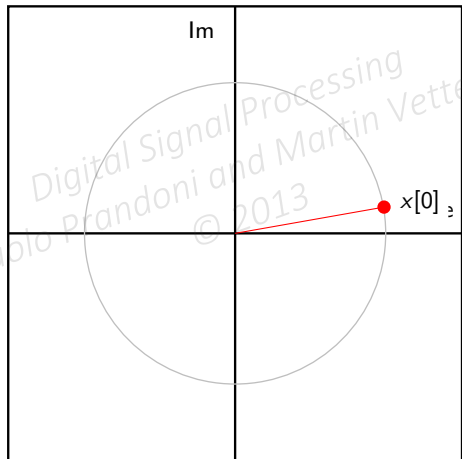
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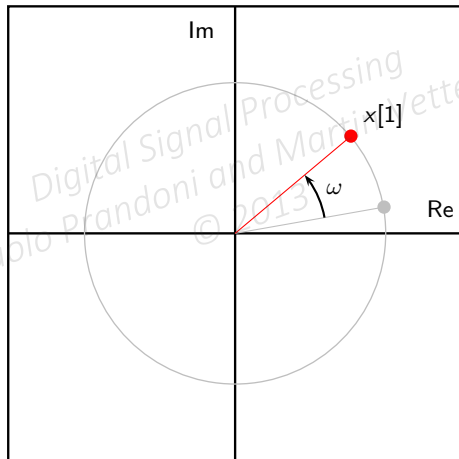
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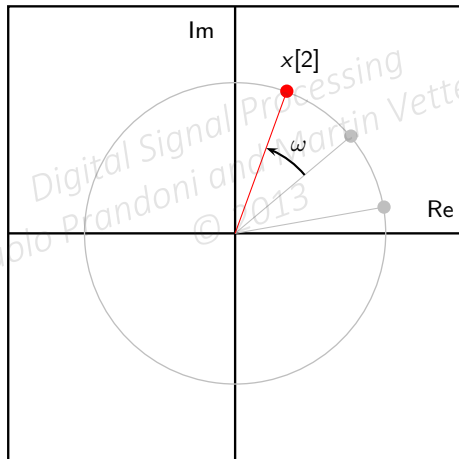
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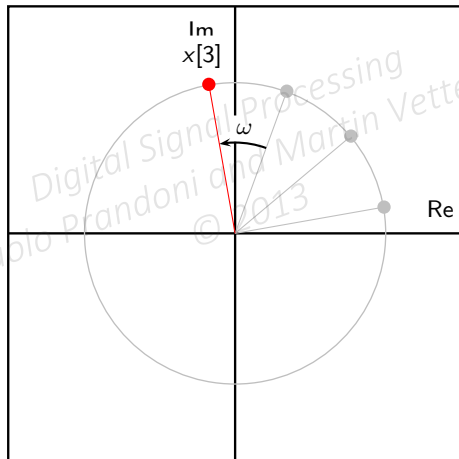
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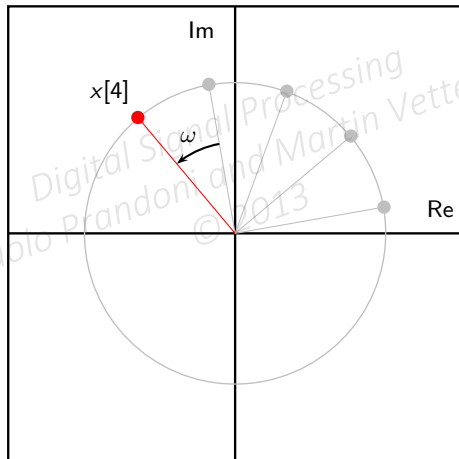
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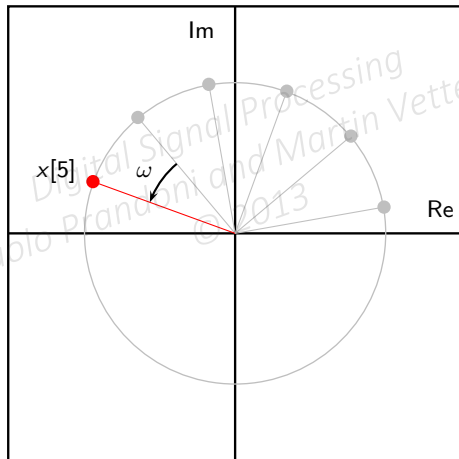
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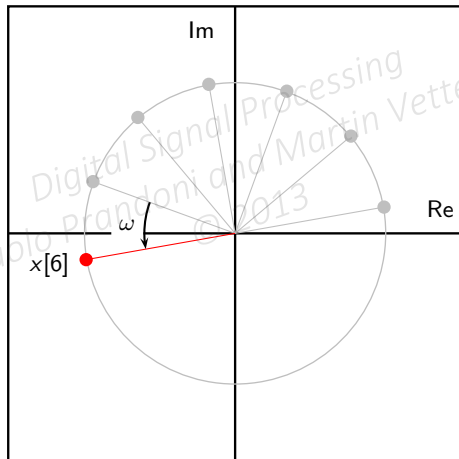
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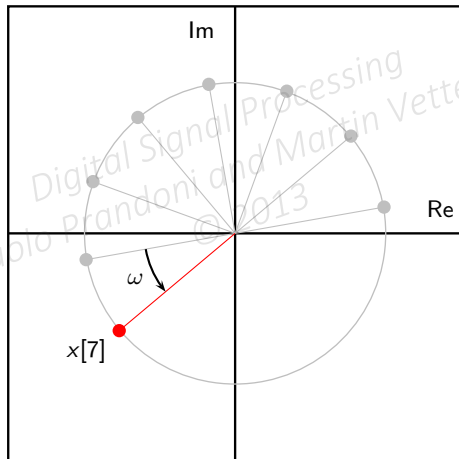
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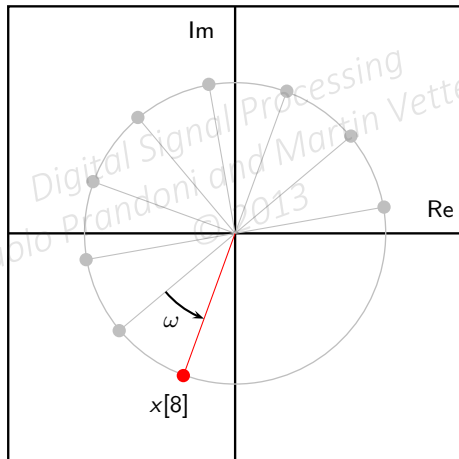
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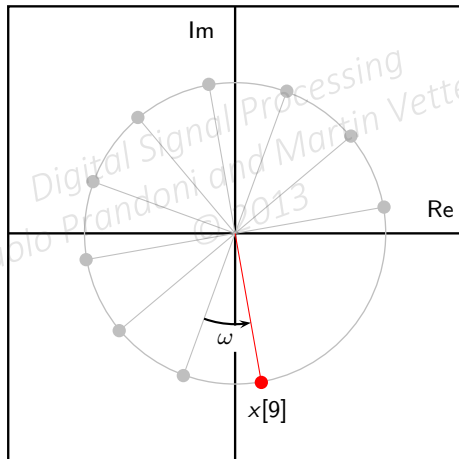
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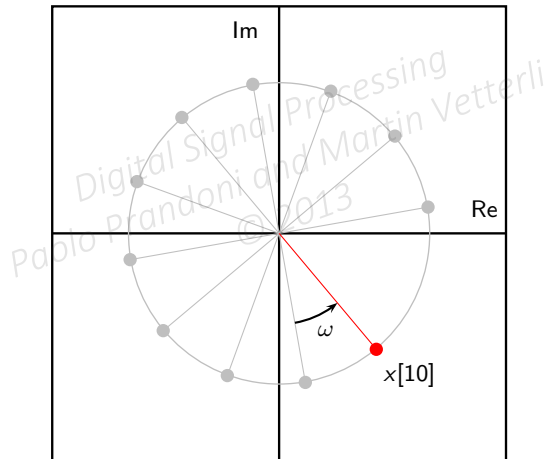
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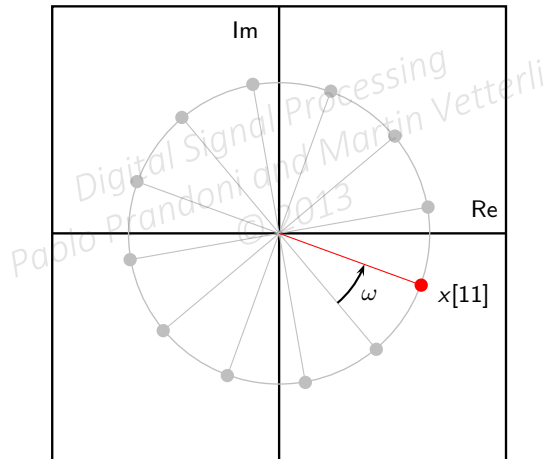
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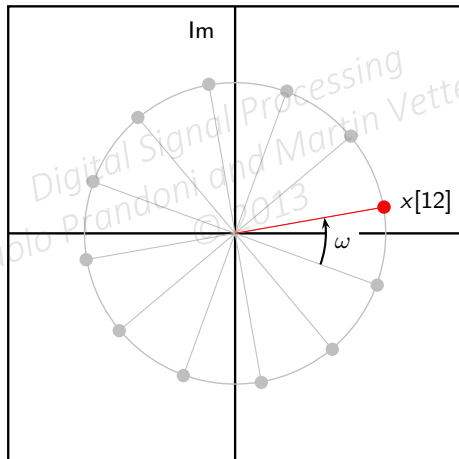
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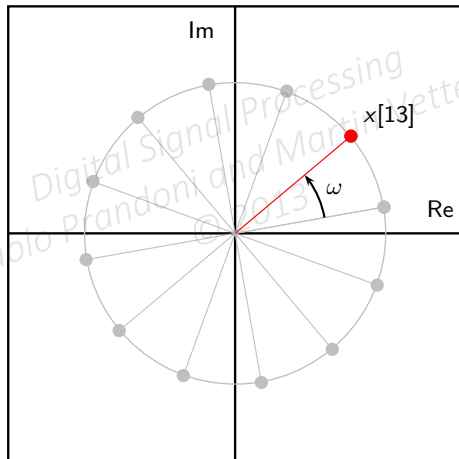
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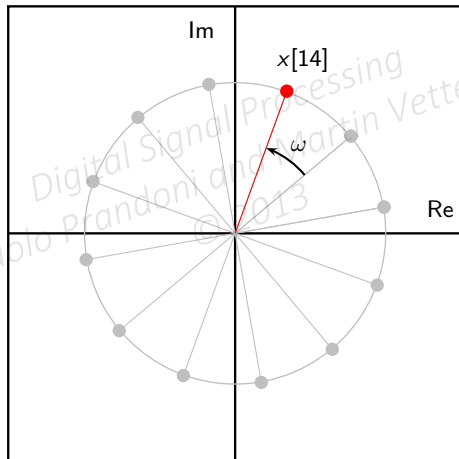
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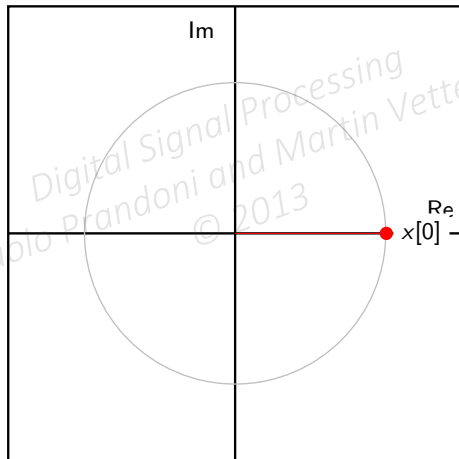


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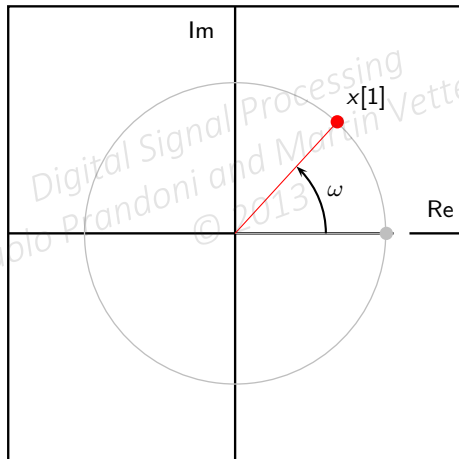
Careful: not every sinusoid is periodic in discrete time

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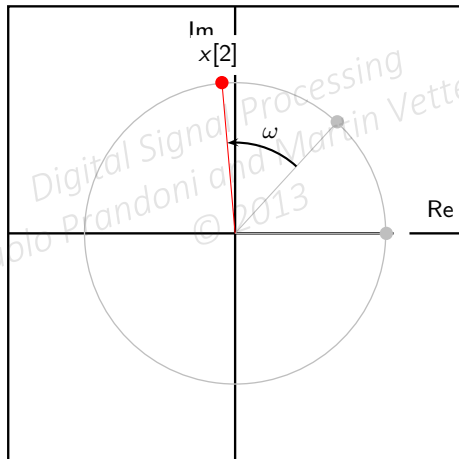
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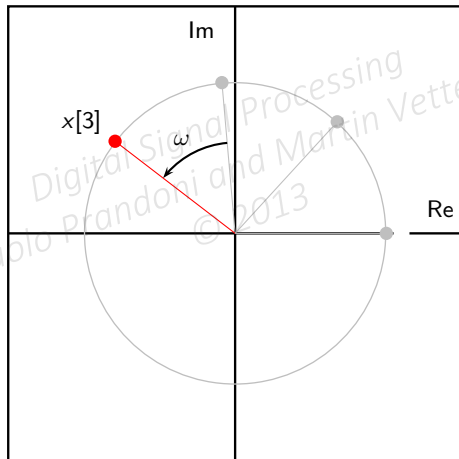
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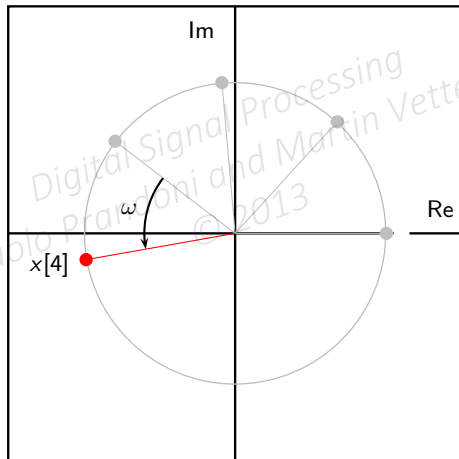
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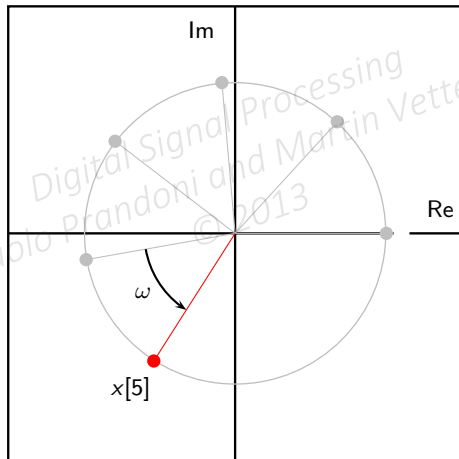
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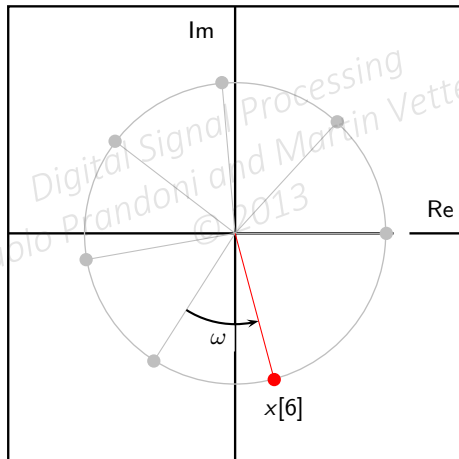
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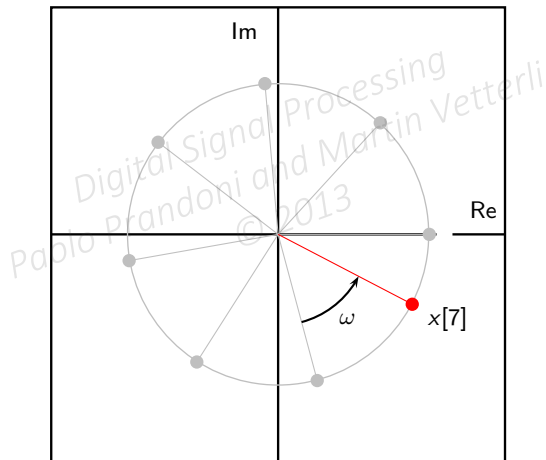
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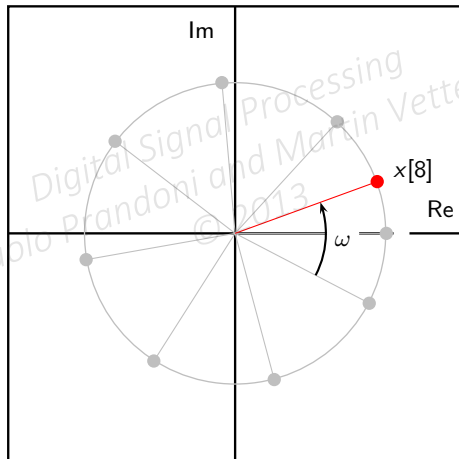
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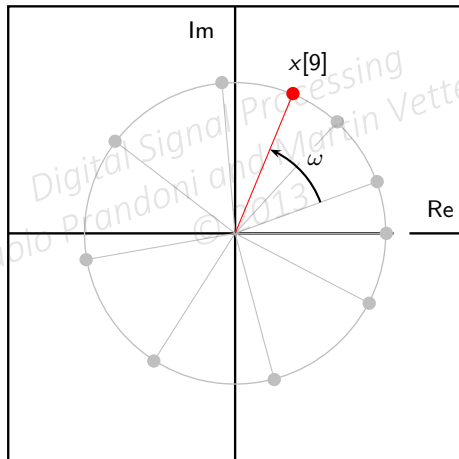
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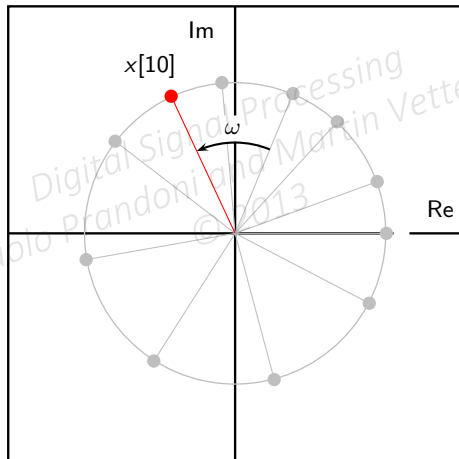
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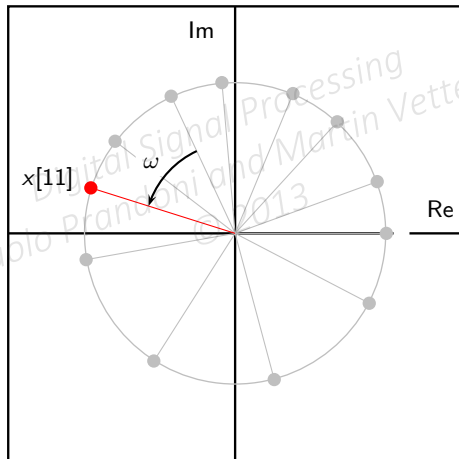
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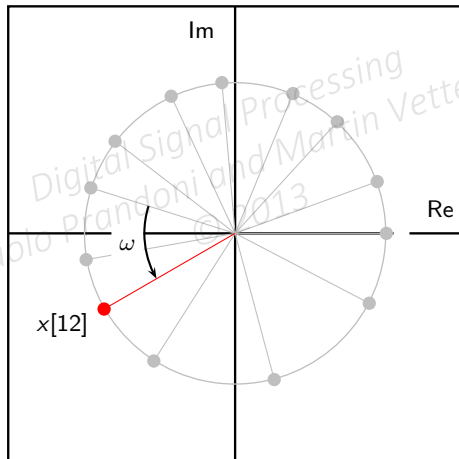
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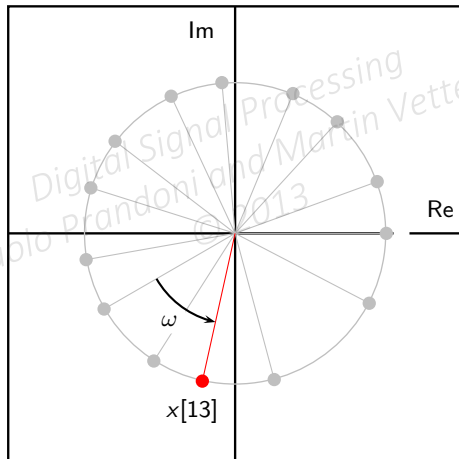
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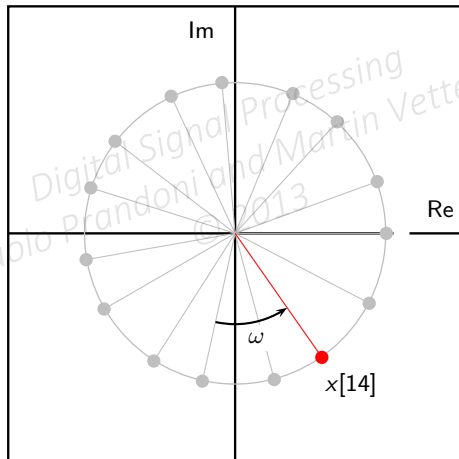
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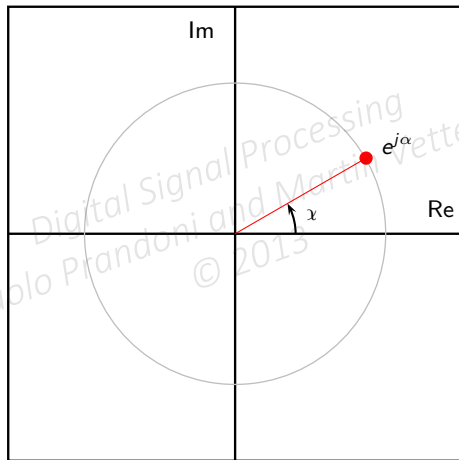
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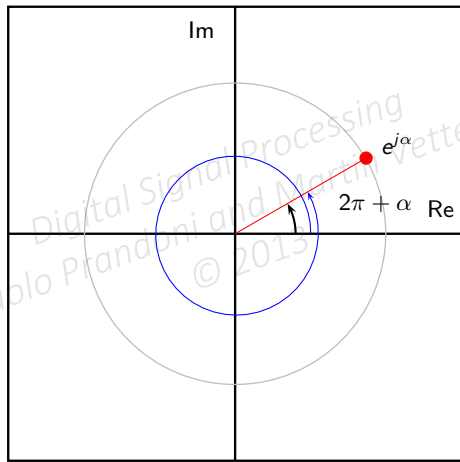
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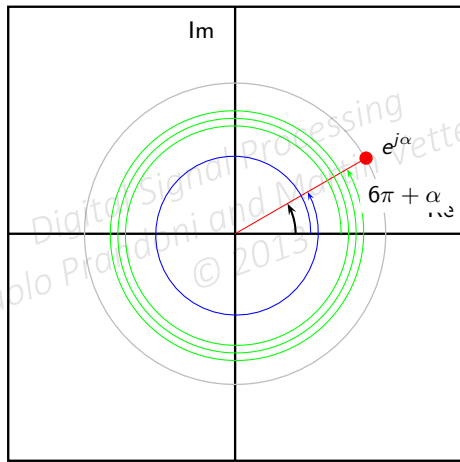
One point, many names



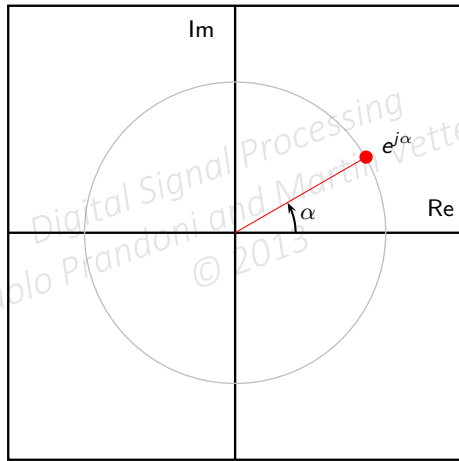
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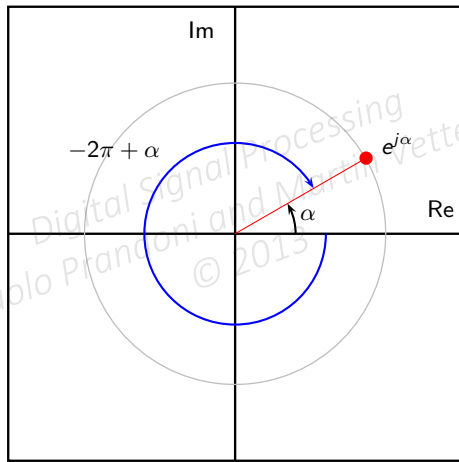
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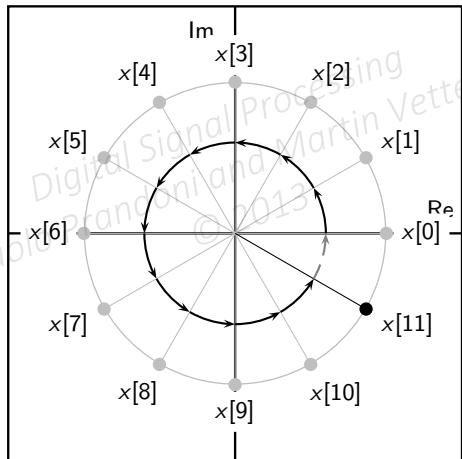
How “fast” can we go?



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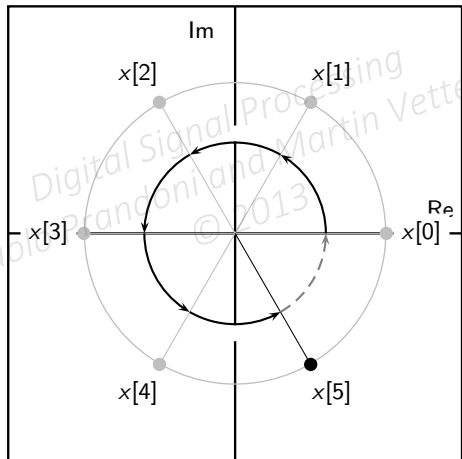
How “fast” can we go?

$$\omega = 2\pi/12$$



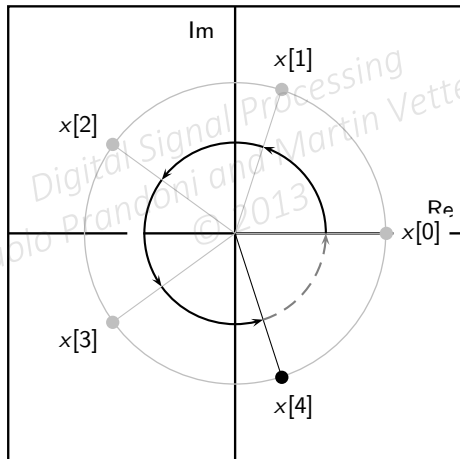
How “fast” can we go?

$$\omega = 2\pi/6$$



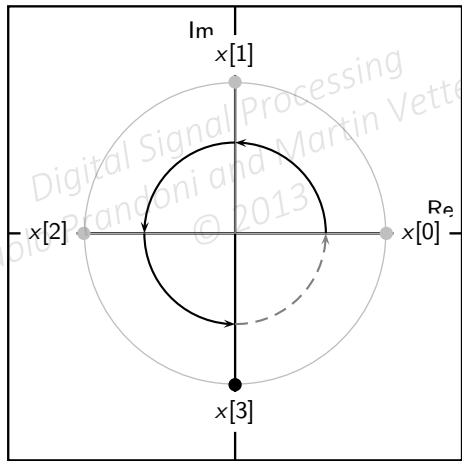
How “fast” can we go?

$$\omega = 2\pi/5$$



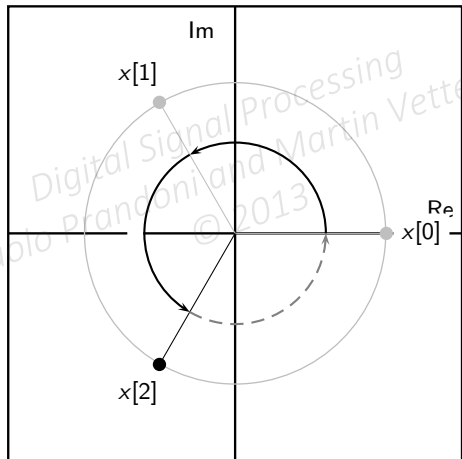
How “fast” can we go?

$$\omega = 2\pi/4$$



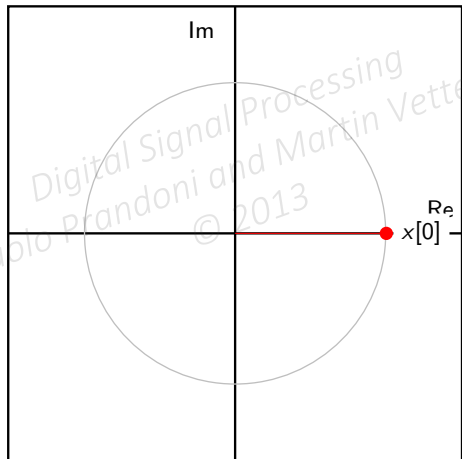
How “fast” can we go?

$$\omega = 2\pi/3$$



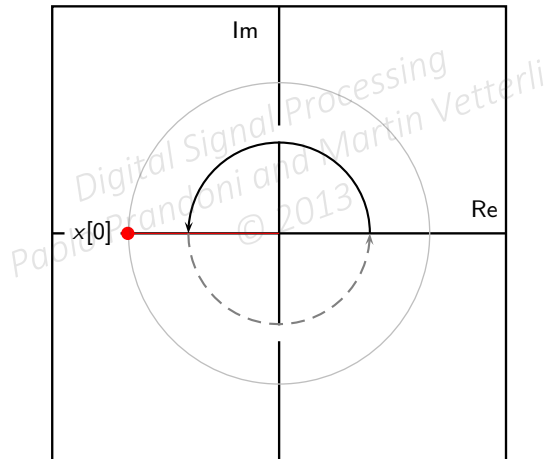
How “fast” can we go?

$$\omega = 2\pi/2 = \pi$$



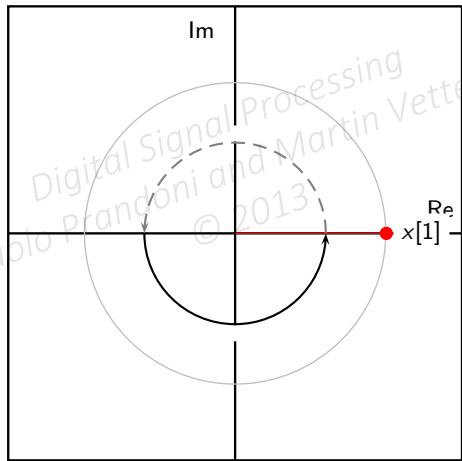
How “fast” can we go?

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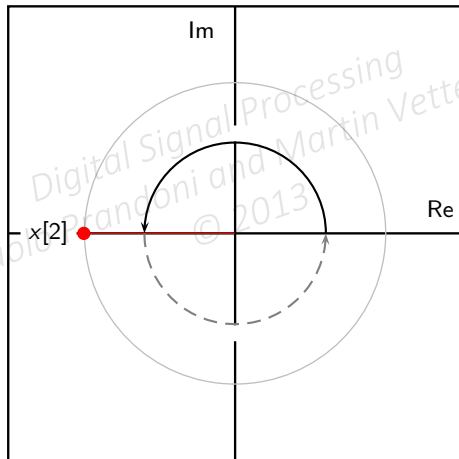
How “fast” can we go?

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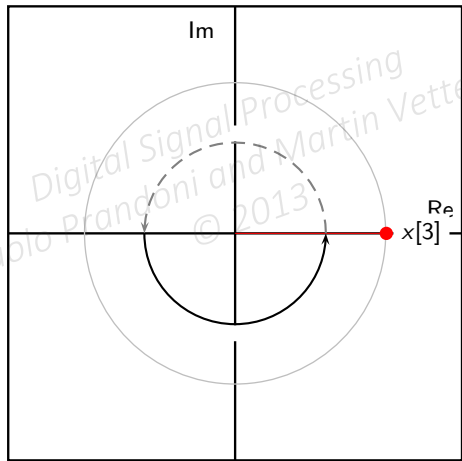
How “fast” can we go?

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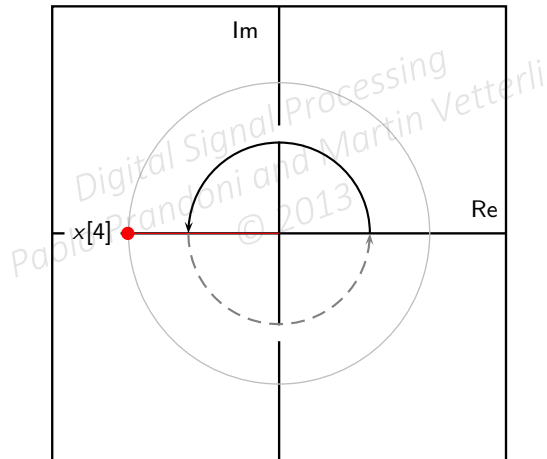
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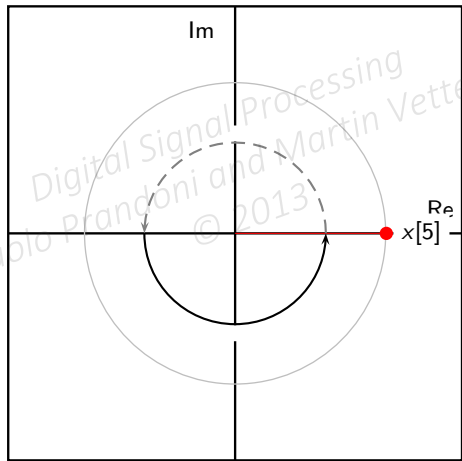
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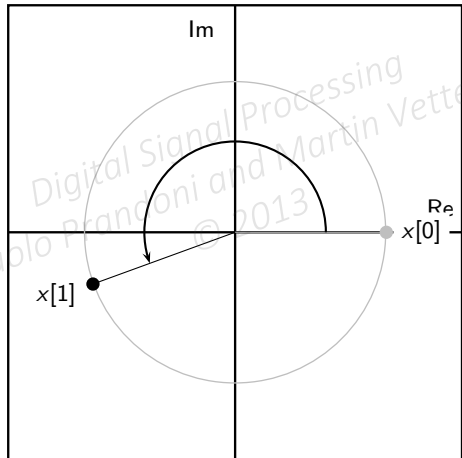
How “fast” can we go?

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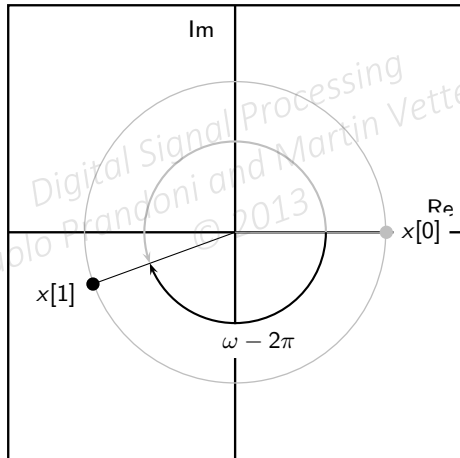
What if we go “faster”?

$$\pi < \omega < 2\pi$$

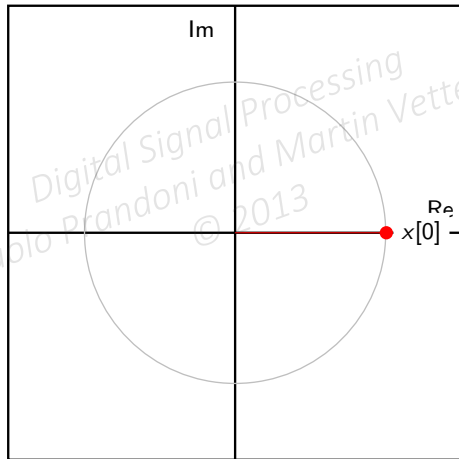


What if we go “faster”?

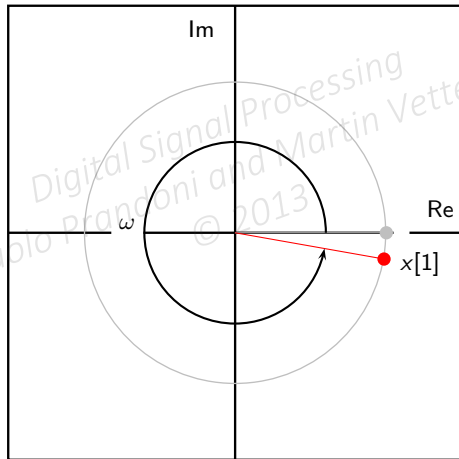
$$\pi < \omega < 2\pi$$



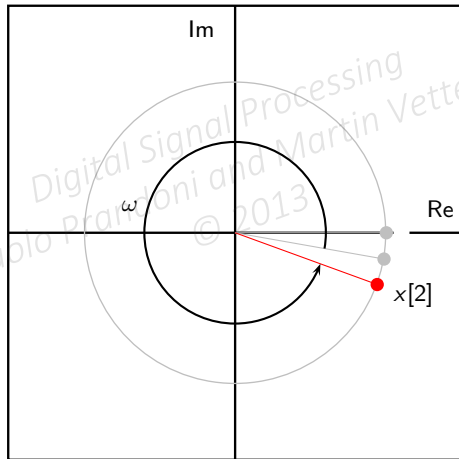
$$\omega = 2\pi - \alpha, \quad \alpha \text{ small}$$



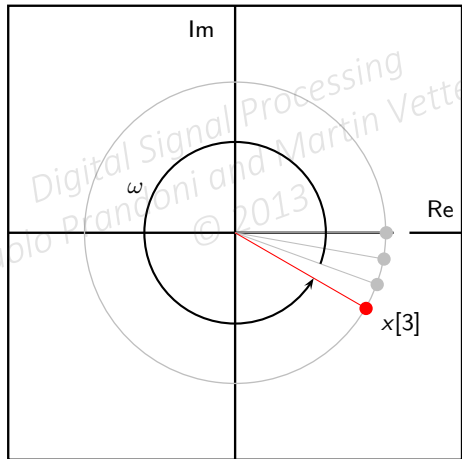
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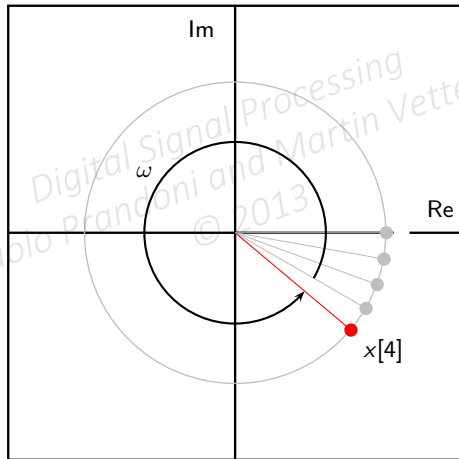
$$\omega = 2\pi - \alpha, \quad \alpha \text{ small}$$



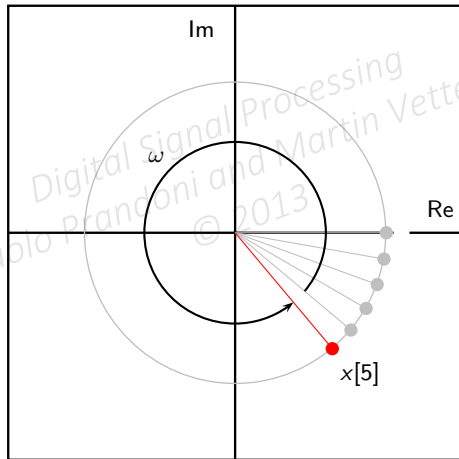
$$\omega = 2\pi - \alpha, \quad \alpha \text{ small}$$



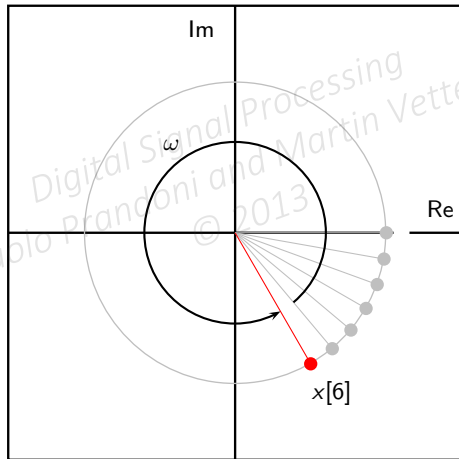
$$\omega = 2\pi - \alpha, \quad \alpha \text{ small}$$



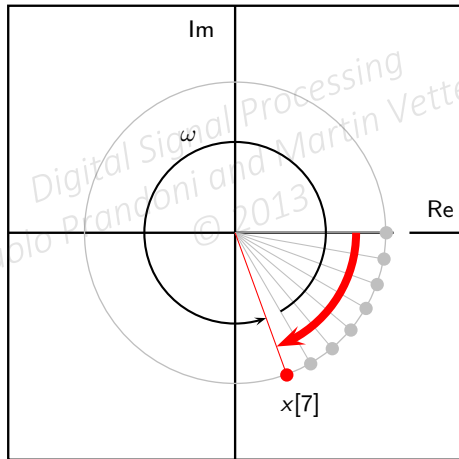
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► Discrete time:

- n : no physical dimension (just a counter)
- periodicity: how many samples before pattern repeats

► “Real world”:

- periodicity: how many *seconds* before pattern repeats
- frequency measured in Hz (s^{-1})

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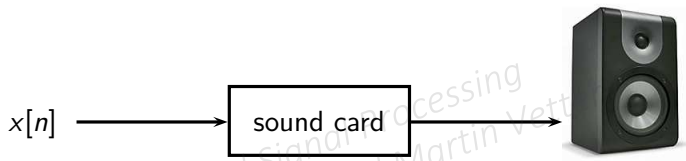
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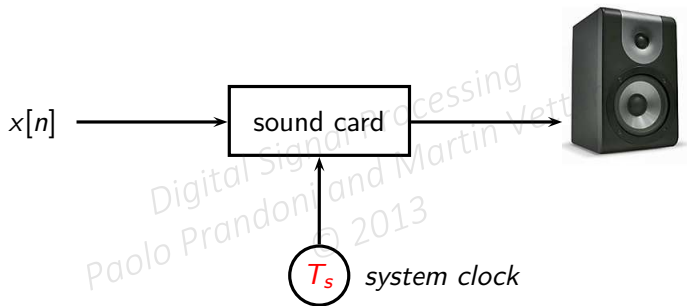
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- ▶ set T_s , time in seconds between samples
- ▶ periodicity of M samples \longrightarrow periodicity of MT_s seconds
- ▶ real world frequency:

$$f = \frac{1}{MT_s}$$

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END OF MODULE 2.2

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Module 2.3: the Karplus-Strong algorithm

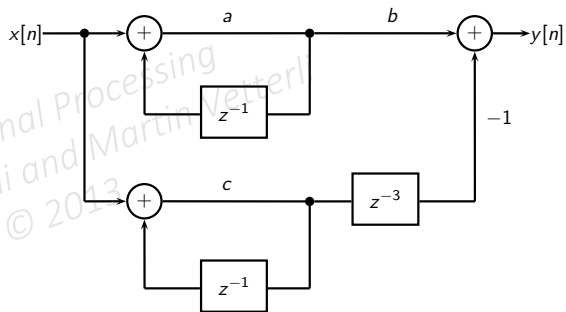
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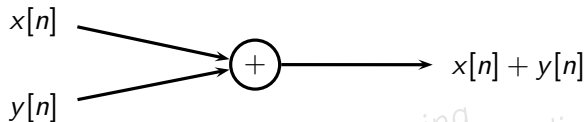
- ▶ DSP building blocks
- ▶ moving averages and simple feedback loops
- ▶ a sound synthesizer

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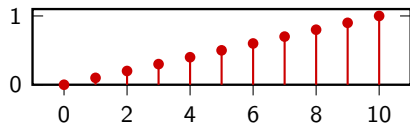
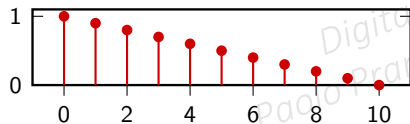
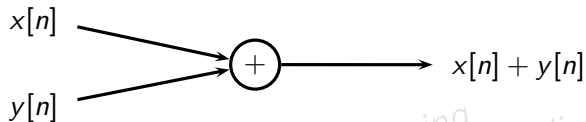
- ▶ DSP as Lego: The fundamental building blocks
- ▶ Averages and moving averages
- ▶ Recursion: Revisiting your bank account
- ▶ Building a simple recursive synthesizer
- ▶ Examples of sounds

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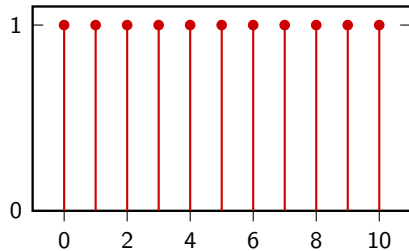
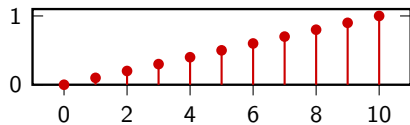
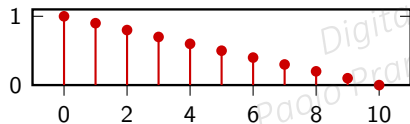
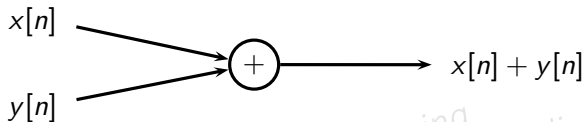




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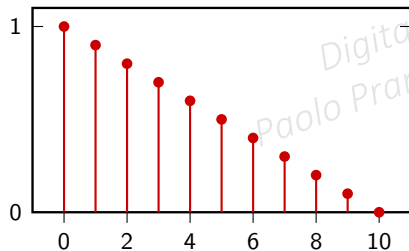


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$$x[n] \xrightarrow{\alpha} \alpha x[n]$$

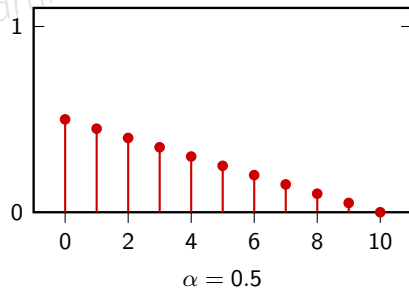
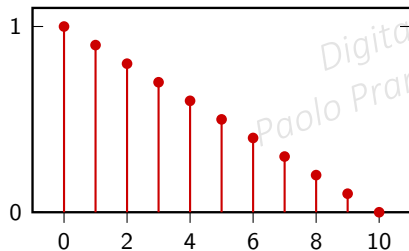
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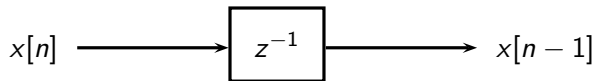
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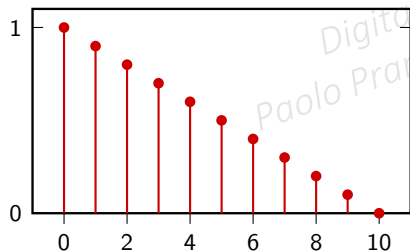
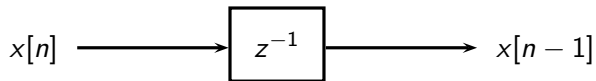
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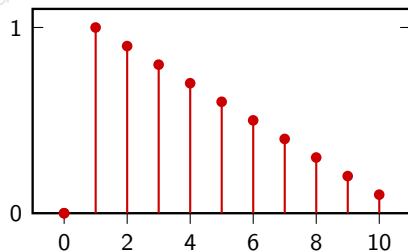
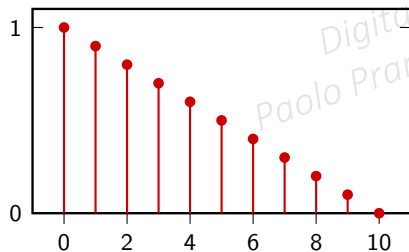
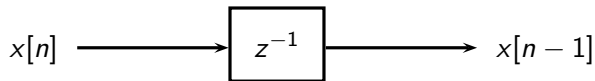




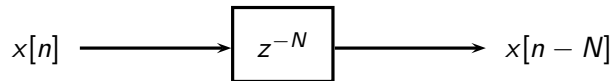
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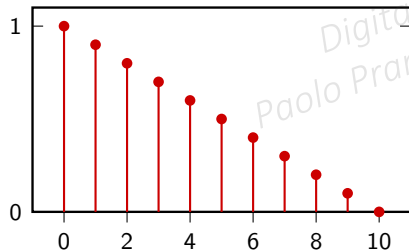
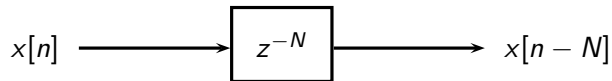
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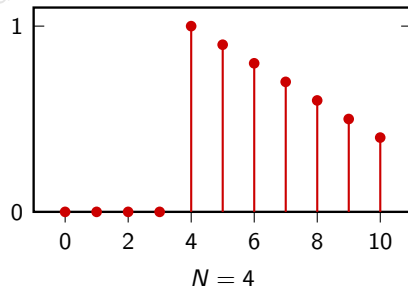
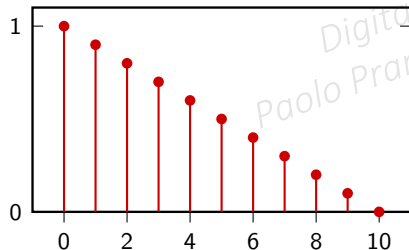
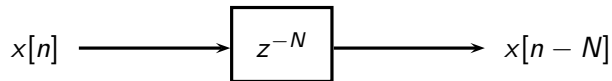
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- ▶ simple average:

$$m = \frac{a + b}{2}$$

- ▶ moving average: take a “local” average

$$y[n] = \frac{x[n] + x[n-1]}{2}$$

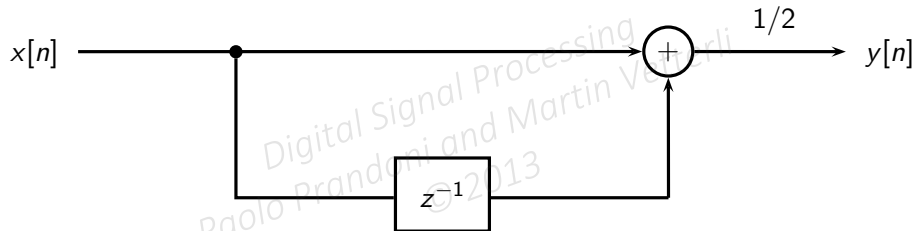
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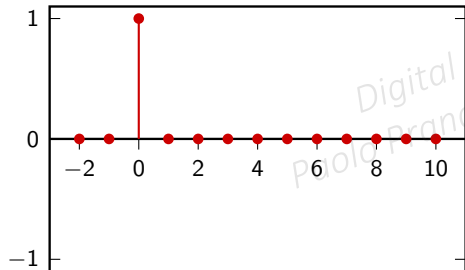
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The 2-point Moving Average Using Lego



Let's average...

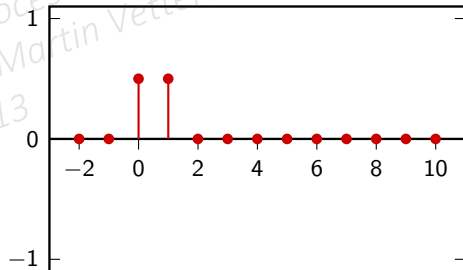
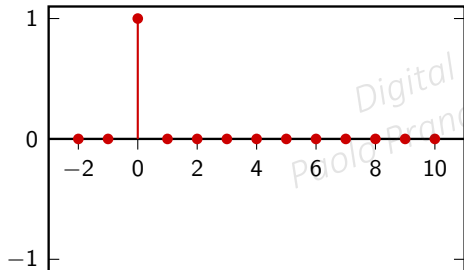
$$x[n] = \delta[n]$$



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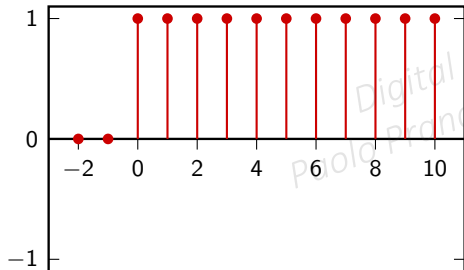
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Let's average...

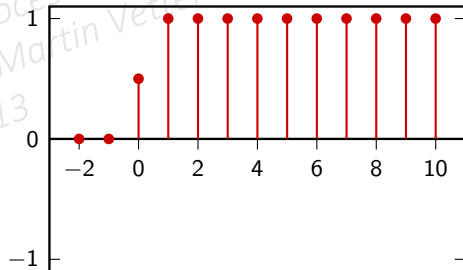
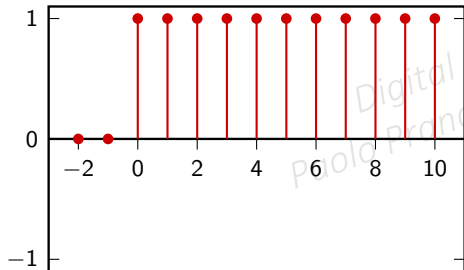
$$x[n] = u[n]$$



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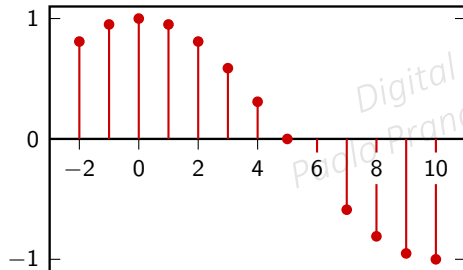
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Let's average...

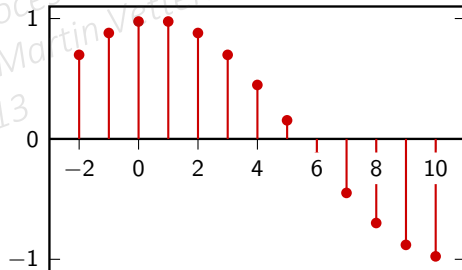
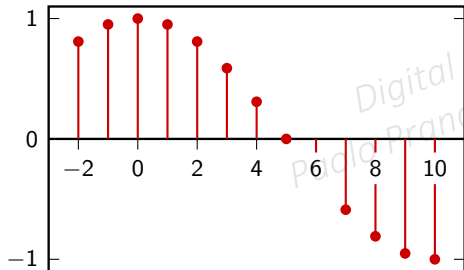
$$x[n] = \cos(\omega n), \quad \omega = \pi/10$$



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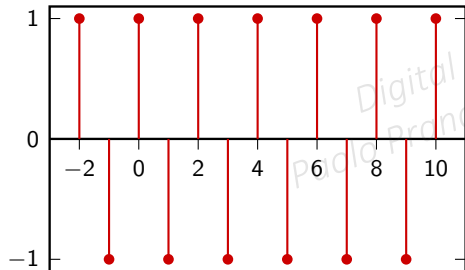
Let's average...

$$x[n] = \cos(\omega n), \quad \omega = \pi/10$$



Let's average...

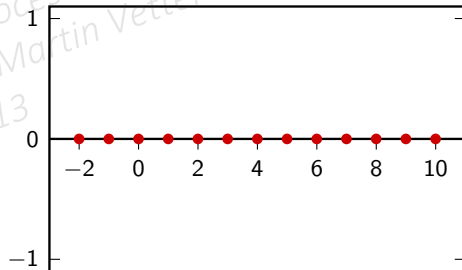
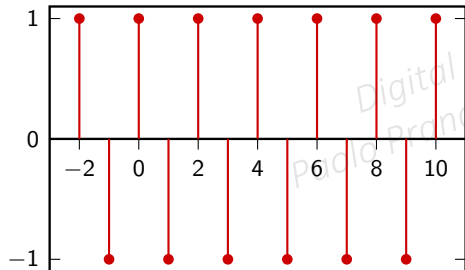
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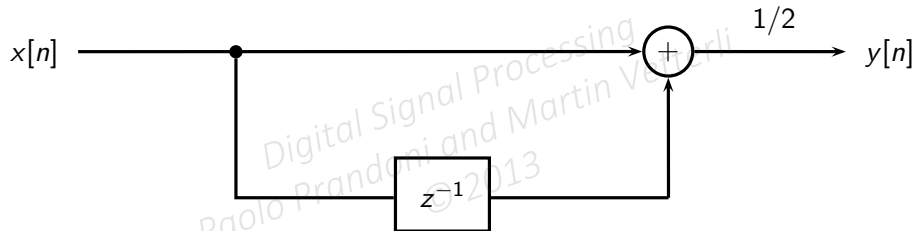
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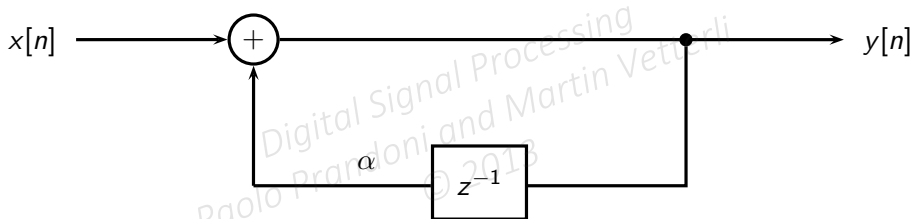
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What if we reverse the loop?



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A simple equation to describe compound interest:

- ▶ constant interest/borrowing rate of 5% per year
- ▶ interest accrues on Dec 31
- ▶ deposits/withdrawals during year n : $x[n]$
- ▶ balance at year n :

$$y[n] = 1.05 y[n-1] + x[n]$$

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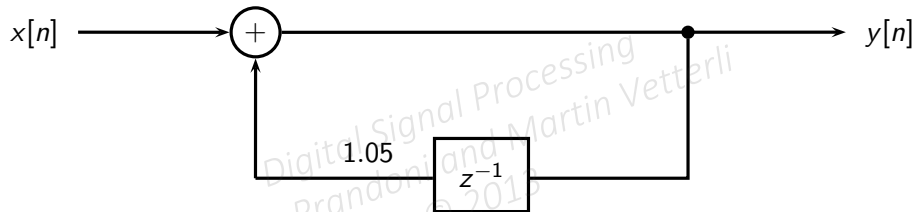
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Example: the one-time investment

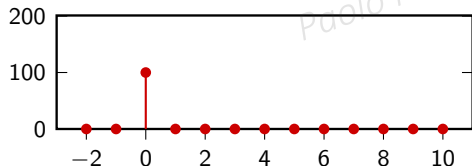
$$x[n] = 100 \delta[n]$$

▶ $y[0] = 100$

▶ $y[1] = 105$

▶ $y[2] = 110.25$, $y[3] = 115.7625$ etc.

▶ In general: $y[n] = (1.05)^n 100 u[n]$



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Example: the one-time investment

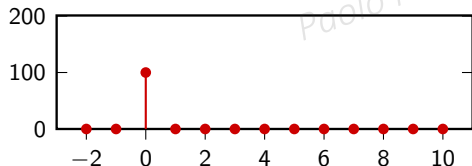
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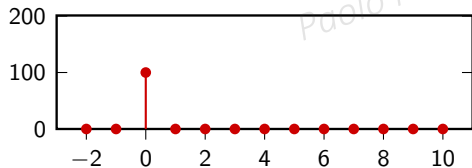
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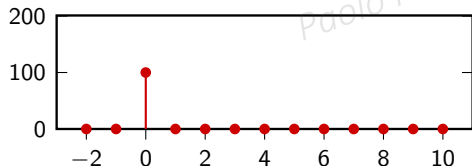
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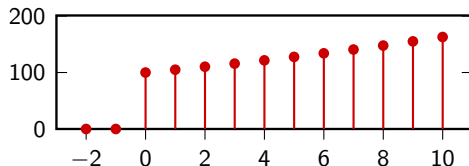
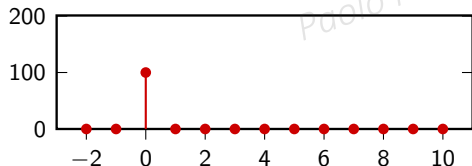
$$x[n] = 100 \delta[n]$$

► $y[0] = 100$

► $y[1] = 105$

► $y[2] = 110.25$, $y[3] = 115.7625$ etc.

► In general: $y[n] = (1.05)^n 100 u[n]$



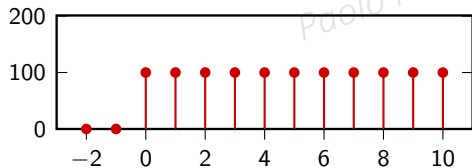
$$x[n] = 100 u[n]$$

▶ $y[0] = 100$

▶ $y[1] = 205$

▶ $y[2] = 315.25$, $y[3] = 431.0125$ etc.

▶ In general: $y[n] = 2000 ((1.05)^{n+1} - 1) u[n]$



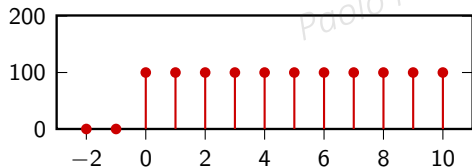
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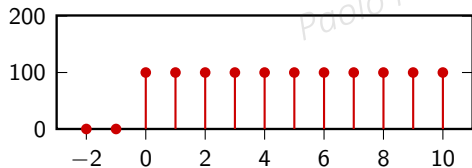
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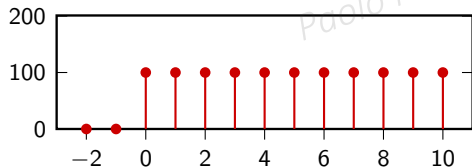
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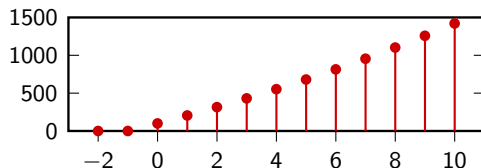
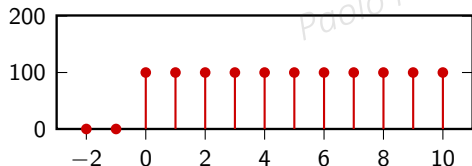
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Example: The independently wealthy

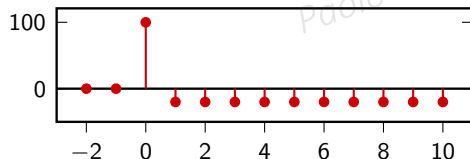
$$x[n] = 100 \delta[n] - 5 u[n - 1]$$

▶ $y[0] = 100$

▶ $y[1] = 100$

▶ $y[2] = 100, y[3] = 100$ etc.

▶ In general: $y[n] = 100 u[n]$



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Example: The independently wealthy

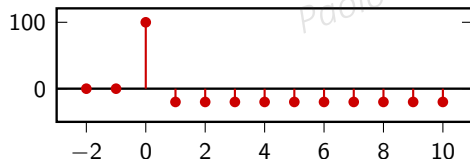
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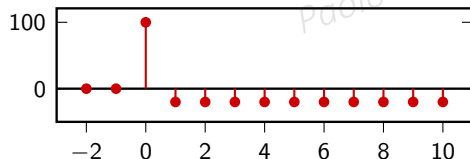
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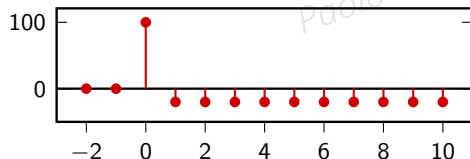
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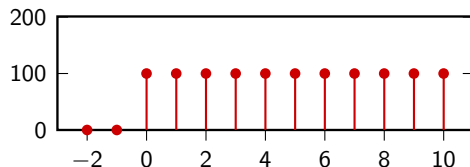
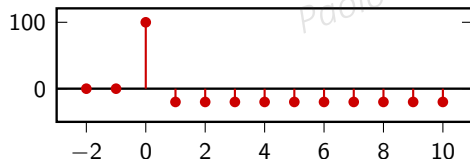


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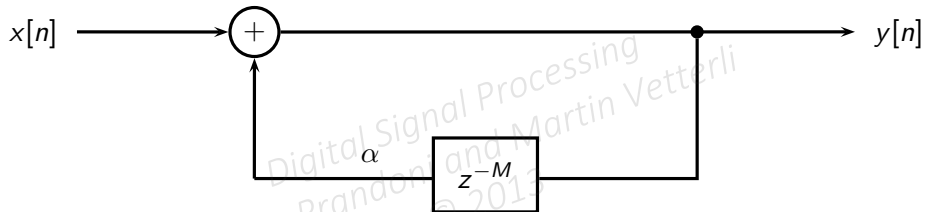
Example: The independently wealthy

$$x[n] = 100 \delta[n] - 5 u[n - 1]$$

- ▶ $y[0] = 100$
- ▶ $y[1] = 100$
- ▶ $y[2] = 100, y[3] = 100$ etc.
- ▶ In general: $y[n] = 100 u[n]$



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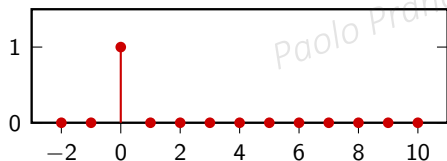
$$y[n] = \alpha y[n - M] + x[n]$$

$$M = 3, \alpha = 0.7, x[n] = \delta[n]$$

► $y[0] = 1, y[1] = 0, y[2] = 0$

► $y[3] = 0.7, y[4] = 0, y[5] = 0$

► $y[6] = 0.7^2, y[7] = 0, y[8] = 0, \text{ etc.}$



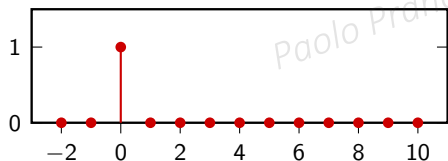
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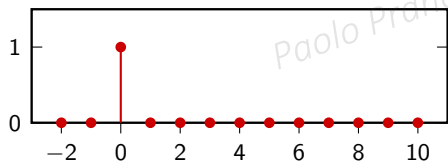
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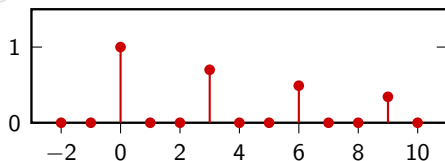
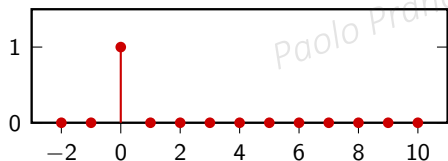
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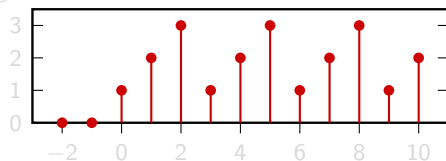
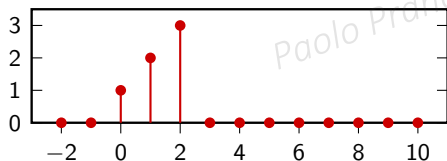
- ▶ $y[0] = 1, y[1] = 0, y[2] = 0$
- ▶ $y[3] = 0.7, y[4] = 0, y[5] = 0$
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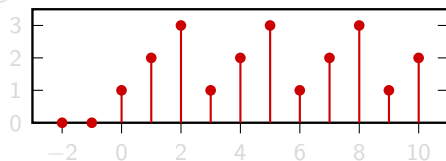
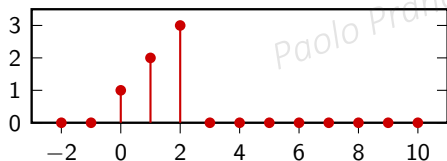
$$M = 3, \alpha = 1, x[n] = \delta[n] + 2\delta[n-1] + 3\delta[n-2]$$

- ▶ $y[0] = 1, y[1] = 2, y[2] = 3$
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- ▶ $y[6] = 1, y[7] = 2, y[8] = 3, \text{ etc.}$



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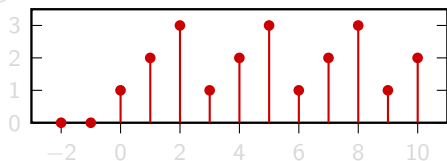
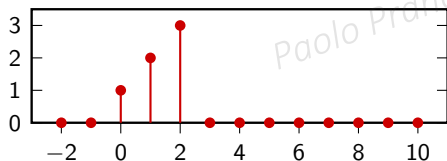


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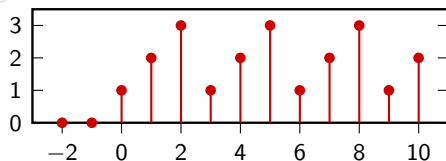
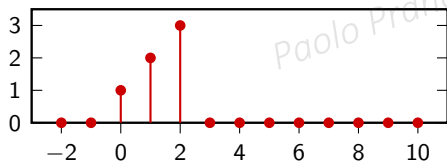
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- ▶ build a recursion loop with a delay of M

- ▶ choose a signal $\bar{x}[n]$ that is nonzero only for $0 \leq n \leq M$

- ▶ choose a decay factor

- ▶ input $\bar{x}[n]$ to the system

- ▶ play the output

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- ▶ build a recursion loop with a delay of M
- ▶ choose a signal $\bar{x}[n]$ that is nonzero only for $0 \leq n < M$
- ▶ choose a decay factor
- ▶ input $\bar{x}[n]$ to the system
- ▶ play the output

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We can make music with that!



- ▶ build a recursion loop with a delay of M
- ▶ choose a signal $\bar{x}[n]$ that is nonzero only for $0 \leq n < M$
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- ▶ M -tap delay \rightarrow M -sample “periodicity”

- ▶ associate time T to sample interval

- ▶ periodic signal of frequency

- ▶ example: $T = 22.7\mu\text{s}$, $M = 100$

$$f \approx 440\text{Hz}$$

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by Protoni and Martin Vetterli
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- ▶ associate time T to sample interval
- ▶ periodic signal of frequency

$$f = \frac{1}{MT} \text{ Hz}$$

▶ example: $T = 22.7 \mu\text{s}$, $M = 100$

$$f \approx 440 \text{ Hz}$$

Digital Signal Processing
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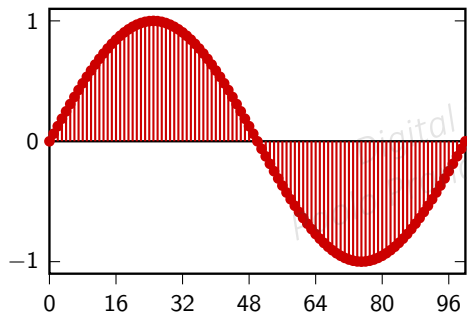
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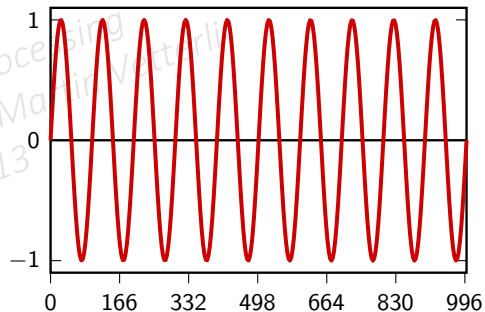
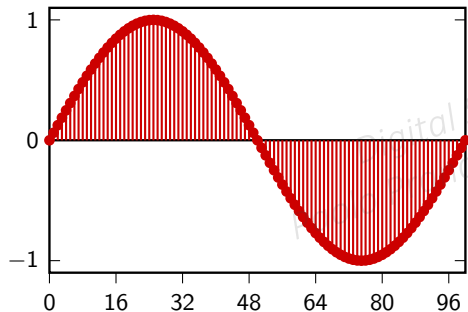
$$f \approx 440 \text{ Hz}$$

$M = 100$, $\alpha = 1$, $\bar{x}[n] = \sin(2\pi n/100)$ for $0 \leq n < 100$ and zero elsewhere



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$M = 100$, $\alpha = 1$, $\bar{x}[n] = \sin(2\pi n/100)$ for $0 \leq n < 100$ and zero elsewhere



- ▶ M controls frequency (pitch)

- ▶ α controls envelope (decay)

- ▶ $\bar{x}[n]$ controls color (timbre)

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- ▶ M controls frequency (pitch)

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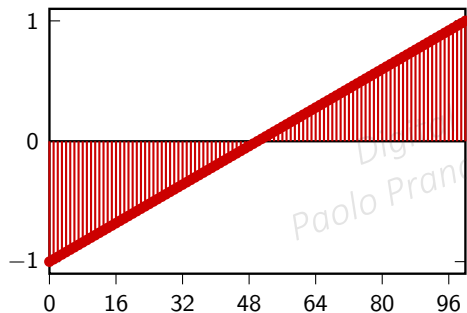
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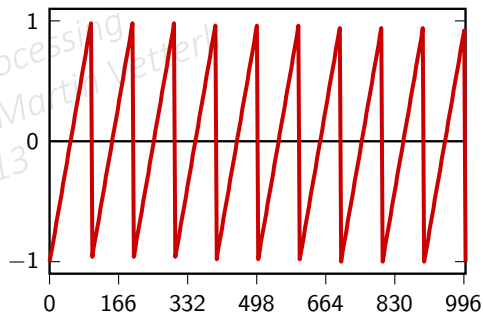
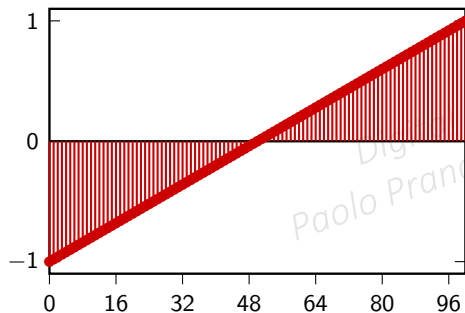
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$M = 100$, $\alpha = 0.95$, $\bar{x}[n]$: zero-mean sawtooth wave between 0 and 99, zero elsewhere



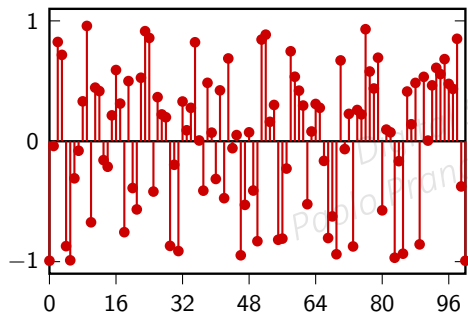
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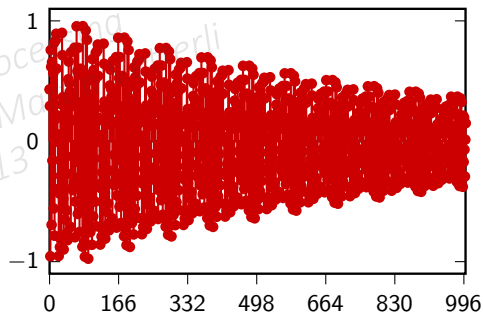
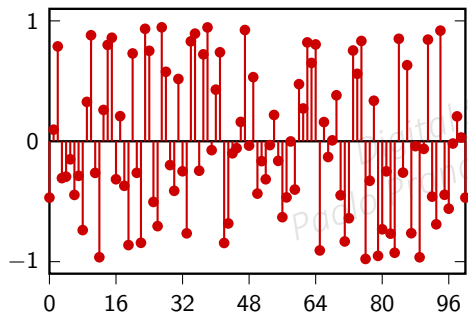


The Karplus-Strong Algorithm

$M = 100$, $\alpha = 0.9$, $\bar{x}[n]$: 100 random values between 0 and 99, zero elsewhere



$M = 100$, $\alpha = 0.9$, $\bar{x}[n]$: 100 random values between 0 and 99, zero elsewhere



- ▶ We have seen basic elements:
 - adders
 - multipliers
 - delays
- ▶ We have seen two systems
 - moving averages
 - recursive systems
- ▶ We were able to build simple systems with interesting properties
- ▶ to understand all of this in more details we need a mathematical framework!

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END OF MODULE 2.3

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