

# Experimental Methods

## Lecture 12

October 15<sup>th</sup>, 2020

# Kramers-Kronig dispersion relations

- In conclusion, for a causal system

$$P \int_{-\infty}^{\infty} \frac{h(\omega')}{\omega - \omega'} d\omega' \equiv \mathcal{H}(h(\omega)) = i\pi h(\omega)$$

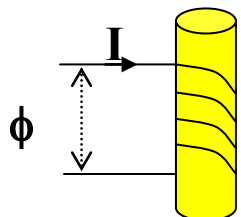
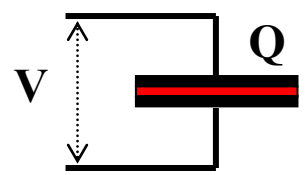
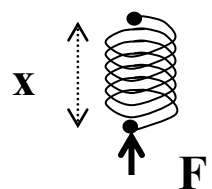
- or

$$Im\{h(\omega)\} = -\frac{1}{\pi} P \int_{-\infty}^{\infty} \frac{Re\{h(\omega')\}}{\omega - \omega'} d\omega' \equiv -\mathcal{H}(Re\{h(\omega)\})$$

$$Re\{h(\omega)\} = \frac{1}{\pi} P \int_{-\infty}^{\infty} \frac{Im\{h(\omega')\}}{\omega - \omega'} d\omega' \equiv \mathcal{H}(Im\{h(\omega)\})$$

# One example from condensed matter

**A general phenomenon in condensed matter**  
a frequency independent imaginary part in the “response”  $\chi(\omega)$

	<b>Magnetism</b>	<b>Dielectrics</b>	<b>Mechanics</b>
<b>Generalized response</b> $\chi(\omega)$	Inductance $L(\omega)$ $\phi(\omega) = L(\omega)I(\omega)$ $L(\omega) = L' - iL_0$	Capacitance $C(\omega)$ $Q(\omega) = C(\omega)V(\omega)$ $C(\omega) = C'(\omega) - iC_0$	Spring constant $k(\omega)$ $F(\omega) = -k(\omega)x(\omega)$ $k(\omega) = k'(\omega) + ik_0$
<b>Nyquist formula</b>	$S_\phi(\omega) = 4k_B T \frac{L_0}{\omega}$	$S_Q(\omega) = 4k_B T \frac{C_0}{\omega}$	$S_F(\omega) = 4k_B T \frac{k_0}{\omega}$
			

# Consider the magnetic case

the magnetic flux-current relation:  $\phi(\omega) = L(\omega)I(\omega)$



- The magnetic susceptibility  $\chi$  of the core material is connected to inductance thru:

$$L(\omega) = \chi(\omega)L_0$$

- Magnetic susceptibility of real material is complex:

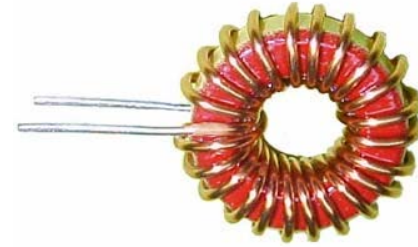
$$\chi(\omega) = \chi'(\omega) - i\chi''(\omega)$$

- Notice that impedance  $Z(\omega) = i\omega L(\omega)$  acquires a real part, i.e. a resistive component

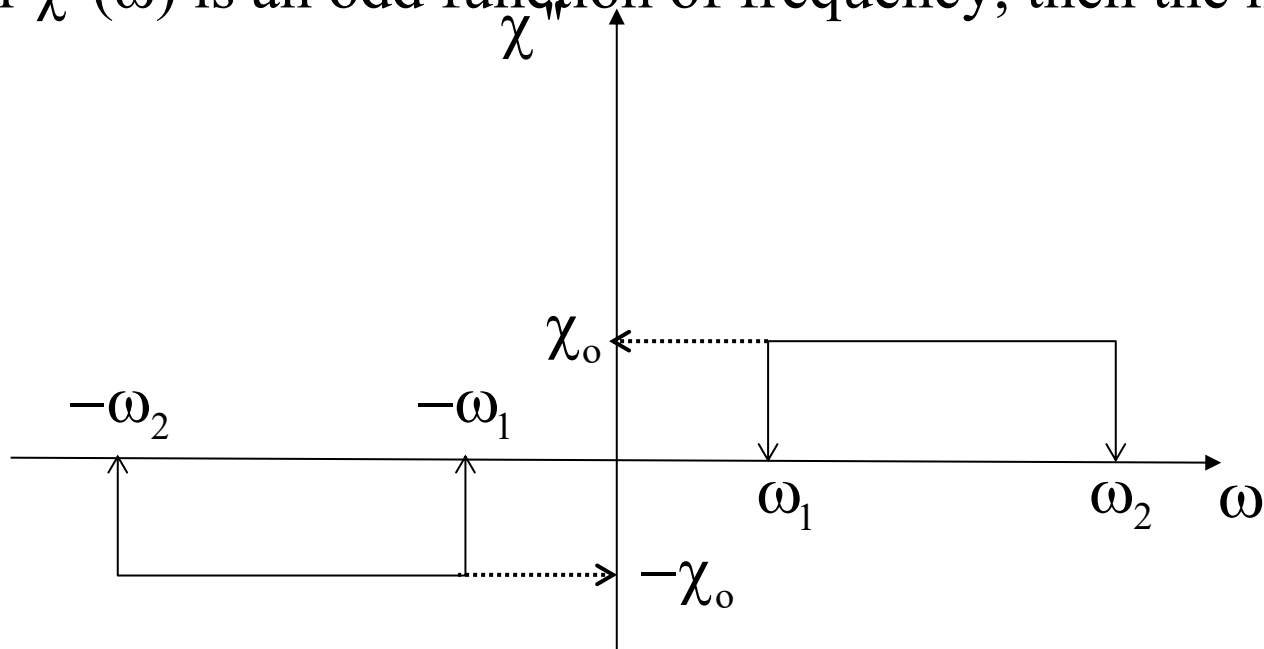
$$\text{Re}\{Z(\omega)\} = \omega\chi''(\omega)L_0$$

- At low frequency  $\chi''(\omega) \simeq \chi_0$  in many materials

# A popular model



- $\chi''(\omega) \approx \chi_o$ . Assume this is true in some decade-wide frequency interval, outside which  $\chi''(\omega) = 0$ . (Can't be infinite otherwise there is no Fourier transform in strict sense)
- Remember  $\chi''(\omega)$  is an odd function of frequency, then the model is:



**Characteristic Function**

$$\chi_{[a,b]}(t)$$

$$\frac{1}{\pi} \ln \left| \frac{t-a}{t-b} \right|$$

# Real part from Kramers Kronig

## Characteristic Function

$$\chi_{[a,b]}(t)$$

$$\frac{1}{\pi} \ln \left| \frac{t-a}{t-b} \right|$$

- Use Hilbert transform

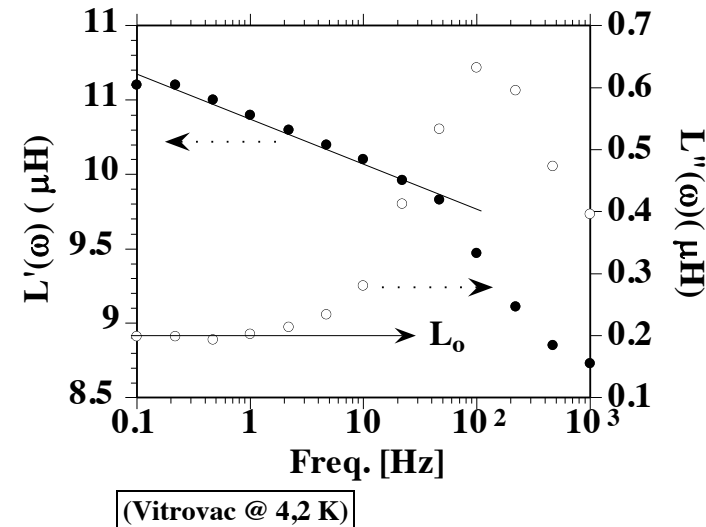
$$\chi'(\omega) = \mathcal{H}\{-\chi''\} = -\frac{\chi_o}{\pi} \ln \left| \frac{\omega-\omega_1}{\omega-\omega_2} \right| + \frac{\chi_o}{\pi} \ln \left| \frac{\omega+\omega_2}{\omega+\omega_1} \right| = -\frac{\chi_o}{\pi} \ln \left| \frac{\omega^2-\omega_1^2}{\omega^2-\omega_2^2} \right|$$

- For  $\omega_1 \ll \omega \ll \omega_2$

$$\chi'(\omega) = -\frac{2\chi_o}{\pi} (\ln(\omega) - \ln(\omega_2))$$

- Notice

$$\frac{1}{\chi''} \frac{d\chi'}{d\ln(\omega)} = -\frac{2}{\pi}$$



# Complex index of refraction and absorption

[\[edit\]](#)

See also: [Mathematical descriptions of opacity](#)

When light passes through a medium, some part of it will always be [absorbed](#). This can be conveniently taken into account by defining a complex index of refraction,

$$\tilde{n} = n + i\kappa.$$

Here, the real part of the refractive index  $n$  indicates the phase speed, while the imaginary part  $\kappa$  indicates the amount of absorption loss when the electromagnetic wave propagates through the material.

That  $\kappa$  corresponds to absorption can be seen by inserting this refractive index into the expression for [electric field](#) of a [plane](#) electromagnetic wave traveling in the  $z$ -direction. We can do this by relating the [wave number](#) to the refractive index through  $k = \frac{2\pi n}{\lambda_0}$ , with  $\lambda_0$  being the vacuum wavelength. With complex wave number  $\tilde{k}$  and refractive index  $n + i\kappa$  this can be inserted into the plane wave expression as

$$\mathbf{E}(z, t) = \text{Re}(\mathbf{E}_0 e^{i(\tilde{k}z - \omega t)}) = \text{Re}(\mathbf{E}_0 e^{i(2\pi(n + i\kappa)z/\lambda_0 - \omega t)}) = e^{-2\pi\kappa z/\lambda_0} \text{Re}(\mathbf{E}_0 e^{i(kz - \omega t)}).$$

Here we see that  $\kappa$  gives an exponential decay, as expected from [Beer–Lambert law](#).

$\kappa$  is often called the **extinction coefficient** in physics although this has a [different definition within chemistry](#). Both  $n$  and  $\kappa$  are dependent on the frequency. In most circumstances  $\kappa > 0$  (light is absorbed) or  $\kappa = 0$  (light travels forever without loss). In special situations, especially in the [gain medium](#) of [lasers](#), it is also possible that  $\kappa < 0$ , corresponding to an amplification of the light.

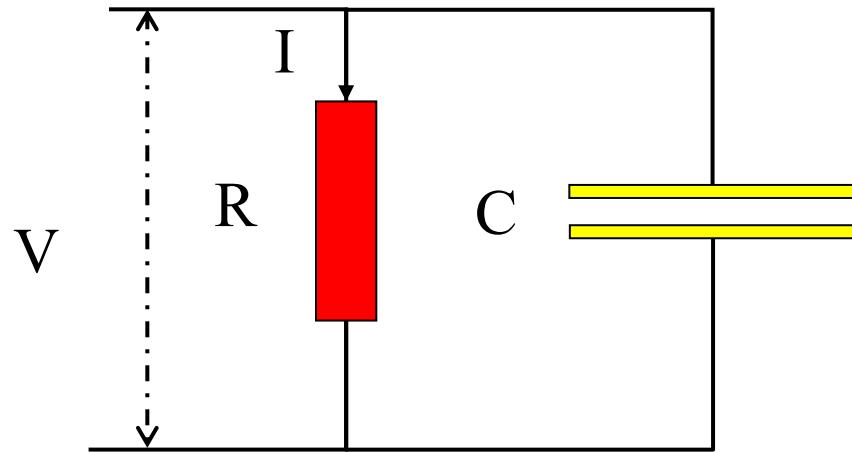
An alternative convention uses  $\tilde{n} = n - i\kappa$  instead of  $\tilde{n} = n + i\kappa$ , but where  $\kappa > 0$  still corresponds to loss. Therefore these two conventions are inconsistent and should not be confused. The difference is related to defining sinusoidal time dependence as  $\text{Re}(e^{-i\omega t})$  versus  $\text{Re}(e^{+i\omega t})$ . See [Mathematical descriptions of opacity](#).

Dielectric loss and non-zero DC conductivity in materials cause absorption. Good dielectric materials such as glass have extremely low DC conductivity, and at low frequencies the dielectric loss is also negligible, resulting in almost no absorption ( $\kappa \approx 0$ ). However, at higher frequencies (such as visible light), dielectric loss may increase absorption significantly, reducing the material's [transparency](#) to these frequencies.

The real and imaginary parts of the complex refractive index are related through the [Kramers–Kronig relations](#). For example, one can determine a material's full complex refractive index as a function of wavelength from an absorption spectrum of the material.

For [X-ray](#) and [extreme ultraviolet](#) radiation the complex refractive index deviates only slightly from unity and usually has a real part smaller than 1. It is therefore normally written as  $\tilde{n} = 1 - \delta + i\beta$  (or  $\tilde{n} = 1 - \delta - i\beta$ ).<sup>[14]</sup>

# Exercise: verify that



$$Z(\omega) = \frac{V(\omega)}{I(\omega)}$$

$$Z(\omega) = \frac{R/i\omega C}{R + 1/i\omega C} = \frac{R}{1 + i\omega\tau} \quad \tau = RC$$

Fulfil Kramers-Kronig



# What should you know on systems

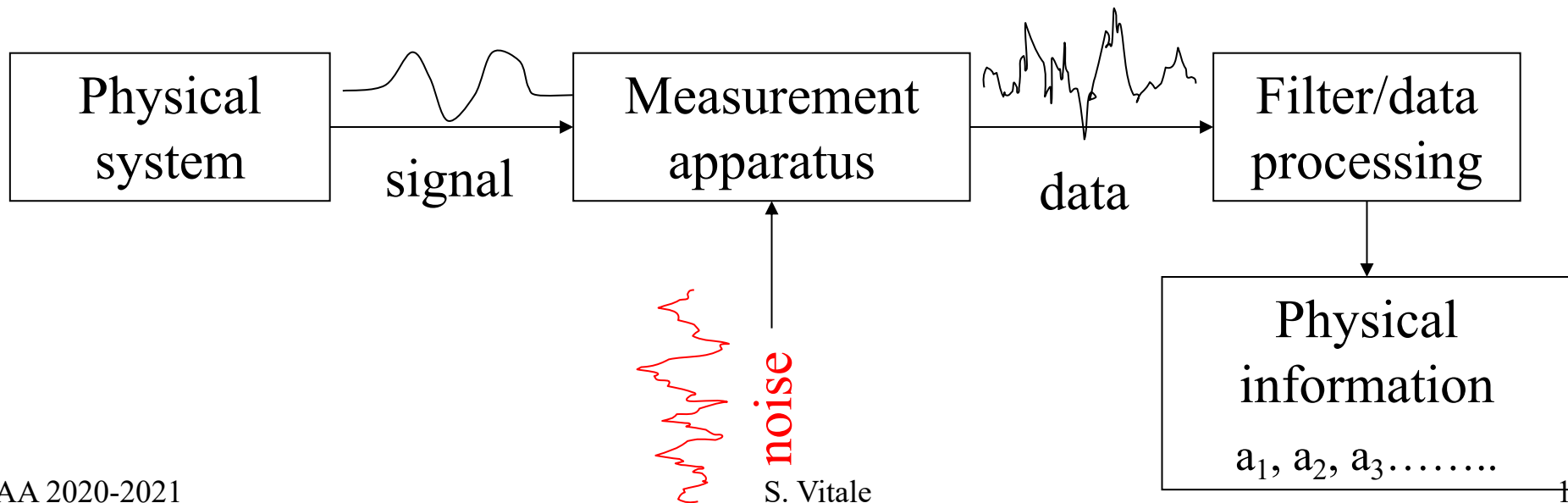
- Impulse response
- Step response
- Frequency response and transfer function
- Role of free evolution
- Linear damped oscillator, simple pole (low pass) and archetypal transfer functions
- Feedback, stability and linearization
- Dispersion relations

# Noise in physical systems

- This part describes disturbances of statistical nature in physical measurements.
- To describe noise we are going to set up the basis of the mathematical theory of *stochastic processes*.
- A good reference text for this part is “Probability, random variables, and stochastic processes” / Athanasios Papoulis, 4. ed Boston, Mass. [etc.] : McGraw-Hill, c2002. (available in Italian)

# Noise

Noise in physical experiments is described as a random signal  $x(t)$ :  
Independent repetitions of the same experiment produce different functions of time  $x(t)$



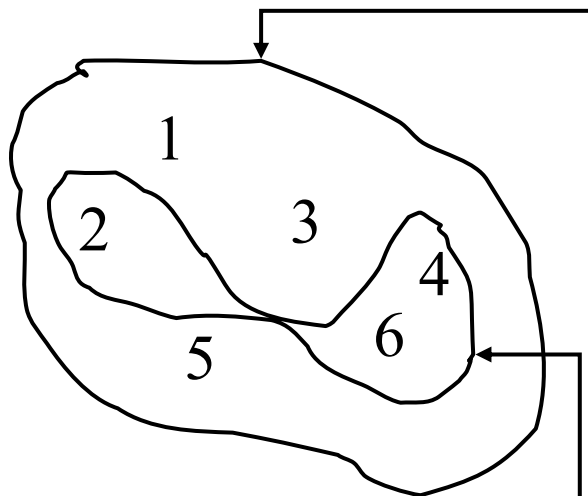
# What you should know about probability

## 1. The concept of experiment:

- a. A set  $S$  of all possible results
- b. A class of subsets of  $S$  called “the events”
- c. A probability  $P(A)$  associated to events

## 1. Various constraints hold

- a.  $0 \leq P(A) \leq 1$
- b.  $S$  is an event (certain)
- c.  $\emptyset$  is an event (impossible)
- d. Etc.....

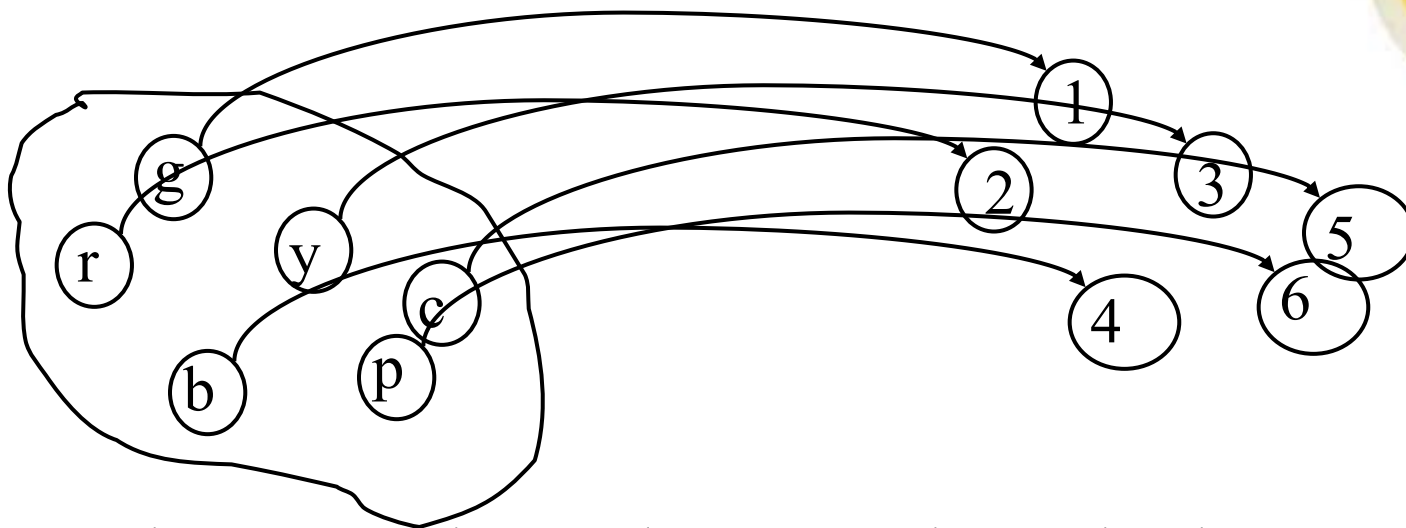


Set  $S$  of possible outcomes (example throwing a dice)

Event  $A$ , a subset of  $S$ . (Example: “I got an even number”)

# What you should know about probability

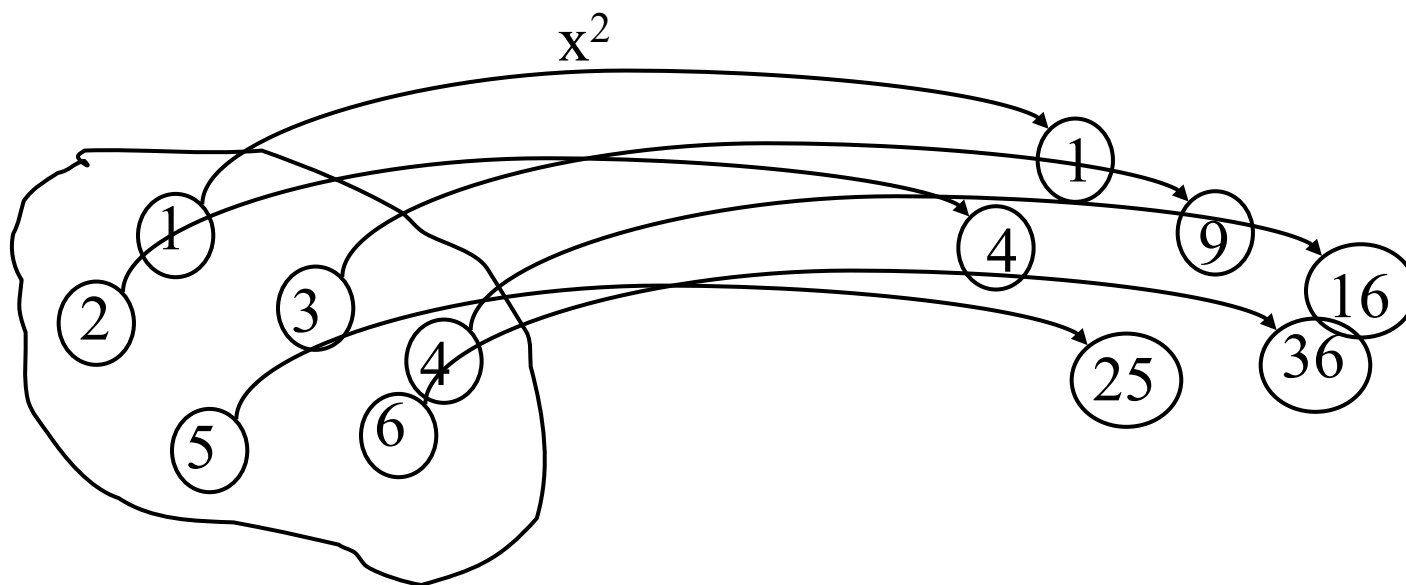
- Random variable (rv): a rule that associates numbers to experiment outcomes



- For each outcome the rv takes one value and only one

# What you should know about probability

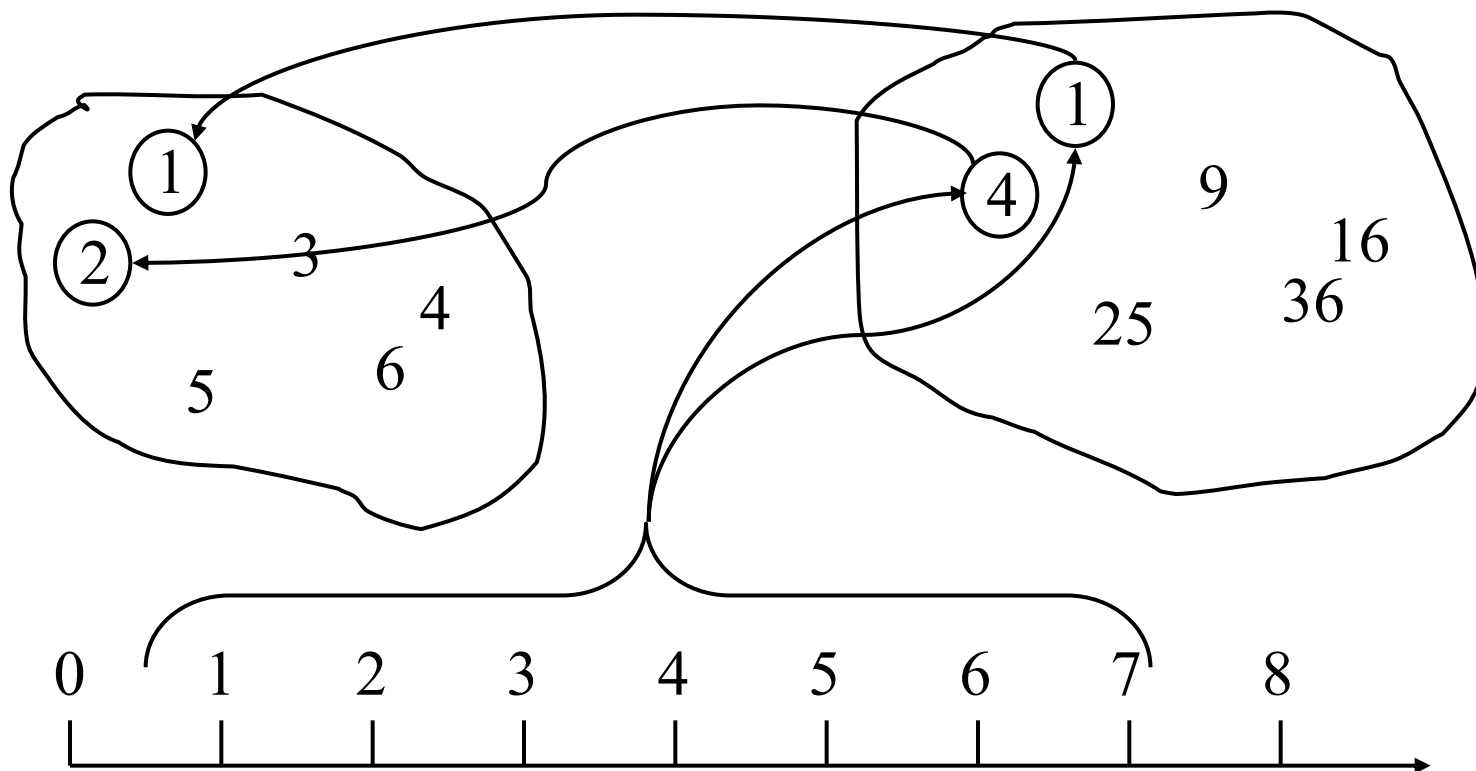
- Random variable (rv): a rule that associates numbers to experiment outcomes



- For each outcome the rv takes one value and only one

# What you should know about probability

- The rv associates real intervals to events



- Thus a probability value is associated to each interval of the real axis

# What you should know about probability

- Probability of the event corresponding to outcomes for which the random variable takes a value  $x_0 \leq x \leq x_1$ .

$$P\{x_0 \leq x \leq x_1\} = \int_{x_0}^{x_1} f(x) dx$$

- $f(x)$  is the *probability density*
- Notice, as for each possible outcome the rv associates a number, the event  $-\infty \leq x \leq \infty$  is certain.

$$P\{-\infty \leq x \leq \infty\} = \int_{-\infty}^{\infty} f(x) dx = 1$$

- For continuous variables

$$P\{x \equiv x_0\} = \int_{x_0}^{x_0} f(x) dx = 0$$

- In discrete experiments (like coin  <sup>$x_0$</sup> tossing), finite probability may be associated to single numbers. Around one of these numbers,  $x_0$ ,  $f(x)$  is

$$f(x) = P\{x_0\} \delta(x - x_0)$$



# What you should know about probability

- From the probability density function

$$P\{x_o \leq x \leq x_1\} = \int_{x_o}^{x_1} f(x) dx$$

- One can define the *cumulative distribution function*

$$P\{-\infty \leq x \leq x_o\} = \int_{-\infty}^{x_o} f(x) dx \equiv F(x_o)$$

- You can easily demonstrate that

$$f(x) = \frac{dF(x)}{dx}$$

# What you should know about probability

- Basic parameters of distribution. Let's call  $x$  the random variable and  $f_x(\chi)$  its probability density

- Mean value

$$\langle x \rangle = \int_{-\infty}^{\infty} \chi f_x(\chi) d\chi$$

- Mean value of any function of the  $g(x)$  of the  $x$

$$\langle g(x) \rangle = \int_{-\infty}^{\infty} g(\chi) f_x(\chi) d\chi$$

- Variance

$$\sigma^2 \equiv \left\langle \left( x - \langle x \rangle \right)^2 \right\rangle = \int_{-\infty}^{\infty} \left( \chi - \langle x \rangle \right)^2 f_x(\chi) d\chi$$

- A classical result:

$$\sigma^2 = \int_{-\infty}^{\infty} \chi^2 f_x(\chi) d\chi + \langle x \rangle^2 \int_{-\infty}^{\infty} f_x(\chi) d\chi - 2 \langle x \rangle \int_{-\infty}^{\infty} \chi f_x(\chi) d\chi$$

- then 
$$\sigma^2 = \langle x^2 \rangle + \langle x \rangle^2 - 2 \langle x \rangle^2 = \langle x^2 \rangle - \langle x \rangle^2$$

# What you should know about probability

- In general

- Moment of order  $m$

$$\mu'_m = \langle x^m \rangle = \int_{-\infty}^{\infty} x^m f_x(x) dx$$

- Central moment of order  $m$

$$\mu_m = \langle (x - \langle x \rangle)^m \rangle = \int_{-\infty}^{\infty} (x - \langle x \rangle)^m f_x(x) dx$$

- Mean value is equal to  $\mu_1'$ , variance is equal to  $\mu_2$
- If all moments are known  $f_x(x)$  is also known

# Key distributions

- Normal

- Density

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-x_0)^2}{2\sigma^2}}$$

- Cumulative distribution

$$F(x) = \text{Erf}\left[\frac{x - x_0}{\sigma}\right]$$

- Mean

$$\int_{-\infty}^{\infty} \frac{x}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-x_0)^2}{2\sigma^2}} dx = x_0$$

- Variance

$$\int_{-\infty}^{\infty} \frac{(x - x_0)^2}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-x_0)^2}{2\sigma^2}} dx = \sigma^2$$



# Key distributions

- Rectangular

$$f(x) = \Pi(x)$$

- Density

- Cumulative distribution

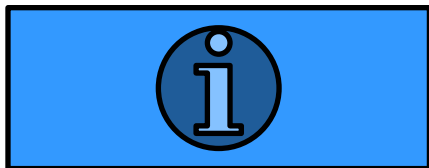
$$F(x) = (x + 1/2)\Pi(x) + \Theta(x - 1/2)$$

- Mean

$$\int_{-\frac{1}{2}}^{\frac{1}{2}} x \, dx = 0$$

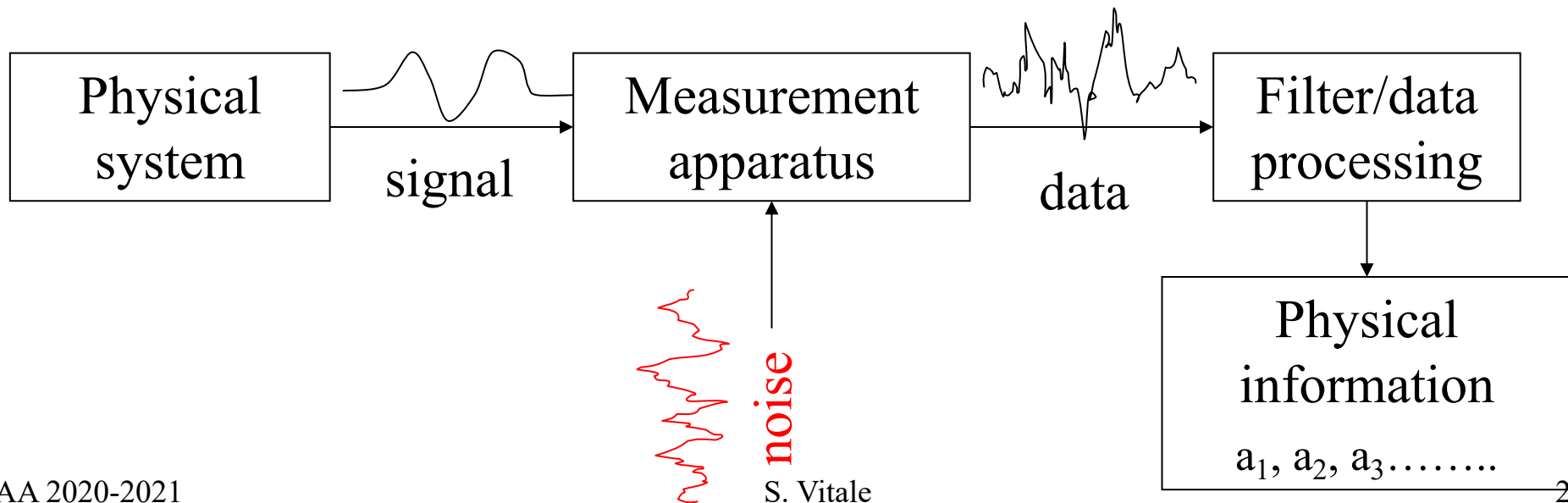
- Variance

$$\int_{-\frac{1}{2}}^{\frac{1}{2}} x^2 \, dx = \frac{1}{12}$$



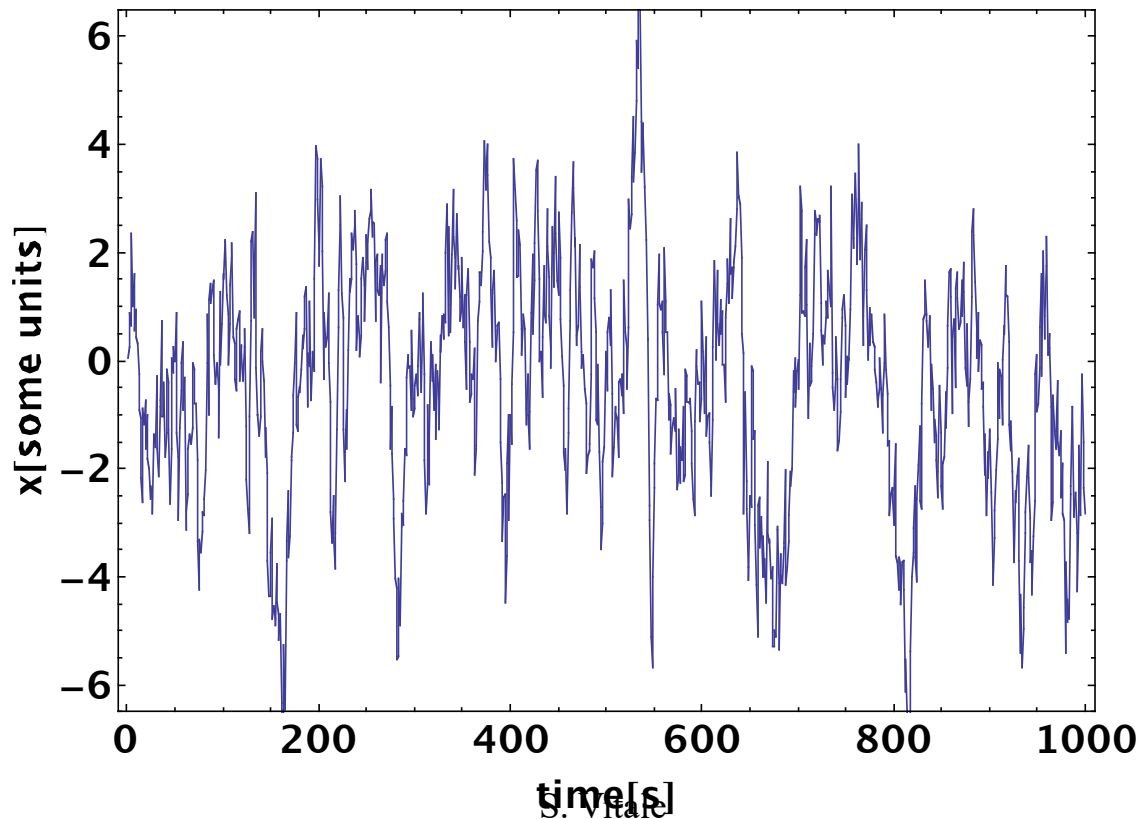
# Noise

Noise in physical experiments is described as a random signal  $x(t)$ :  
Independent (ensemble) repetitions of the same experiment produce different functions of time  $x(t)$



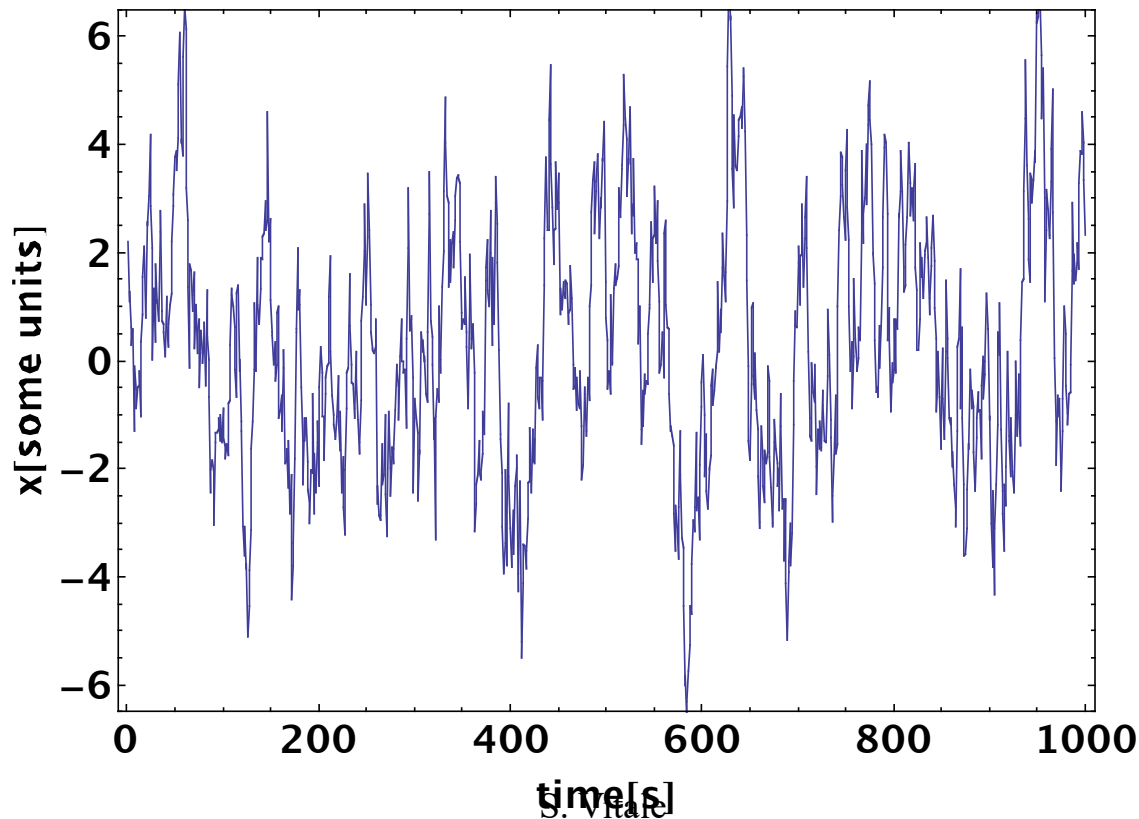
# Random signals

- At each repetition of a given experiment I get a different function of time  $x(t)$



# Random signals

- At each repetition of a given experiment I get a different function of time  $x(t)$

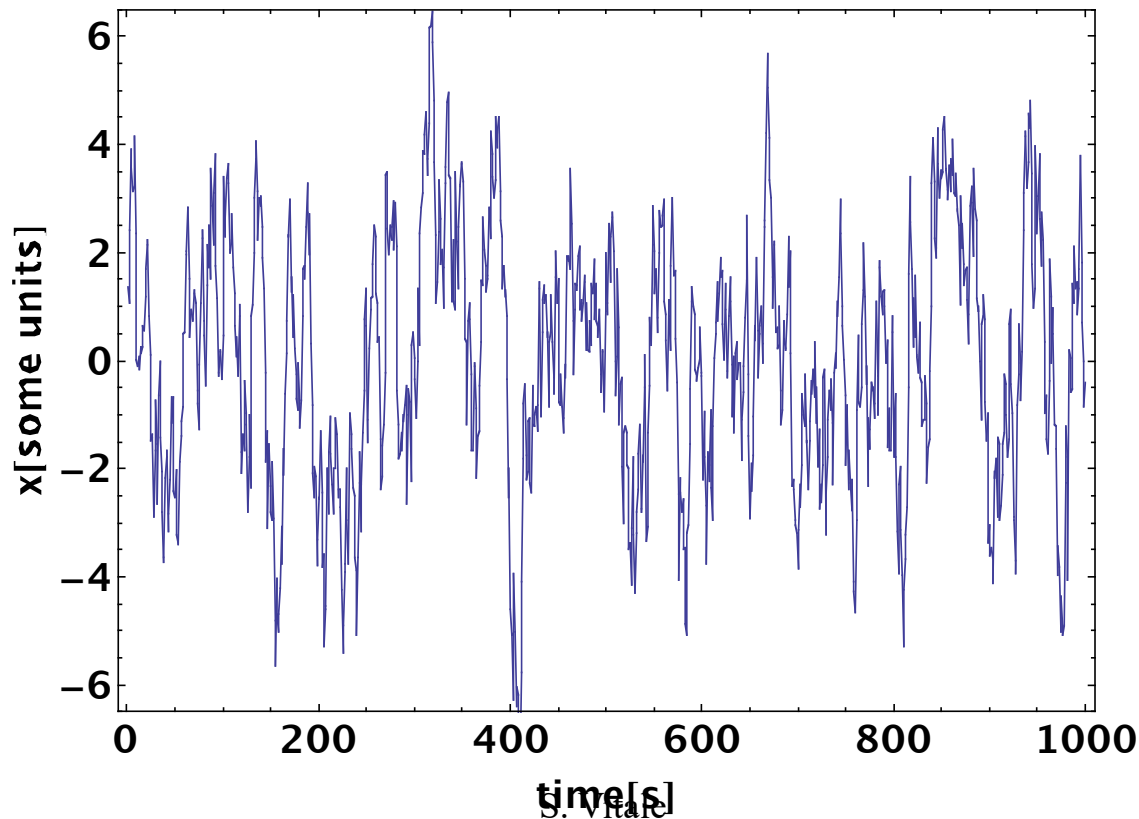


Trial 2



# Random signals

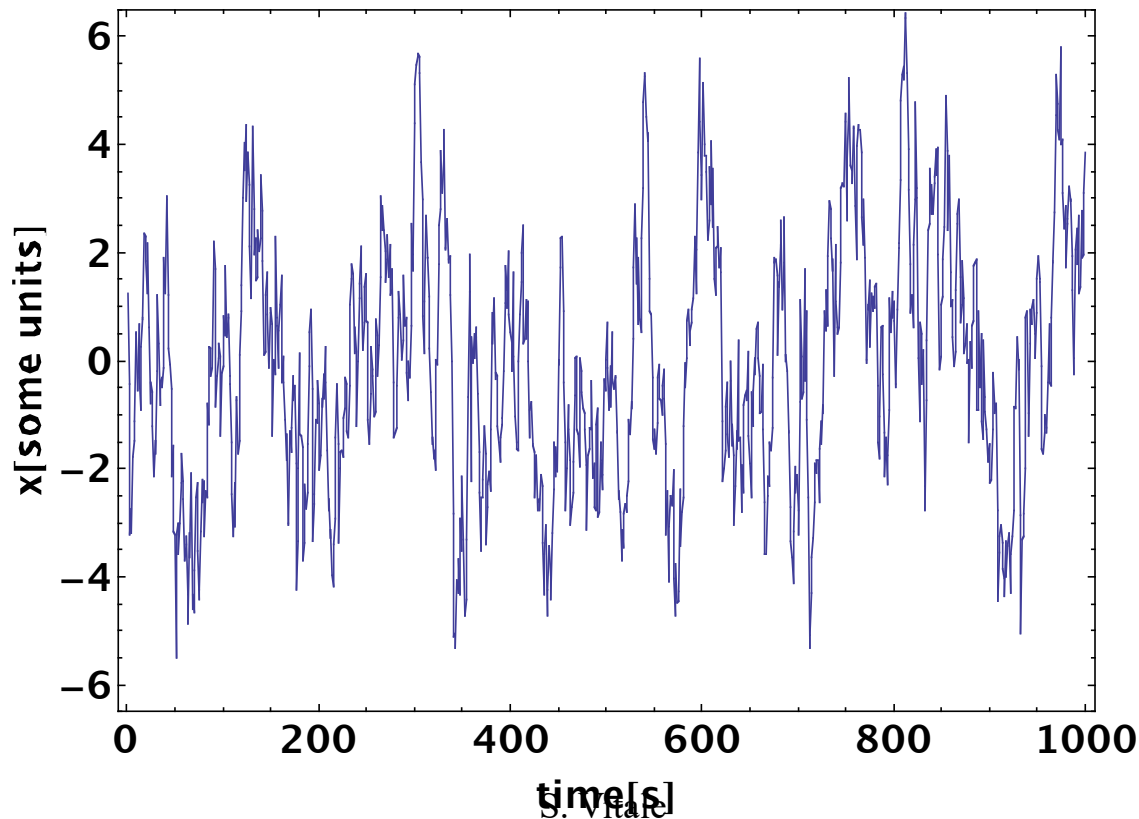
- At each repetition of a given experiment I get a different function of time  $x(t)$



Trial 3

# Random signals

- At each repetition of a given experiment I get a different function of time  $x(t)$

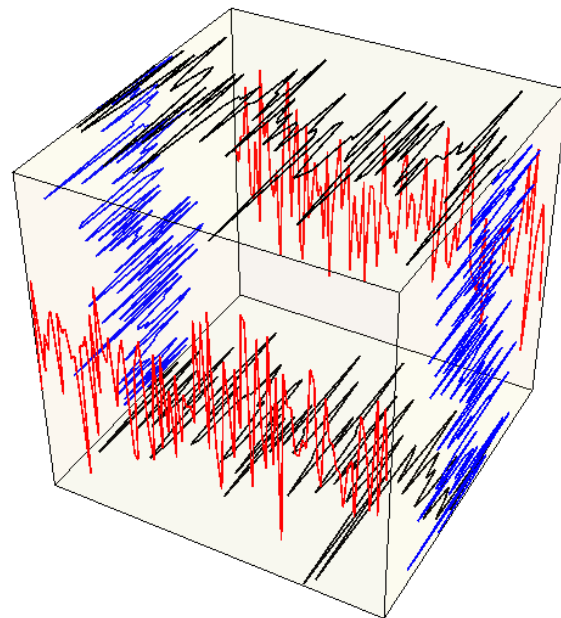


# A stochastic process

- At each outcome of an experiment I get a function of time  $x(t)$



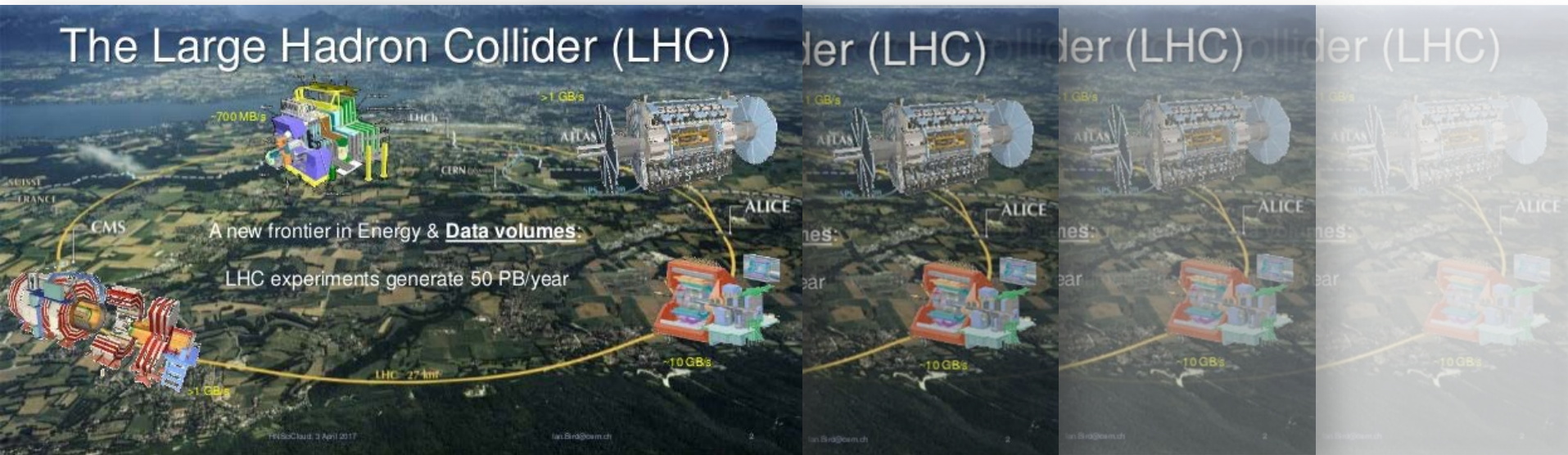
dice



Stochastic dice

# Probability ensemble and repetitions

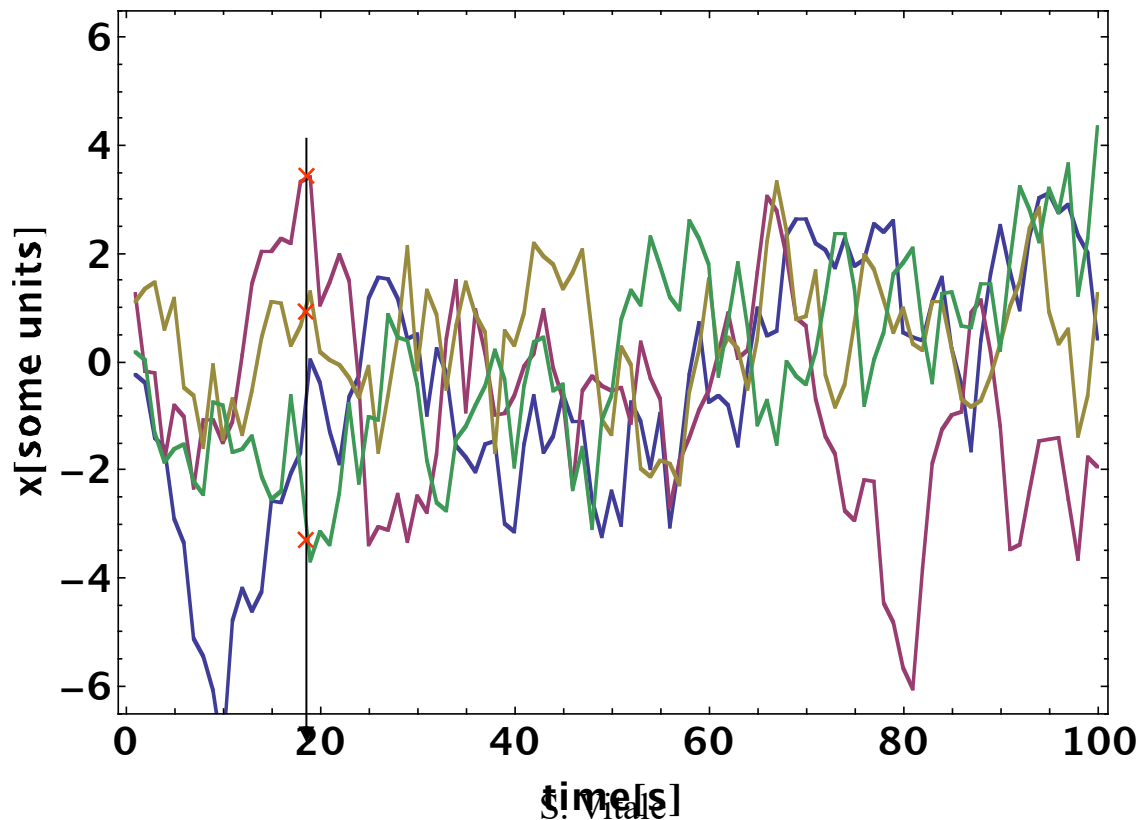
- The theory of probability describes only one experiment at the time



- Probability: prediction on the frequency of occurrence of same experiment in another “universe” (ensemble element)
- Not a prediction on frequency of occurrence in multiple repetitions over time
- An experiment repeated in time is a new single experiment

# A stochastic process

- For each outcome a function of time  $x(t)$
- For each time  $t$ : a random variable



# A stochastic process

- As  $x(t_1), x(t_2), x(t_3), \dots$  are all different rv, a stochastic process defines infinite many rv.
- Each rv will have its own density function

$$f_{x(t)}(\chi)$$

- With mean value

$$\eta(t) \equiv \langle x(t) \rangle = \int_{-\infty}^{\infty} \chi f_{x(t)}(\chi) d\chi$$

- This is called the mean value of the process and is in general a function of time.
- Same for the variance of the process

$$\sigma^2(t) \equiv \langle x^2(t) \rangle - \eta(t)^2 = \int_{-\infty}^{\infty} \chi^2 f_{x(t)}(\chi) d\chi - \eta(t)^2$$

# Multiple random variables

- The statistical properties of multiple random variables like the  $x(t_1)$ ,  $x(t_2)$ ,  $x(t_3)$ ..... are described by joint probabilities. Take two rv  $x$  and  $y$ . Their joint probability density  $f_{x,y}$  is defined by:

$$P\{x_o \leq x \leq x_1 \text{ and } y_o \leq y \leq y_1\} = \int_{x_o}^{x_1} \int_{y_o}^{y_1} f_{x,y}(\chi, \psi) d\chi d\psi$$

- Also this density needs to be normalized

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{x,y}(\chi, \psi) d\chi d\psi = 1$$

- The relation between the joint density  $f_{x,y}$  and the marginal densities  $f_x$  and  $f_y$  is

$$\begin{aligned} \int_{x_o}^{x_1} f_x(\chi) d\chi &= P\{x_o \leq x \leq x_1\} = \\ &= P\{x_o \leq x \leq x_1 \text{ and } -\infty \leq y \leq \infty\} = \int_{x_o}^{x_1} \int_{-\infty}^{\infty} f_{x,y}(\chi, \psi) d\chi d\psi \end{aligned}$$

- Then:  $f_x(\chi) = \int_{-\infty}^{\infty} f_{x,y}(\chi, \psi) d\psi$        $f_y(\psi) = \int_{-\infty}^{\infty} f_{x,y}(\chi, \psi) d\chi$