



**KAPITAŁ LUDZKI**  
NARODOWA STRATEGIA SPÓJNOŚCI

**UNIA EUROPEJSKA**  
EUROPEJSKI  
FUNDUSZ SPOŁECZNY



## **„SIGNAL PROCESSING”**

**Prezentacja multimedialna współfinansowana przez  
Unię Europejską w ramach  
Europejskiego Funduszu Społecznego w projekcie pt.  
*„Innowacyjna dydaktyka bez ograniczeń - zintegrowany  
rozwój Politechniki Łódzkiej - zarządzanie Uczelnią,  
nowoczesna oferta edukacyjna i wzmacniania zdolności  
do zatrudniania osób niepełnosprawnych”***

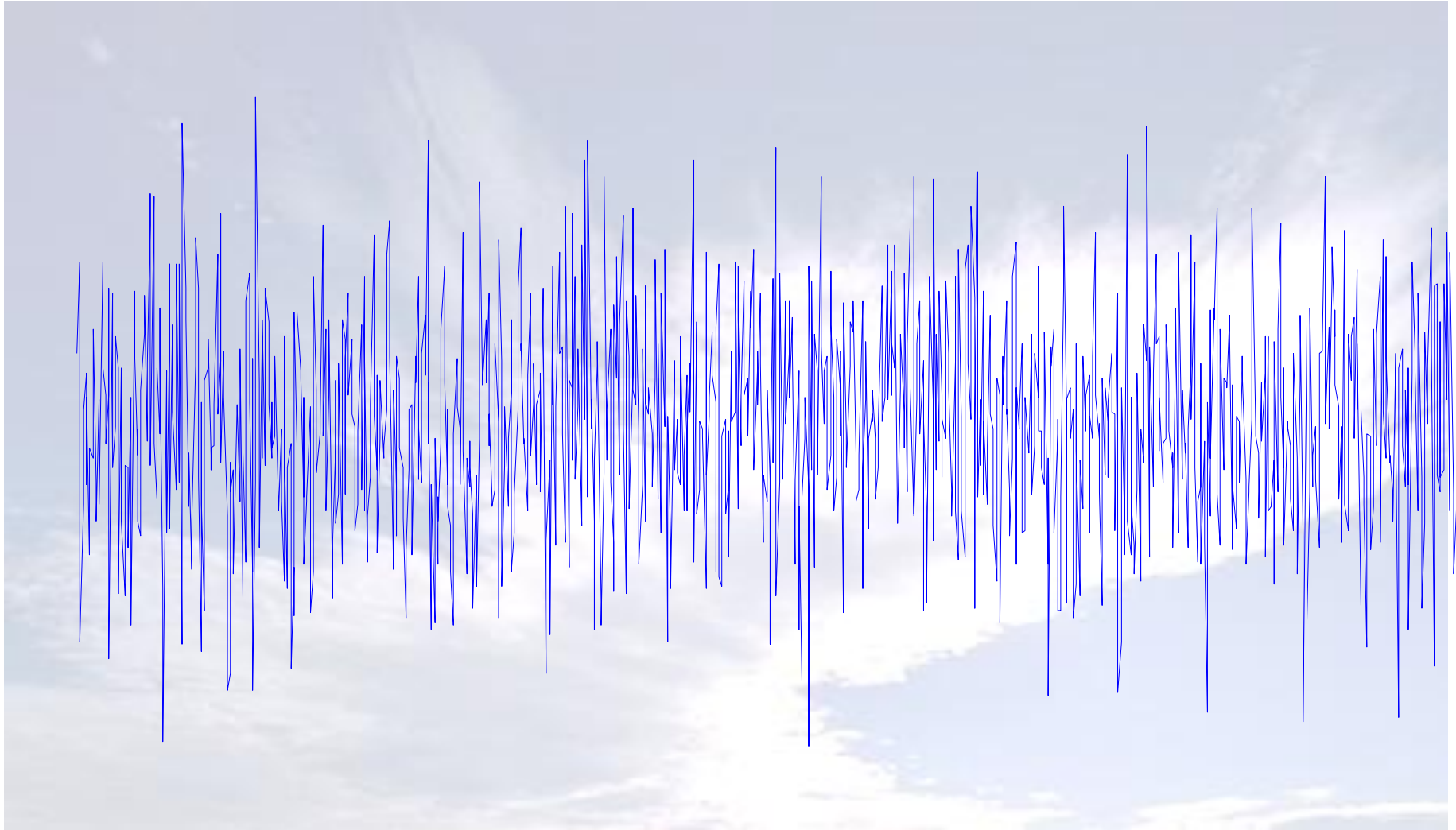


Politechnika Łódzka

Politechnika Łódzka, ul. Żeromskiego 116, 90-924 Łódź, tel. (042) 631 28 83  
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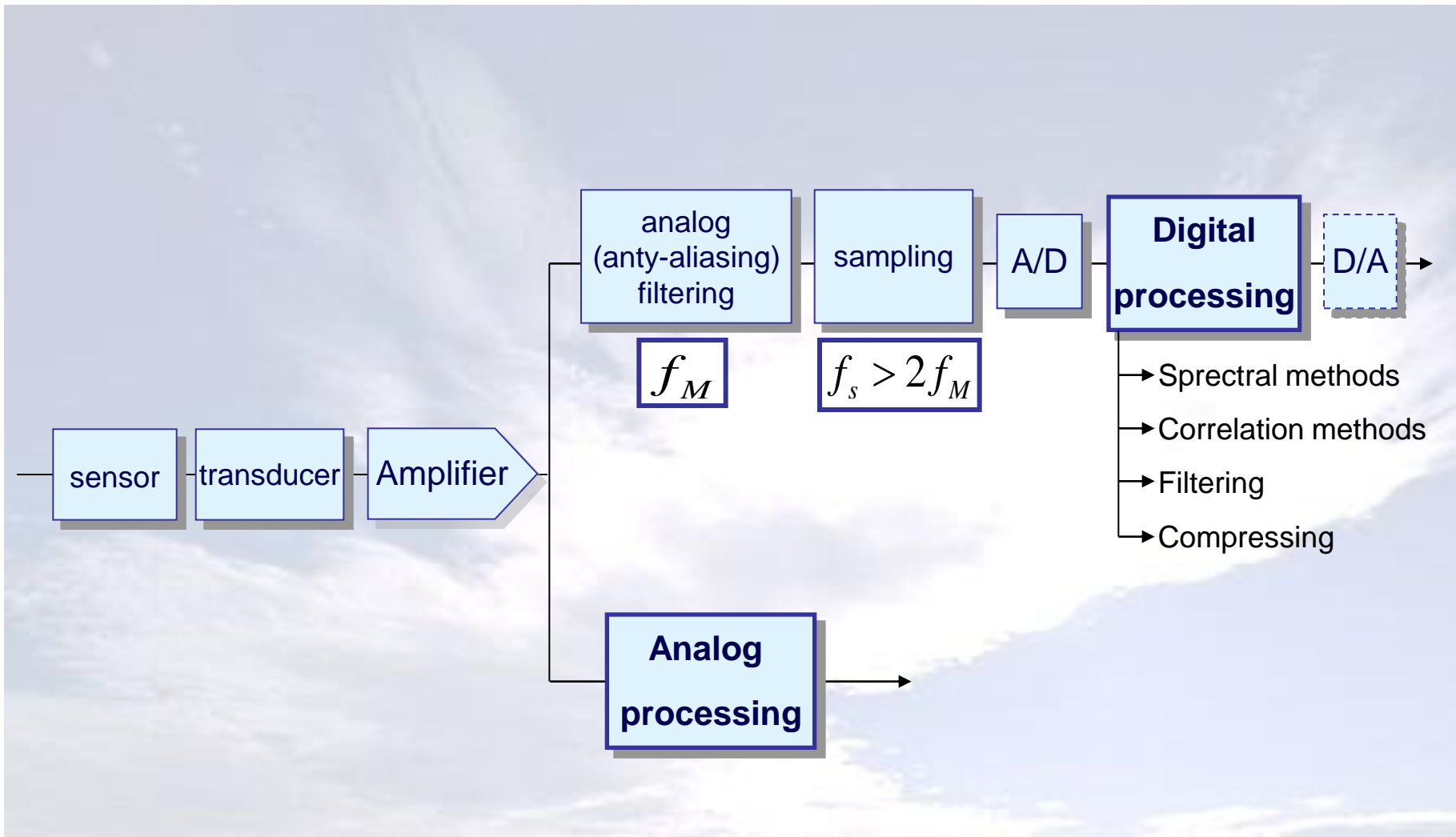


# Analysis of Random signals

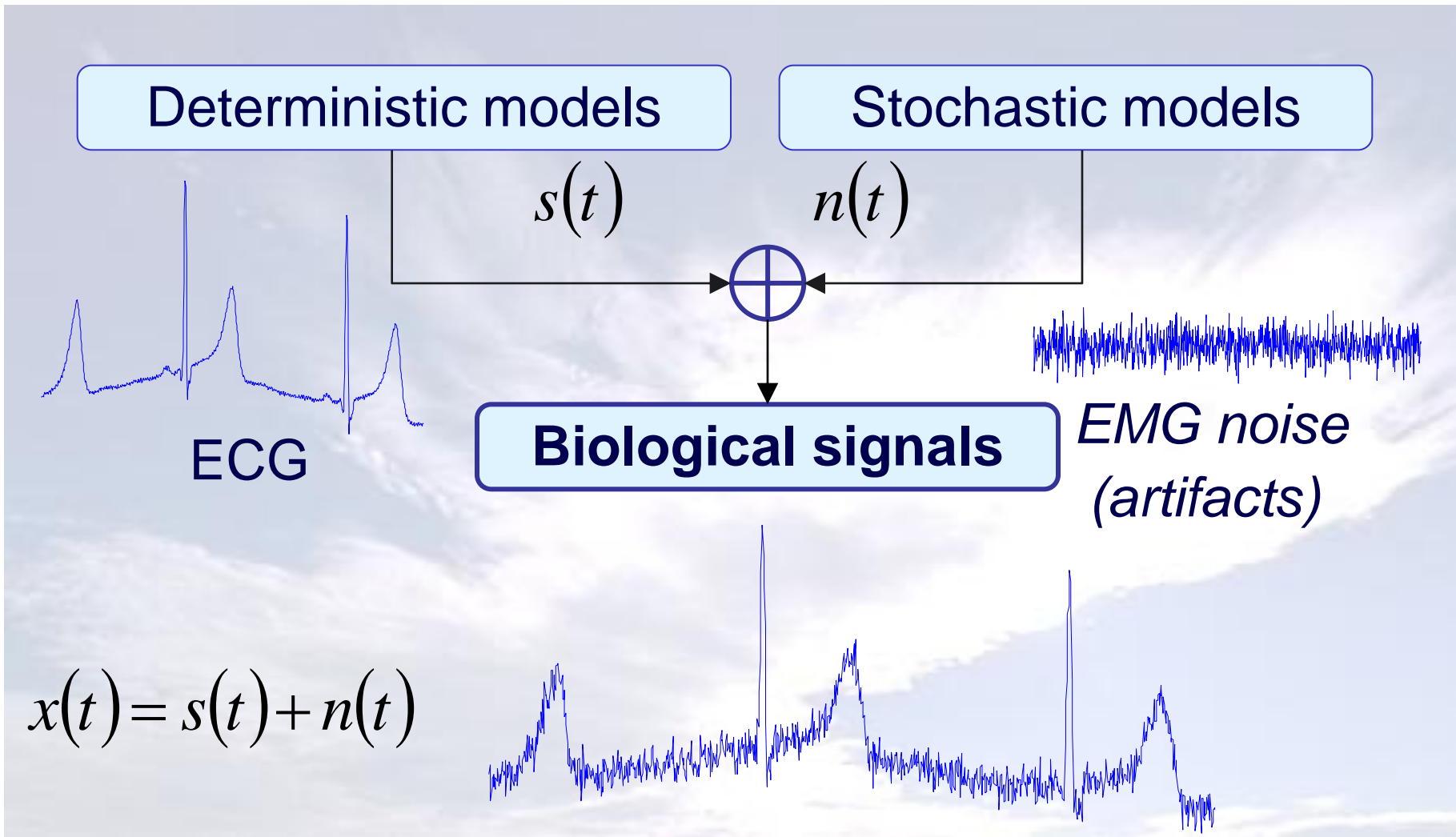




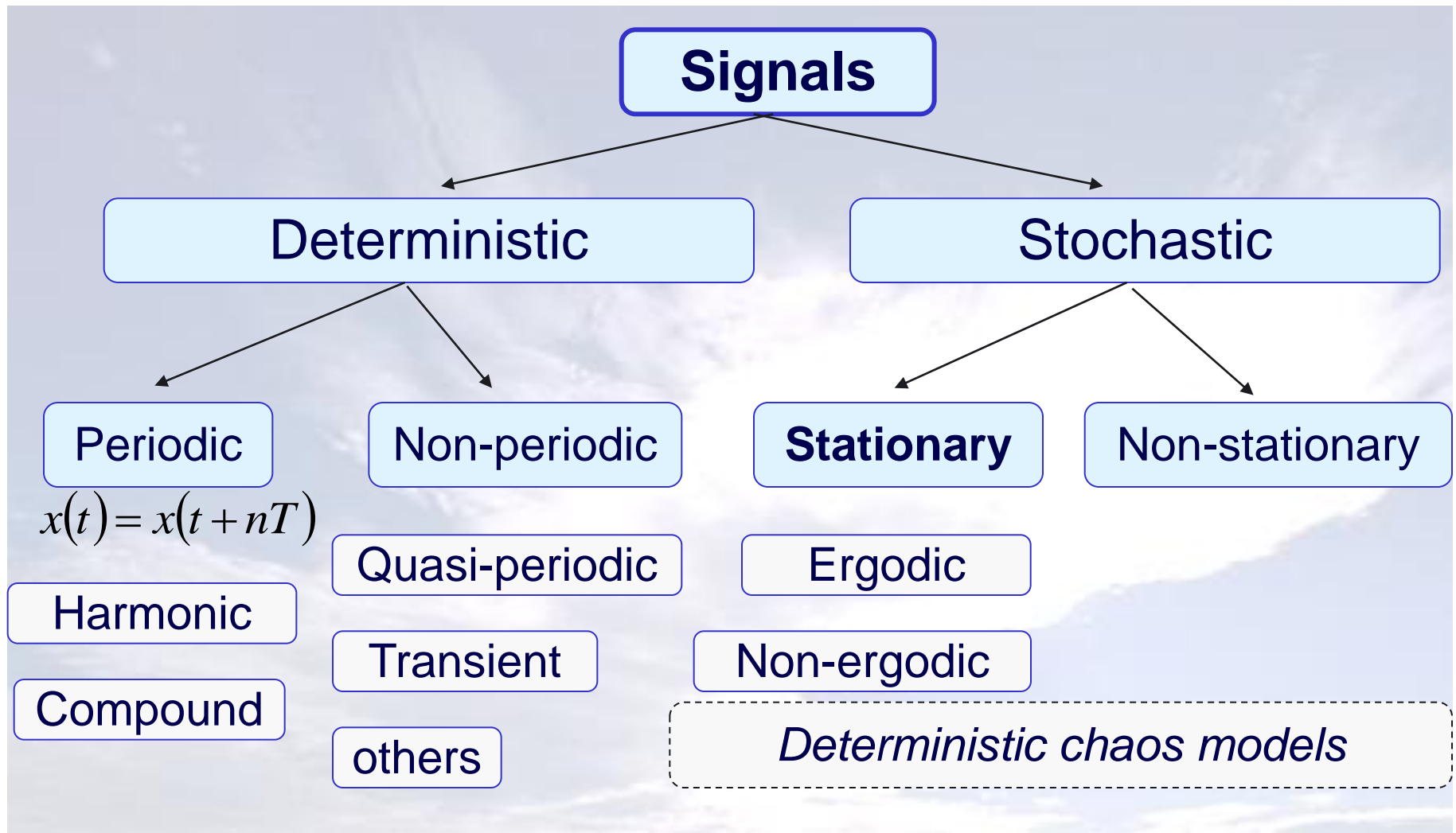
# Signal processing



# Biological signals



# Signal models



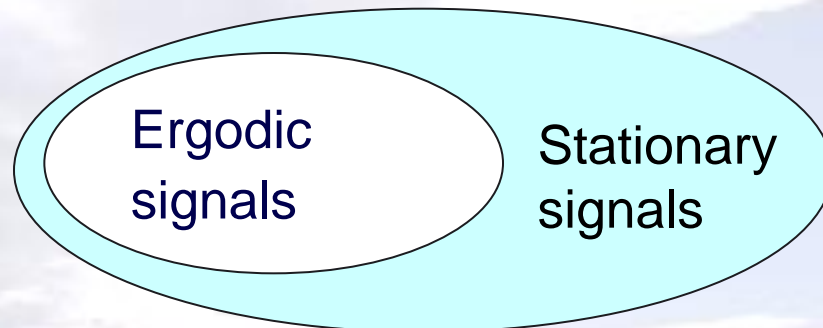
# Stationary and ergodic signals (processes)

## Stationary signals

Statistical moments calculated over time (e.g. mean, variance) of a stochastic signal do not change with time.

## Ergodic signals

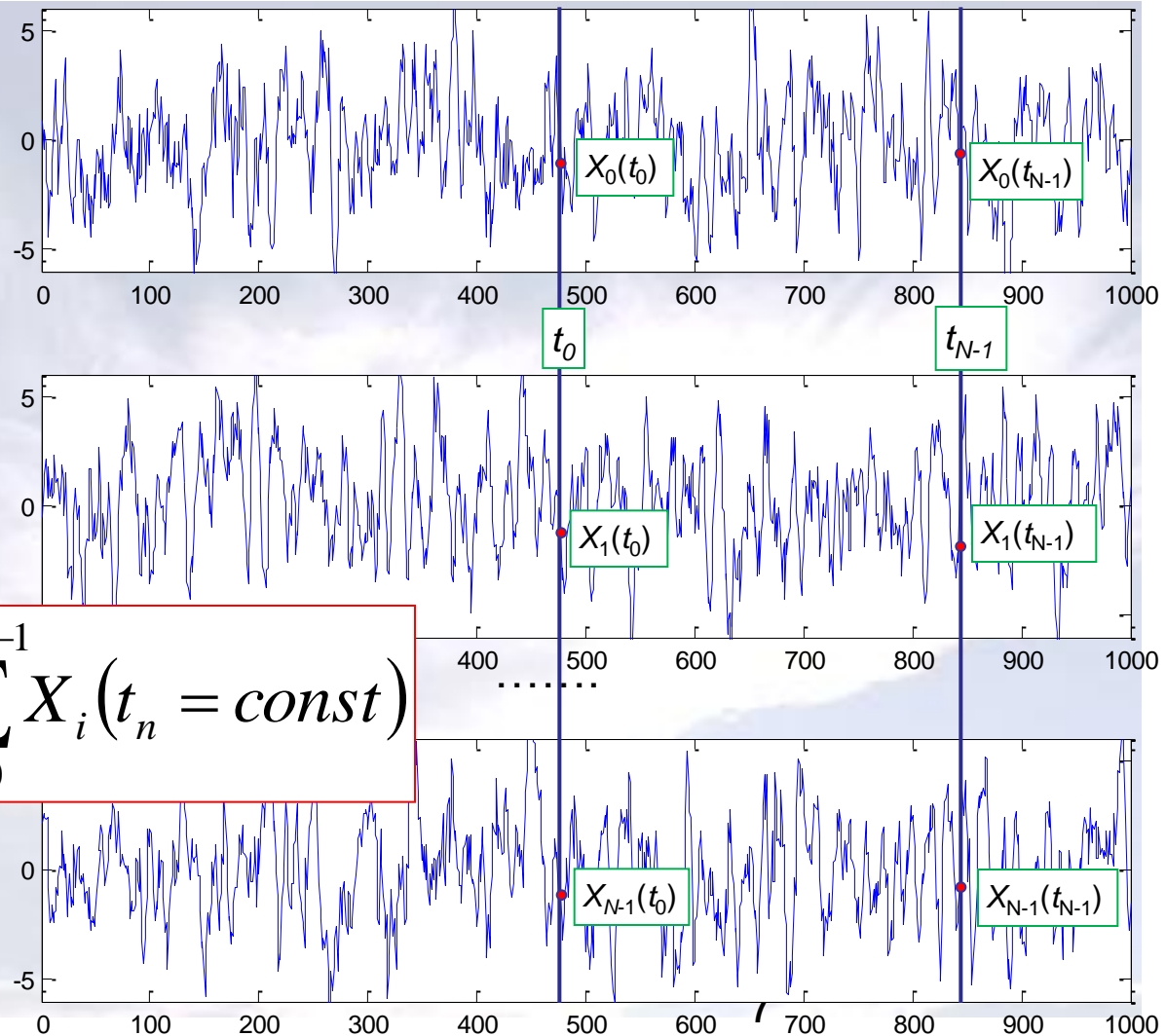
Statistical moments of a signal (e.g. mean, variance) calculated over time and calculated over realizations (ensemble moments) are correspondingly equal.



# Stationary and ergodic signals (processes)

A stochastic signal  
is ergodic if:

$$\frac{1}{N} \sum_{t=0}^{t=N-1} X_{i=const}(t) = \frac{1}{N} \sum_{i=0}^{i=N-1} X_i(t_n = const)$$



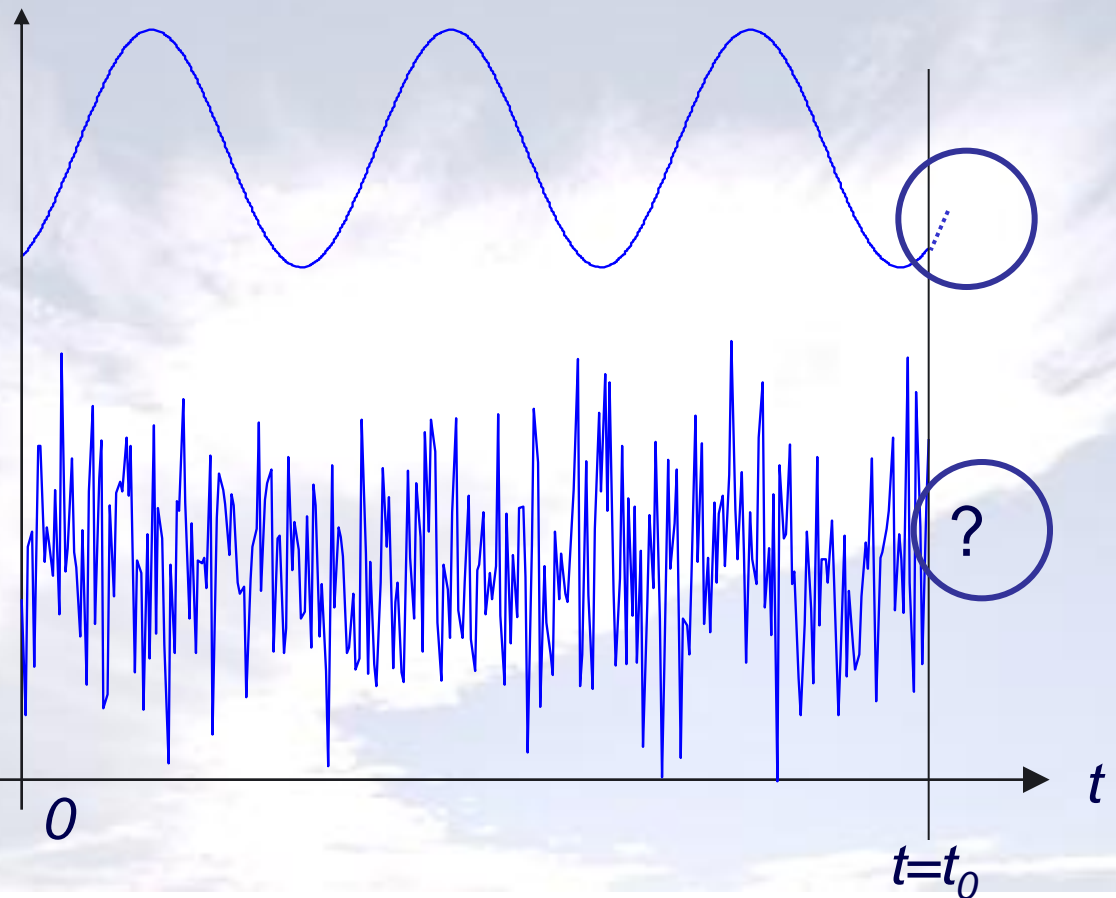
# Random vs stochastic signal

## **deterministic**

(samples may be predicted with high accuracy)

## **random**

(unpredictable values of samples, only statistical parameters may be established)







## Random signal (parameters)

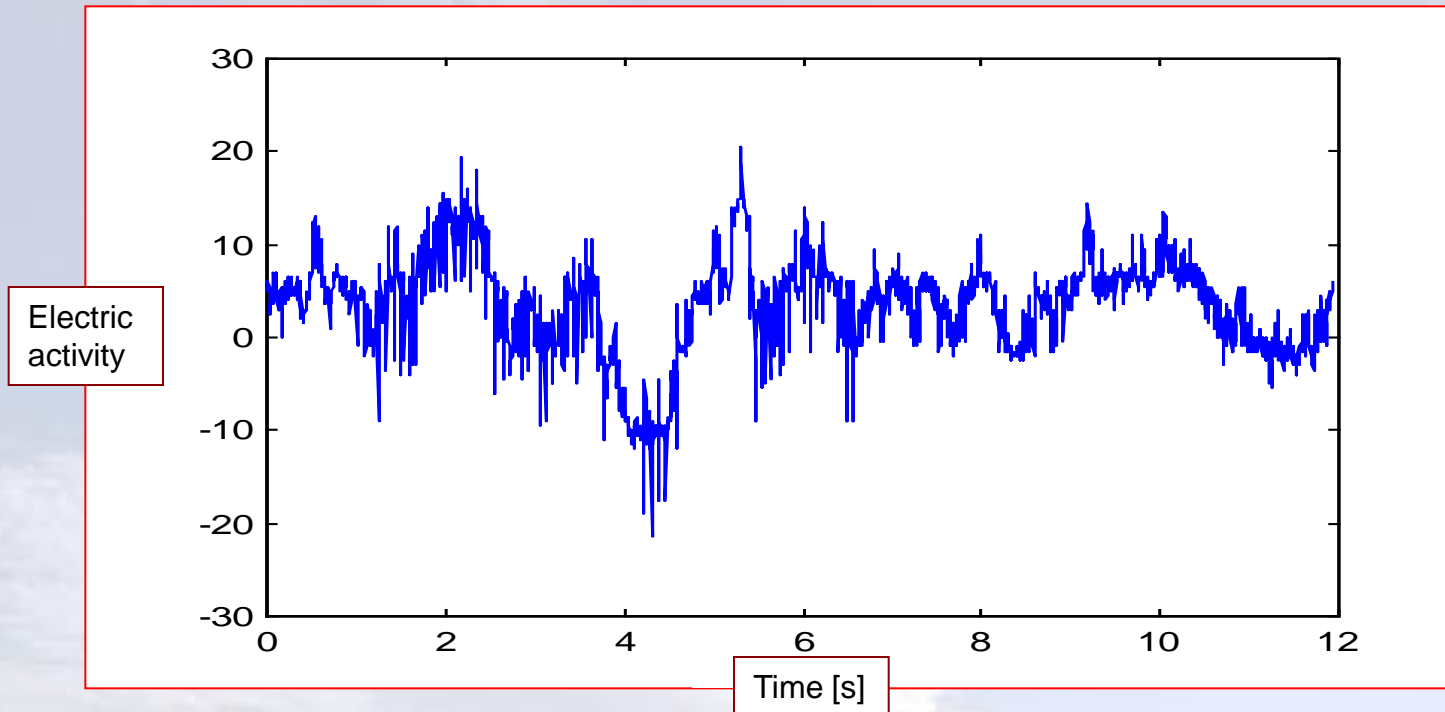
- exact values unpredictable
- can be interpreted as random variable  
(and expressed in terms of the following parameters):
  - expected value (**mean**)
  - **variance** and **standard deviation**
  - **autocorrelation** function

For stationary random signals these parameters are time invariant.



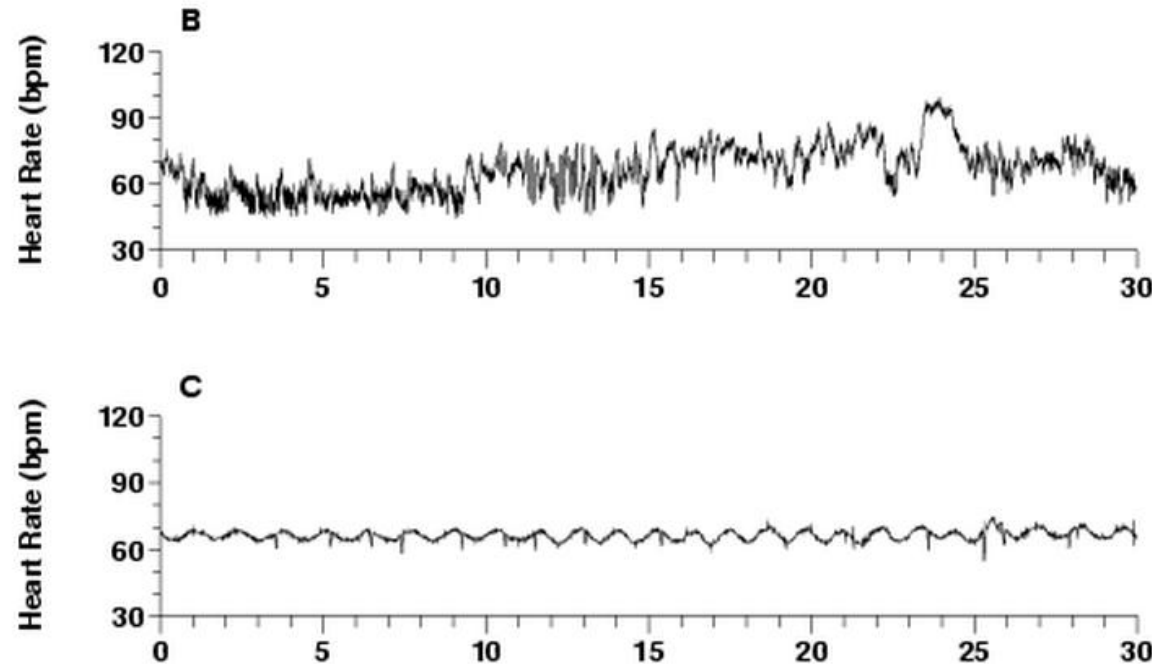
## Example – EEG signal

Stochastic model may be applied to electroencephalographic signal (EEG)  
EEG is a result of electric activity of large amount \* of neurons.



\*) It is being proven that neural perception is based on synchronization of specialised populations of neurons

# Heart rate (HR) in health and disease

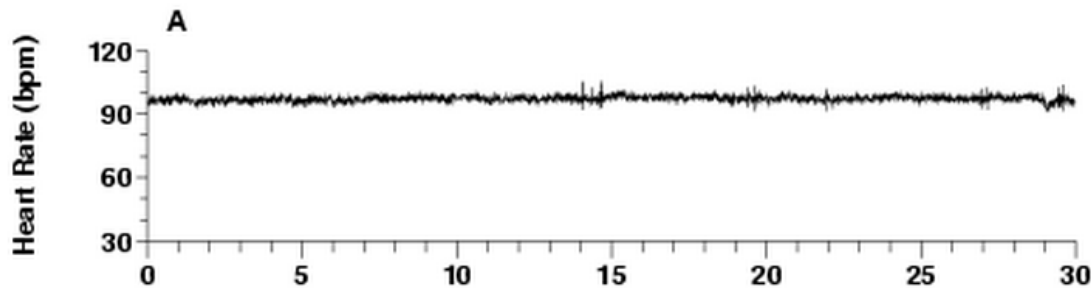


Normal sinus rhythm  
(nonstationarity, „fractality”)

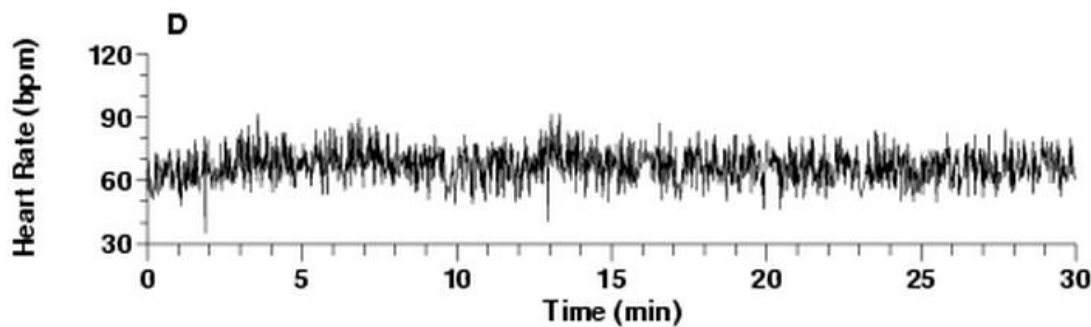
HR in severe congestive heart  
failure (periodic oscillations  
presents ( $\sim 1/\text{min}$ ) due to  
Cheyne-Stokes breathing)

[www.physionet.org](http://www.physionet.org)

# Heart rate (HR) in health and disease



HR in severe congestive heart failure – excessive regularity of the heart rate



HR at cardiac arrhythmia, (atrial fibrillation), uncorrelated randomness present

[www.physionet.org](http://www.physionet.org)

## Random signal – expected value

**Continuous signal:**

$$E\{x(t)\} = \int_{-\infty}^{+\infty} xp(x)dx = x$$

```
#Python  
#see also random?  
x=random.randn(1000)  
expected=x.mean()
```

**Discrete signal:**

$$E\{x(n)\} = \sum_{n=-N}^{n=N} x(n)p(n) = \frac{1}{2N+1} \sum_{n=-N}^{n=N} x(n) = x$$

successive samples are  
equally probable



## Random signal - variance

**Variance-** *square mean of deviations from the expected value*

**Continuous signal:**

$$\sigma^2 = E\{[x(t) - x]^2\} = \int_{-\infty}^{+\infty} [x(t) - x]^2 p(x) dx$$

**Discrete signal:**

$$\sigma^2 = \frac{1}{2N+1} \sum_{n=-N}^{n=N} [x(n) - x]^2 = \frac{1}{2N+1} \sum_{n=-N}^{n=N} x^2(n) - x^2$$

**#Python**

```
x=random.randn(1000)
```

```
std_dev=x.std()
```

```
Variance=x.var()
```



# Random signal – autocorrelation function

**Continuous signal:**

$$R(\tau) = \int_{-\infty}^{\infty} x(t)x(t + \tau)dt$$

**Discrete signal:**

$$R(m) = \sum_{n=-\infty}^{n=\infty} x(n)x(n + m)$$

**#Python**

```
x=random.randn(1000)
R=correlate(x,x,mode='full')
plot(R)
```

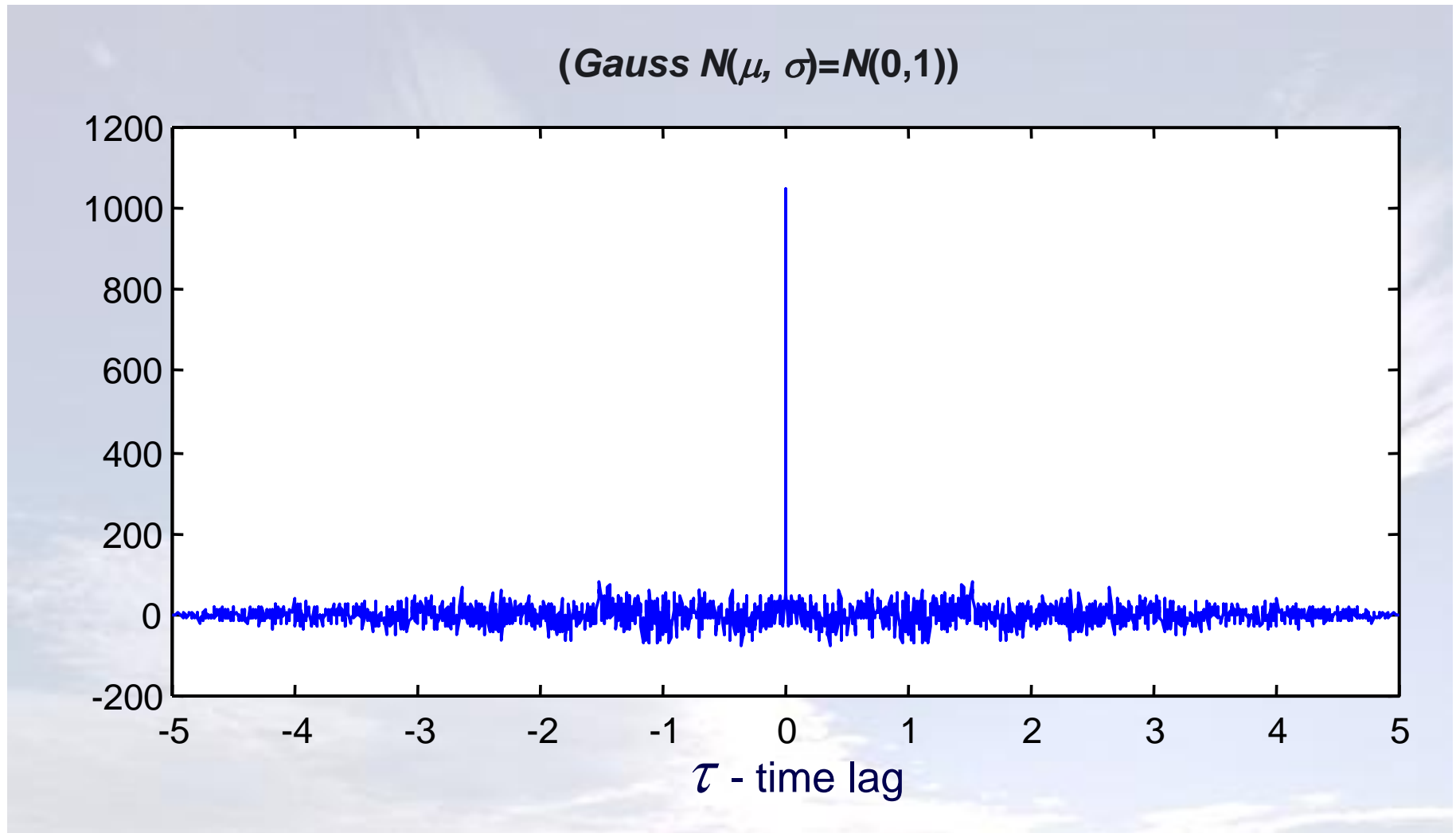
**#better way, remove DC**

```
x=x-x.mean()
R=correlate(x,x,mode='full')
plot(R)
```

**Don't forget to remove the mean from the signal!**

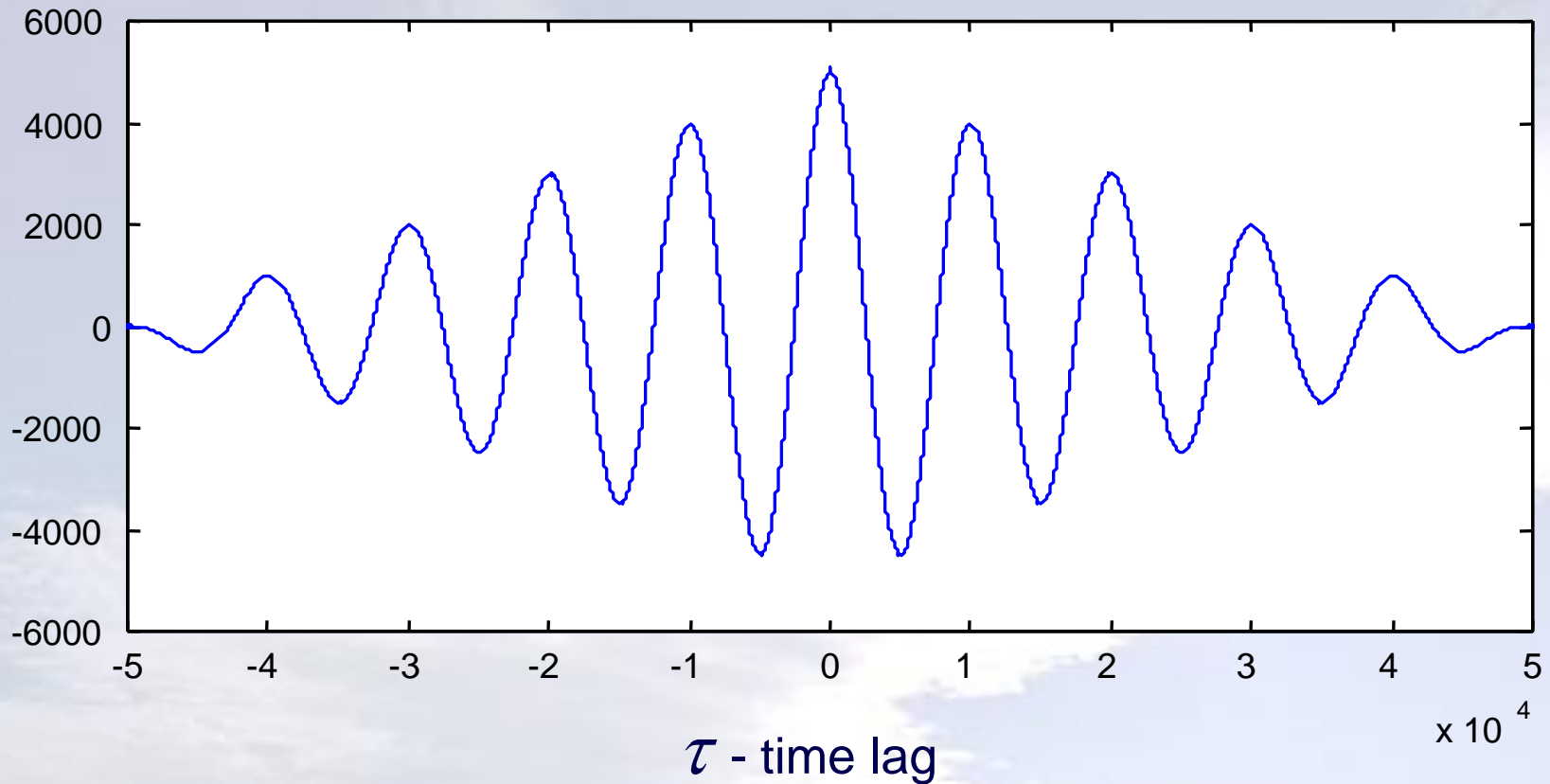


# Autocorrelation function of a random signal





# Autocorrelation function of harmonic signal





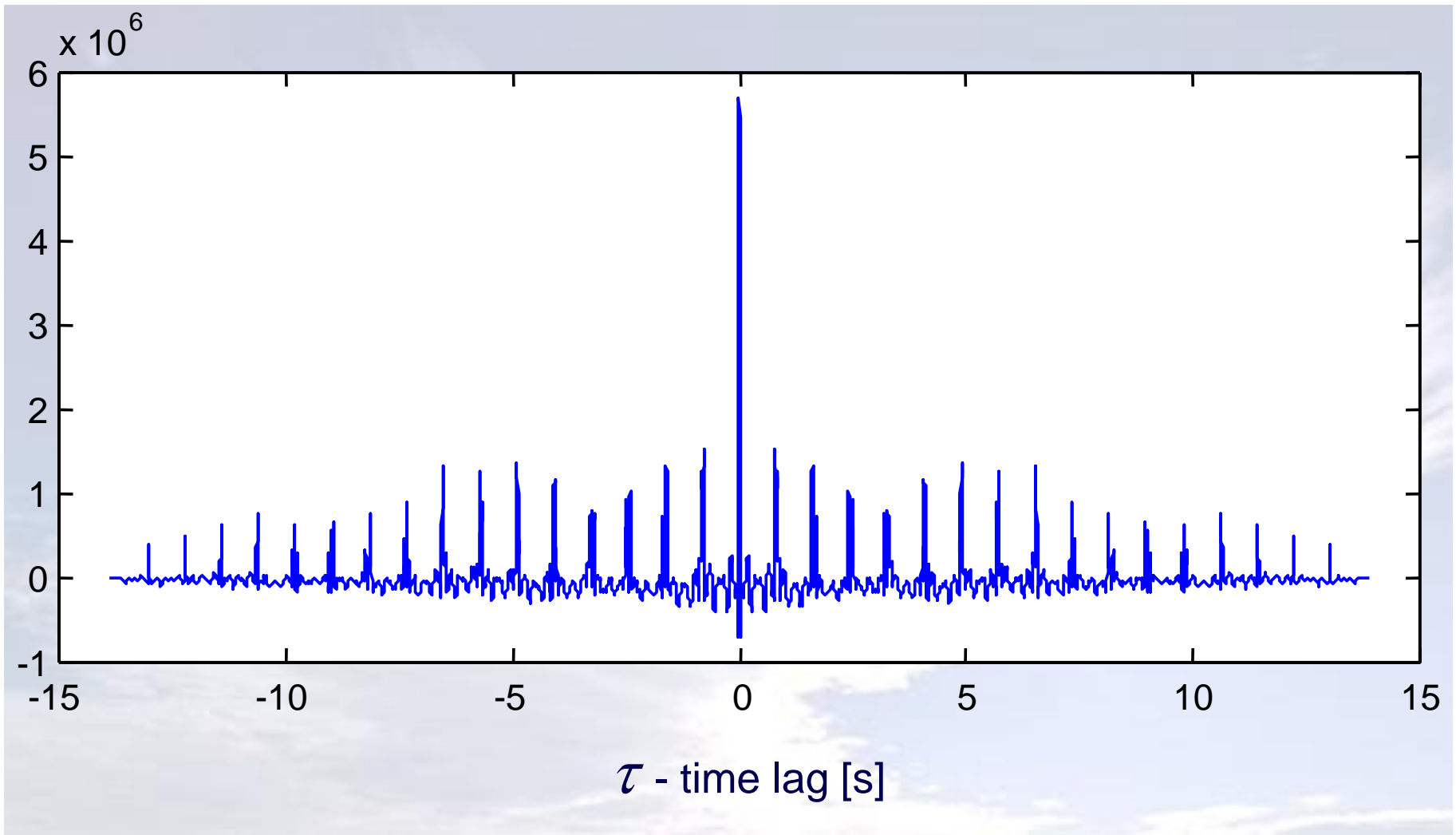
# Autocorrelation function – open questions

1. Compare the length of the signal with the length of autocorrelation function.
2. What properties of autocorrelation can you notice?
3. What can be the application of autocorrelation function?





# ECG signal autocorrelation



# Cross-correlation function

Continuous signals:

$$R_{xy}(\tau) = \int_{-\infty}^{\infty} x(t)y(t + \tau)dt$$

Discrete signals:

$$R_{xy}(m) = \sum_{n=-\infty}^{n=\infty} x(n)y(n + m)$$

**#Python**

```
R=correlate(x,y,mode='full')
```

Don't forget to remove the mean from the signals!



# Cross-correlation function – example1

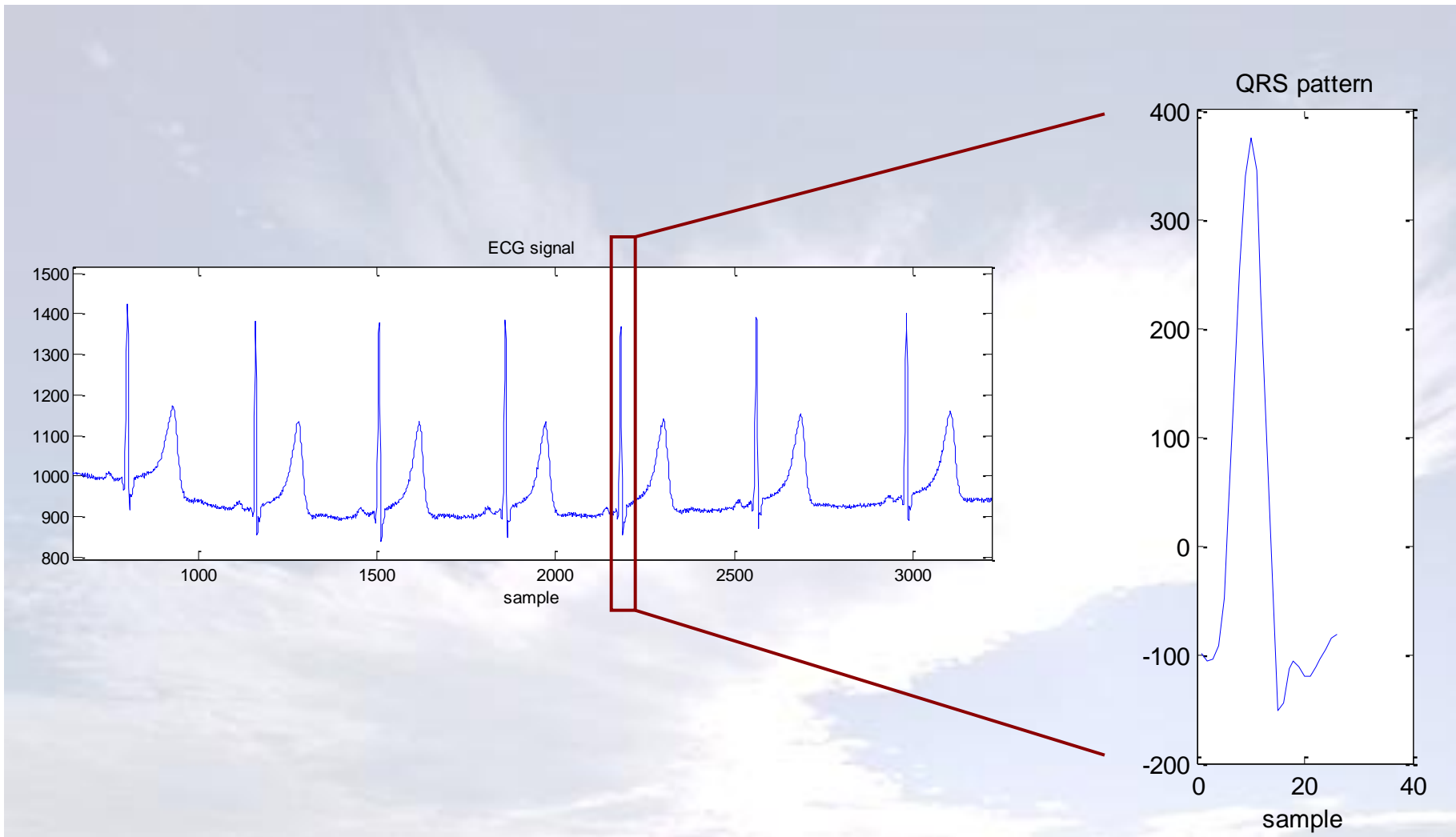
## QRS detection by correlation with QRS pattern

1. Determine the QRS pattern of the sample ECG signal.
2. Use the pattern to detect other QRS complexes
3. Keep the results in order to compare them with the other detection methods



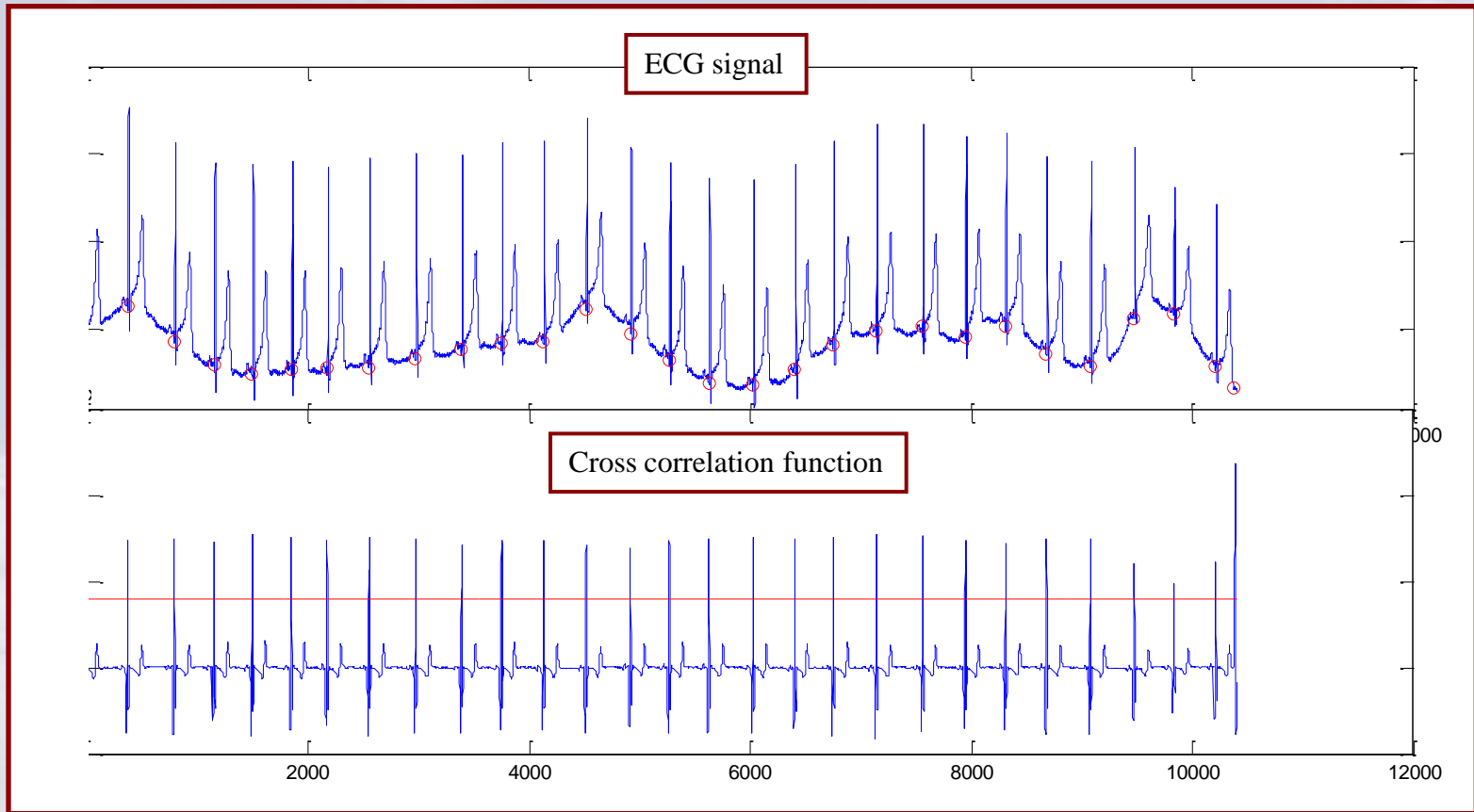


# Cross-correlation function – example1



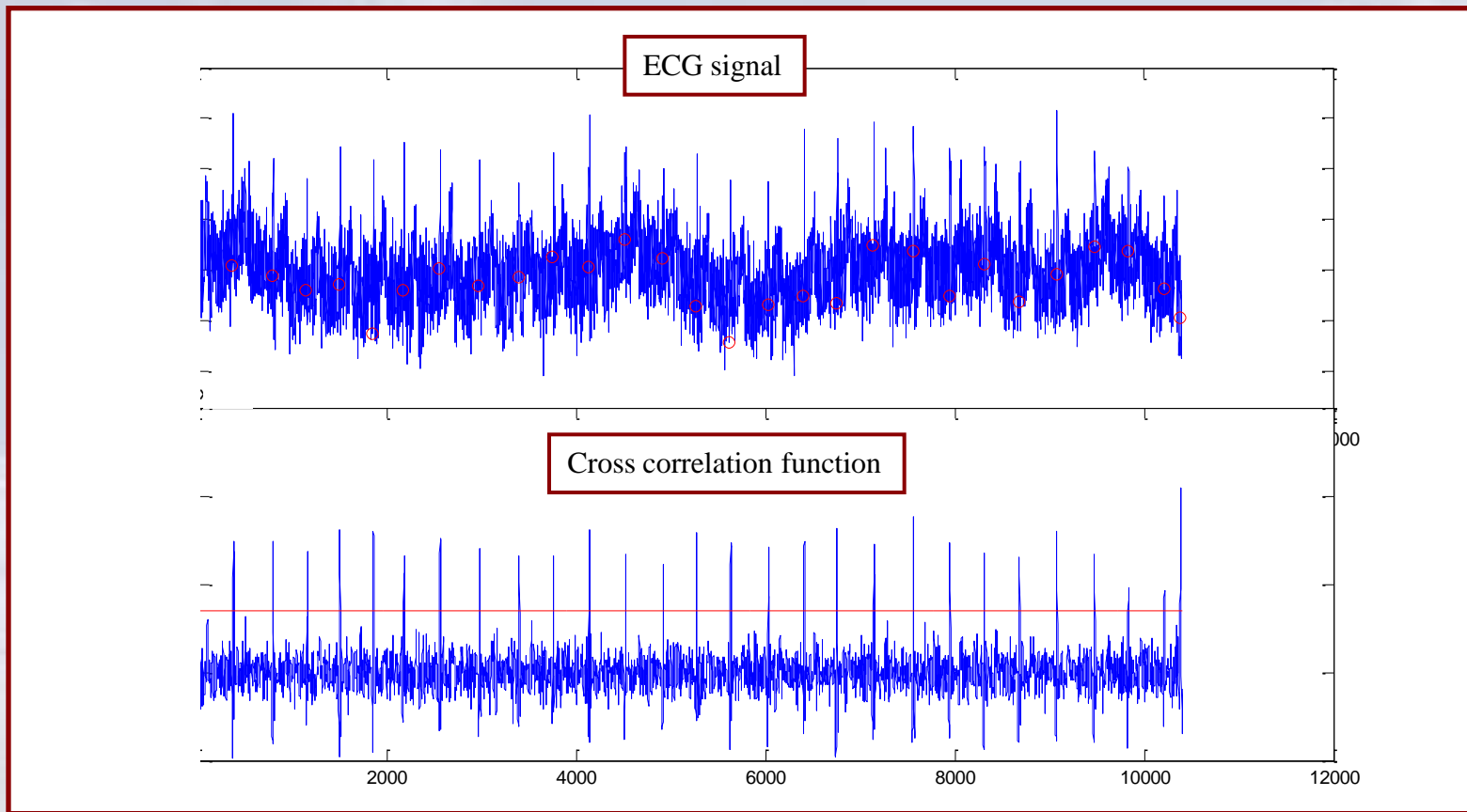
# Cross-correlation function – example1

```
kor=xcorr(ECGsignal,QRSpattern);
```



# Cross-correlation function – example1

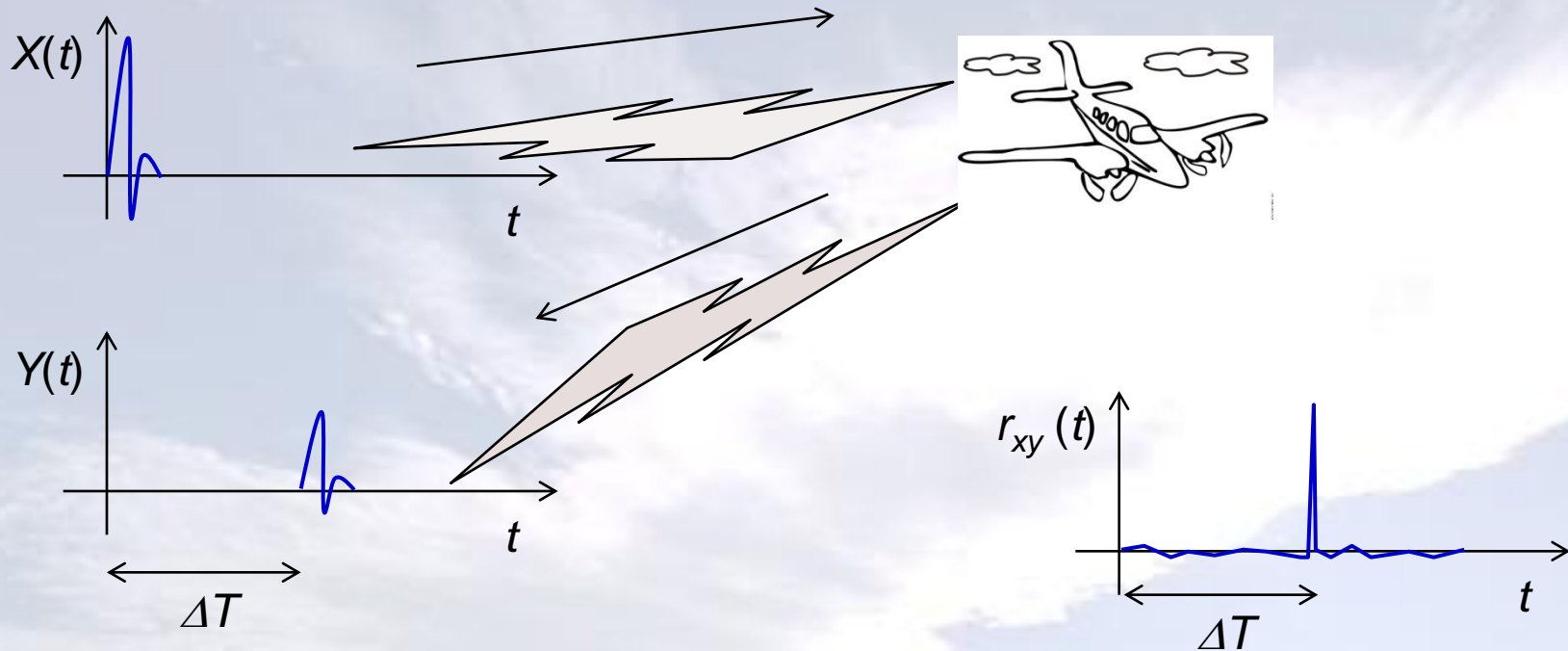
```
kor=xcorr(ECGsignal,QRSpattern)
```





# Cross-correlation function – example2

Delay estimation of radar signals:





## Power Spectral Density - PSD

- **Autocorrelation function** of a random signal is deterministic  $\Rightarrow$  its Fourier spectrum may be computed

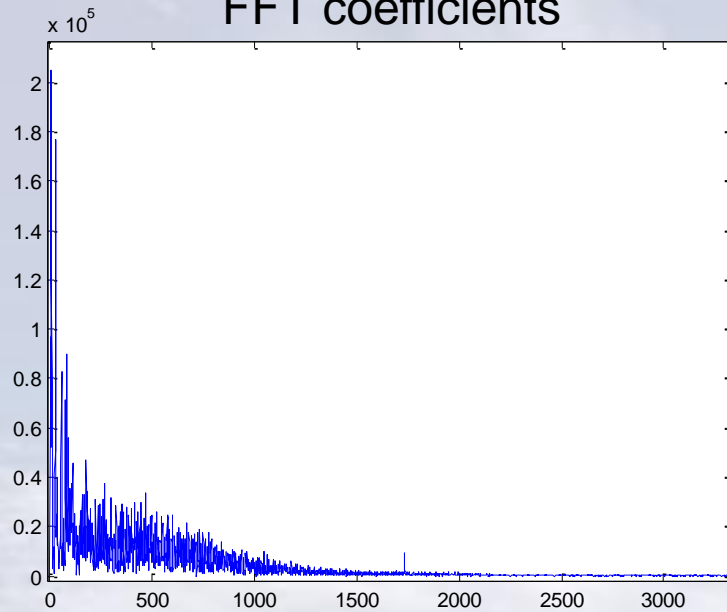
$$FT\{R_{xx}(\tau)\} = X(j\omega)X^*(j\omega) = |X(j\omega)|^2$$



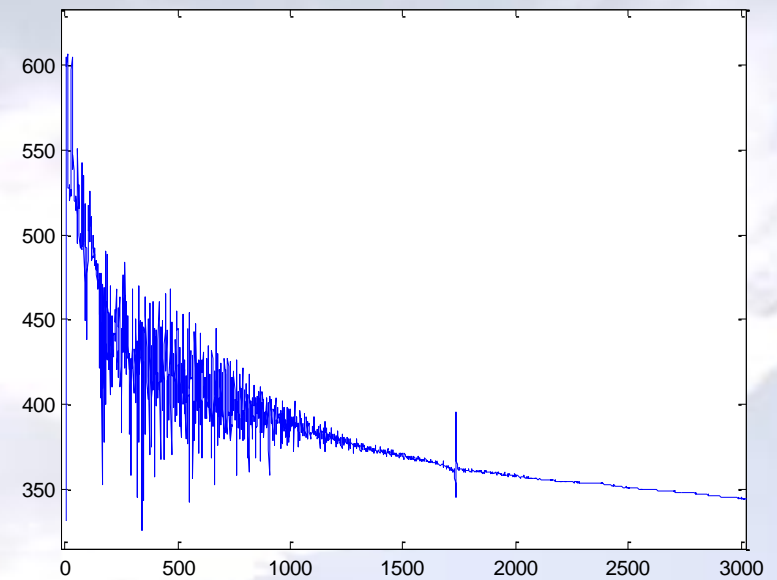


# Power Spectral Density - PSD

## FFT coefficients



## PSD coefficients



# Correlation coefficients

Similarity of two signals may be assessed with correlation coefficients.

$$r_{xy} = \frac{\sigma_{xy}}{\sqrt{\sigma_x \sigma_y}} = \frac{\sum_{n=1}^N [x(n) - \bar{x}][y(n) - \bar{y}]}{\sqrt{\sum_{n=1}^N [x(n) - \bar{x}]^2 \sum_{n=1}^N [y(n) - \bar{y}]^2}}$$

The term  $\sigma_{xy}$  in the numerator of the first fraction is circled in blue. An arrow points from a box labeled "covariance" to this circled term.

$$-1 \leq r_{xy} \leq 1$$

#Python

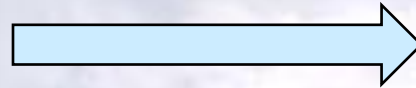
`corrcoef?`

## Correlation coefficient - example

Compute correlation coefficient between signals  $a$  and  $b$ :

$$a = [1, 2, 3]$$

$$b = [-3, 4, 2]$$



$$r_{xy} \cong 0.6934$$

$$r_{xy} = \frac{\sum_{n=1}^N [x(n) - \bar{x}][y(n) - \bar{y}]}{\sqrt{\sum_{n=1}^N [x(n) - \bar{x}]^2 \sum_{n=1}^N [y(n) - \bar{y}]^2}}$$



# Correlation coefficients

**EXERCISE:** Create 3 vectors:

1. Vector  $x_1$  whose elements are heights of the members of the group in meters
2. Vector  $x_2$  – the weights in the same order as in  $x_1$  (do not cheat!)
3. Vector  $x_3$  – shoe size

Determine correlation coefficients of the vectors and point the most correlated pair of vectors.



# Probability distribution functions

**Uniform:**

$$p(x) = \begin{cases} \frac{1}{x_1 - x_0} & \text{for } x_0 \leq x \leq x_1 \\ 0 & \text{for other } x \end{cases}$$

**Laplace:**

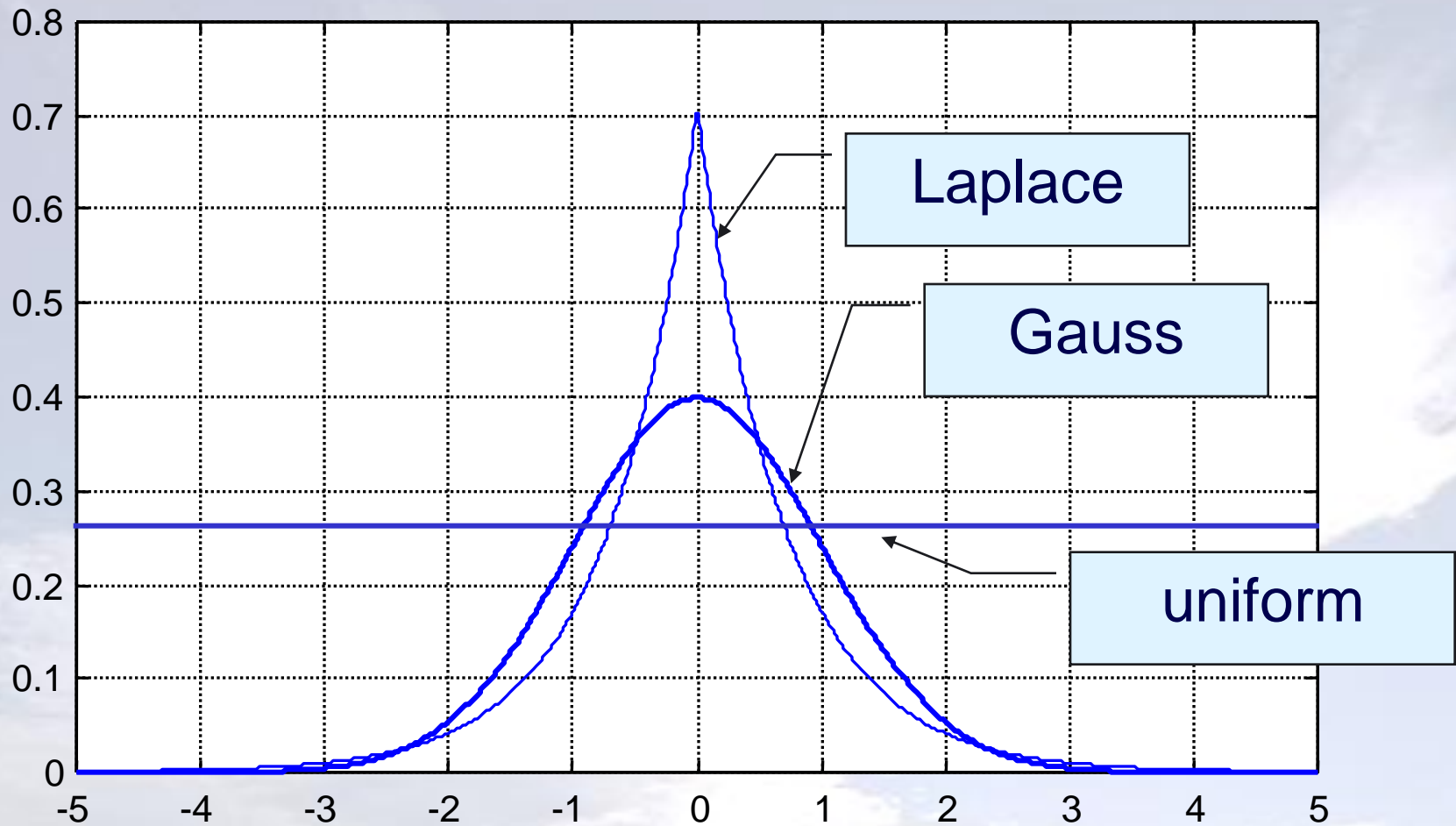
$$p(x) = \frac{1}{\sigma\sqrt{2}} e^{-\frac{\sqrt{2}|x-\mu|}{\sigma}}$$

**Normal (Gauss):**

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

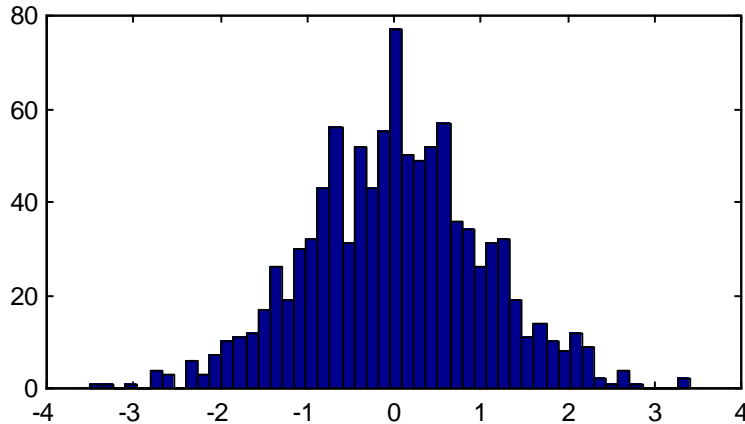


# Probability distribution functions

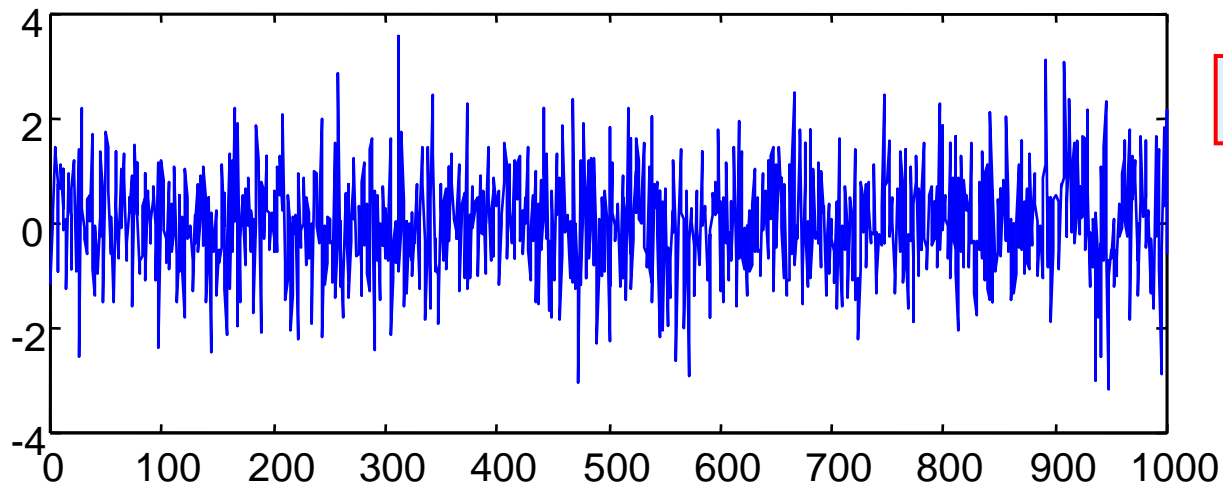




# Gaussian noise

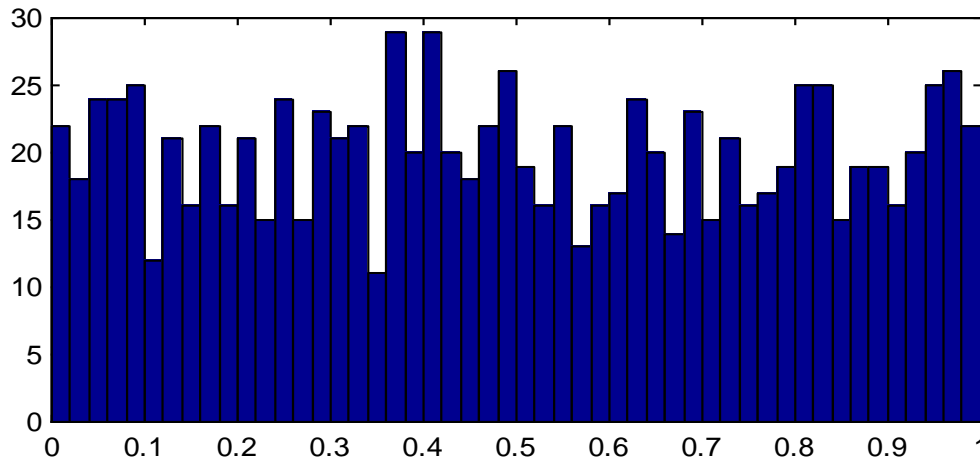


```
#Python  
sigma=1  
x=sigma*random.randn(1000)  
bins=linspace(-5,5,50)  
hist(x,bins, normed='True')
```

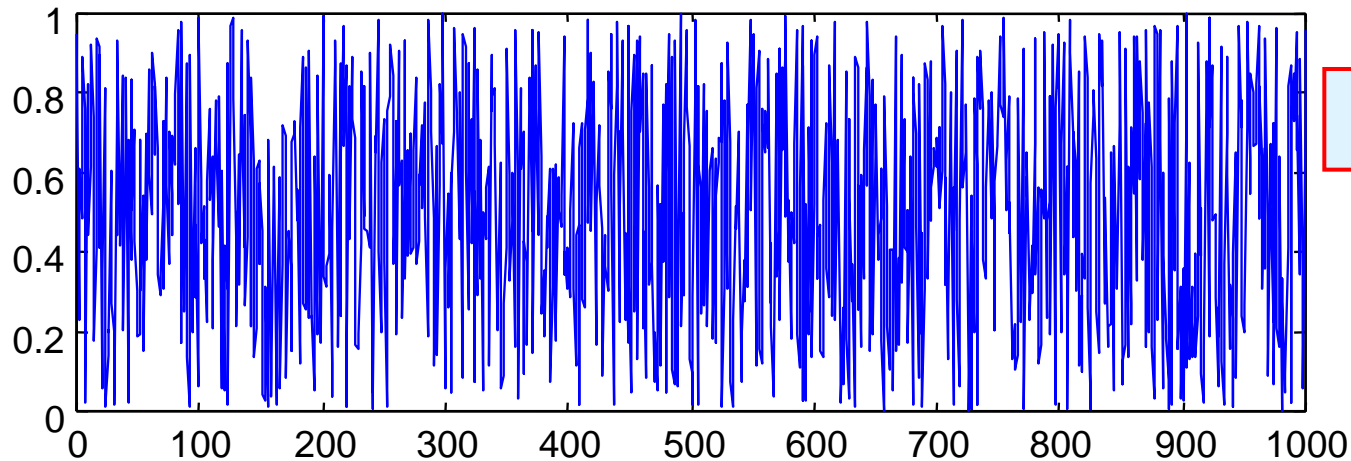


$$x = 0, \sigma = 1$$

# Noise with uniform distribution

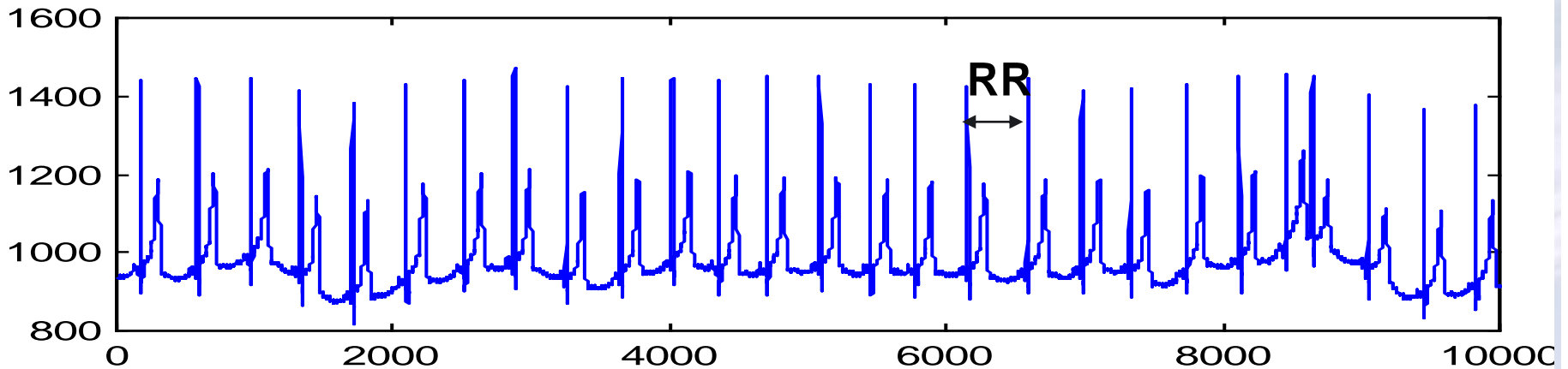


```
#Python
range=1
y=range*random.rand(1000)
bins=linspace(-1,1.5,50)
hist(y,bins, normed='True')
```



# Markov Process

$$p[x(n)/x(n-1), x(n-2), \dots, x(1)] = p[x(n)/x(n-1)]$$



Is the series  $\{RR_1, RR_2, \dots, RR_n, \dots\}$  a Markov process?

## Poisson discrete distribution

Poisson distribution is applied in observations of independent phenomena with low probability of success (eg radioactive decay)

Probability of  $j$  events in time  $\Delta t$ :

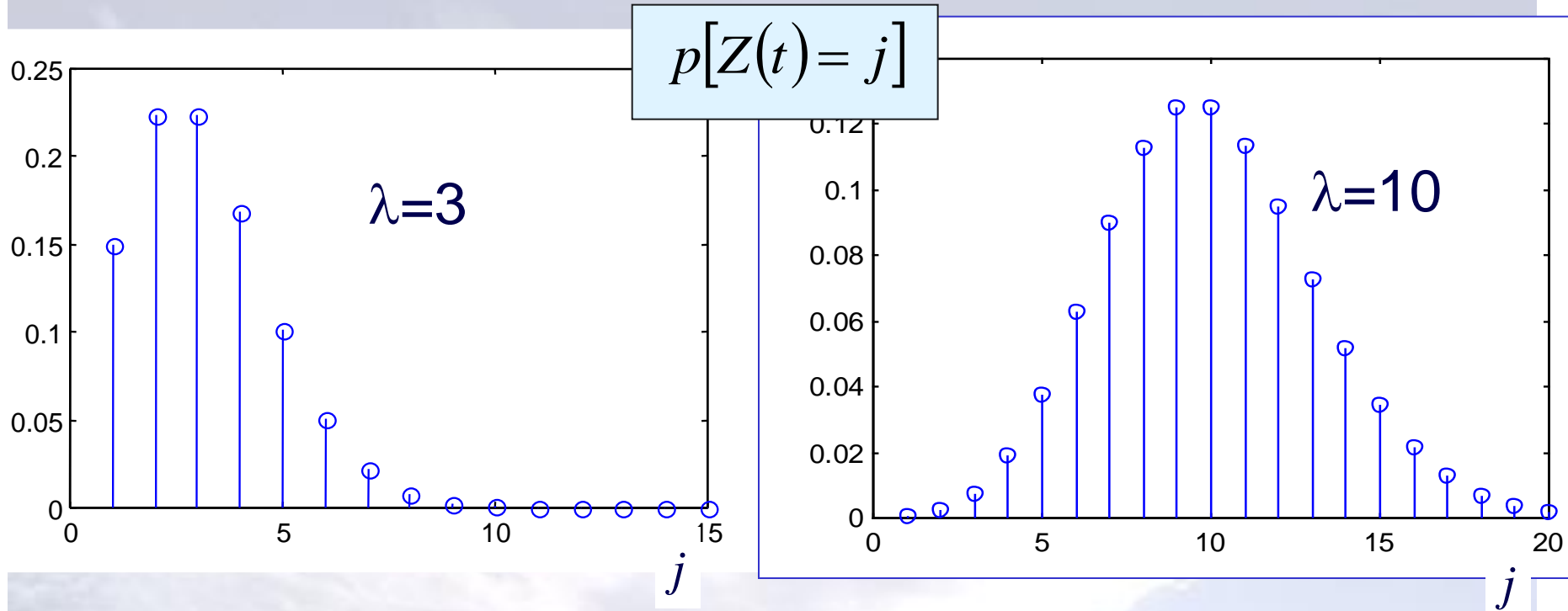
$$p[Z(\Delta t) = j] = \frac{(\lambda \Delta t)^j}{j!} e^{-\lambda \Delta t}, \quad \lambda > 0 \quad \Rightarrow \quad \begin{aligned} E(Z) &= \lambda \\ D^2(Z) &= \lambda \end{aligned}$$

where  $\lambda$  is a mean number of events in  $\Delta t$ .

### Application:

RR intervals modeling, EMG signal, neural impulses, quality control

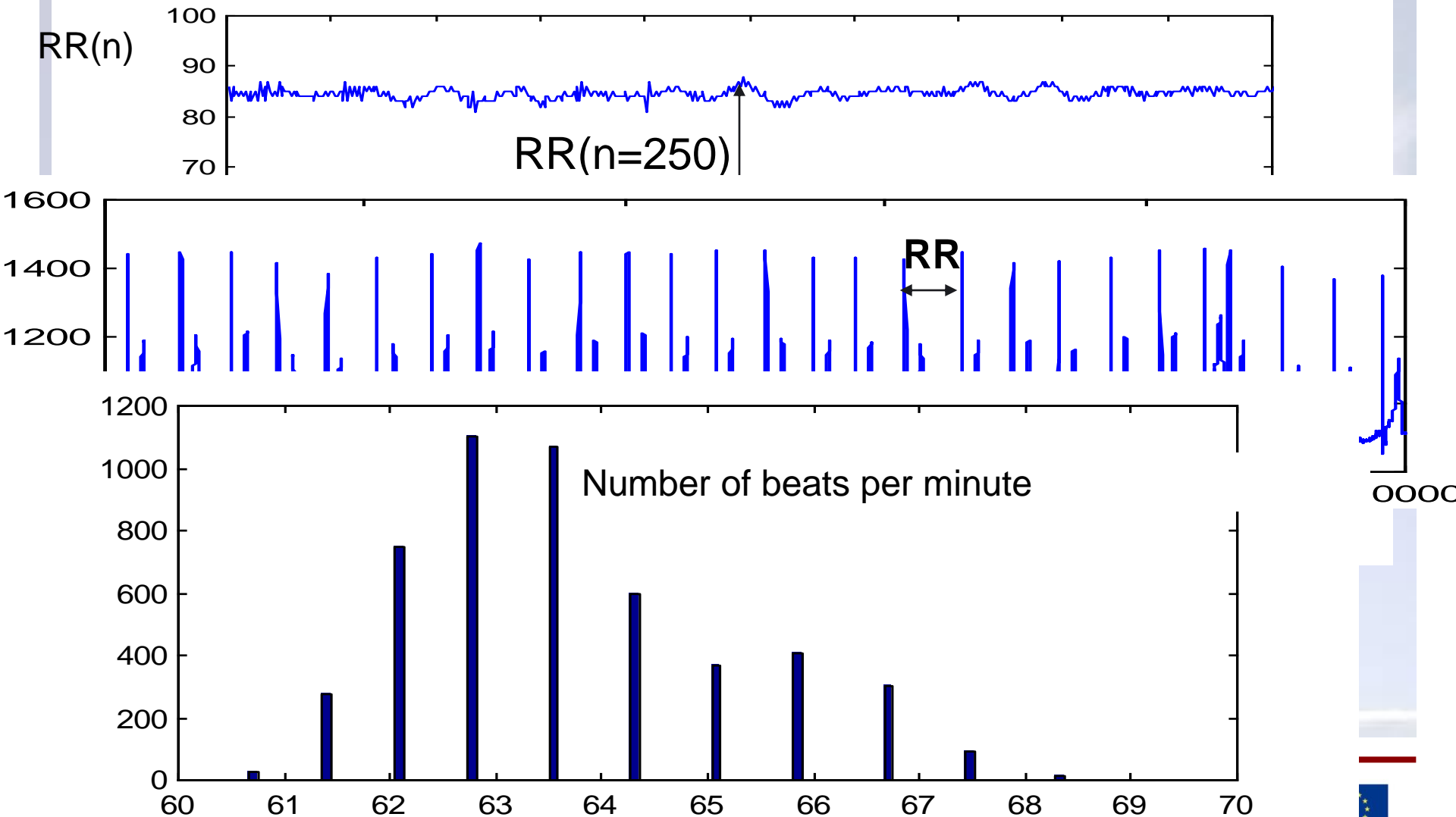
# Poisson distribution



Poisson process describes so called memoryless process, where the number of events in time is independent on the past.



# Heart rate variability - HRV



# Linear regression

Let's consider:

$$y = \alpha_0 + \alpha_1 x + \varepsilon$$

where:

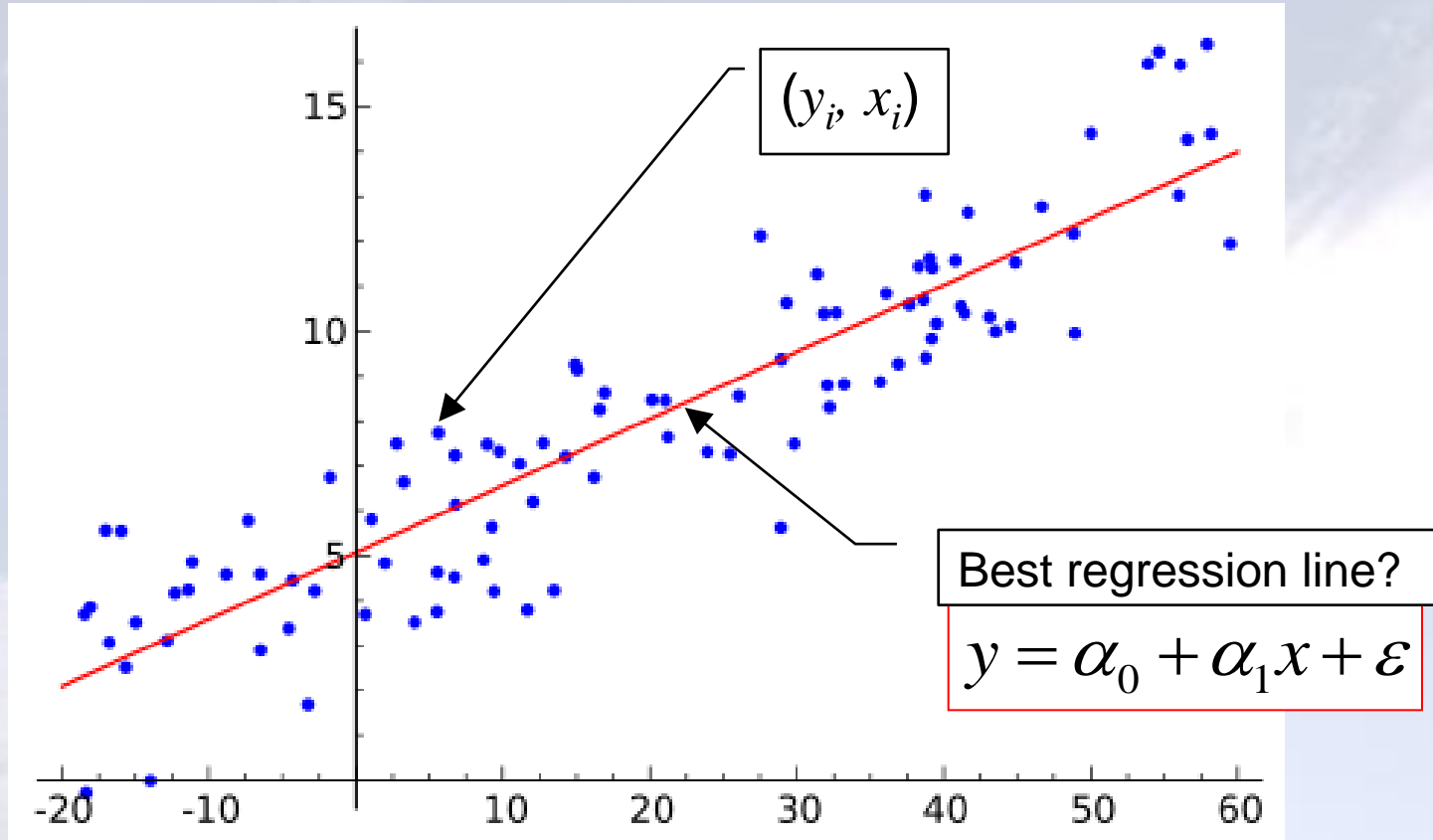
$x$  – independent variable,

$y$  – dependent variable,

$\varepsilon$  - noise (eg. measurement error),

$\alpha_0, \alpha_1$  – regression coefficients.

# Linear regression





# Linear regression - estimation

## Aim:

Suppose  $P$  observations  $(x_1, y_1), (x_2, y_2), \dots, (x_P, y_P)$  are given. Determine  $\alpha_0, \alpha_1$ , which most accurately relate variables  $x$  i  $y$ .

## Solution:

Estimates of the coefficients may be derived by means of the least square method.

$$\varepsilon_i = y_i - (\alpha_0 + \alpha_1 x) = y_i - \hat{y}_i$$

observation

Linear regression model

# Linear regression - estimation

## summed squared error - SSE

$$SSE = \sum_{i=1}^P \varepsilon_i^2 = \sum_{i=1}^P (y_i - (\hat{\alpha}_0 + \hat{\alpha}_1 x_i))^2$$

According to the idea of least squares method proposed by Gauss in XIX-th century.

The minimum of the function is to be established.

# Linear regression - estimation

The minimum is determined by setting the partial derivatives to zero.

$$\left\{ \begin{array}{l} \frac{\partial(SEE)}{\partial \hat{\alpha}_0} = \frac{\partial}{\partial \hat{\alpha}_0} \left( \sum_{i=1}^P (y_i - \hat{\alpha}_0 - \hat{\alpha}_1 x_i)^2 \right) = 0 \\ \frac{\partial(SEE)}{\partial \hat{\alpha}_1} = \frac{\partial}{\partial \hat{\alpha}_1} \left( \sum_{i=1}^P (y_i - \hat{\alpha}_0 - \hat{\alpha}_1 x_i)^2 \right) = 0 \end{array} \right\}$$

So called normal set of equations

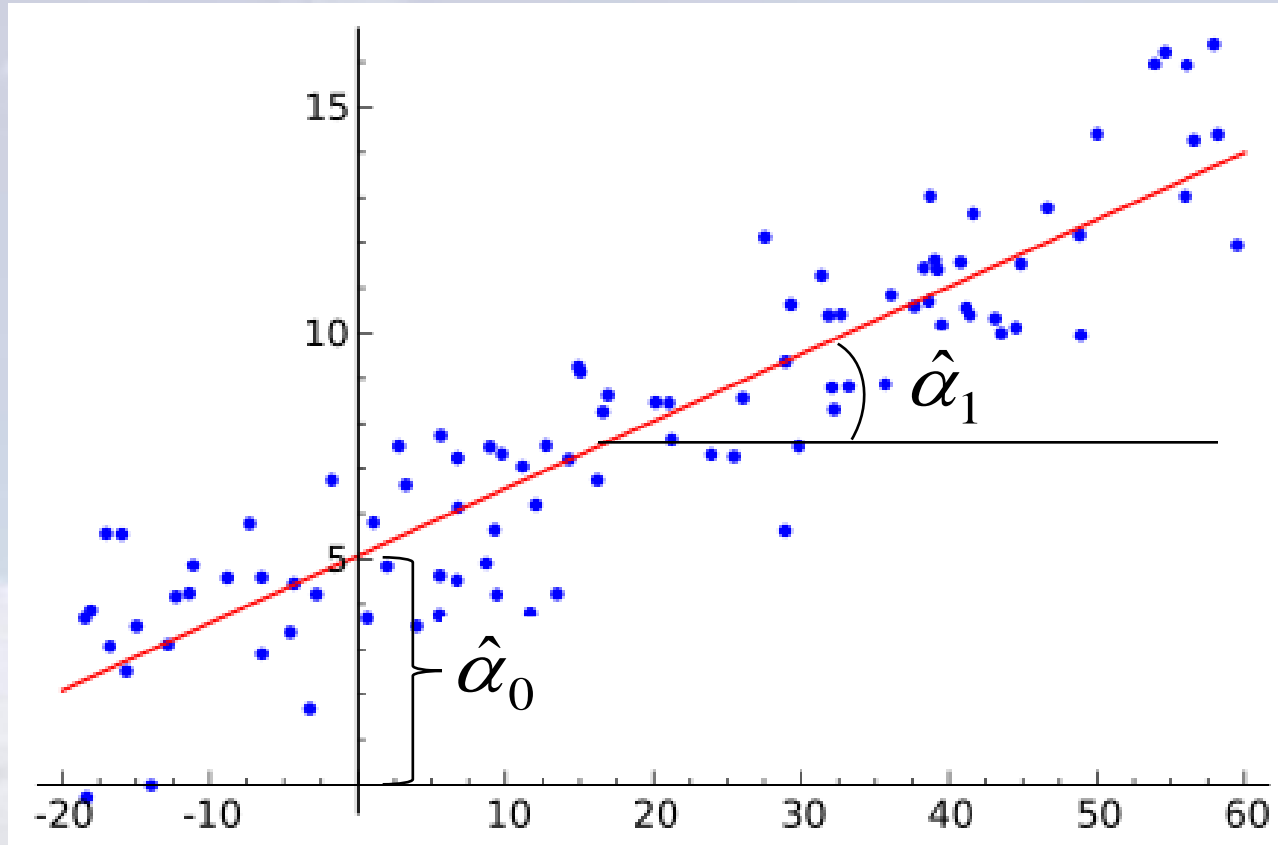
# Linear regression - estimation

The solution to the normal set of equations:

$$\hat{\alpha}_0 = \bar{y} - \hat{\alpha}_1 \bar{x}$$

$$\hat{\alpha}_1 = \frac{\sum_{i=1}^P (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^P (x_i - \bar{x})^2}$$

# Linear regression - estimation



Source: Wikipedia

# Linear regression

Higher order regression:

$$y_i = \alpha_0 + \alpha_1 x_{i1} + \alpha_1 x_{i1} + \dots + \alpha_N x_{iN}, \quad i = 1, \dots, P$$

$$y_i = \alpha_0 + \sum_{j=1}^N \alpha_j x_{ij}, \quad i = 1, \dots, P$$

Analogous solution:

$$\frac{\partial(SEE)}{\partial \hat{\alpha}_j} = 0, \quad j = 0, \dots, N$$

( $N+1$  equations with  $N+1$  variables)

## Other regression models

Polynomial regression:

$$y_i = \alpha_0 + \alpha_1 x_{i1}^1 + \alpha_1 x_{i2}^2 + \dots + \alpha_N x_{iN}^N, \quad i = 1, \dots, P$$

Nonlinear regression:

$$y_i = \alpha_0 e^{\alpha_1 x_i}$$

$$y_i = \alpha_0 \alpha_1^{x_i}$$

$$y_i = \frac{\alpha_0}{1 + \alpha_1 e^{x_i}}$$

# Trajectory reconstruction

Takens (1981) proved, the properties of the trajectories of the dynamic system may be reconstructed from the samples of a 1-dimensional registration of the activity of the given system.

$$\mathbf{y}(k) = [y(k), y(k - \Delta t), \dots, y(k - (D - 1)\Delta t)]^T$$

For the appropriate length  $D$  of the vector  $\mathbf{y}(k)$  (*time delay embedding*)



# Trajectory reconstruction

In practice Takens theorem means that there exists a dependance:

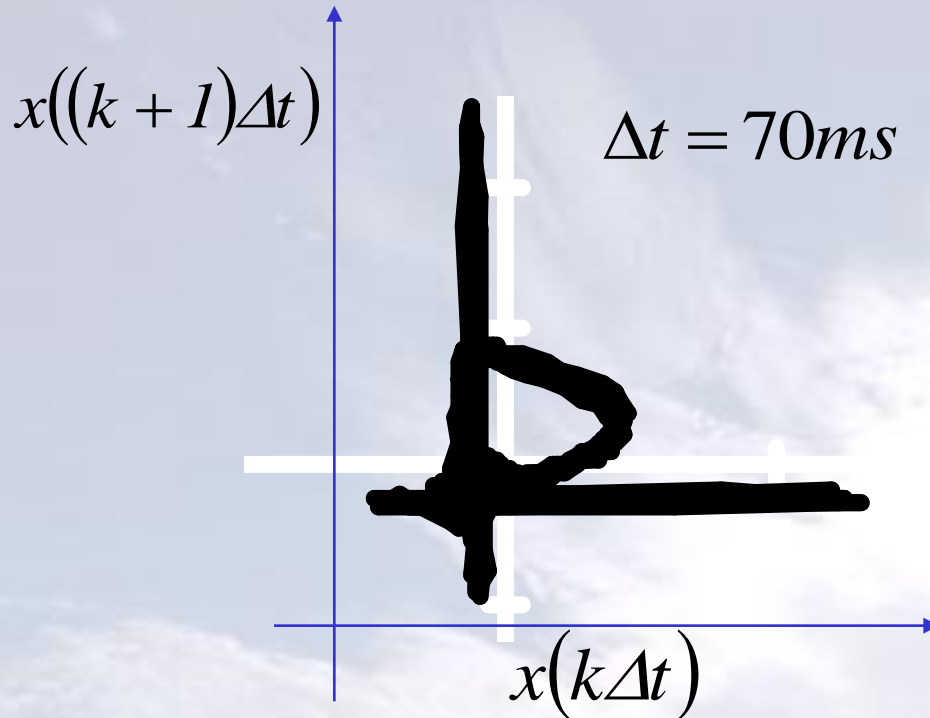
$$y((k + 1)\Delta t) = F(y(k))$$



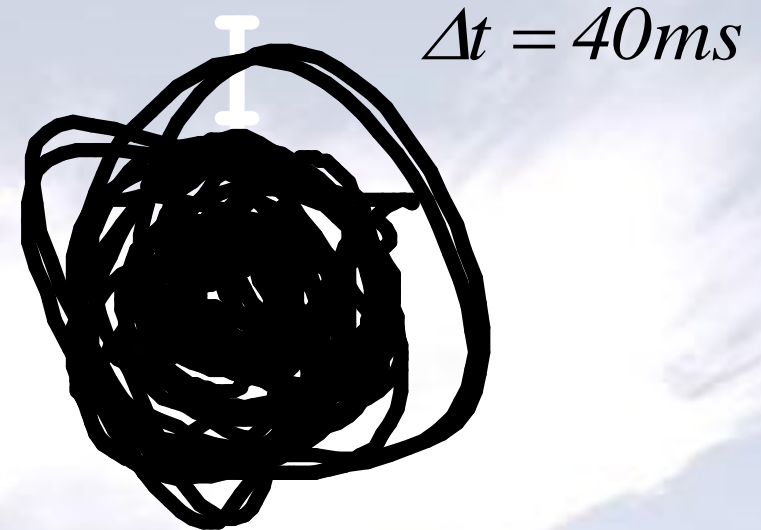
?

due to which it is possible to forecast the sample  $y(k+1)\Delta t$  of the signal reflecting the behaviour of the given system.

# Trajectory reconstruction for heart rhythm



*Normal heart rhythm*



*Ventricle fibrillation*

$\Delta t$  – first zero of the autocorrelation function of the series  $y(k)$

# Deterministic chaos

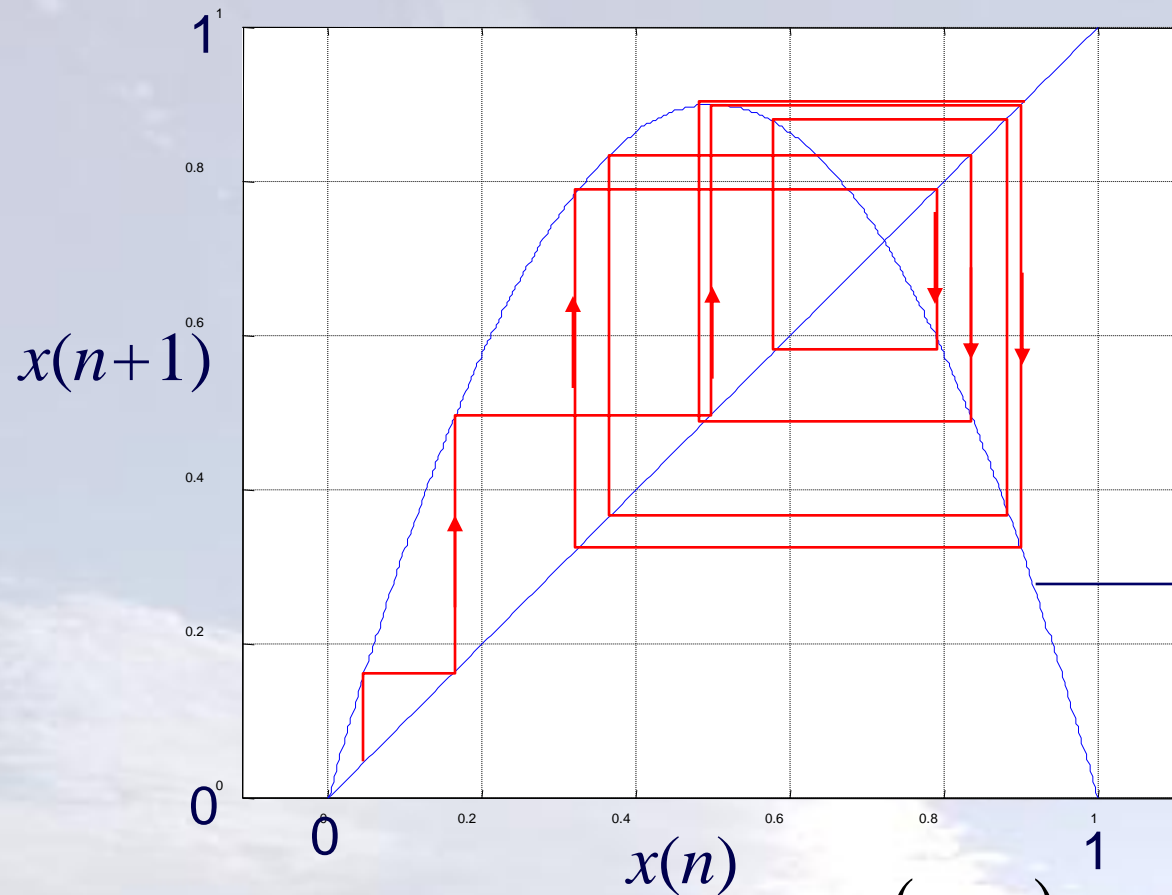
Consider the simple dynamic system:

$$x(n+1) = ax(n)(1-x(n))$$

Simulating the change of number of population  $x$  in limited environment. This number is normalized to  $\langle 0,1 \rangle$ , therefore the value of parameter  $a$  belongs to  $\langle 0,4 \rangle$ .

This is the, so called, **logistic equation**.

# Deterministic chaos

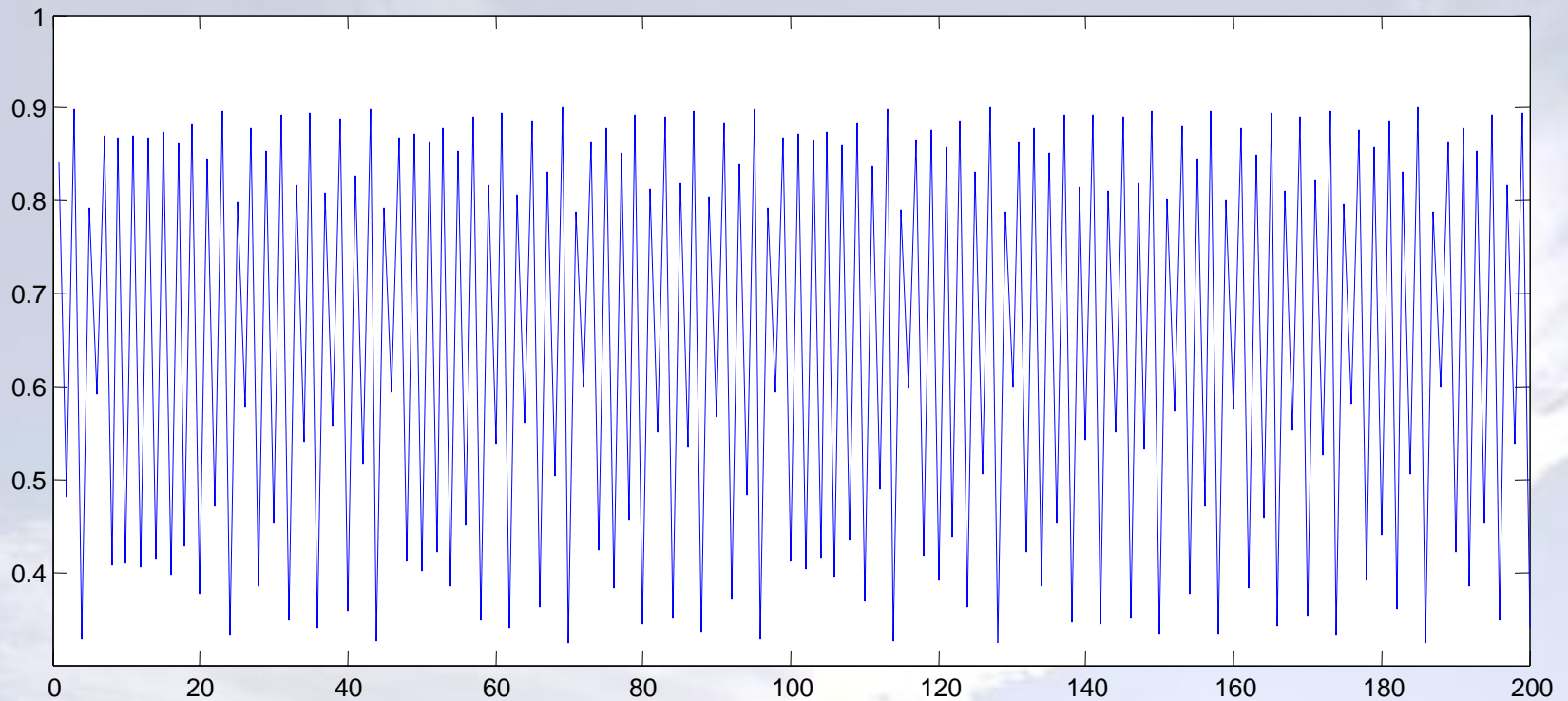


So called „square projection”

$$x(n+1) = 3.6x(n)(1 - x(n))$$



# Deterministically chaotic process



# Deterministic chaos

The properties of the deterministic, chaotic processes:

- may be generated by simple dynamic systems \*,
- the dynamics has „unpredictable” random character (so called **strange attractors**) → demo Matlab
- Dynamics highly dependent on the initial conditions („**butterfly effect**”)

\*) R, May: „ ... *life would be easier if it was taken into account, that simple dynamic systems not always lead to simple dynamic evolution*”



# Analysis of random signals - summary

1. Deterministic vs stochastic signals
2. Mean, variance, standard deviation
3. Autocorrelation function
4. Cross-correlation function
5. Correlation coefficient
6. Power Spectral Density
7. Distributions
8. System dynamics





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