





# Discrete Time Signal Processing

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#### **INTRODUCTION**



Difference between analog and digital signals?

**Pros and Cons** for using digital signal processing?

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Digital signal => sequence

Scilab software to compute data

# **INTRODUCTION**

Applications:

Medical signals (EEG, MRI,

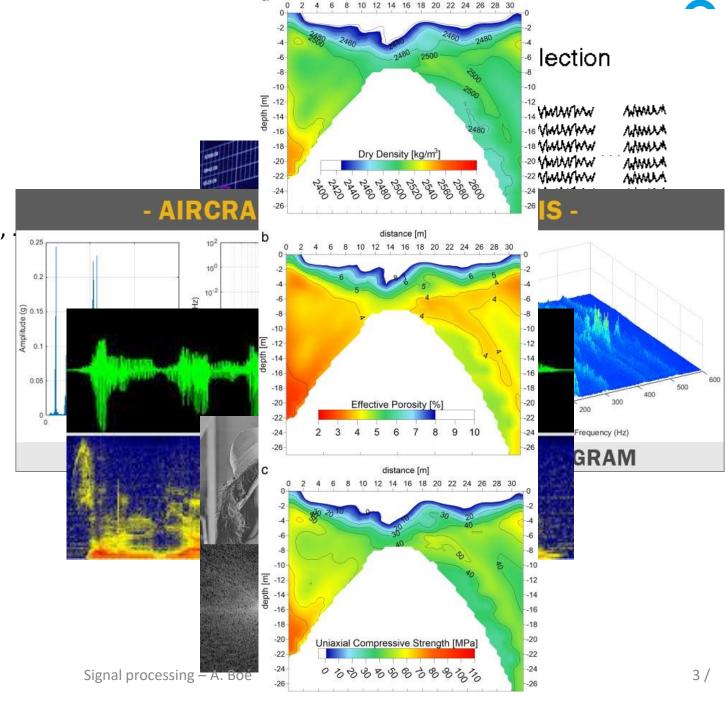
Financial engineering

Data analysis

Speech processing

Image processing

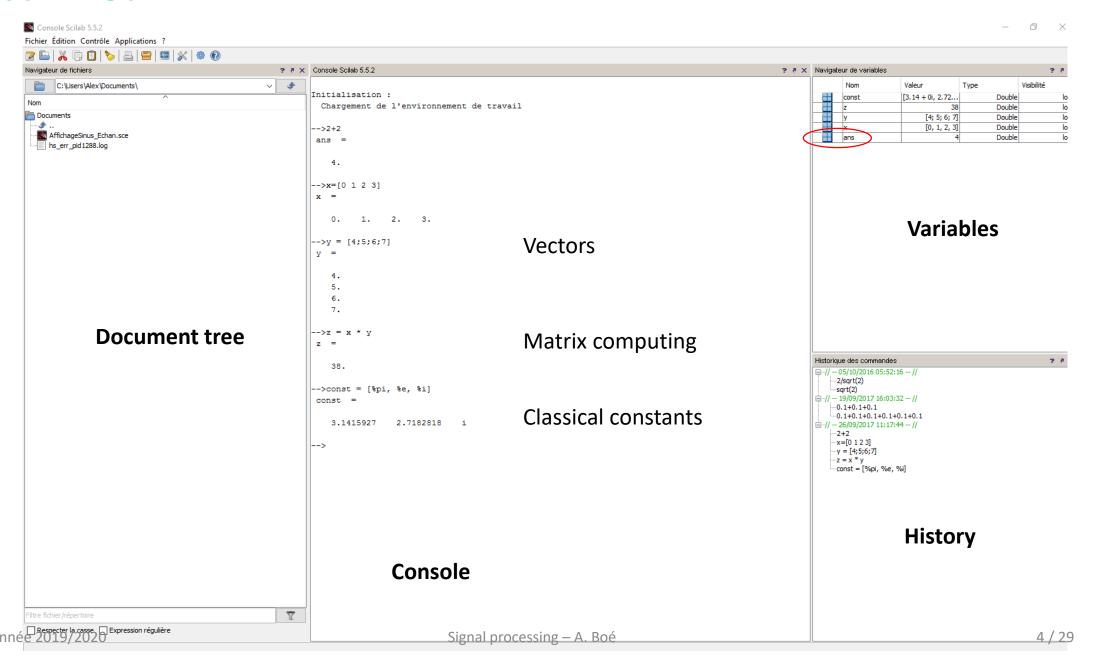
Seismic survey



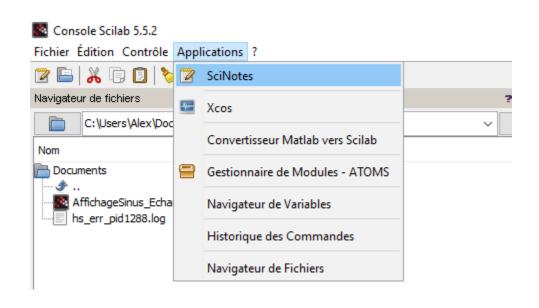
distance [m]

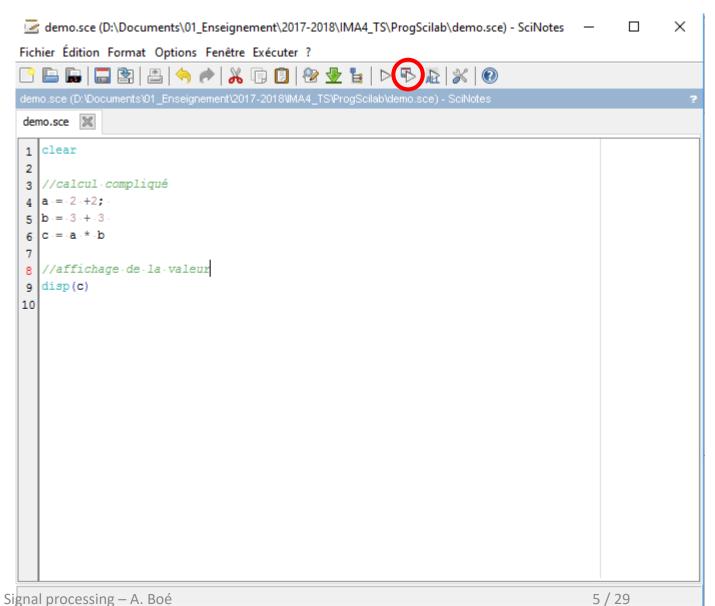
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Numerical computing

Programming language:

variables

comments

structures

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Display

**Useful fonction: help** 



# Double float by default (64 bit)

Computing on matrices

Display

Programming

Command			Description		
abs	sign		Module and sign		
real	imag		Real and imaginary part		
exp	log	log10			
cos	sin	tan			
acos	asin	atan			
cosh	acosh	•••			
sqrt			Square root		
floor			Round down		
round			Round to nearest integer		
ceil			Round up		
int			Round towards zero		
rand()			Random number generator		



#### Double float by default (64 bit)

Computing on matrices
Display
Programming

#### Exercice 1.

display 2^53, 2^53 +1, ..., 2^53+10 conclusion ?

#### Exercice 2.

compute sinus of 180° and Pi radians conclusion?

#### Voir %eps/2

**Exercice 3.** IEE-754 64 bit floating point number



compute 1+%eps and 1+%eps/2 conclusion?

display 0.1 with maximum number of digits conclusion?

associativity with respect to addition / multiplication distributivity of multiplication over addition

Assess derivative of sin(x) at x=1 and conclude on aproximation and convergence



Double float by default (64 bit)

# Computing on matrices

Display

Programmin	Command	Description	
	[0, 1, 2]	Row vector	
	[0; 1; 2]	Column vector	
	[0,1,2;0,1,2]	2*2 matrix	
	A(i, j)	A <sub>ij</sub> element	
	A(i1:i2, j1:j2)	Sub matrix	
	\$	Means last index	
	+ - * ^	Operations on matrices	
	.+* .^	Elementwise operations	
	,	Transpose and conjugate	
		Transpose (no conjugate)	
	one(x,y) zero(x,y) eye(x,y)	Predefined matrices	
Année 2019/202	min:step:max Signal proce	Vector from min to max with step	



Double float by default (64 bit)

# Computing on matrices

Display

Programmir

Command		Description	
size()		Size of the matrix	
det()		Determinant of the matrix	
rank()		Rank of the matrix	
inv()		Inverse of the matrix	
sum() prod(	()	Sum and product of vector coefficients	
min() max()	)	Maximum or minimum coefficient	
norm()		Norm of a vector	
spec()		Eigenvalue / eigenvectors	
bdiag		Diagonalized matrix	
det()		Determinant of the matrix	



Double float by default (64 bit) Computing on matrices

Display

Programming

Simple graph: two vectors using plot2d

style (negative numbers for different points, positive numbers for colors)

*subplot* for multiplots

xtitle for titles and axis captions

and many other functions ... go to visit help!

**Exercice 4.** Plot the function  $\frac{\sin x}{x}$ 



Double float by default (64 bit) Computing on matrices Display

Programming

#### Looping and branching

if ... then ... else ... end

select ... case ... else ... end

for ... end

while ... end

break and continue

#### Exercise 5.

Determine the Fourier's series of a square signal (periodic) Display on a graph the sum of harmonics for n = 1, 3, 5, 7, 9 ...



$$x[n] = x_c(nT_S)$$
 avec  $-\infty < n < +\infty$ 

 $T_S$  is the sampling period  $F_S = \frac{1}{T_S}$  is the sampling frequency

Experiment the sampling of periodic signals and expose the Shannon criterion



Effect of sampling:

$$x_{ei}(t) = x(t) \cdot \delta_{Te}(t) = \sum_{n=\infty}^{\infty} x(t) \delta(t - nT_e)$$

Dirac:

$$e_i(t) = \delta_{Te}(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT_e) = \sum_{n=-\infty}^{\infty} \delta(t - \frac{n}{f_e})$$

Fourier's Transform:

$$\label{eq:deltaTe} \begin{tabular}{ll} $\mathcal{F}\left\{\delta_{Te}(t)\right\}$ & = \begin{tabular}{ll} $\mathcal{F}\left\{\sum_{n=-\infty}^{\infty}$ $\delta(t-\frac{n}{f_e})$\right\}$ & = f_e \; \delta_{fe}(f) = f_e \; \sum_{k=-\infty}^{\infty}$ $\delta(f-k \; f_e)$ \\ \end{tabular}$$

$$X_e(f) = X(f) * f_e \delta_{fe}(f) = f_e \sum_{k=-\infty}^{\infty} X(f - kf_e)$$



Considering a continuous signal x(t) and its sampled version  $x_S(t)$ :

$$(t) = \sum_{k=-\infty}^{+\infty} \delta(t - kT_S)$$

Apply the Fourier's transform to both parts:

$$TF[x_S(t)] = TF[x(t)] * F_S \sum_{k=-\infty}^{+\infty} \delta(f - kF_S)$$

We can note that:

$$x(t) * \delta(t - a) = x(t - a)$$

Then:

$$TF[x_S(t)] = F_S \sum_{k=-\infty}^{+\infty} X(f - kF_S)$$



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Signal reconstruction:

$$x_S(t) = x(t) \sum_{k=-\infty}^{+\infty} \delta(t - kT_S) = \sum_{k=-\infty}^{+\infty} x(kT_S)\delta(t - kT_S)$$

The sampling repects the Shannon criterion, and applying a low pas filter ( $F_{stop} = \frac{F_S}{2}$ ). The transfer function of the filter is given by:  $\frac{1}{F_S} \prod \left( \frac{f}{F_S} \right)$ .

At the output of the filter:

$$\left[X(f) * F_S \sum_{k=-\infty}^{+\infty} \delta(f - kF_S)\right] \frac{1}{F_S} \prod \left(\frac{f}{F_S}\right)$$

Apply inverse FT:

$$x(t) = TF^{-1} \left[ \left[ X(f) * F_S \sum_{k=-\infty}^{+\infty} \delta(f - kF_S) \right] \frac{1}{F_S} \prod \left( \frac{f}{F_S} \right) \right] = x_S(t) * sinc(\pi F_S t)$$

Then:

$$x(t) = \left[\sum_{k=-\infty}^{+\infty} x(kT_S)\delta(t-kT_S)\right] * sinc(\pi F_S t) = \sum_{k=-\infty}^{+\infty} x(kT_S) sinc[\pi F_S t(t-KT_S)]$$
Signal processing – A. Boe



$$X(k) = \frac{1}{N} \sum_{n=0}^{N-1} s(n) e^{-2i\pi n \frac{k}{N}}$$

(Non unique definition ...)

Demonstration:

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Exercice:

DFT of  $x_{k \in [0,15]} = \{1,1,1,0,...,0\}$ 

DFT inverse

$$x(n) = \frac{1}{N} \sum_{k=0}^{N-1} X(k) e^{2i\pi n \frac{k}{N}}$$



#### Properties:

- Linearity
- Delay:  $x(n n_0) = X(k)e^{-i2\pi k\frac{n_0}{N}}$
- $(X)_{n\in\mathbb{Z}} = TFD((x_k)_{k\in\mathbb{Z}})$ 
  - $TFD(x_{-k}) = X_{-n}$
  - $-TFD(\overline{x_k}) = \overline{Y_{-n}}$
  - $-TFD(\overline{y_{-k}}) = \overline{Y_N}$
- Convolution:

$$(x_k)_{k\in\mathbb{Z}}$$
 and  $(y_k)_{k\in\mathbb{Z}}$ 

 $(x_k)_{k\in\mathbb{Z}}$  and  $(y_k)_{k\in\mathbb{Z}}$  The circular or cyclic convolution is z=x\*y defined by  $\forall k\in\mathbb{Z}$ ,  $z_k=x*y=\sum x_qy_{k-q}$ 

$$X = TFD(x), Y = TFD(y), Z = TFD(z) Z_k = NX_kY_k$$

$$TFD(x \cdot y) = X * Y$$

• Energy: 
$$\sum_{k=0}^{N-1} |y_k|^2 = N \sum_{n=0}^{N-1} |Y_n|^2$$



Link with Fourier's Transform

#### **Exercice 1. DFT**

On considère un signal rampe échantillonné avec un pas de 1s.

- 1. Ecrire la suite x[n] pour N (nombre d'échantillon) = 4.
- 2. Calculer 'à la main' la transformée de Fourier discrète du signal.
- 3. À partir de cet exemple, écrire un algorithme simple de calcul de la TFD d'un signal échantillonné.
- 4. Donner le nombre d'opérations à effectuer.



#### FFT algorithm

#### **Exercice 2. FFT**

On considère maintenant la fonction  $x(t) = e^{-j2\pi F_0 t}$ 

- 1. Donner le spectre de cette fonction par calcul
- 2. Vérifier à l'aide de Scilab le calcul, en utilisant la fonction écrite précédemment et en utilisant la fonction fft
- 3. Comparer le temps de calcul des deux méthodes



#### Exercice 3. Propriétés de la TFD sur un signal périodique

- 1. Calculer la TFD d'un signal sinusoïdal sur un nombre entier de périodes
- 2. Faire de même sur un nombre non entier de périodes
- 3. Conclusion?
- 4. Faire le même exercice en considérant une somme de deux sinusoïdes d'amplitude égale et de fréquences proches puis de fréquences « éloignées » avec des amplitudes très différentes
- 5. Conclusions ?

TFD gives exact results for: a multiple integer of the signal period

a window length equal to a multiple of sampling period (always true on pratical)<sup>2</sup>

# ww.wikipedia.fr

#### LESSON 2. DISCRETE FOURIER TRANSFORM AND FAST FOURIER TRANSFORM



Fenêtre rectangulaire, qui conduit à l'approximation sigma :

$$h(t) = \left\{egin{array}{ll} 1 & ext{si } t \in [0,T] \ 0 & ext{sinon.} \end{array}
ight.$$

Fenêtre triangulaire (de Bartlett) :

$$h(t) = egin{cases} rac{2t}{T} & ext{si } t \in [0, rac{T}{2}[\ rac{2(T-t)}{T} & ext{si } t \in [rac{T}{2}, T] \ 0 & ext{sinon}. \end{cases}$$

Fenêtre de Hann:

$$h(t) = egin{cases} rac{1}{2} - rac{1}{2}\cos(2\pirac{t}{T}) & ext{si } t \in [0,T] \ 0 & ext{sinon}. \end{cases}$$

Fenêtre de Hamming:

$$h(t) = egin{cases} 0,54-0,46\cos(2\pirac{t}{T}) & ext{si } t \in [0,T] \ 0 & ext{sinon}. \end{cases}$$

Fenêtre de Blackman :

$$h(t) = egin{cases} 0,42-0,5\cos(2\pirac{t}{T})+0,08\cos(4\pirac{t}{T}) & ext{si } t \in [0,T] \ 0 & ext{sinon}. \end{cases}$$

Et d'autres : fenêtres de Kaiser (en) (de paramètre  $\alpha$ ), gaussienne, flat top, en cosinus relevé...

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Fenêtre	Lobe 2aire (dB)	Pente (dB/oct)	Bande passante (bins)	Perte au pire des cas (dB)	
Rectangulaire	-13	-6	1,21	3,92	
Triangulaire	-27	-12	1,78	3,07	•
Hann	-32	-18	2,00	3,18 3,10	D D J
Hamming	-43	-6	1,81	3,10	「 -
Blackman-Harris 3	-67	-6	1,81		· × ×



How to choose the pertinent window?

- For signal with distant frequencies, use low secondary lobe window
- •For a highest resolution in close frequencies, use low main lobe window.
- For a highest resolution on amplitude (with lower resolution on frequency), use large main lobe window.
- With a wideband signal, use flat window or no window (= rectangular)
- Hann window is the most versatile, with a good frequency resolution. In case of unknown signal, it should be a good starting choice



#### Window choice:

#### Type de signal

Transitoires dont la durée est inférieure à la longueur de la fenêtre

Transitoires dont la durée est supérieure à la longueur de la fenêtre

Applications standards
Analyse spectrale (mesures de réponses fréquentielles)

Séparation de deux tons dont les fréquences sont très proches mais dont les amplitudes sont très différentes

Séparation de deux tons dont les fréquences sont très proches mais dont les amplitudes sont presque identiques

Mesures précises de l'amplitude d'un ton unique Signal sinusoïdal ou combinaison de signaux sinusoïdaux

Signal sinusoïdal avec nécessité de précision de l'amplitude
Signal aléatoire à bande étroite (données de vibration)
Signal aléatoire à large bande (bruit blanc)
Signal sinusoïdal avec courbes rapprochées
Signaux d'excitation (coup de marteau)
Signaux de réponse
Signaux dont le contenu est inconnu

#### Fenêtre

Rectangulaire

Exponentielle, Hann

Hann
Hanning (pour excitation aléatoire), rectangulaire
(pour excitation pseudo-aléatoire)

Kaiser-Bessel

Rectangulaire

À profil plat Hann

À profil plat
Hann
Uniforme
Uniforme, Hamming
Force
Exponentielle
Hann



#### **Exercice 4. Fenêtrage**

On considère un signal sinusoïdal observé sur un nombre entier de périodes ou sur un nombre non entier de périodes, puis deux signaux de même amplitude avec des fréquences proches, puis deux signaux avec des amplitudes très différentes :

- 1. Tracer les fenetres Hann, Hamming, Blackman
- 2. Appliquer au signal une fenêtre de Hann, de Hamming et de Blackman
- 3. Calculer dans les trois cas le spectre du signal obtenu
- 4. Conclusions?



#### **Exercice 5. Zero padding**

On veut artificiellement augmenter la résolution de la FFT pour obtenir un affichage de meilleure qualité :

- 1. Représenter un signal temporel avec deux sinusoïdes de même amplitude et proches en fréquence et une troisième avec une amplitude faible comparée aux deux autres
- 2. Calculer le spectre du signal
- 3. Ajouter des zéros sur le signal et observer le spectre obtenu



#### **Summary:**

- 1. Complexity : DFT =  $N^2$ , FFT algorithm =  $N \log N$  when  $N = 2^P$
- 2. FFT gives exact results for a multiple integer of period
- 3.  $k^{th}$  sample represent the frequency  $F_k = k \frac{F_S}{N}$
- 4. Frequency resolution is roughly proportionnal to  $NF_s$
- 5. Window

#### LESSON 4. RANDOM SIGNALS

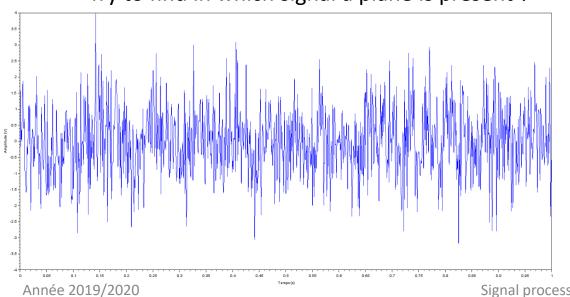


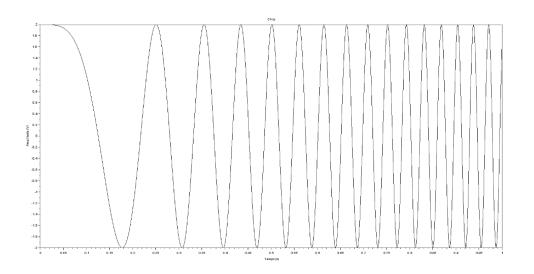
#### Exercise 4.1:

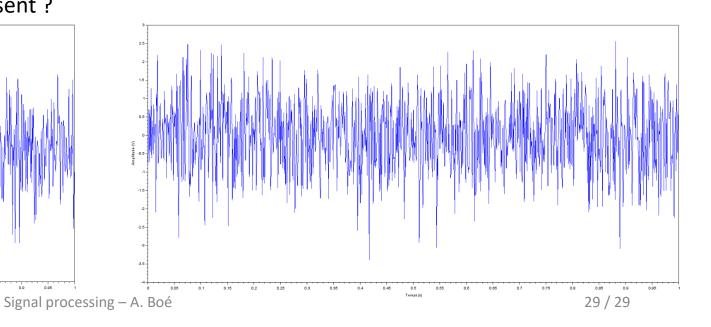
Extraction of a signal: simulation of RADAR signals

Simulate a chirp emission of a RADAR for plane detection

- Add noise and a return signal if a plane is present
- Have a look at received signals
- Try to find in which signal a plane is present?







#### LESSON 4. RANDOM SIGNALS



Exercice 4.2: Random signal and Wiener-Khintchine's theorem.

- We consider a Gaussian process (electronics noise)
- Write a random function and draw the probability graph of a Gaussian process
- Assess the power of a signal by summing the samples and by using Wiener-Kintchine theorem

 In real analysis, it is generally not possible to get as many samples or to do mean on different signals. Using a fixed number of samples and working on subset, do the same analysis and conclude

### LESSON 4. RANDOM SIGNALS



Exercice 4.3: Intercorrelation for leak detection.

- We consider a leak on a pipe, generating noise. By using two tranducers (e.g. microphones) on the pipe, we collect data.
- Give the expression of the signal received at the two transducers
- Write the intercorrelation functions
- Knowing the propagation speed of the signal, deduce the localization of the leak
- Application on data 'leak'

# **LESSON 5. FILTERING**



#### **BIBLIOGRAPHY**

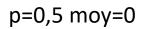


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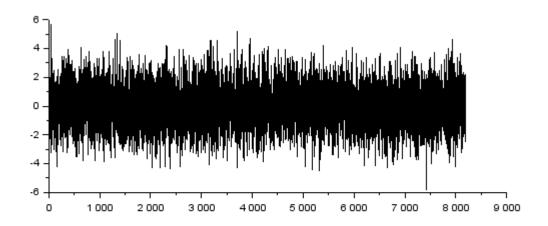
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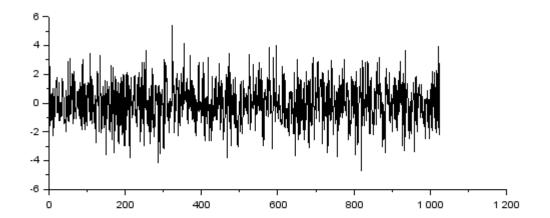
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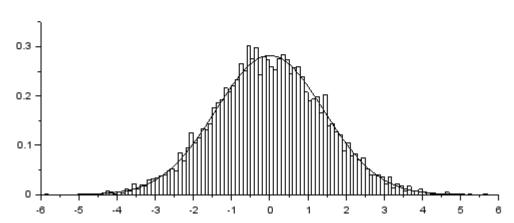
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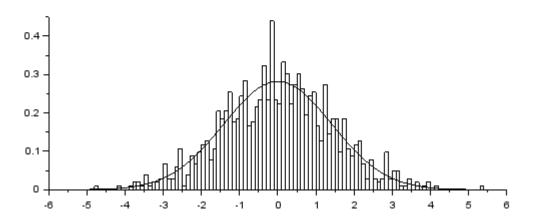








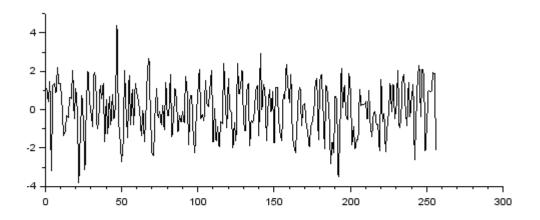


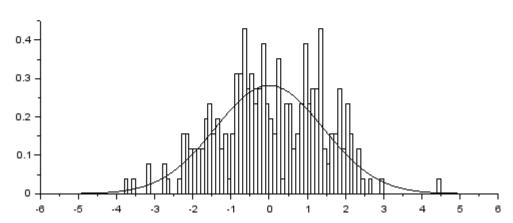


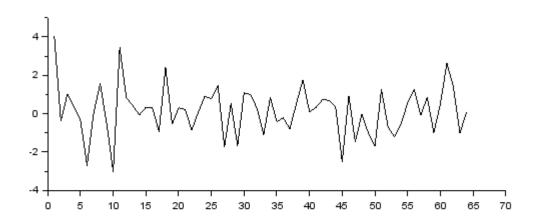
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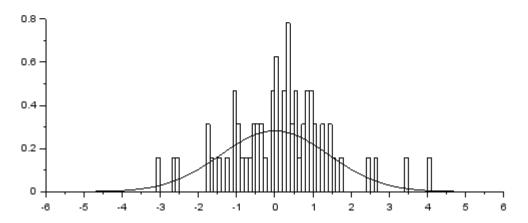












N=256 N=64



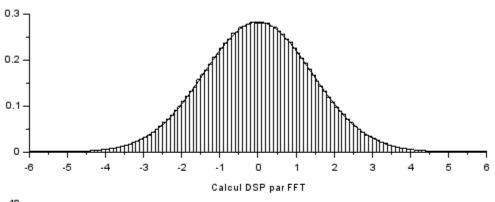


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