

NEED FOR A TIME-FREQUENCY REPRESENTATION

Jean VILLE, 1948

'Théorie et application de
la notion de signal analytique;
Cables et transmissions.'

'If we consider a musical piece
and that a note, for instance
an 'A', appears at least once in
it, the harmonic analysis will
represent the corresponding
frequency with a certain
amplitude and a certain phase,
but without localizing the 'A'
in time. However, it is obvious
that during this musical
piece there are instants for which

we do not hear the 'A' note.
Although the Fourier representation
is mathematically correct, because
the phases of nearby notes are
organized in such a way that
they destroy by interference the 'A'
when we do not hear it or
reinforce it, also by interference
when we hear it; but, if there
is in this conception a skill which
honours mathematical analysis,
we should not hide the fact
that there is a deformation of reality:
indeed, when we do not hear the
'A', the genuine reason is that
it has not been emitted.

To analyze a signal we have to
look for a mixed definition,
comparable to the one proposed.

By Gabor: at each instant we have a certain number of frequencies, which give the pitch and the timbre of the sound emitted such as we hear it; to each frequency is associated a certain time distribution which corresponds to the instants when the note is emitted.

We are then led to define an instantaneous spectrum, function of time, which gives the structure of the signal at a given instant; the signal spectrum, in the usual Fourier sense, which gives the frequency structure of the signal for its entire duration, is then obtained by adding all the instantaneous spectra'.

BRIEF HISTORY

Theory:

- 1984 Continuous wavelet transform, (CWT)
Alex Grossmann & Jean Morlet
- 1986 orthogonal wavelet transform,
Yves Meyer & Pierre-Gilles Lemarie'
- 1987 Compactly supported wavelets,
Ingrid Daubechies
- 1988 Fast wavelet transform,
Stephane Mallat
- 1989 CWT in n dimensions,
Romain Murenzi

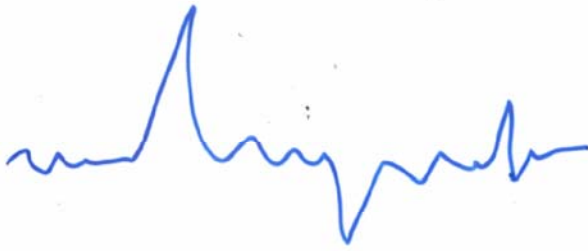
Applications:

- 1984 Geophysics, Morlet
- 1986 Marie, Risset & Krouland-Martinet
- 1986 quantum mechanics, Paul & Daubechies
- 1987 Fractals, Holschneider & Jaffard
- 1988 Image processing, Mallat
- 1988 Turbulence, Farge
- 1990 PDE's, Liandrato & Tchamitchian
- 1991 Numerical analysis, Coifman & Beylkin
- 1996 Navier-Stokes, Schneider, Perrier

Reference Book

Barbara Burke
The world according to wavelets
A. K. Peters, Wellesley, 1996

INTEGRAL TRANSFORM PRINCIPLE



Signal $f(x)$
to be analyzed



Analyzing
functions $\psi_k(x)$

Analysis

$$\hat{f}(k) = \int f(x) \psi_k(x) dx$$

Synthesis

$$f(x) = \frac{1}{c} \int \hat{f}(k) \psi_k(x) dk$$

Energy conservation (Parseval)

$$\int |f(x)|^2 dx = \frac{1}{c} \int |\hat{f}(k)|^2 dk$$

DIFFERENT FOURIER TRANSFORMS

Used for Signal Processing

$$\hat{f}(k) = \int_{-\infty}^{+\infty} f(x) e^{-2i\pi kx} dx$$

$$f(x) = \int_{-\infty}^{+\infty} \hat{f}(k) e^{+2i\pi kx} dk$$

$$e^{-\pi x^2} \xrightarrow{\mathcal{F}} e^{-\pi k^2}$$

$$\int |f(x)|^2 dx = \int |\hat{f}(k)|^2 dk$$

$$\widehat{f(x) * g(x)} = \hat{f}(k) \cdot \hat{g}(k)$$

$$\frac{\partial^n f(x)}{\partial x^n} = (2i\pi k)^n \hat{f}(k)$$

Used for Harmonic Analysis

$$\hat{f}(k) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} f(x) e^{-ikx} dx$$

$$f(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} \hat{f}(k) e^{ikx} dk$$

$$e^{-\frac{x^2}{2}} \xrightarrow{\mathcal{F}} e^{-\frac{k^2}{2}}$$

$$\int |f(x)|^2 dx = \int |\hat{f}(k)|^2 dk$$

$$\widehat{f(x) * g(x)} = \frac{1}{2\pi} \hat{f}(k) \cdot \hat{g}(k)$$

$$\frac{\partial^n f(x)}{\partial x^n} = (ik)^n \hat{f}(k)$$

Used for Group Theory

$$\hat{f}(k) = \int_{-\infty}^{+\infty} f(x) e^{-ikx} dx$$

$$f(x) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} \hat{f}(k) e^{ikx} dk$$

$$e^{-\pi x^2} \xrightarrow{\mathcal{F}} e^{-\frac{k^2}{4\pi}}$$

$$\int |f(x)|^2 dx = \frac{1}{2\pi} \int |\hat{f}(k)|^2 dk$$

$$\widehat{f(x) * g(x)} = \hat{f}(k) \cdot \hat{g}(k)$$

$$\frac{\partial^n f(x)}{\partial x^n} = (ik)^n \hat{f}(k)$$

HEISENBERG'S UNCERTAINTY PRINCIPLE

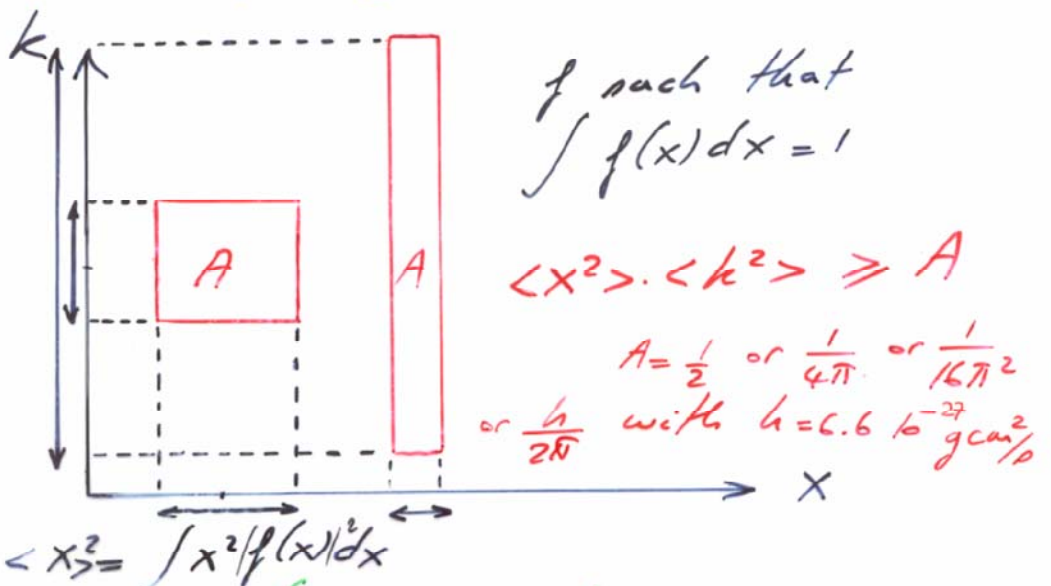
A signal cannot be concentrated in both time and frequency, or position and wavenumber.

This is not a limitation of our perception of reality. This is just by definition of the frequency or wavenumber.

$$\langle k^2 \rangle = \int k^2 |\hat{f}(k)|^2 dk$$

$$\hat{f}(2x) = \frac{1}{2} \hat{f}\left(\frac{k}{2}\right)$$

Space dilation \leftrightarrow Wavenumber contraction



Time-frequency plane,
 or position-wavenumber plane,
 or position-impulsion ($p = \frac{hk}{2\pi}$, h Planck's constant),
 or phase space,
 or information plane.

INTEGRAL TRANSFORM PRINCIPLE

Analysis

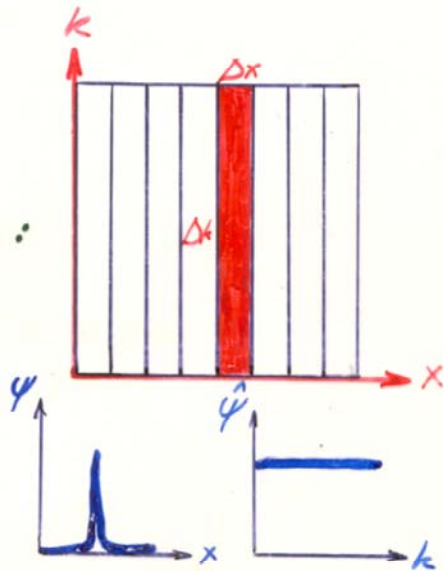
$$T_f(k) = \int f(x) \psi_k(x) dx$$

Synthesis

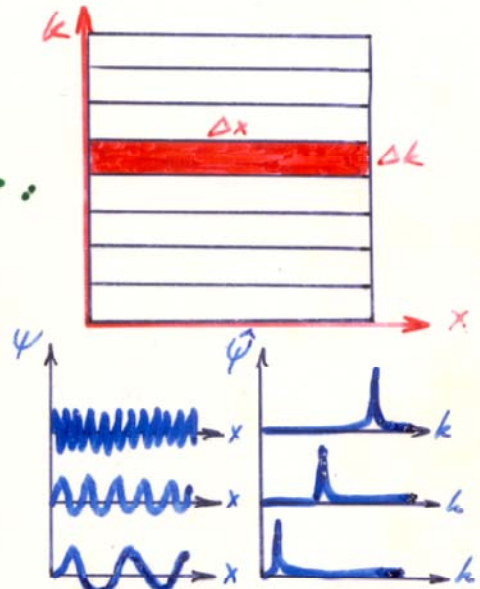
$$f(x) = \frac{1}{c} \int T_f(k) \psi_x(k) d\mu(k)$$

$$\Delta x \cdot \Delta k \geq C \text{cte}$$

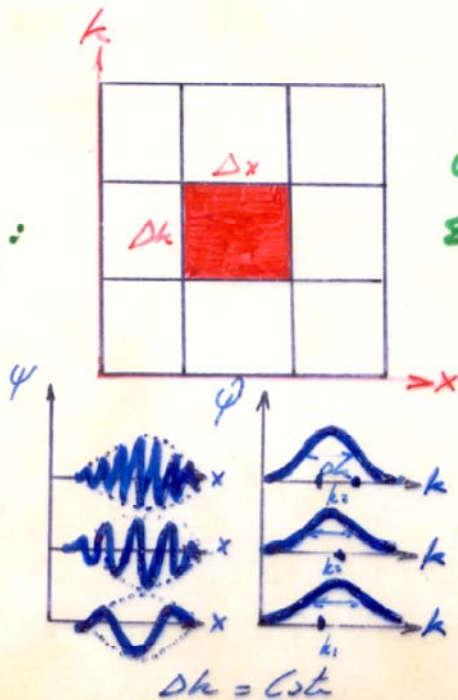
Shannon:



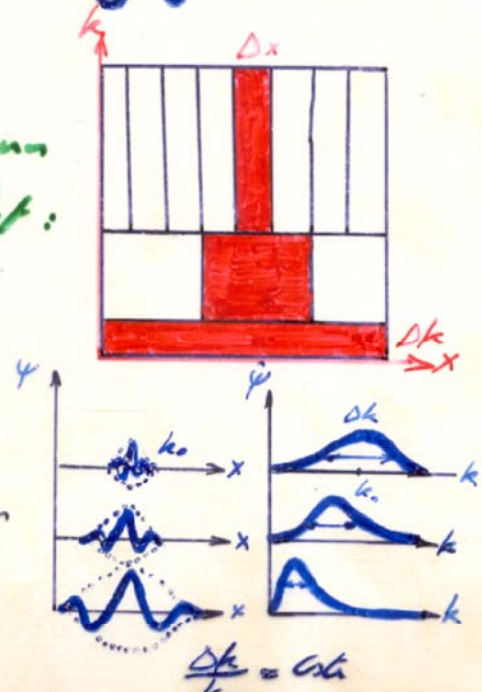
Fourier:
(1807)



Gabor:
(1946)



Grossmann
& Morlet:
(1981)



Balian's
destruction
(1981)

CONTINUOUS WAVELET TRANSFORM

Choice of the 'mother wavelet'

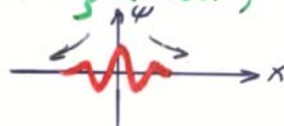
Admissibility condition: $\psi \in \{L^1 \cap L^2\}$

$$C = \int |\hat{\psi}(k)|^2 \frac{dk}{|k|} < \infty \Rightarrow \hat{\psi}(0) = 0 \text{ i.e. } \int \psi(x) dx = 0$$

And, if possible: energy reproducing kernel

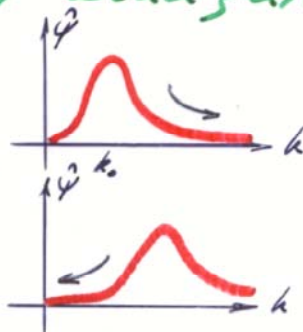
Good space (time) localization,

$$|\psi(x)| < \frac{1}{1+|x|^n}$$



Good scale (frequency) localization,

$$|\hat{\psi}(k)| < \frac{1}{1+|k-k_0|^n}$$



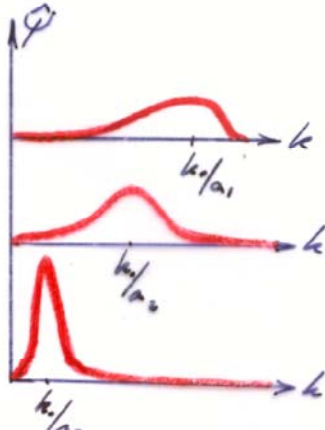
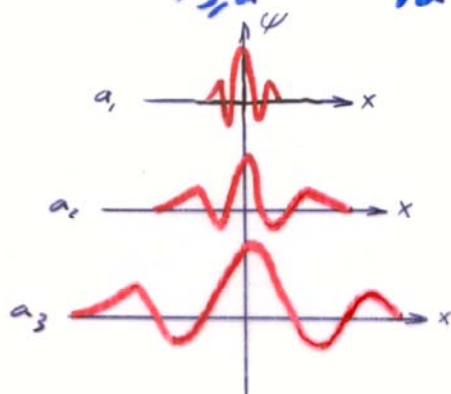
Zero high order moments,

$$\int \psi(x) x^m dx = 0 \\ = \left. \left(\frac{d}{dk} \right)^m \hat{\psi}(k) \right|_{k=0}$$

Generation of the 'wavelet family'

Affine Group $\left\{ \begin{array}{l} \text{by translation (parameter } b) \\ \text{and dilatation (parameter } a): \end{array} \right.$

$$\psi_{b,a}(x) = \frac{1}{\sqrt{a}} \psi\left(\frac{x-b}{a}\right)$$

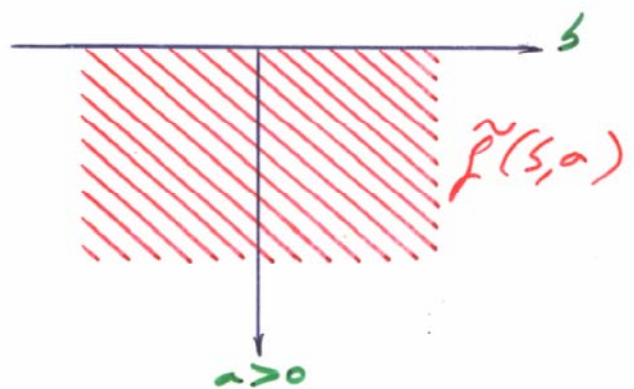


ANALYSIS/SYNTHESIS

Analysis: $\bar{\Psi}$ complex conjugate

$$\begin{aligned} \tilde{f}(b, a) &= \int_{a, b} \bar{\Psi}(x) f(x) dx \\ &= \frac{1}{\sqrt{a}} \int_{-\infty}^{+\infty} \bar{\Psi}\left(\frac{x-b}{a}\right) f(x) dx \\ &= \int_{-\infty}^{+\infty} \underbrace{\bar{\Psi}(ak)}_{\text{Filter with } \frac{\Delta k}{|k|} = Csk} e^{ibk} \hat{f}(k) dk \end{aligned}$$

The wavelet coefficients are defined on the open half-plane $(b, a) \in \mathbb{R} \times \mathbb{R}^+$



Synthesis:

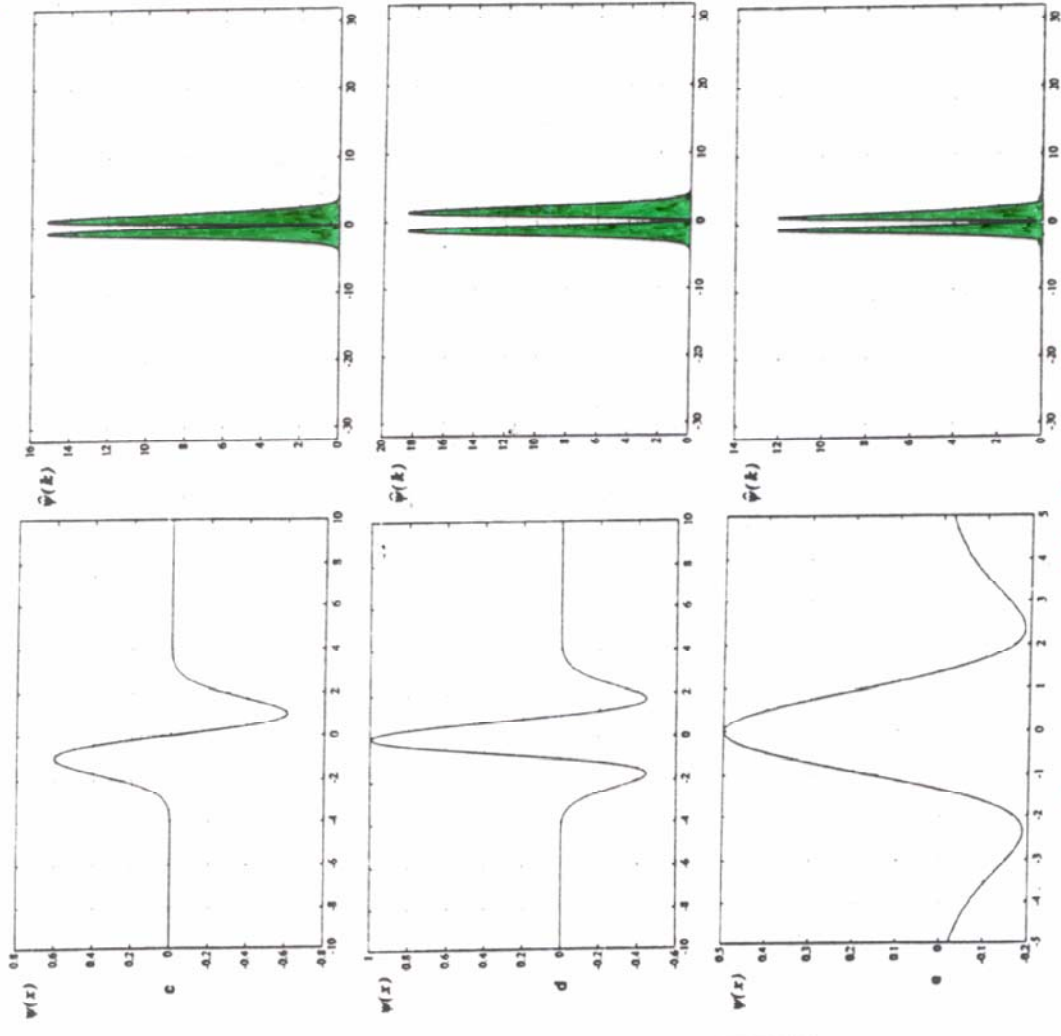
$$\begin{aligned} f(x) &= \frac{1}{C} \iint_{a, b} \Psi(x) \tilde{f}(b, a) \frac{da db}{a^2} \\ &= \frac{1}{C} \int_0^{+\infty} \frac{1}{\sqrt{a}} \int_{-\infty}^{+\infty} \Psi\left(\frac{x-b}{a}\right) \tilde{f}(b, a) \frac{da db}{a} \\ \text{avec } C &= 2\pi \int_{-\infty}^{+\infty} |\hat{\Psi}(k)|^2 \frac{dk}{|k|} \end{aligned}$$

Energy conservation (Parseval):

$$\int |f(x)|^2 dx = \frac{1}{C} \iint |\tilde{f}(b, a)|^2 \frac{da db}{a^2}$$

Haar Measure

Derivation of a Gaussian

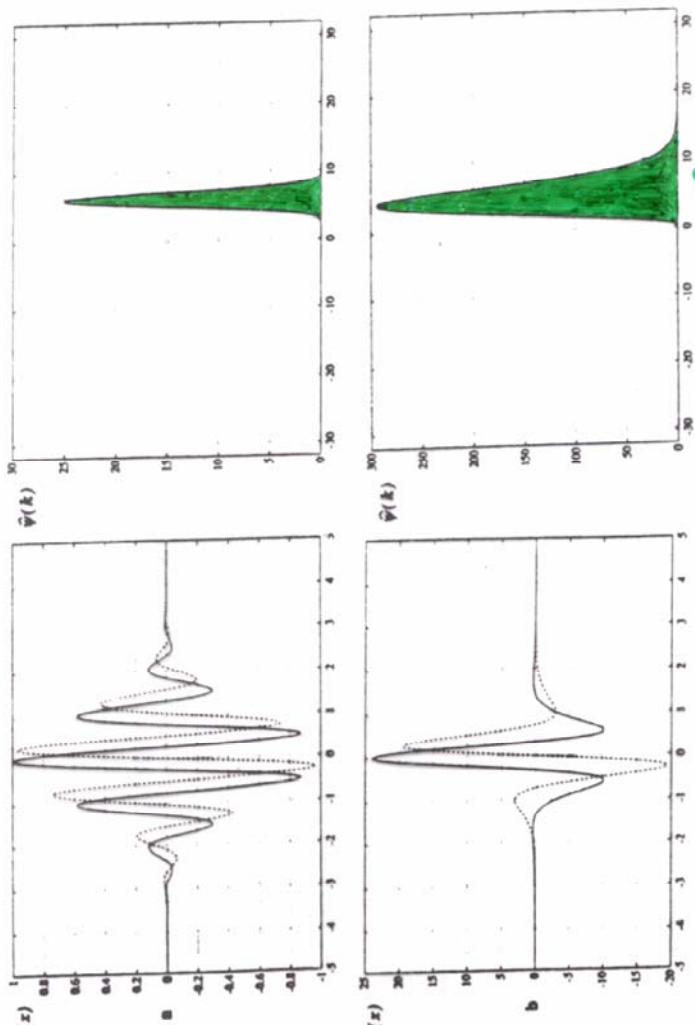


Difference of two Gaussians

ψ

Real-valued wavelets

Morlet

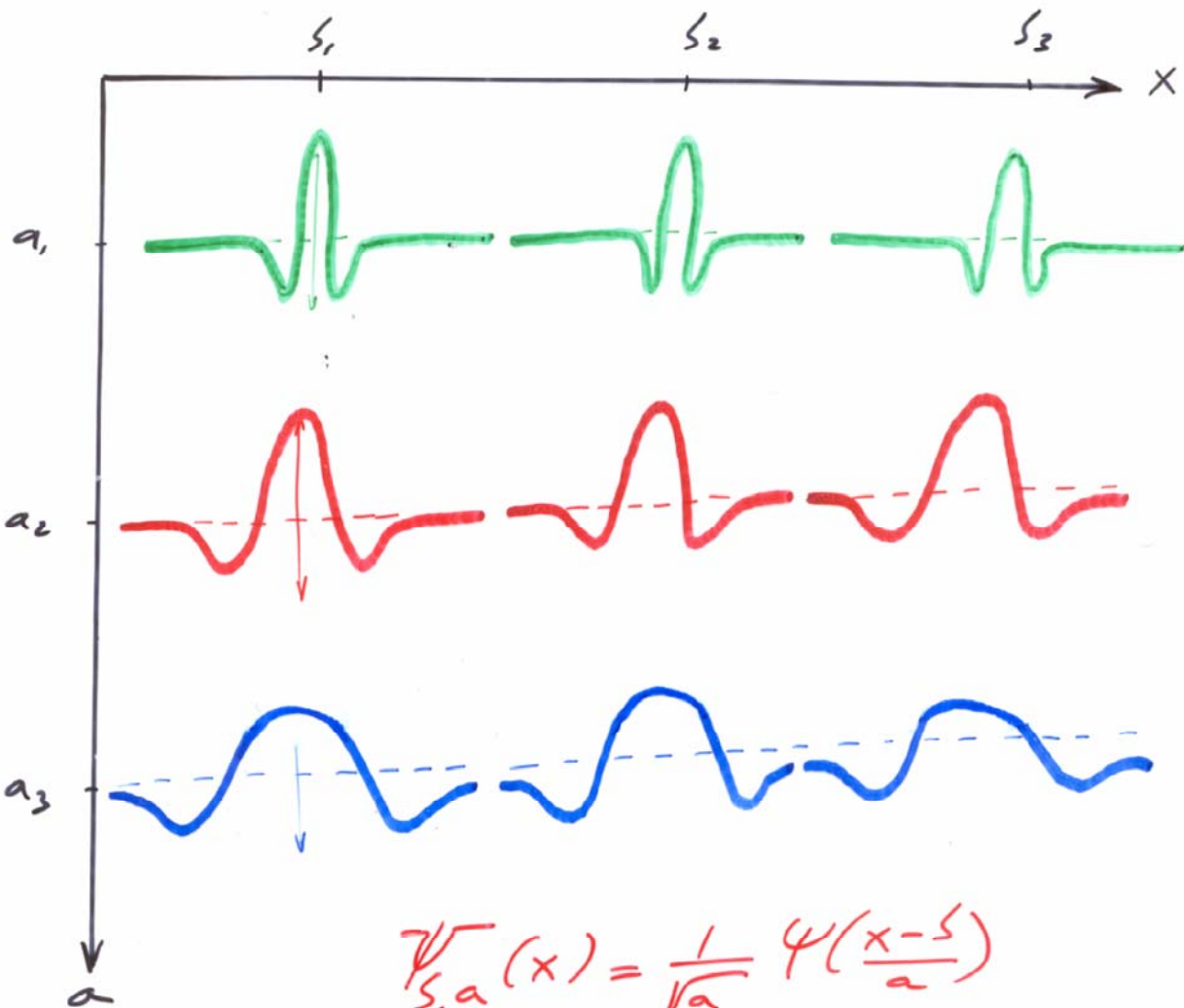


ψ

Pauli

Complex-valued wavelets

WAVELET FAMILY IN L^2 -NORM

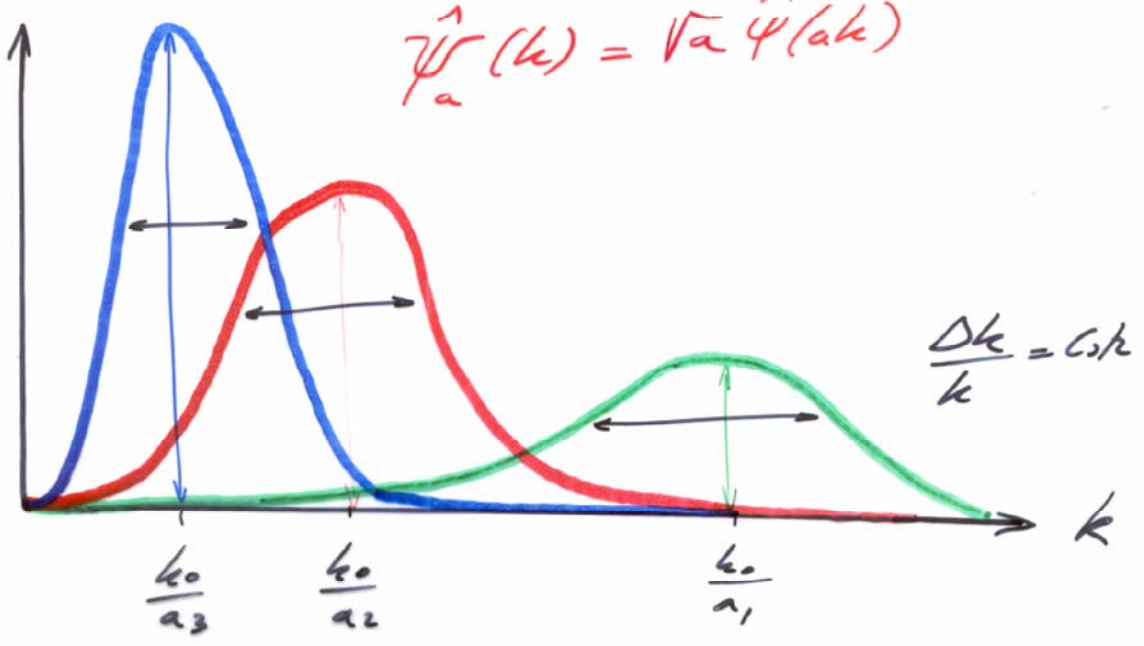


$\psi_{s_1, a}$

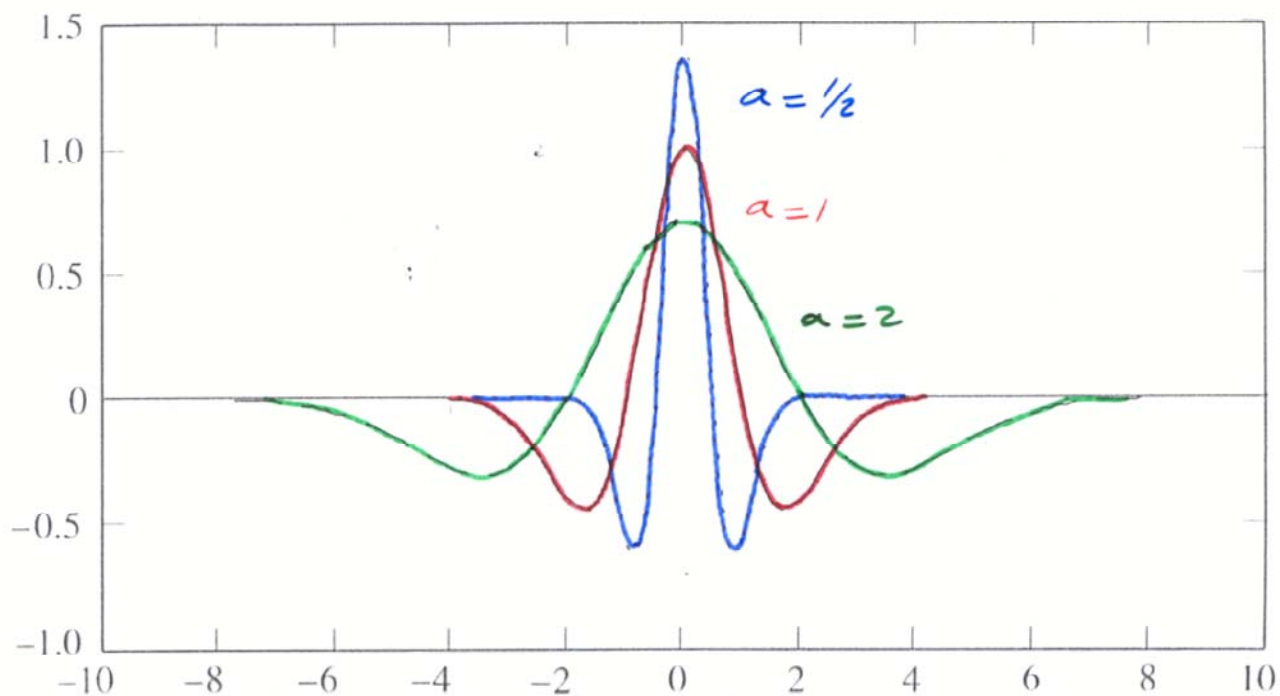
$$\psi_{s_1, a}(x) = \frac{1}{\sqrt{a}} \psi\left(\frac{x-s_1}{a}\right)$$

$$\hat{\psi}_a(k) = \sqrt{a} \hat{\psi}(ak)$$

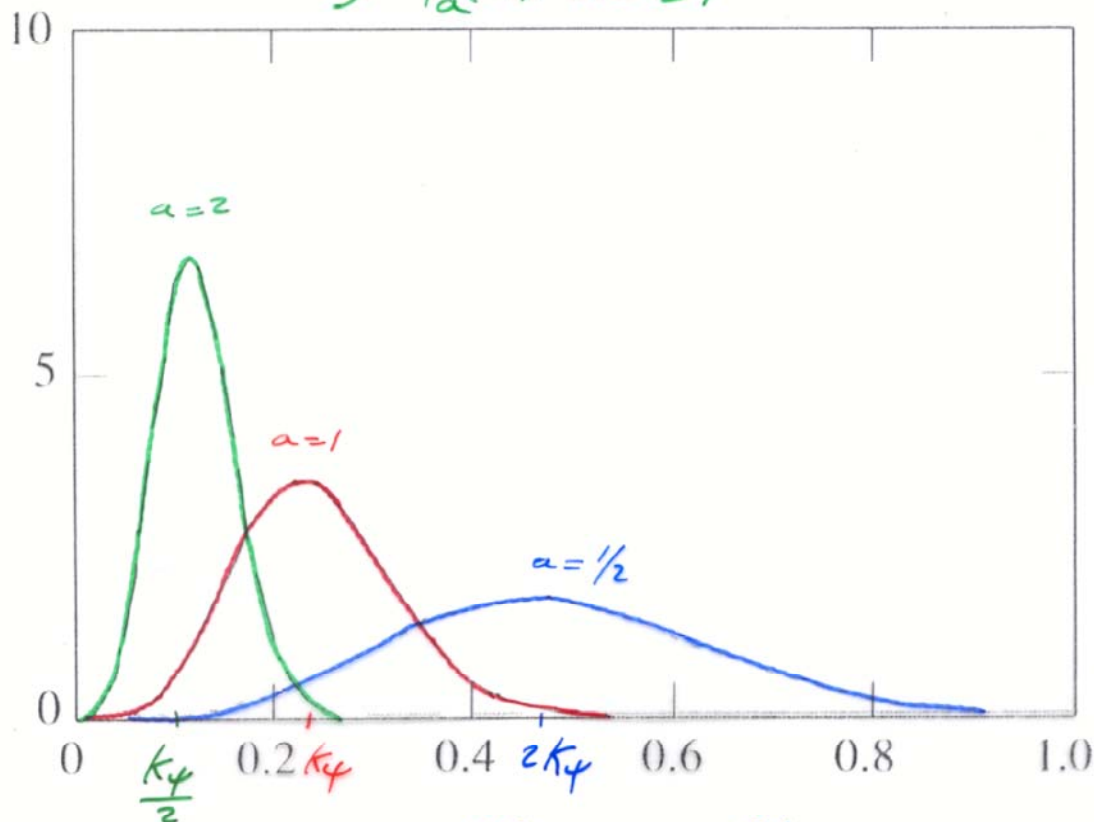
$\hat{\psi}_a$



WAVELET FAMILY WITH L^2 -NORM

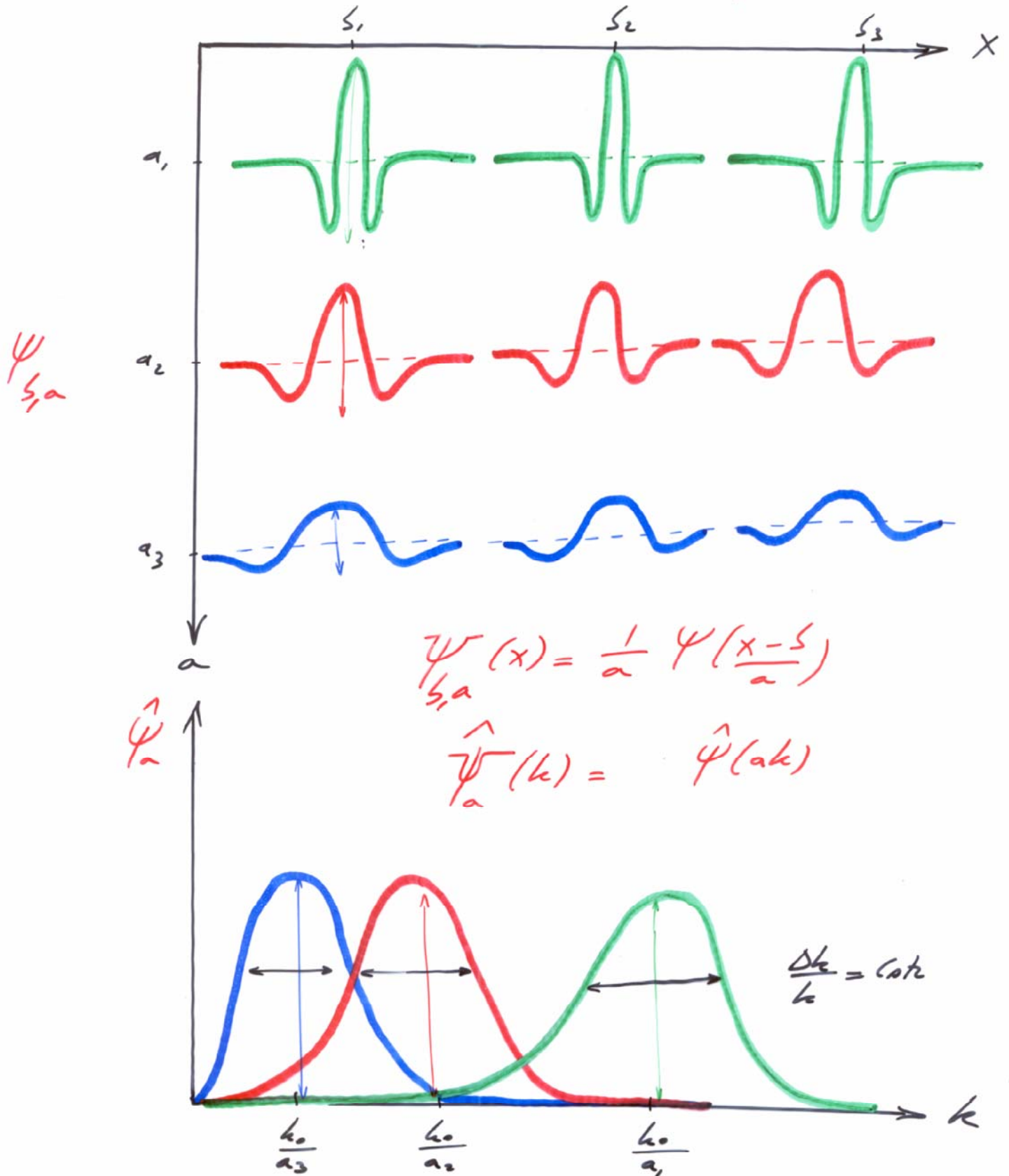


$$\int \psi_a^2(t) dt = 1$$

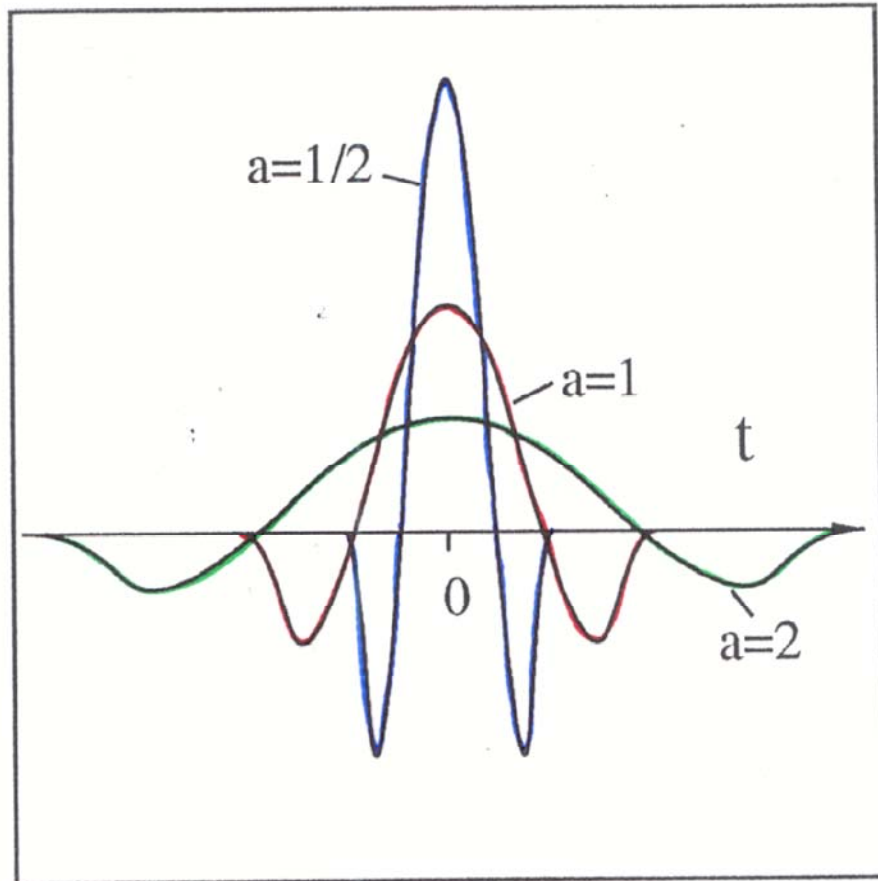


$$\Rightarrow \tilde{f}_{L^2} \approx \sqrt{a} \tilde{f}_{L^1}$$

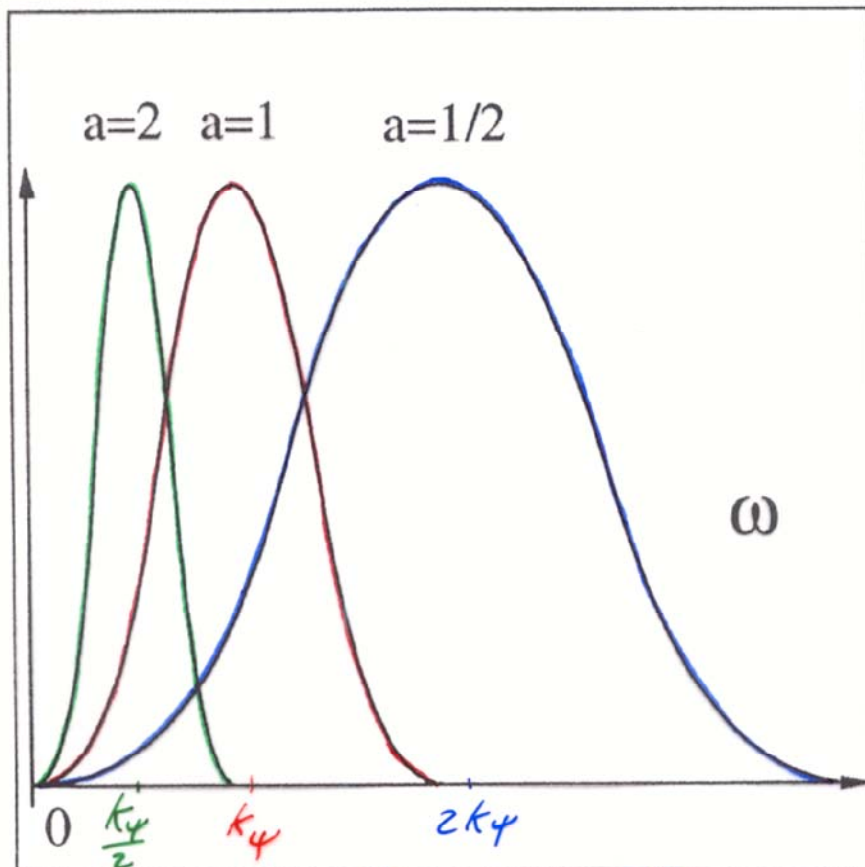
WAVELET FAMILY IN L^1 -NORM



WAVELET FAMILY WITH L1-NORM



$$\int |\psi_a(t)| dt = 1$$



$$\Rightarrow \tilde{f}_1 = \frac{1}{\sqrt{a}} \tilde{f}_2$$

Real part

Parameters:

Sampling frequency: 44.1 kHz

Number of voice per octave = 1

Highest voice: index = 1 frequency = 6000 Hz

Lowest voice: index = 7 frequency = 93.7 Hz

Time window = 18 ms

Signal de parole (1983)
Jean Perlot

Analysis:

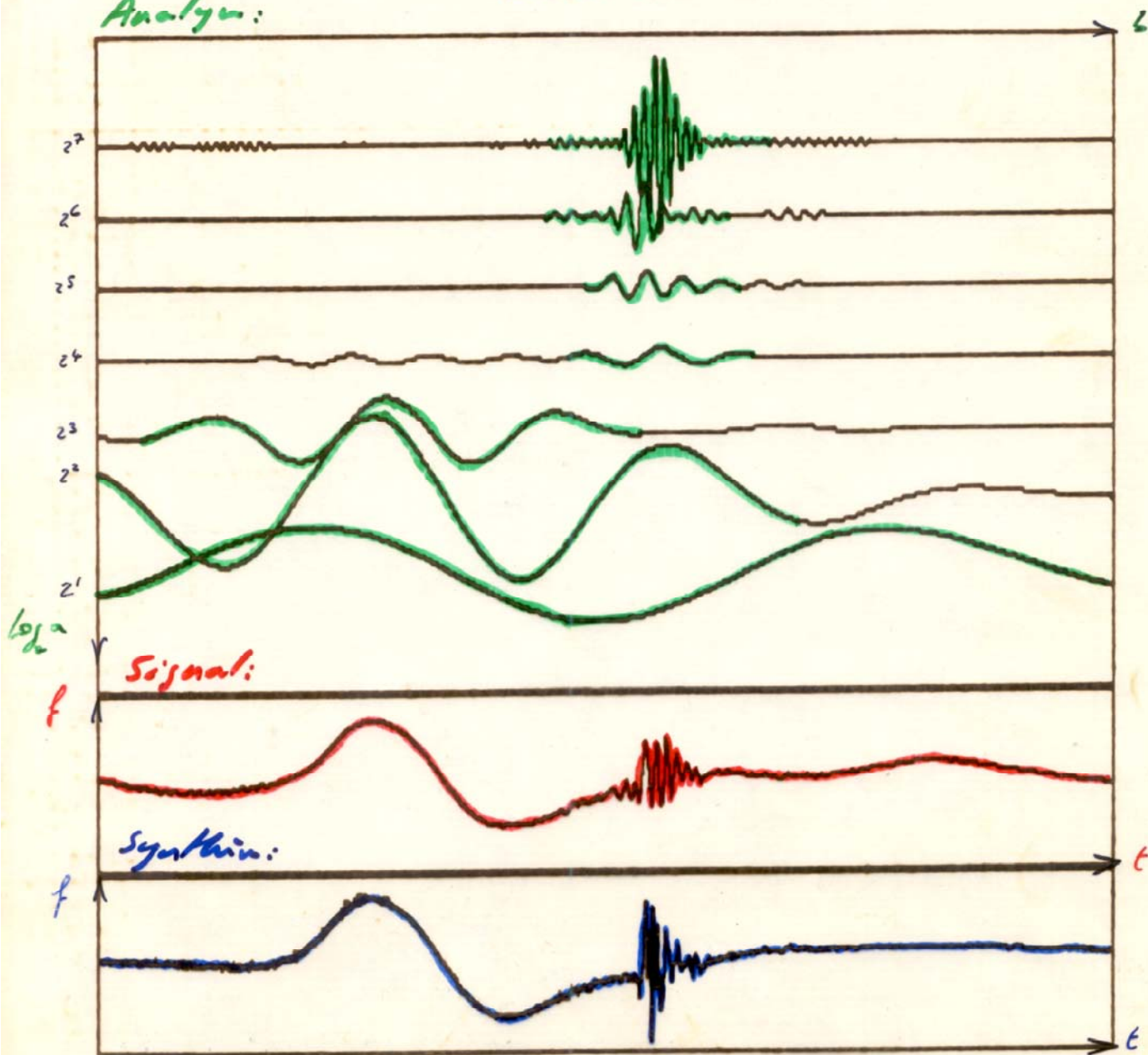


Figure 5

TWO WAVELET FORMULA

We can use different wavelets for analysis Ψ_A and synthesis Ψ_S .

Admissibility condition:

$$C_\Psi = \int_0^\infty \overline{\hat{\Psi}_A(k)} \hat{\Psi}_S(k) \frac{dk}{k} = \int_0^\infty \overline{\hat{\Psi}_A(-k)} \hat{\Psi}_S(-k) \frac{dk}{k} < \infty$$

Then isometry:

$$f(x) = \frac{1}{C_\Psi} \int_0^\infty \int_{-\infty}^{\infty} \tilde{f}(s, a) \Psi_S(s, a) \frac{da ds}{a^2}$$

$$\text{with } \tilde{f}(s, a) = \frac{1}{\sqrt{a}} \int f(x) \overline{\hat{\Psi}_A(s, a)} dx$$

Morlet's reconstruction formula:

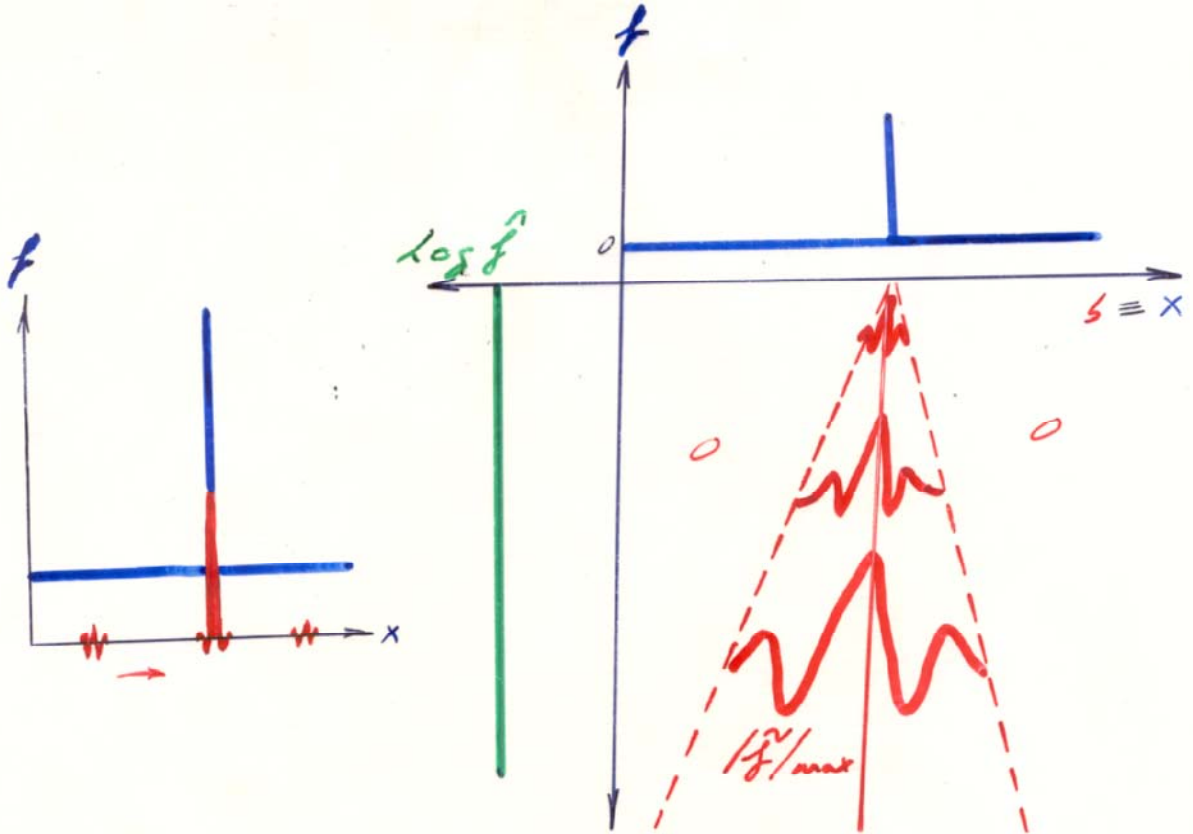
$$\Psi_S(x) = \delta(x)$$

$$\text{then } f(x) = \frac{1}{C_\Psi} \int_0^\infty \frac{1}{\sqrt{a}} \tilde{f}(s, a) \frac{da}{a}$$

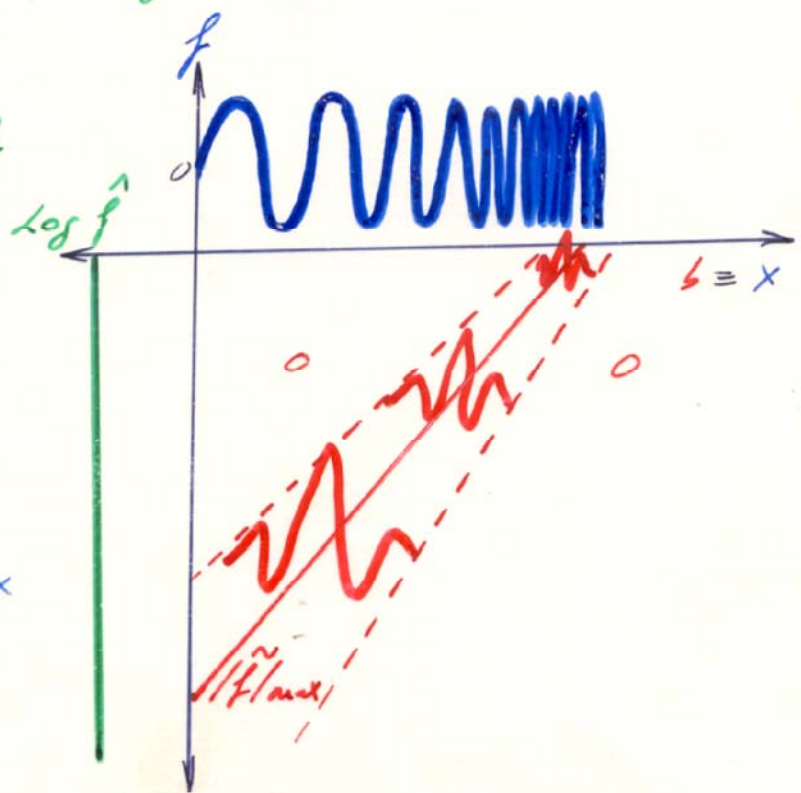
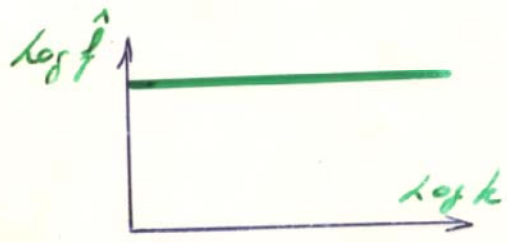
$$\text{with } C_\Psi = \int_0^\infty \overline{\hat{\Psi}_A(k)} \frac{dk}{k} = \int_0^\infty \overline{\hat{\Psi}_A(-k)} \frac{dk}{k} < \infty$$

We need only one integration to reconstruct.

T_e O : EXAMPLES



$$\log \hat{f}(x) \approx \log a$$



$$\log \hat{f}(x) \approx \log a$$

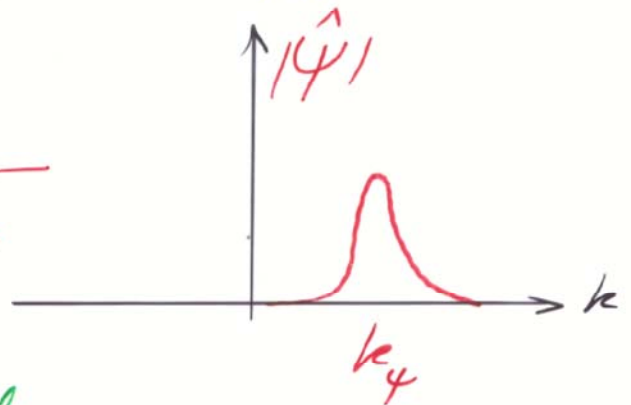
WAVELET TRANSFORM OF HARMONIC SIGNALS

Progressive wavelet $\psi(x) \in \mathcal{H}$
Hardy space
 $\Leftrightarrow \psi(x) \in \mathcal{C} / \hat{\psi}(k \leq 0) = 0,$

i.e. $\mathcal{R}(\psi) \xrightarrow{H} \mathcal{J}(\psi)$
H Hilbert transform,

with maximum of $\psi(x)$ at

$$k_\psi = \frac{\int_{-\infty}^{+\infty} k^2 \psi^2(k) dk}{\int_{-\infty}^{+\infty} \psi^2(k) dk}$$



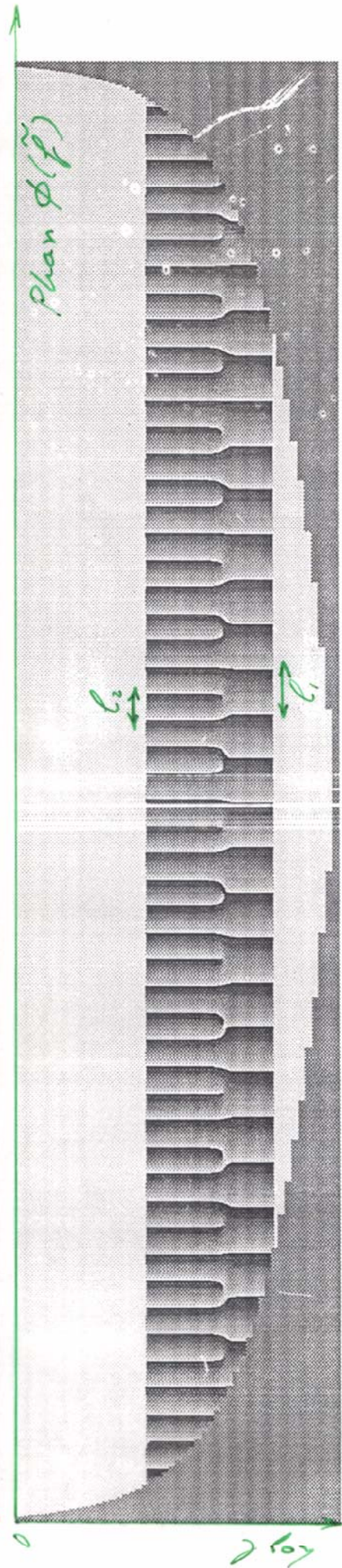
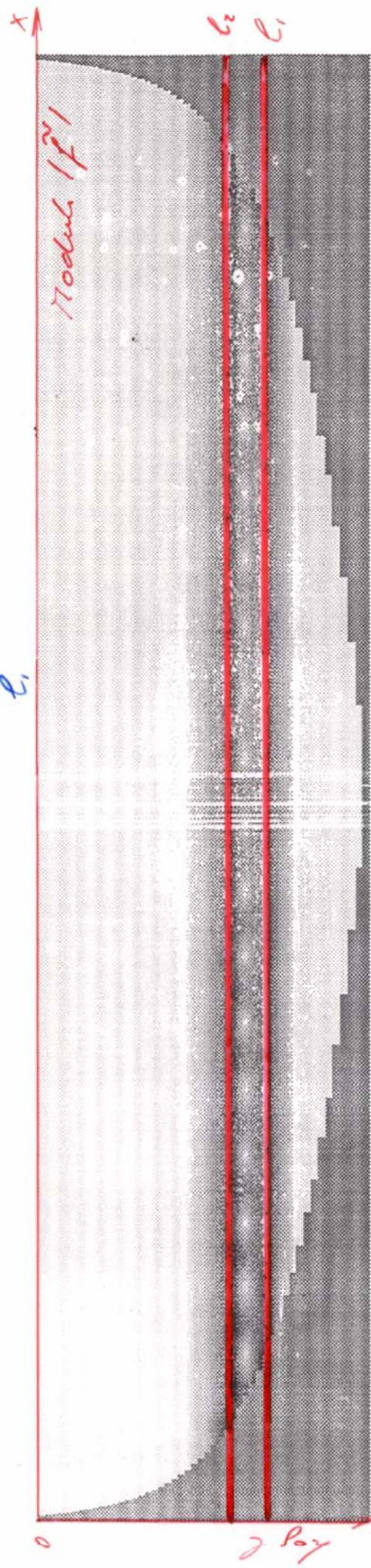
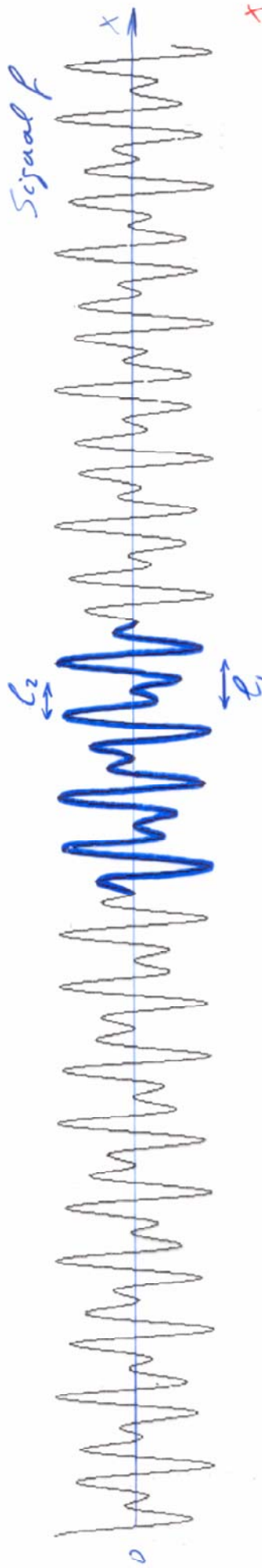
Harmonic signal

$$f(x) = \cos k_0 x = \frac{e^{ik_0 x} + e^{-ik_0 x}}{2}$$

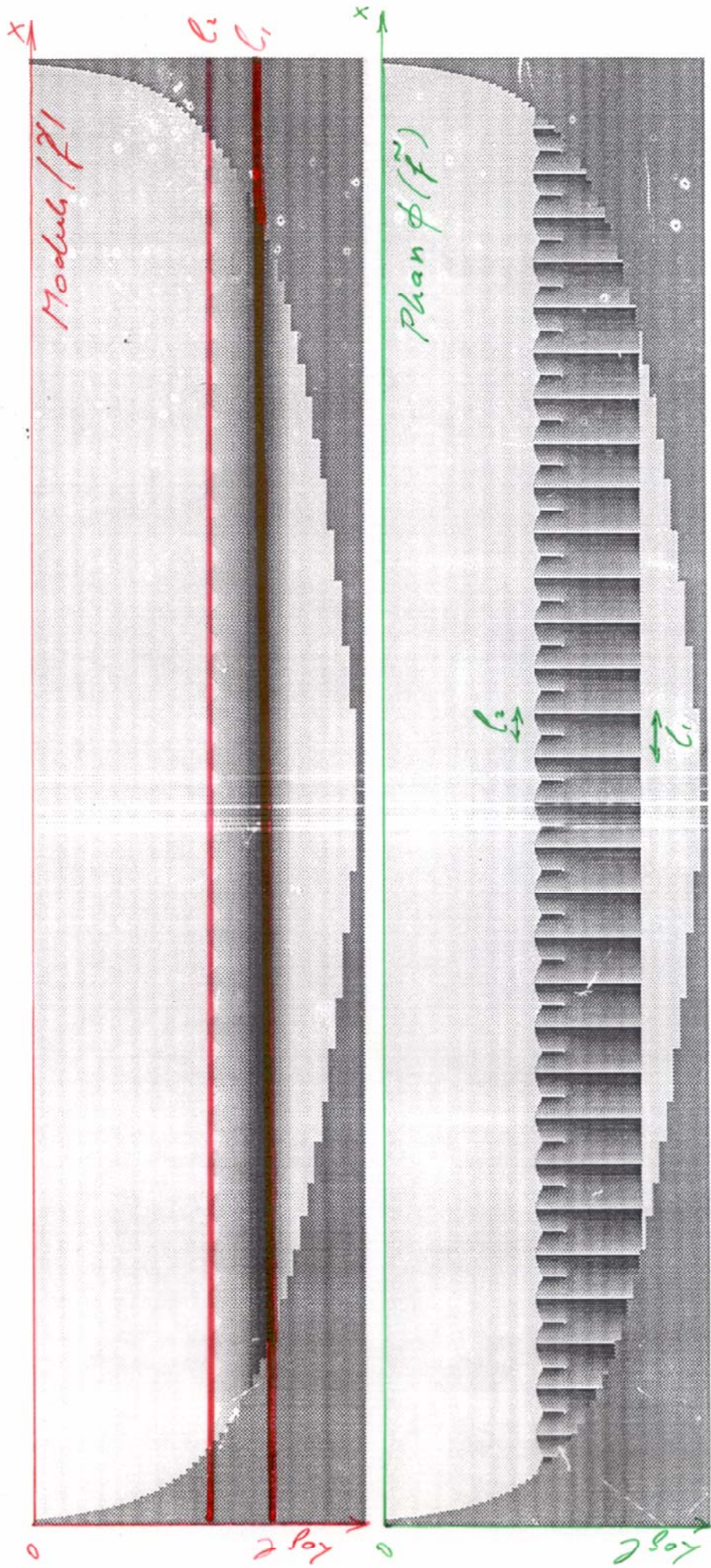
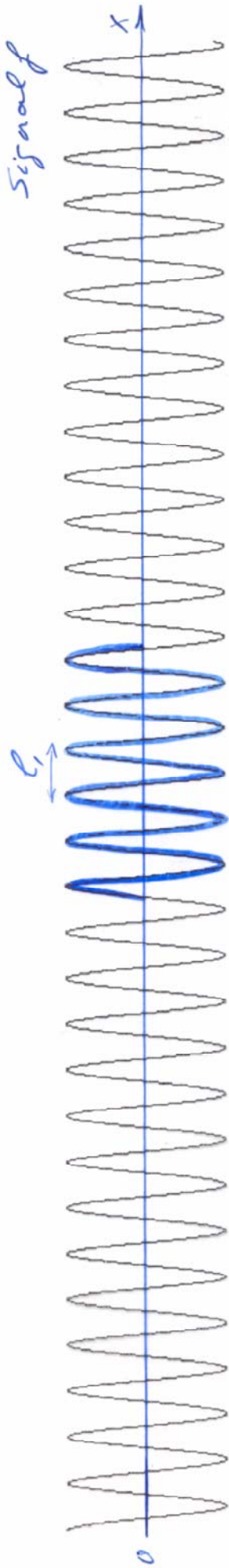
$$\tilde{f}(s, a) = \sqrt{a} e^{i s k_0 \frac{\pi}{a}} \hat{\psi}(a \nu_0)$$

- Modulus behaves as $\hat{\psi}(a \nu_0)$
which is maximal for $\hat{\psi}(\nu_\psi) \Rightarrow a = \frac{k_\psi}{k_0}$
- Phase varies linearly with s
and therefore unfolds the
signal phase in space $\Rightarrow \frac{\partial \varphi}{\partial s} = \frac{k_0}{k_\psi}$

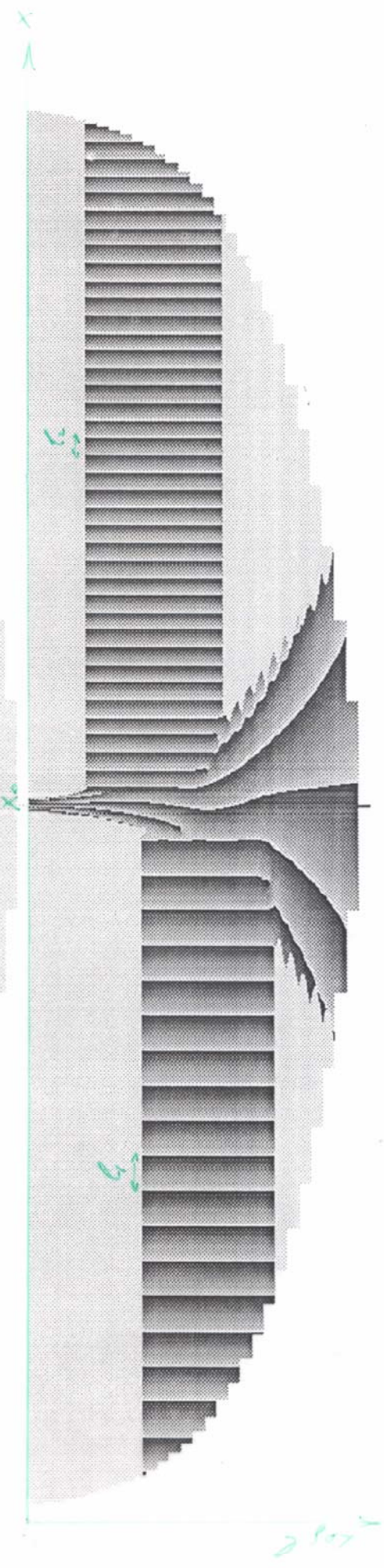
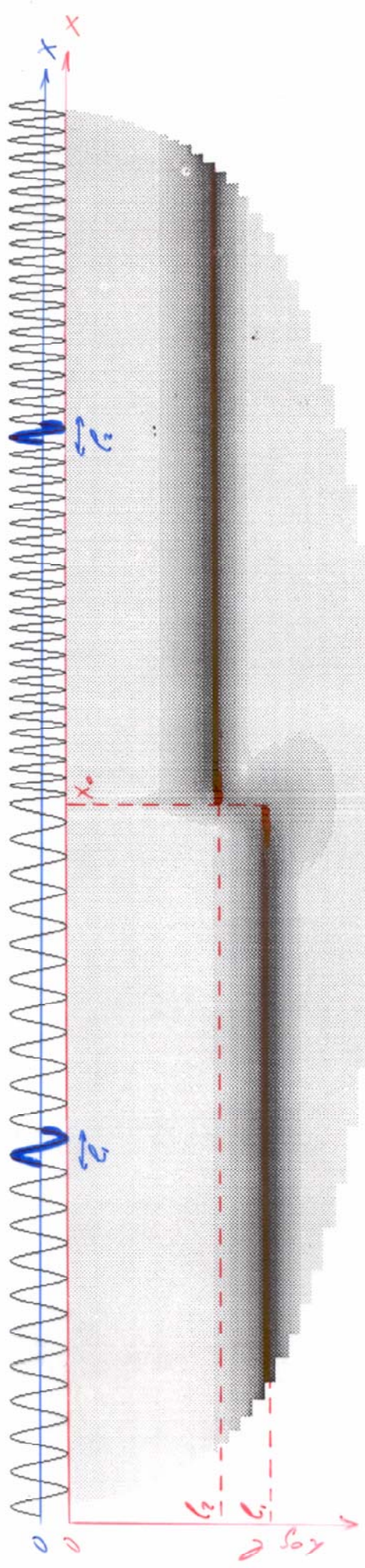
$$\cos(t) + \cos(1.68t)$$



$$\cos(t) + 0.02\cos(2t)$$

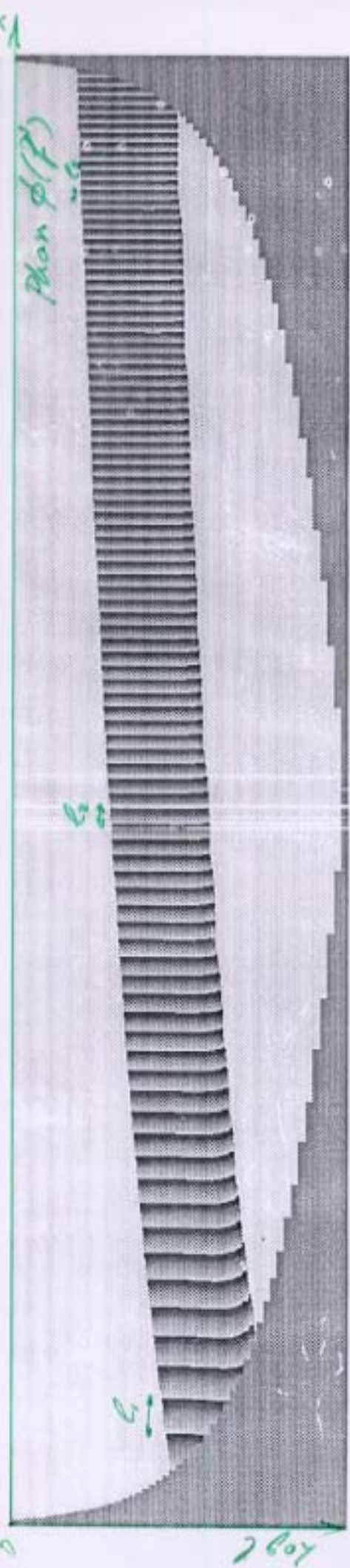
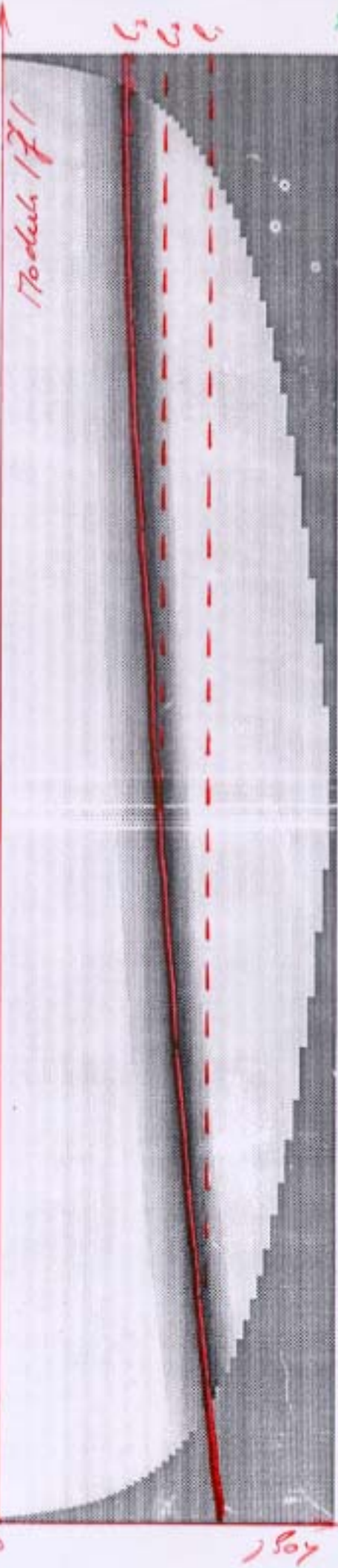
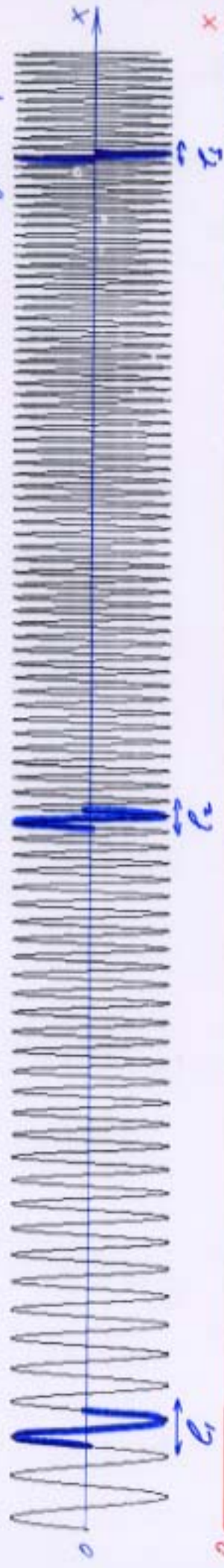


SIN(t) si t < 512 SIN(2t) sinon



$\sin(\xi)$

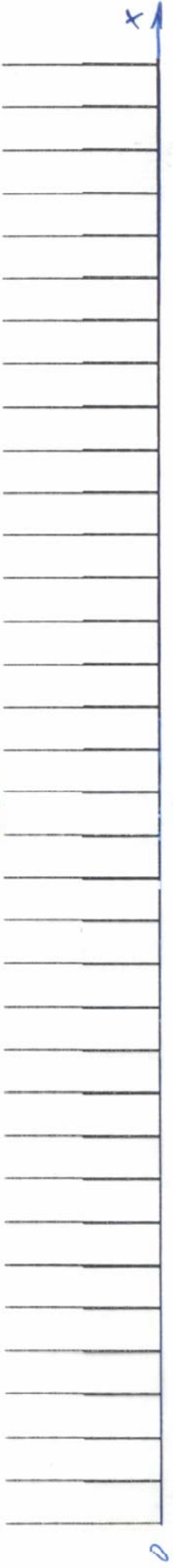
Signal f



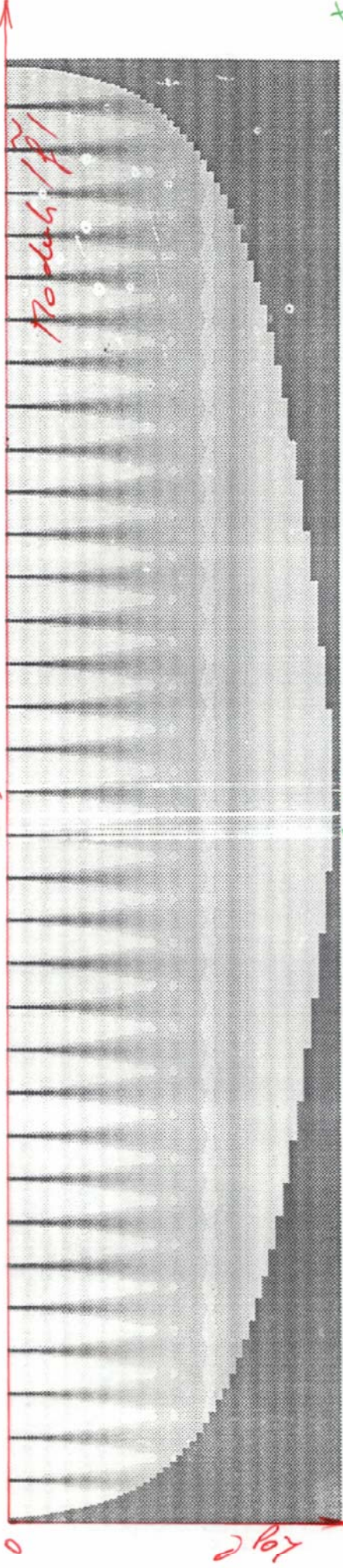
DELTA

Signal f

Δ

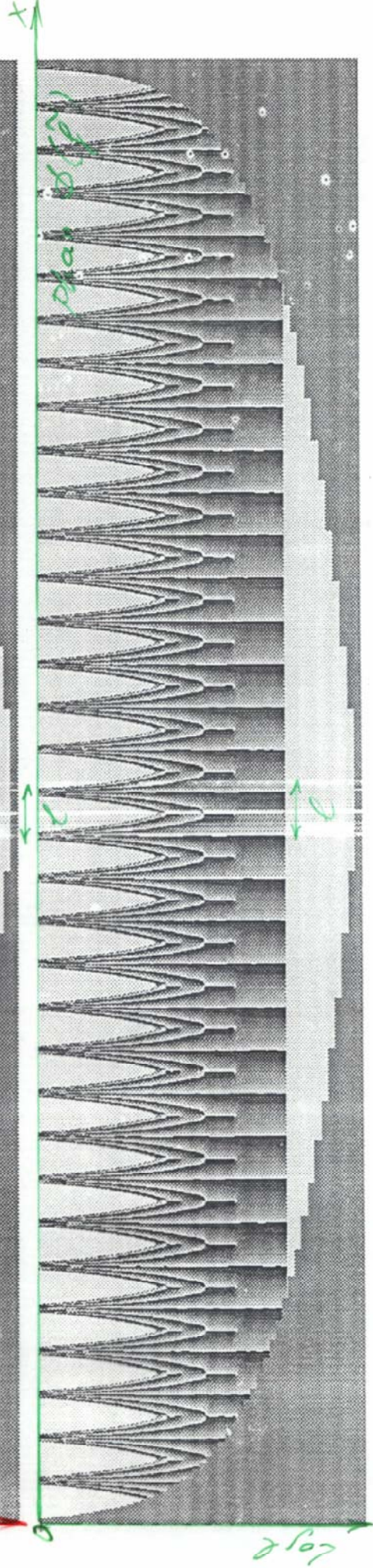


Δ



Noch Δ

f

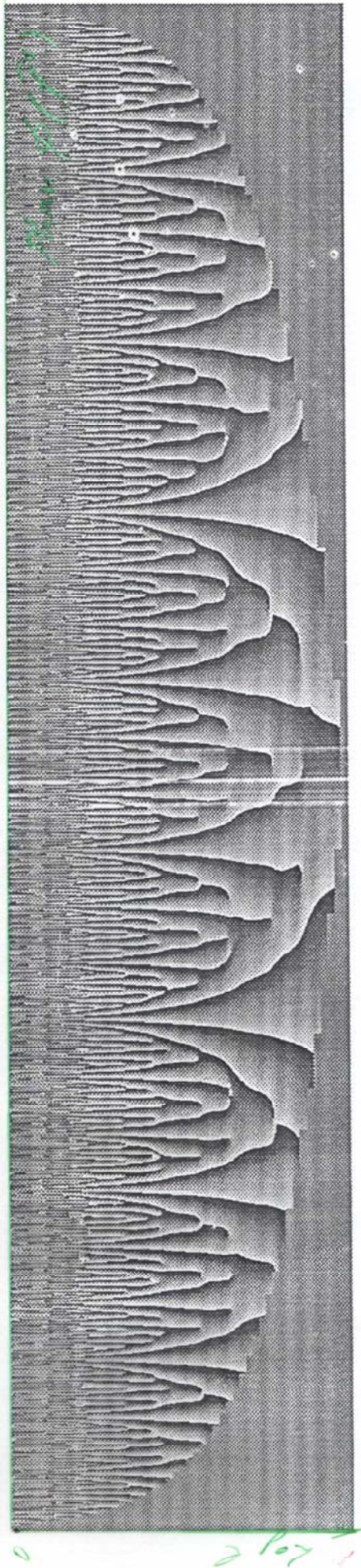


Noch Δ

f

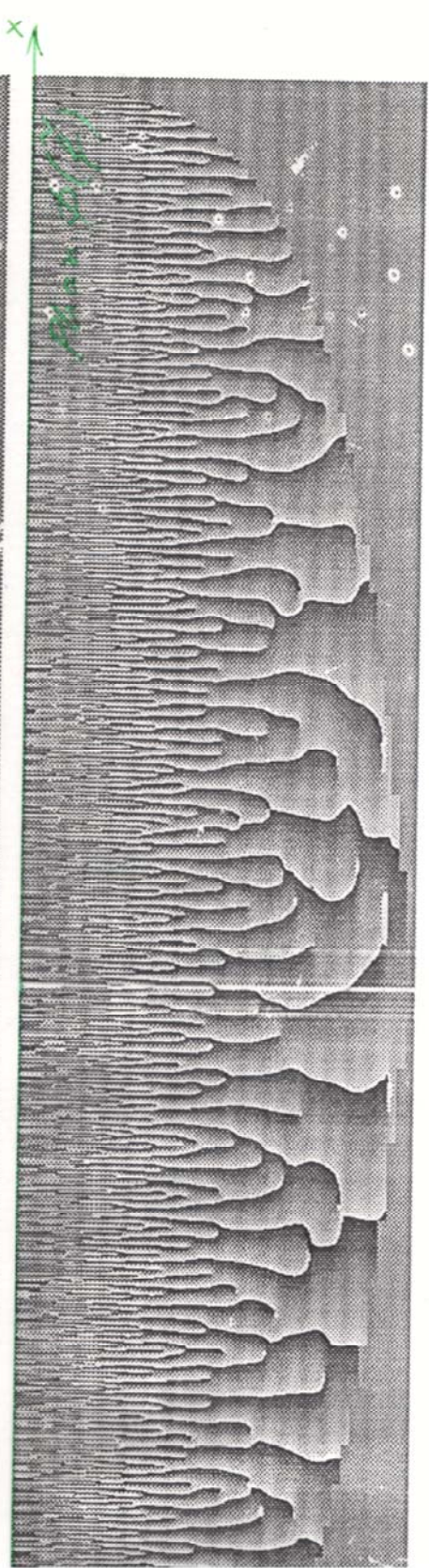
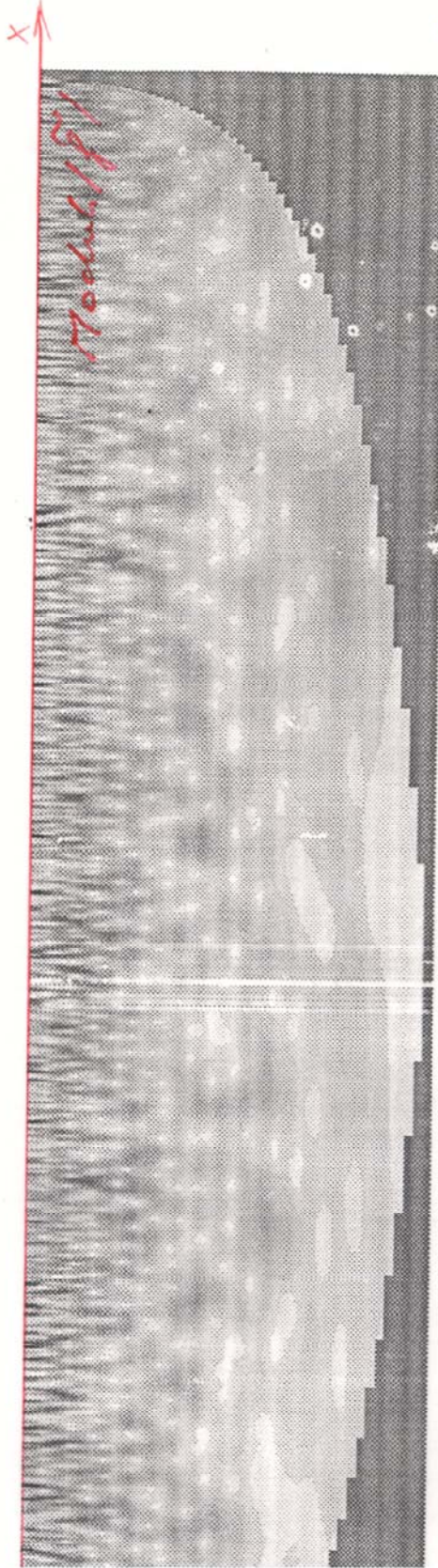
SAUTS

Signal f



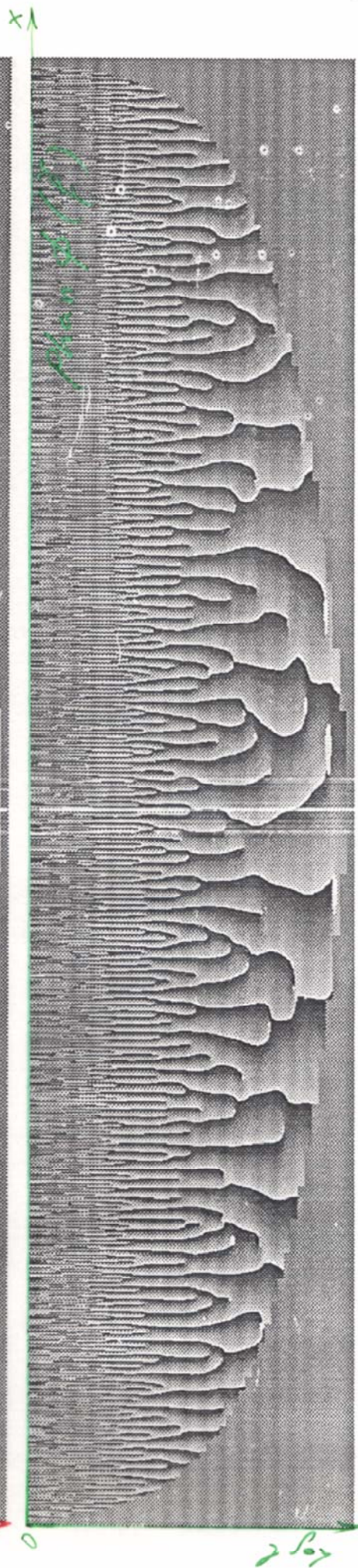
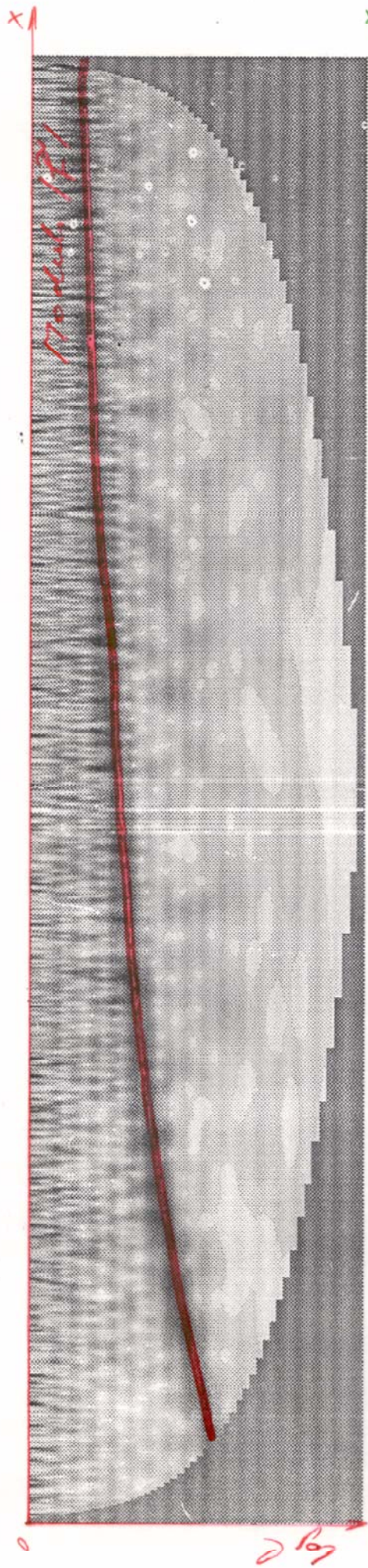
RANDOM

Signal f



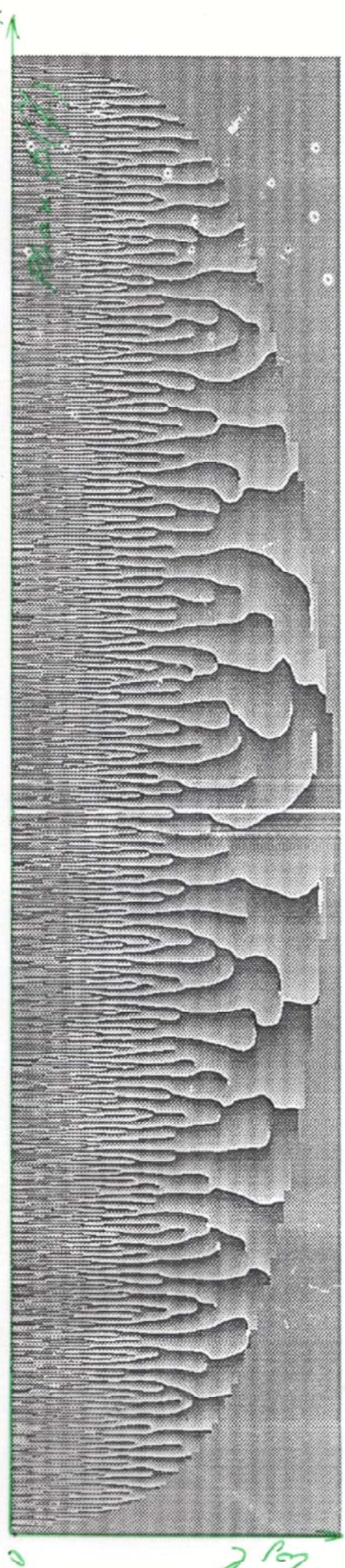
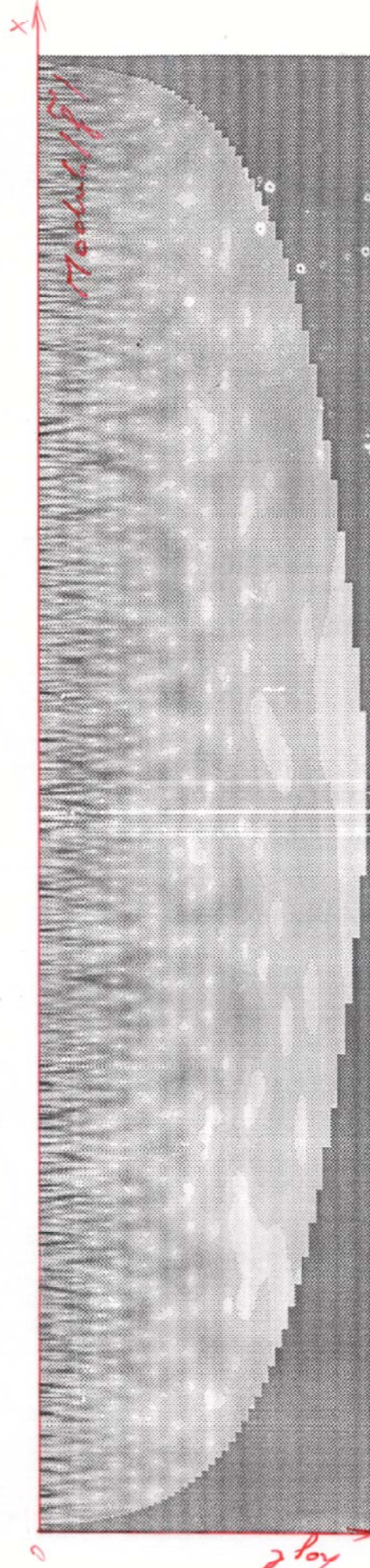
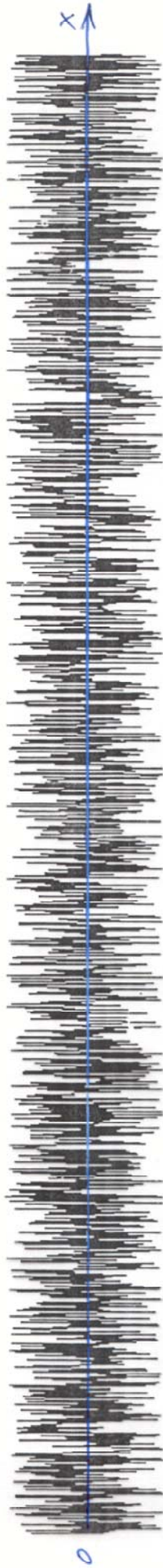
$\sinus(t^2) + \text{random}$

Signal f



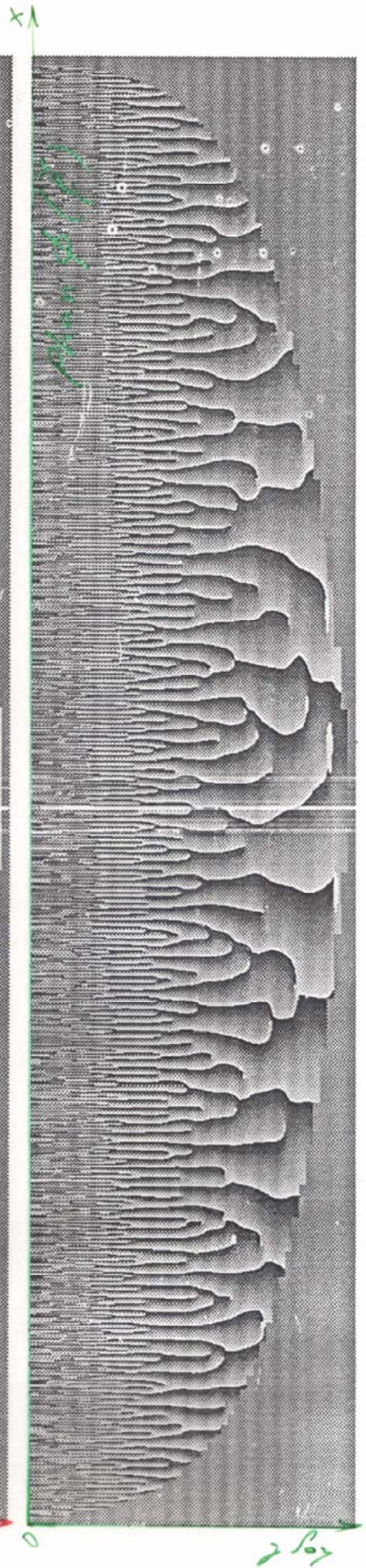
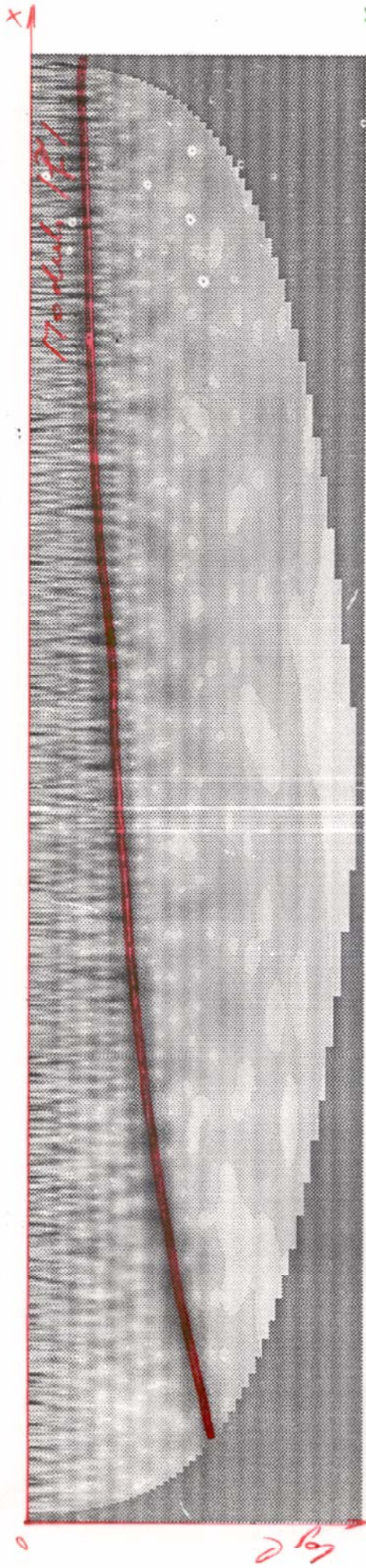
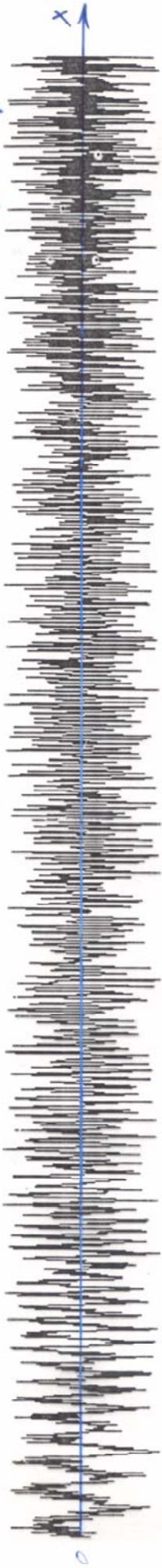
RANDOM

Signal f



$\sinus(t^2) + \text{random}$

Signal f



LINEARITY

\mathcal{W} is the continuous wavelet transform operator

$$\mathcal{W}[f(x) + g(x)] = \tilde{f}(\xi, a) + \tilde{g}(\xi, a)$$

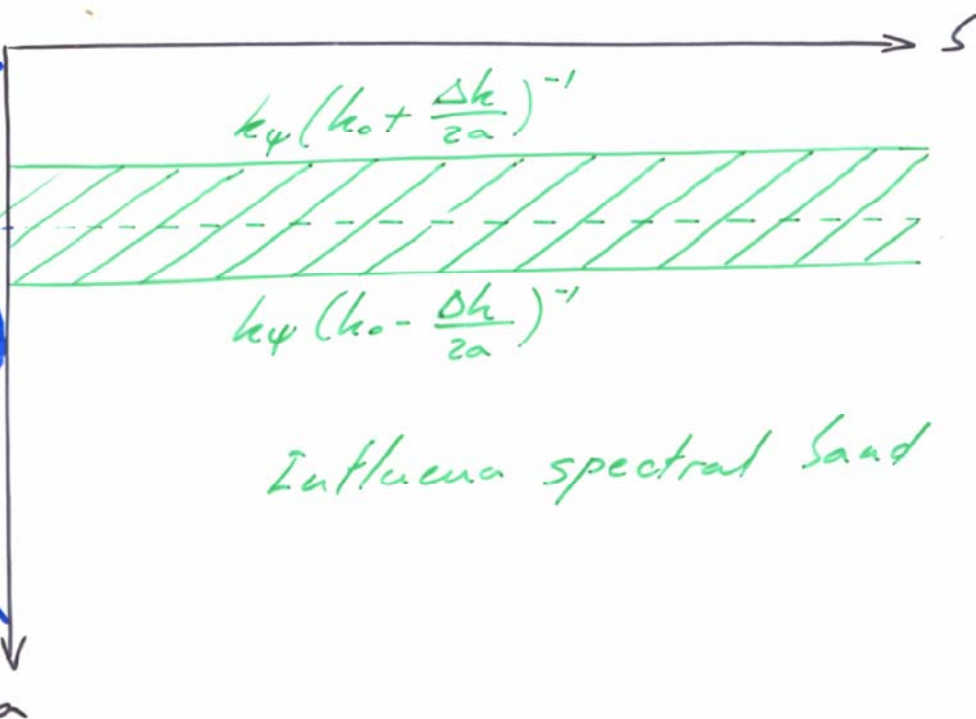
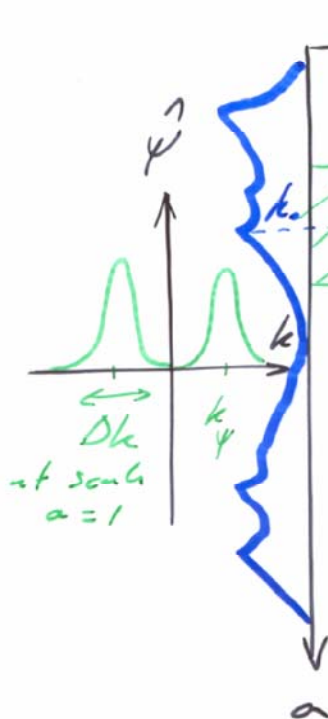
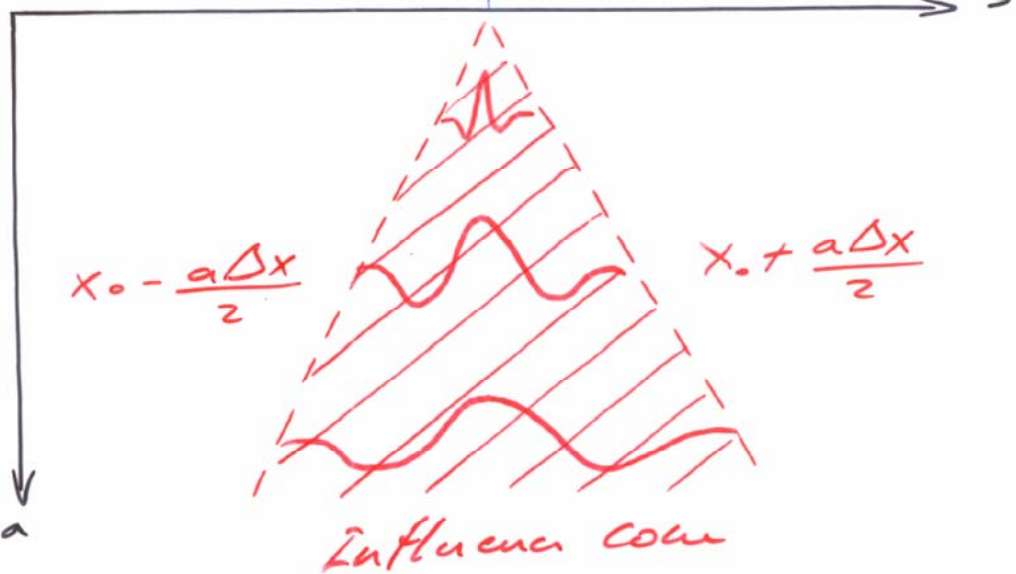
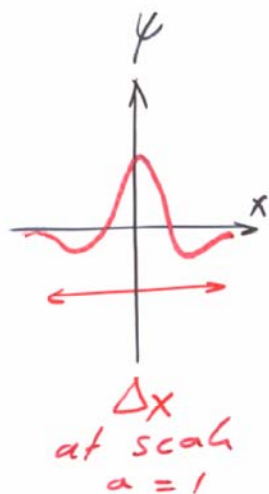
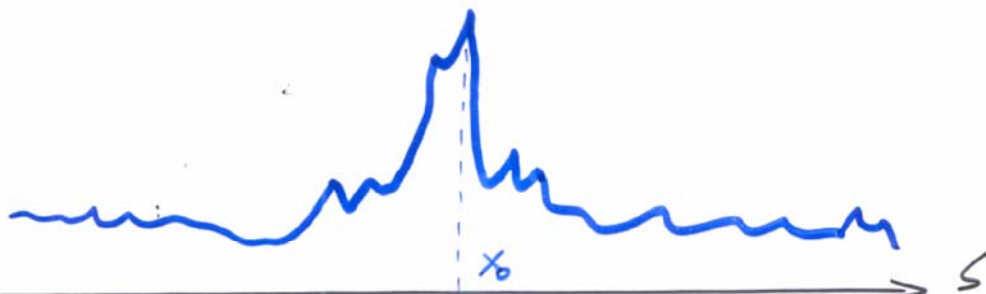
$$\Rightarrow \mathcal{W}[\vec{v}] = \vec{\tilde{v}} \begin{array}{l} \tilde{f}(x) \\ \tilde{f}(y) \\ \tilde{f}(z) \end{array}$$

with $\vec{v} \begin{array}{l} f(x) \\ f(y) \\ f(z) \end{array}$

Differentiation

$$\mathcal{W}\left[\frac{\partial^n f}{\partial x^n}\right] = (-i)^n \int_{-\infty}^{+\infty} f(x) \frac{\partial^n}{\partial x^n} \left[\frac{\psi}{\xi a} (x) \right] dx$$

CONSERVATION OF THE SPACE-SCALE LOCALITY



COVARIANCE BY TRANSLATION

$$\mathcal{W}[f(x-x_0)] = \tilde{f}(a, \omega - \omega_0)$$

This implies that differentiation commutes with \mathcal{W} :

$$\frac{\partial}{\partial x} \mathcal{W}(f) = \mathcal{W}\left(\frac{\partial f}{\partial x}\right)$$

$$\nabla[\mathcal{W}(f)] = \mathcal{W}(\nabla f)$$

$$\nabla \cdot [\mathcal{W}(f)] = \mathcal{W}(\nabla \cdot f)$$

A consequence of the covariance by translation is that the frequency of an harmonic signal can be read off from the phase of the wavelet coefficients. It corresponds to the number of zeros of the phase for $a = \text{const}$.

COVARIANCE BY DILATION

$$W[f(ax)] = \frac{1}{|a|} \tilde{f}(\frac{\omega}{a}, \omega)$$

This is not the same as for
the Fourier transform F :

$$F[f(ax)] = \frac{1}{|a|} \hat{f}(\frac{k}{a}) = \frac{1}{|a|} \hat{f}(\frac{2\pi}{\lambda a})$$

λ wavelength

A consequence of the dilation covariance is that the wavelet transform of a power-law function is fully determined by its restriction to any line $a = \text{const}$.

As a consequence, the lines of constant phase point out onto the singularities of the function f .

ENERGY CONSERVATION

The continuous wavelet transform
is an isometry between

$$L^2(\mathbb{R}) \quad \text{and} \quad H_{\psi} \subset L^2(\mathbb{R}^+ \times \mathbb{R})$$

therefore it conserves energy:

$$\int_{-\infty}^{+\infty} |f(x)|^2 dx = \frac{1}{C_{\psi}} \int_0^{+\infty} \int_{-\infty}^{+\infty} \tilde{f}(s, a) \overline{\tilde{f}(s, a)} \frac{da ds}{a^2}$$

$$\text{with } C_{\psi} = 2\pi \int_{-\infty}^{+\infty} \frac{|\hat{\psi}(k)|^2 dk}{|k|}$$

It conserves energy both globally
and locally for all coefficients
inside the influence cone.

The total energy can also be splitted
among contributions at different scales:

$$E(a) = \frac{1}{C_{\psi}} \int_{-\infty}^{+\infty} |\tilde{f}(s, a)|^2 \frac{ds}{a^2}$$

REPRODUCING KERNEL

The projection $L^2(\mathbb{R}^+ \times \mathbb{R}) \rightarrow H_\psi$
is an integral operator with
kernel:

$$k(s', a', s, a) = \langle \psi_{s', a'} / \psi_{s, a} \rangle$$

autocorrelation function of ψ .

Therefore $\tilde{f}(s, a) \in L^2(\mathbb{R}^+ \times \mathbb{R})$ is
the continuous wavelet transform
of a function f iff it satisfies
the reproducing kernel equation:

$$\tilde{f}(s', a') = \int_{0^+}^{+\infty} \int_{-\infty}^{+\infty} k(s', a', s, a) f(s, a) \frac{da ds}{a^2}$$

INSTANTANEOUS FREQUENCY

Frequency is a characteristic of a wave-like (harmonic) signal, which can be measured experimentally, to describe the rapidity of the wave oscillations.

If this wave has a variable amplitude we can still define a frequency iff the amplitude varies much slower than the oscillations, stationary phase hypothesis, namely:

$$f(t) = \underbrace{A(t)}_{\text{amplitude}} e^{i\varphi(t)} + \underbrace{u(t)}_{\text{noise}}$$

$$\text{with } \frac{\partial A}{\partial t} \text{ and } \frac{\partial \varphi}{\partial t} \ll |e^{i\varphi}| = 1$$

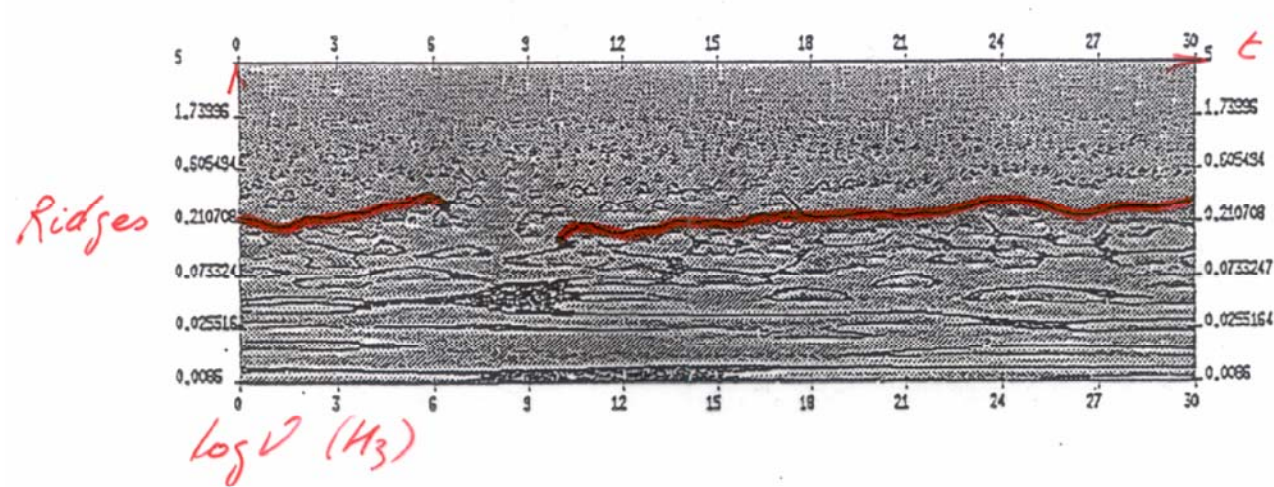
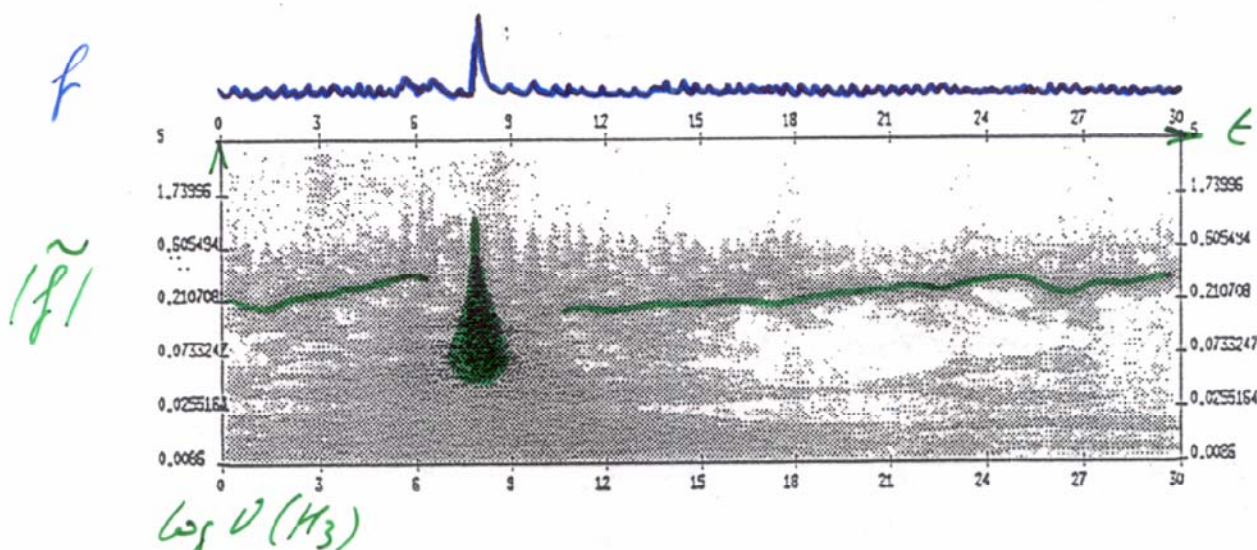
The instantaneous frequency is

$$\nu(t) = \frac{\nu_\varphi}{2\pi} \frac{\partial \varphi}{\partial t} \text{ which defines}$$

the ridge of the wavelet transform, curve in phase-space (β, a) where $\frac{\partial \varphi}{\partial t} = \omega t$

1) Figure 1: Calcul de transformées en ondelettes (signal respiratoire) **MEDICAL SIGNAL**

De haut en bas, le signal, le module, et la fréquence instantanée de la transformée en ondelettes (les lignes continues noires représentent les arêtes de la transformée) (On s'intéresse à l'évolution temporelle des fréquences instantanées).



© Science & Tec
 Ex. Virgo: $10 \log \frac{P_s}{P_B} = -40 \text{ dB}$ soit $\frac{P_s}{P_B} = 10^{-4}$
 Référence:

Carmoua, Hwang and Torresani
 Characterization of signals by the ridges
 of their wavelet transforms, 1995
 anonymous ftp from chelsea.math.uci.edu
 http://cvt.sxt.univ-mrs.fr

CHARACTERIZATION OF THE LOCAL SCALING OF A SIGNAL

The signal f can be a function,
a distribution or a measure,
that we want to study in x_0 .

$$\| \tilde{f}(b, a) \|_{L^1} \leq M_1 |b - x_0|^\alpha \| \psi \|_{L^1} + M_2 M_\alpha a^\alpha$$

with $M_\alpha = \int_{-\infty}^{+\infty} x^\alpha \psi(x) dx$

α is the degree of differentiability
of f at x_0 if $\alpha \geq 1$, i.e. f regular in x_0
or α is the Lipschitz exponent of
the singularity in x_0 if $-1 < \alpha < 1$.

If we want to eliminate the most
regular (polynomial) contribution of f
we have to choose ψ with cancellations
up to order m (then $M_m = 0$).

In this case $\| \tilde{f}(b, a) \|_{L^1}$ will only react
to regions where f is less smooth as order m .

CWT ALGORITHM (1)

1D periodic signal $f(x)$

1D periodic wavelet $\psi(x)$

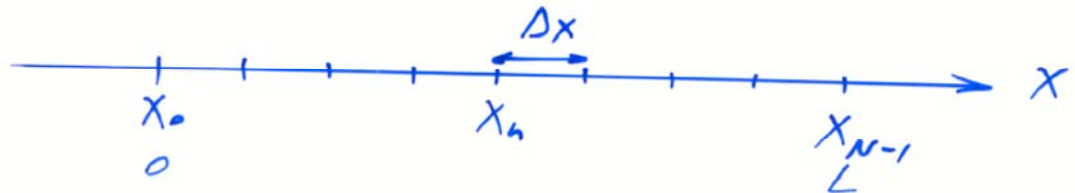
$x \in [0, L]$ with period L

Space discretization

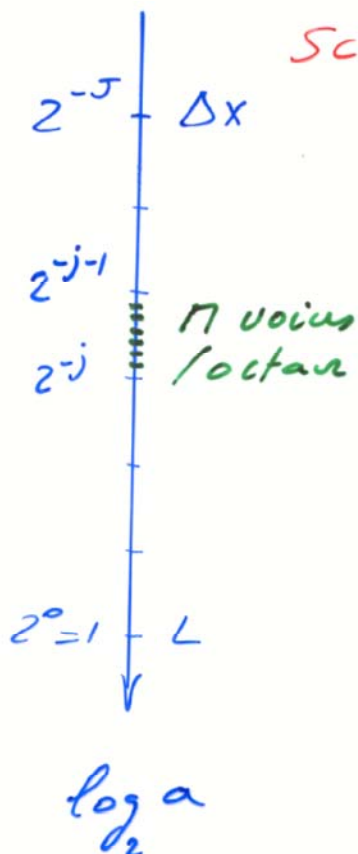
Sampling on a regular grid

$\Delta x = \frac{L}{N}$ with $N = 2^J$ samples

$\Rightarrow x_n = \frac{nL}{N}$ and $s_n = \frac{nL}{N}$



Scale discretization



Logarithmic sampling

$a_j = a_0^{-j}$ with $j \geq 0$

$a_0 = 2^{1/M}$ M number of voices/octave

$\Rightarrow M \cdot J$ samples for scale

Largest scale $L \rightarrow a_0 = 1$

Smallest scale $\Delta x \rightarrow a_J = 2^{-J}$

CWT ALGORITHM (2)

L2-norm wavelet family

$$\psi_{a,s}(x) = \frac{1}{\sqrt{a}} \psi\left(\frac{x-s}{a}\right)$$

$$\Rightarrow \hat{\psi}_{a,s}(k) = \sqrt{a} \hat{\psi}(ak) e^{-2\pi i s k}$$

Wavelet coefficients

$$\tilde{f}(a,s) = \int_{-\infty}^{+\infty} f(x) \overline{\psi_{a,s}(x)} dx$$

$$\stackrel{\text{Parseval}}{=} \int_{-\infty}^{+\infty} \hat{f}(k) \overline{\hat{\psi}_{a,s}(k)} dk$$

$$= \sqrt{a} \int_{-\infty}^{+\infty} \hat{f}(k) \hat{\psi}(ak) e^{2\pi i k s} dk$$

Discrete wavelet coefficients

$$\tilde{f}(a_j, b_n) = a_j^{1/2} \sum_{k=0}^{N-1} \hat{f}_k \hat{\psi}(a_j k) e^{2\pi i b_n k}$$

$$\text{where } \hat{f}_k = \frac{1}{N} \sum_{n=0}^{N-1} f(x_n) e^{-2\pi i x_n k}$$

$$\text{with } x_n = \frac{nL}{N}, \quad b_n = \frac{nL}{N}$$

WAVELET FAMILY IN L2-NORTH
WITH REAL-VAUED WAVELETS

Scale in
pixel units

L signal duration or length
 k_{max} Nyquist cut-off frequency

$$a_{min} = \frac{k_0}{k_{max}}$$

2

3

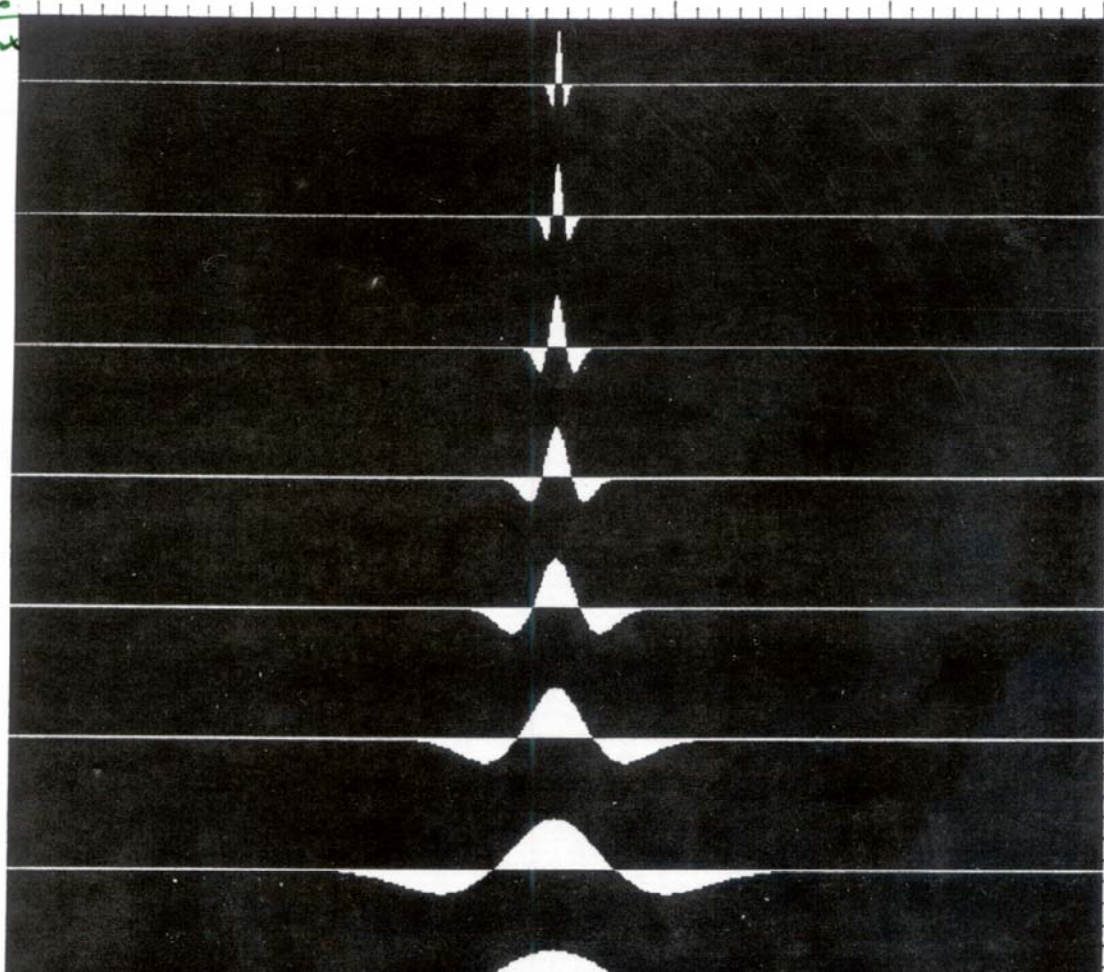
5

8

12

20

32



WAVELET FAMILY IN L2-NORM
WITH COMPLEX-VALUED WAVELETS

scale in
pixel units

L signal duration or length
 k_{max} Nyquist cut-off frequency

$$a_{min} = \frac{k_0}{k_{max}}$$

2

3

5

8

12

20

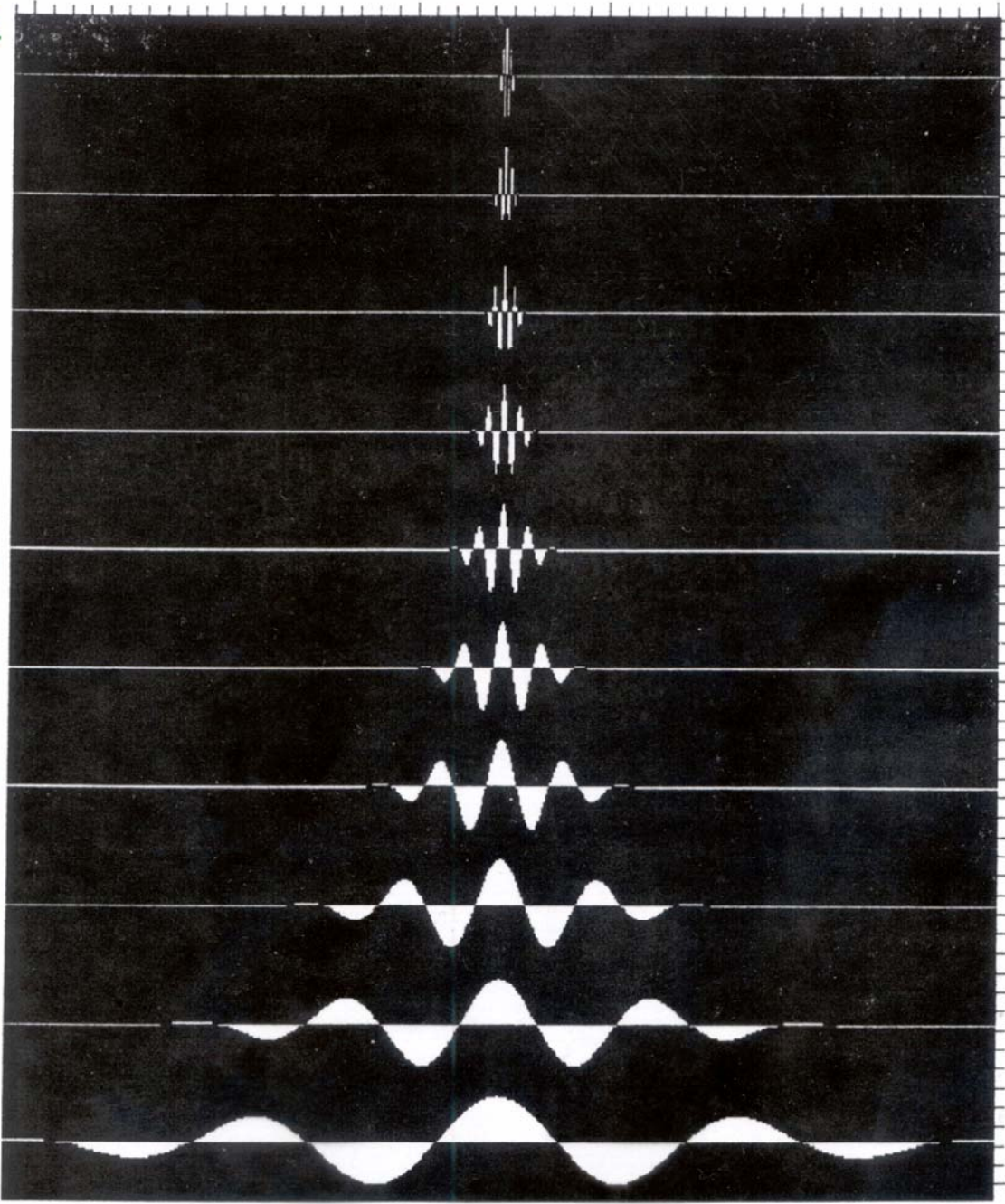
32

50

80

128

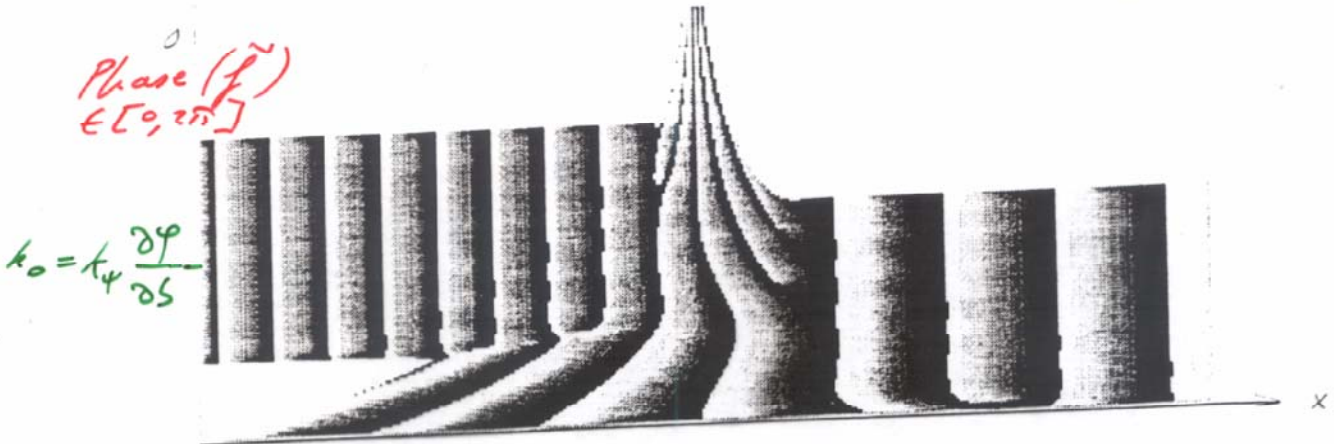
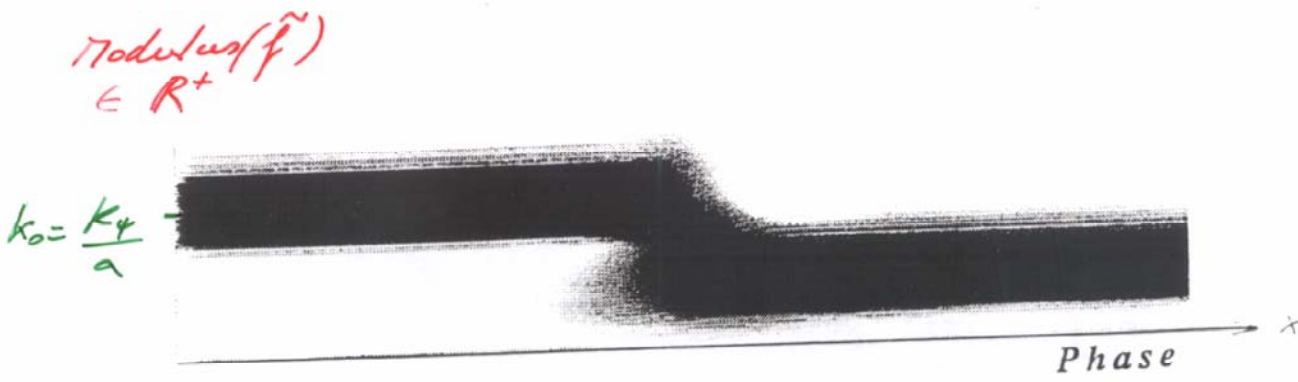
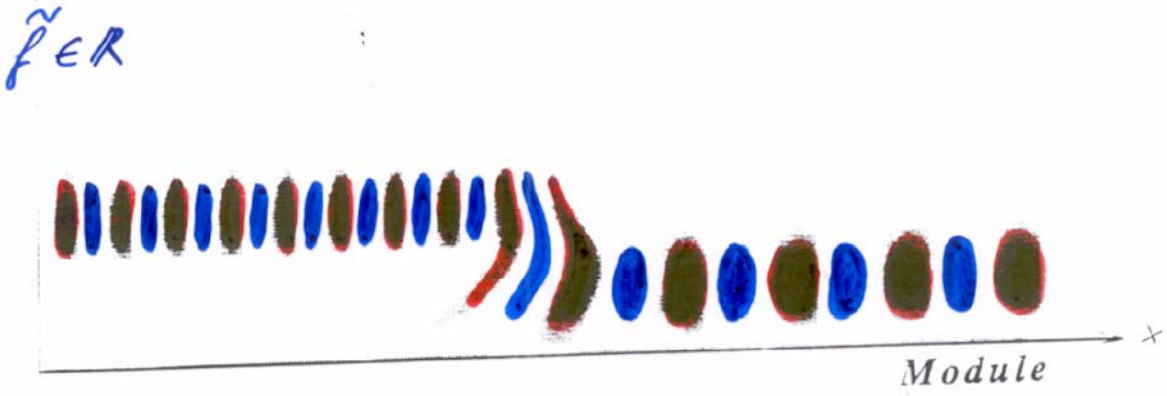
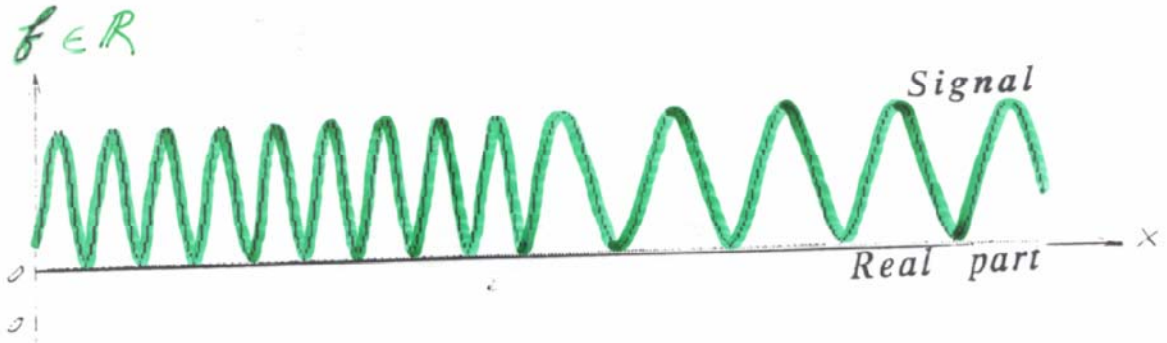
$$a_{max} = L$$



Morlet Wavelet

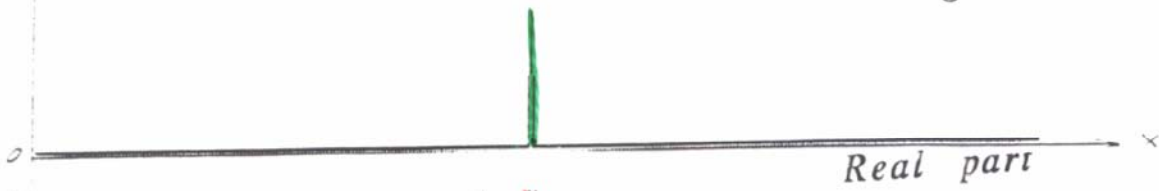
$$\psi(x) = e^{-ik_0x} e^{-\frac{x^2}{2}}$$

$$\psi \in \mathcal{C}$$

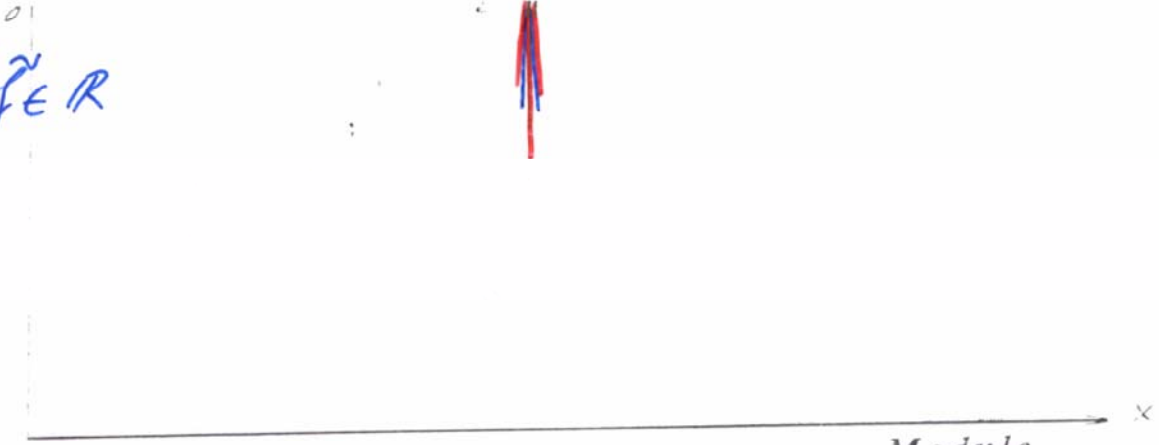


$f \in \mathbb{R}$

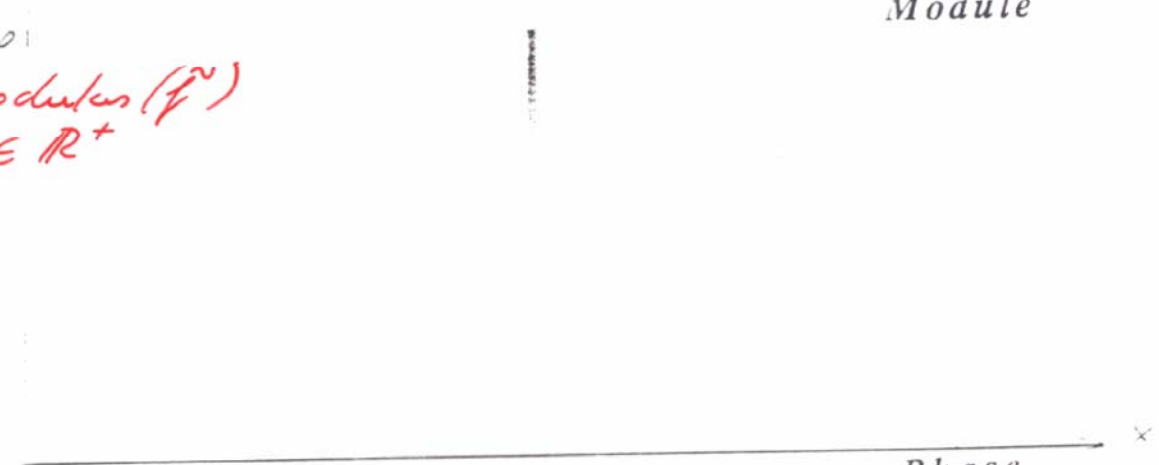
Signal



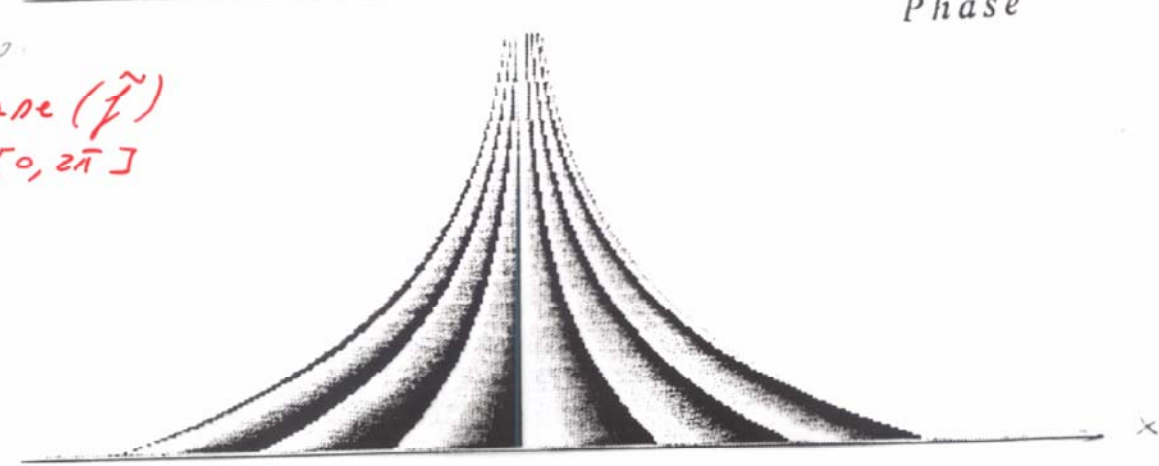
$\tilde{f} \in \mathbb{R}$



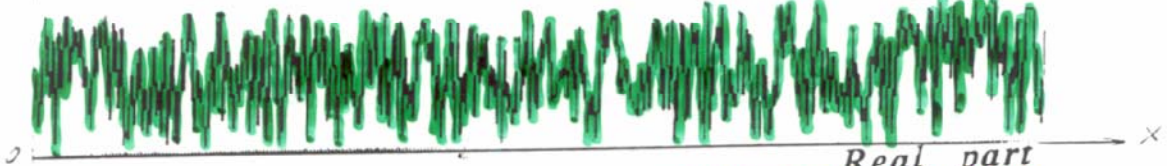
Module (\tilde{f})
 $\in \mathbb{R}^+$



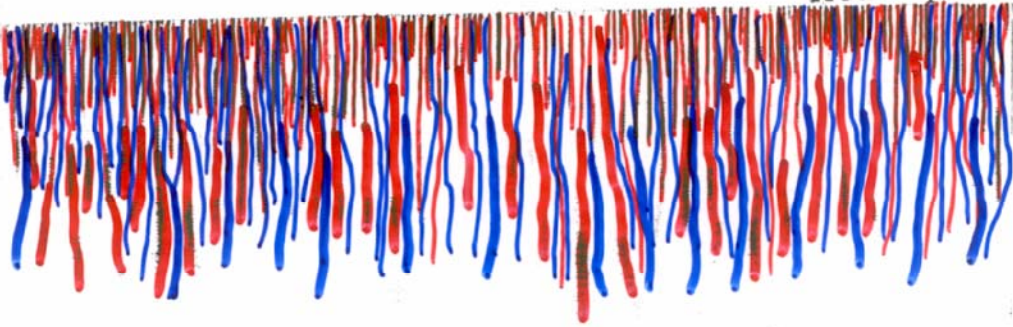
Phase (\tilde{f})
 $\in [0, 2\pi]$



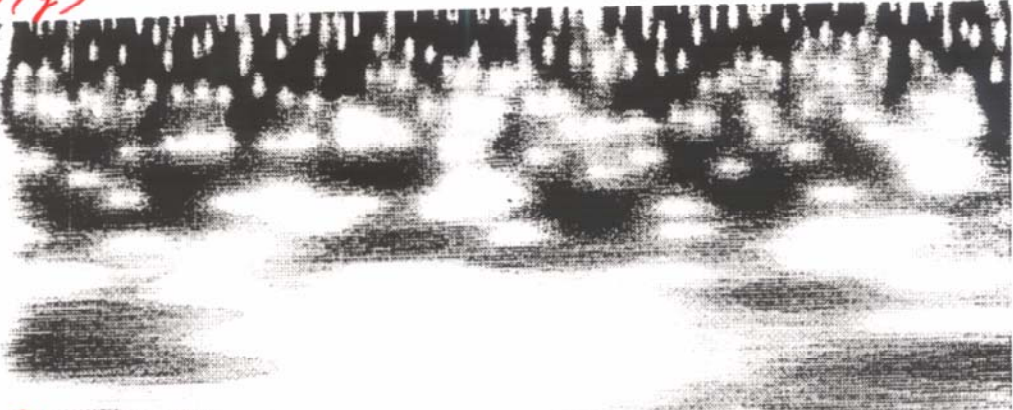
$f \in \mathbb{R}$



$f \in \mathbb{R}$



Modulus (f)
 $\in \mathbb{R}^+$



Phase (f)
 $\in [0, 2\pi]$

