

Different resolutions for different informations

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INTRODUCTION

The existing algorithms for **analysis, synthesis and transformation of sounds** ask the users to have a high competence on signal processing. To make the access to these efficient algorithms easier, it is desirable to **let the algorithms adapt themselves** to the signal and to the transformation needed.

One of the key parameter for nearly all the algorithms which analyze or process the sounds by using time-frequency representations or sinusoidal models is **the size of the analysis window**: here we show some examples about how this parameter is linked to the resolution of a signal analysis, and which informations can be highlighted choosing a certain size instead of another.

We are at present studying and developing algorithms for the **automatic adaptation of the analysis window size**. The first method approached is the parallel evaluation of multiple choices of the parameters and the subsequent combination of the final outcome through a comparison between their features.

ANALYSIS-SYNTHESIS SCHEME

An **audio signal** can be represented as a function

$$s : \mathbb{R} \rightarrow \mathbb{C}, s \in L^2(\mathbb{R}).$$

A system for the analysis and synthesis of a signal is composed of

- a set of functions called **time-frequency atoms**

$$\{\phi_\gamma\}_{\gamma \in \Gamma} \subset L^2(\mathbb{R}) \quad (1)$$

- a **time-frequency transform** operator associated to the above set

$$T_s(\gamma) = \int s(t) \overline{\phi_\gamma(t)} dt = \langle s, \phi_\gamma \rangle \quad (2)$$

- a **reconstruction formula**

$$s(t) = \int_{\gamma \in \Gamma} T_s(\gamma) \phi_\gamma(t) d\gamma \quad (3)$$

FRAMES THEORY

For the reconstruction formula to exist, the set $\{\phi_\gamma\}_{\gamma \in \Gamma}$ has to satisfy a condition obtained by relaxing the definition of an orthonormal system in a Hilbert space.

Definition 1 The set $\{\phi_n\}_{n \in \Gamma}$ is a frame of $L^2(\mathbb{R})$ if A and B exist such that they are positive and for every $s \in L^2(\mathbb{R})$

$$A \|s\|^2 \leq \sum_{n \in \Gamma} |\langle s, \phi_n \rangle|^2 \leq B \|s\|^2, \quad (4)$$

where thanks to the features of the Hilbert space $L^2(\mathbb{R})$ we can suppose Γ to be countable.

WINDOW FUNCTIONS

We consider a **window function** as a function $h \in L^2(\mathbb{R})$ such that $\|h\| = 1$ and $h(t) = h(-t)$; modifying h with a time translation of a factor t_0 and a frequency scaling by a factor ν_0 we obtain the set $\{h_{t_0, \nu_0}\}_{(t_0, \nu_0) \in \mathbb{R}^2}$ where

$$h_{t_0, \nu_0}(t) = h(t - t_0) e^{2\pi i \nu_0 t}. \quad (5)$$

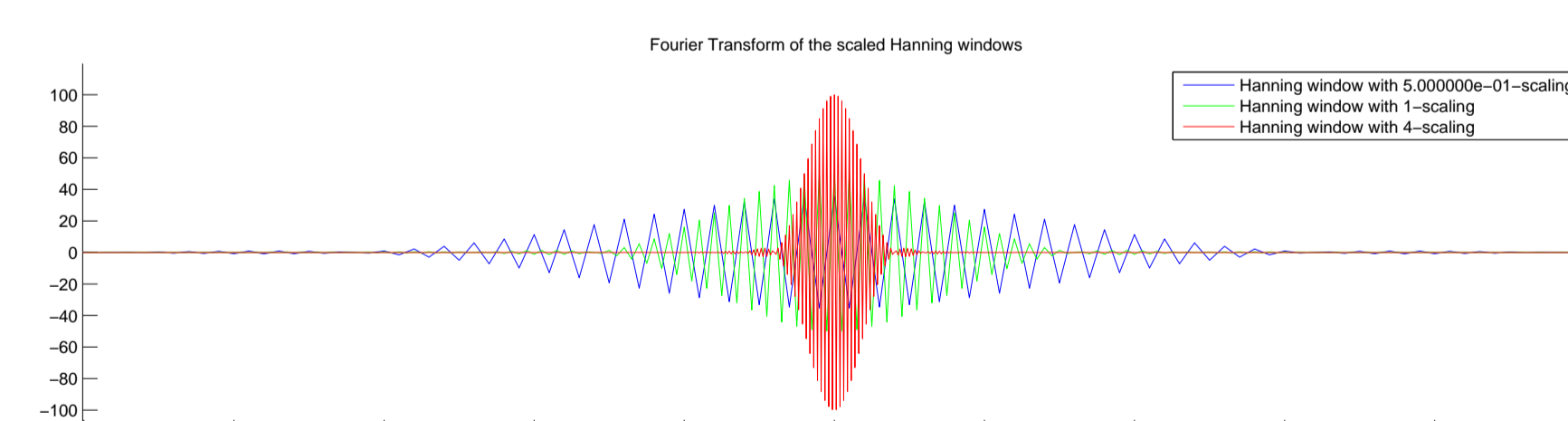
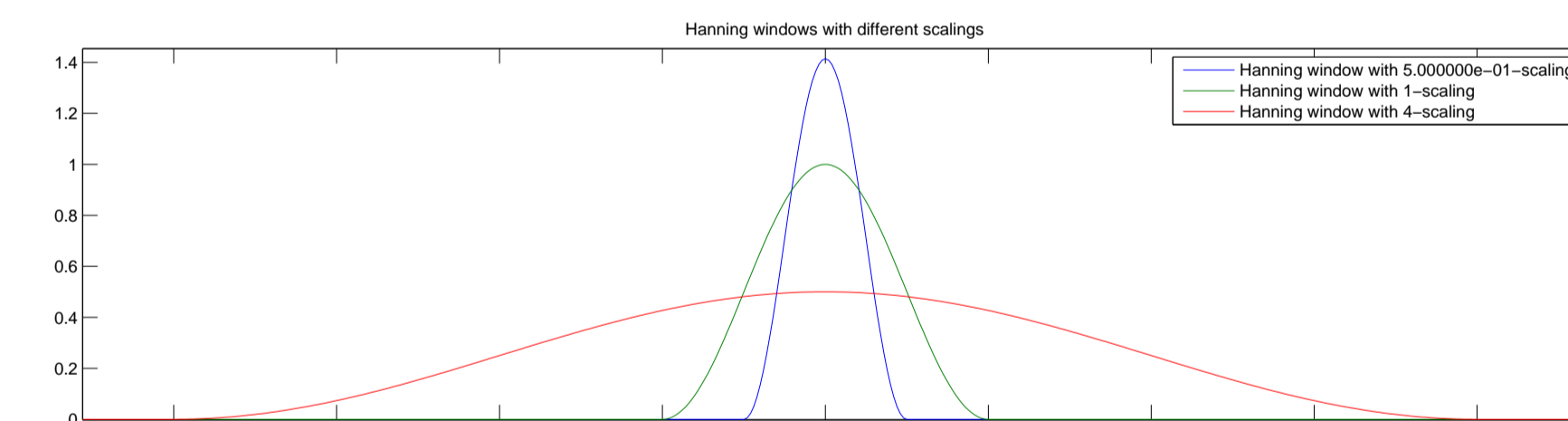
Remark 1

- The **Short Time Fourier Transform** is the time-frequency transform associated to the set $\{h_{t_0, \nu_0}\}_{(t_0, \nu_0) \in \mathbb{R}^2} \subset L^2(\mathbb{R})$
- Such set contains infinite countable frames of window functions for $L^2(\mathbb{R})$

A frame can be built by conveniently combining different sets of atoms derived from different window functions.

We are particularly interested to the case of the scaling of a same window function, which modifies its **time-frequency spread**,

$$h_l(t) = \frac{1}{\sqrt{l}} h\left(\frac{t}{l}\right). \quad (6)$$



ADAPTIVE SPECTROGRAM

The **Spectrogram** of a signal s is the squared modulus of the STFT of the signal

$$PS_s(t_0, \nu_0) = \left| \int s(t) h(t - t_0) e^{-2\pi i \nu_0 t} dt \right|^2. \quad (7)$$

It is always positive and with certain exceptions it can be treated as a probability distribution.

It is possible to perfectly represent the signal s in infinite many different ways, by choosing an appropriate set of multi-index $\Gamma \subset \mathbb{R}^2$ so that the relative frame $\{h_{t_0, \nu_0}\}_{(t_0, \nu_0) \in \Gamma}$ is dense enough in the time-frequency plane.

Given PS_s^h the spectrogram of a signal s through the frame built on the window h , a **sparsity measure** can be defined to evaluate the energy concentration in the analysis PS_s^h .

The notion of **best energy concentration** is not uniquely defined and is linked to the desired trade-off between time and frequency resolution.

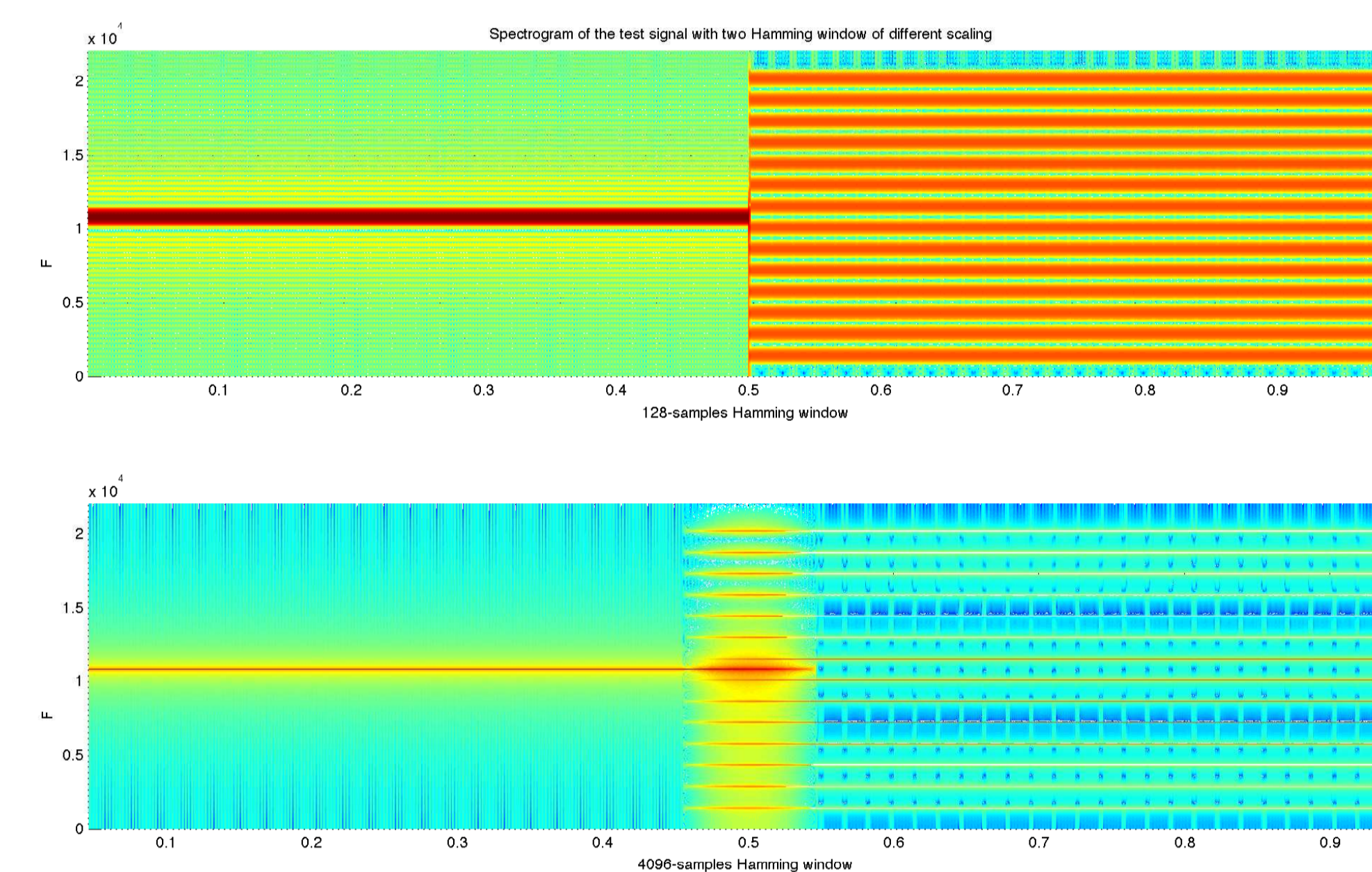


Fig.2 Spectrogram of a test signal with two Hamming window of different scaling.

As shown in [BF01] it is possible to extend the notion of **entropy** to certain time-frequency representations, such as the spectrogram, to define on them a class of sparsity measures.

Definition 2 Given a time-frequency representation $C_s \in L^2(\mathbb{R}^2)$ of the signal s , its Rényi entropy of order $\alpha > 0, \alpha \neq 1$ is

$$H_\alpha(C_s) = \frac{1}{1 - \alpha} \log_2 \iint C_s^\alpha(t, \nu) dt d\nu \quad (8)$$

Given a signal s and n different spectrograms $PS_s^{h_1}, \dots, PS_s^{h_n}$ of the same signal based on different windows h_1, \dots, h_n we choose as the **best spectrogram** the one which minimizes the Rényi entropy measure,

$$l^* = \arg \min_{l=1, \dots, n} H(PS_s^{h_l}). \quad (9)$$

ALGORITHM AND EXAMPLES

- Input and main parameters: the signal s , the set of window functions h_1, \dots, h_n , the Rényi entropy order α
- Computation of the n different spectrograms
- Evaluation of the Rényi entropy values on the global analysis
- Choice of the best spectrogram

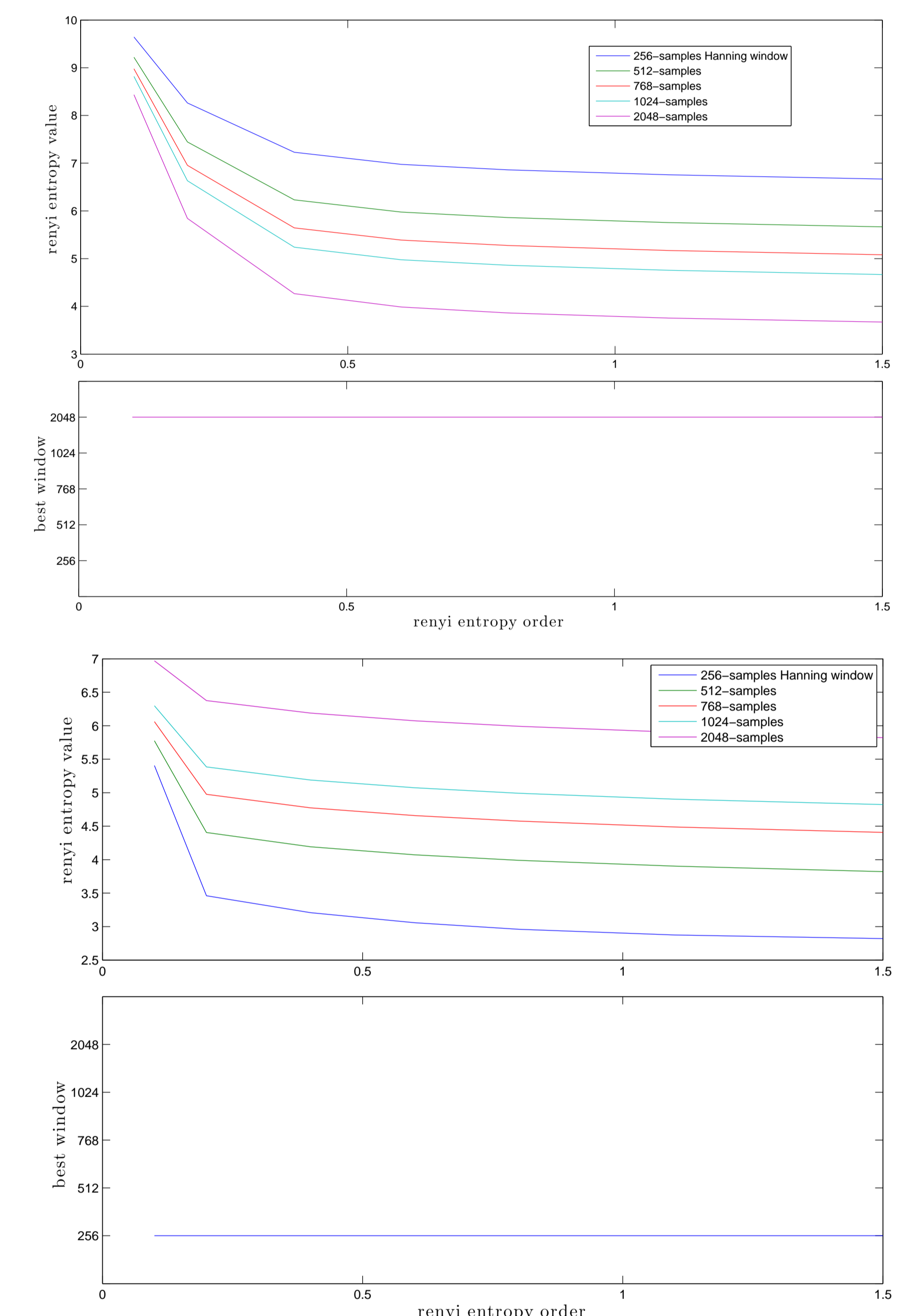


Fig.3 The choice made by the algorithm for two basic signals: two different sinusoids in the upper figure, an impulse in the lower. In these cases, the choice is α -independent

RESEARCH PROGRAM

- Development and investigation on the best strategy to set the hop size of the different spectrograms, in order to enhance the compatibility of the different analysis
- Definition of a strategy for time and frequency localization of the Rényi entropy evaluation, to obtain a local adaptation of the window size to the signal
- Investigation on the relation between the Rényi entropy order α and the choice of the best spectrogram by the algorithm

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