

A software companion for compressively sensed time-frequency processing of sparse nonstationary signals

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Abstract

Compressive sensing is a computational framework for acquisition and processing of sparse signals at sampling rates below the rates mandated by the Nyquist sampling theorem. In this paper, we present seven MATLAB functions for compressive sensing based time-frequency processing of sparse nonstationary signals. These functions are developed to reproduce figures in our companion review paper.

Keywords: Compressive sensing, time-frequency analysis, nonstationary signals.

1. Motivation and significance

Compressive sensing has been a very active research area for the past ten years. Its main idea is to decrease a number of steps during acquisition and compression of signals by combining these two processes into a single step [1], [2]. The main advantage of compressive sensing is that it enables us to acquire signals/images at sampling rates below rates mandated by the Nyquist sampling theorem, but signals under consideration have to be sparse in time or other signal domains [1].

9 Compressively sampling non-stationary signals is not always trivial, as
10 those signals are typically not sparse in time or frequency domains [3], [4],
11 [5]. Furthermore, such compressively sampled nonstationary signals need
12 to be processed using advanced signal processing techniques such as time-
13 frequency representations [6], [7].

14 In our companion review paper [8], we reviewed several algorithms deal-
15 ing with compressive sensing and time-frequency processing of nonstationary
16 signals. In this paper, we briefly describe MATLAB functions developed to
17 produce all seven figures from the companion paper. These figures represent
18 various test cases where compressive sensing and time-frequency analysis can
19 be combined to provide a framework for the analysis of nonstationary sparse
20 signals.

21 These developed functions expand our previous efforts to disseminate soft-
22 ware solutions that provide implementations of advanced signal processing
23 algorithms in various applications [9], [10]. Hence, current functions along
24 with previous software releases aim to increase the implementation of these
25 algorithms in practical applications. This is a significant contributions, as it
26 is a well known fact that many signal processing solutions have never been
27 translated to practice due to difficulties encountered during the algorithmic
28 software implementation.

29 As mentioned before, these functions were developed as standalone MAT-
30 LAB functions and can be run independently of other functions. A typical
31 user will download the function into their working MATLAB directory and
32 run functions as any other MATLAB functions. However, these functions do
33 not require any inputs, and their outputs are figures shown in the companion
34 paper.

35 Lastly, it should be mentioned that these functions implement previously

36 proposed algorithms (e.g., [11], [12], [13], [14]) and no new algorithms were
37 proposed in the companion paper.

38 **2. Software description**

39 Each developed function is a stand-alone function that can be run inde-
40 pendently in a MATLAB environment. No additional functions are needed.
41 It should be mentioned that for function *figure1* additional data is needed,
42 which is also shared via the ZIP file.

43 It should be pointed out that parts of codes were borrowed from the
44 l_1 -magic toolbox [15] developed by Professor Candes and his collaborators.
45 However, those parts are clearly labeled, and no ownership is claimed.

46 *2.1. Software Architecture*

47 No particular software architecture is used here, as each function is a
48 stand-alone function.

49 *2.2. Software Functionalities*

50 The main functionality of the software is to reproduce figures from the
51 companion paper. However, other researchers will be able to use the software
52 and adopt it for their own examples.

53 **3. Impact**

54 The proposed software package represents an excellent tool for researchers
55 and engineers to recreate examples shown in the companion review paper.
56 Moreover, the researchers working in the compressive sensing field will be
57 able to reuse the existing functions in other applications with different prob-
58 lem set-ups. In other words, signal processing practitioners will be able to
59 apply these techniques in diverse applications from medical signal processing

60 to communications applications. Particularly, the idea of using the ambi-
61 guity domain to produce sparse time-frequency representations can be fur-
62 ther explored to cope with the cross-terms in time-frequency representations.
63 Thereby, researchers working in this field can modify the software functions
64 in order to include certain foreknowledge about the desired components lo-
65 cations in the ambiguity domain. Functions developed for the separation of
66 stationary and non-stationary signals can be also reused in various scenarios
67 occurring in communication channels. In our examples, we illustrated the
68 case with overlapping radar data. However, similar challenging situations
69 also occur in wireless communications, where signals of interest are station-
70 ary, while disturbance components can be highly nonstationary signals. In
71 those cases, researchers can adopt the provided functions to eliminate the
72 disturbance and obtain an incomplete representation that needs to be re-
73 covered using the described approach. Furthermore, the proposed package
74 also enables signal processing practitioners, especially at early stages such
75 as upper-year undergraduate students or early-stage graduate students, to
76 understand the nuisances of algorithmic implementations using modern soft-
77 ware packages such as MATLAB.

78 Furthermore, we anticipate that the considered functions can be used in
79 diverse signal processing applications from compressive sensing of nonsta-
80 tionary signals to estimation of instantaneous frequencies of highly complex
81 nonstationary signals corrupted by measurement noise.

82 The developed functions could be also of interest for the practitioners
83 outside the signal processing community. In other words, these functions
84 can be used as a framework for the implementation of missing data recon-
85 struction in certain economic and financial time series analysis, where it is
86 commonly assumed that the discrete Fourier transform is a suitable tool for

87 data representation.

88 **4. Conclusions**

89 In this paper, we presented a set of MATLAB functions that is useful for
90 implementing algorithms considered in the companion review paper. These
91 algorithms were based on the ideas of compressive sensing and time-frequency
92 processing of nonstationary signals. In particular, these proposed functions
93 represent an initial draft of the future comprehensive MATLAB toolbox for
94 compressive sensing of nonstationary signals that will be able to also provide
95 a set of functions for time-frequency processing of such signals.

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Nr.	Code metadata description	Please fill in this column
C1	Current code version	V1.0
C2	Permanent link to code/repository used for this code version	https://pitt.box.com/s/1vnh5dxzauiy6hyy2pd5t5zul0719ht3
C3	Legal Code License	Apache License 2.0
C4	Code versioning system used	None
C5	Software code languages, tools, and services used	MATLAB
C6	None	
C7	If available Link to developer documentation/manual	None
C8	Support email for questions	irenao@ac.me

Table 1: Code metadata (mandatory)

102 **Current code version**

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