

IRootLab: a free and open-source MATLAB toolbox for vibrational biospectroscopy data analysis

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ABSTRACT

Summary: IRootLab is a free and open-source MATLAB toolbox for vibrational biospectroscopy (VBS) data analysis. It offers an object-oriented programming class library, graphical user interfaces (GUIs) and automatic MATLAB code generation. The class library contains a large number of methods, concepts and visualizations for VBS data analysis, some of which are introduced in the toolbox. The GUIs provide an interface to the class library, including a module to merge several spectral files into a dataset. Automatic code allows developers to quickly write VBS data analysis scripts and is a unique resource among tools for VBS. Documentation includes a manual, tutorials, Doxygen-generated reference and a demonstration showcase. IRootLab can handle some of the most popular file formats used in VBS.

License: GNU-LGPL.

Availability: Official website: <http://irootlab.googlecode.com/>.

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1 INTRODUCTION

Vibrational biospectroscopy (VBS) is the application of infrared and Raman spectroscopy to biological studies and biomedical applications (Martin *et al.*, 2010). In the past decades, vibrational spectroscopy data processing was developed on the grounds of signal processing, statistics and pattern recognition, resulting in a large number of computational methods that are routinely applied in the field. However, the field needs further development and establishment of data analysis strategies, partially owing to a continuous push from instrumental advancements, but also motivated by challenges in pre-processing, biomarker extraction and standardization of inter-experimental datasets, and inspired by the increasing power of computers (Trevisan *et al.*, 2012).

VBS data analysis is largely carried out by commercial softwares such as CytoSpec (www.cytospec.com), Neurodeveloper (www.neurodeveloper.com), Unscrambler (www.camo.com), Pirouette (www.infometrix.com), OPUS (www.bruker.com) and

Wire (www.renishaw.com). Popularly used commercial MATLAB toolboxes include the Neural Network and the Bioinformatics Toolbox. Others are free to use, but closed-source, such as PRTools (Duin *et al.*, 2007) and GA_ORs (Nikulin *et al.*, 1998). While these are all high-quality software, the use of closed-source software contradicts the multidisciplinary aspect of the field, as researchers looking for new analysis strategies remain dependant from the software manufacturers. On the other hand, open-source softwares available are largely restricted to a given task, such as LibSVM (Chang and Lin, 2011), Functional Data Analysis toolbox (Ramsay *et al.*, 2009) and various other resources available online.

IRootLab is a free/libre and open-source (FOSS) MATLAB toolbox created to fulfil the need for FOSS software encompassing every stage of VBS data analysis. It is a highly modular software that was created based on existing theory of data analysis (Alpaydin, 2004; Bishop, 2006; Duda *et al.*, 2001; Guyon *et al.*, 2006; Hastie *et al.*, 2007; Kuncheva, 2004), and as such, applied to VBS (Griffiths and Haseth, 2007; Somorjai, 2009), bringing together families of methods such as outlier removal, pre-processing, feature extraction, feature selection, feature construction, classification, classifier aggregation, clustering; concepts such as cross-validation, random sub-sampling, peak detection, biomarker identification; a set of unique visualization options (Supplementary Fig. S1) and a MATLAB code generator. The basic modular design is simple, which makes IRootLab a flexible and intuitive resource for vibrational spectroscopy data analysis developers.

2 CLASS LIBRARY

The majority of IRootLab code lines are attributed to its hierarchically organized object-oriented programming (OOP) classes. The basic framework is constituted of three branches: *datasets*, *blocks* and *logs*. A *dataset* is an object representing point spectra (with their respective classes, patient names, etc.), image maps, clustering data or classifier estimations. A *block* is the basis of data manipulation. It is an object that allows mainly two operations: *training* and *using*. *Training* involves the use of training data to modify a *block* (according to its specific idiosyncrasy). *Using* consists of inputting data into a block to get various results, including output data, estimation data, figures and HTML reports. Finally, *logs* represent the output of special blocks called

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analysis sessions, which perform complex analyses and generate output that does not fit into the usual *dataset* class.

Additionally, there are a few complementary branches, from which the most important are *sub-dataset generation specs* (SGS), *feature sub-set grader* (FSG) and *peak detector*. SGS is an abstraction of the concept of random sub-sampling (e.g. cross-validation, repeated random sub-sampling); FSG is used to evaluate sub-sets of features in feature selection tasks, freeing the feature selection algorithm from having to implement its own evaluation; and *peak detection* (Coombes et al., 2003) is used in visualizations and biomarker identification.

Blocks can be created and combined recursively (e.g. a special block called ‘block cascade’ is composed of other blocks, which in turn can be used as a component of a higher level block cascade) to build analyses as complex as desired.

Many complex operations are already offered as sub-classes of a special class of block called *analysis session*, including grid search optimization (Hsu et al., 2010), repeated train-test using an SGS object to obtain performance estimation of a classifier and cross-calculation of Linear Discriminant Analysis (LDA) scores (Riding et al., 2012).

3 GRAPHICAL USER INTERFACES

Figure 1 shows the interaction of IRootLab graphical user interfaces (GUIs) with the class library and their surrounding environment. IRootLab offers a flexible and intuitive GUI called *objtool* that is an interface to the class library. *Objtool* allows datasets and other objects present in the MATLAB workspace to be browsed and manipulated using new or existing blocks. *Objtool* can handle three types of TXT files, OPUS image maps and a native IRootLab MAT format.

Mergetool is another GUI that allows a collection of single-spectrum files to be merged together to form a dataset. Currently, *mergetool* can import three different types of

single-spectrum files: Pirouette.dat, OPUS single-spectrum files and Wire TXT files. Finally, the *SHEload* GUI accesses an online MySQL database that is part of a chemical database project (Trevisan et al., 2010), importing datasets from there.

4 CODE GENERATION

One of the major features of IRootLab is the ability of *objtool* to generate MATLAB code (similar to macro recording in Microsoft Word). This is a highly efficient resource for MATLAB scripting. In fact, a significant part of IRootLab was written using this resource. Code generation also keeps developers from having to consult the library documentation too often.

5 DOCUMENTATION

IRootLab has a clear documentation project. The official documentation (available at <http://bioph.lancs.ac.uk/irootlabdoc>) is generated directly from the source code using Doxygen software. Apart from this, IRootLab has a manual and a series of tutorials (available at downloads area of the official website) that together provide a friendly start for new users. The official reference can be accessed from MATLAB in two different occasions: from the command line, by using the *help2* command, and from the GUIs, by pressing the F1 key (context-sensitive help). Additionally, a demonstration showcase was prepared (which can be opened by typing the command *browse_demos* at MATLAB command line), and a number of sample datasets are shipped with the toolbox.

6 CONCLUSION

Although the most valuable contributions of IRootLab are its framework and its code generation, there are currently more than 200 implemented OOP classes representing methods, algorithms, concepts and visualizations. One of the advantages of FOSS software is that any user can potentially expand the toolbox to include the methods necessary to their own analysis, should these methods not be already present, and then share the additions, creating a synergic effort among the scientific community. Expansion is facilitated by IRootLab modular structure, which allows developers to target or inherit a specific OOP class. IRootLab is a unique and innovative contribution to the interdisciplinary research field of VBS, providing researchers with a valuable tool for the development of VBS data analysis.

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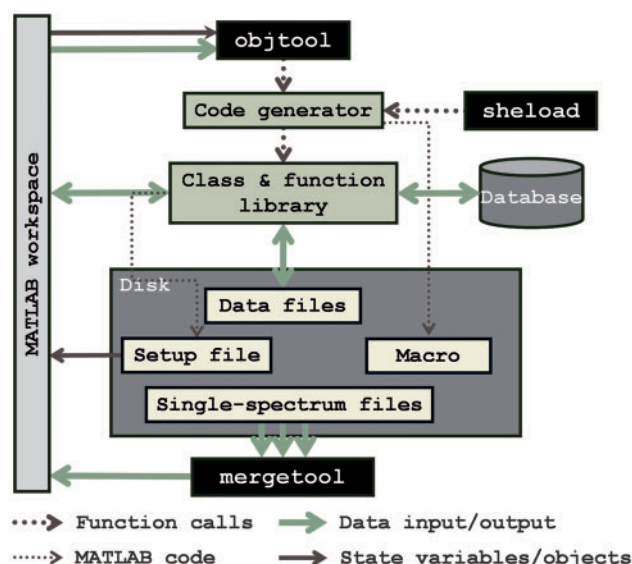


Fig. 1. IRootLab GUIs (*objtool*, *mergetool*, *sheload*) and their surrounding environment

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Conflict of Interest: none declared.

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