

SIGNAL PROCESSING TOOLS FOR SPEECH RECOGNITION¹

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ABSTRACT

This paper describes the design and development of a set of signal processing software tools for speech recognition. The tools were developed for inclusion in a comprehensive public domain speech recognition toolkit. We describe the design philosophy underlying the development of the tools as well as the key features that enable realization of our design goals of modularity, extensibility, and usability. A GUI-based configuration tool is presented that allows complicated, multi-pass front end algorithms to be created using a graphical editor and a library of fundamental algorithm components. We also discuss results of tests to verify the correctness and usability of the tool set, including benchmarks on SWITCHBOARD, WSJ0 and the Aurora Large Vocabulary tasks.

1. INTRODUCTION

The Institute for Signal and Information Processing (ISIP) provides a comprehensive public domain toolkit [1] for performing speech and signal processing research. Several differentiating features are its ease of use, extensibility, and educational components. In this paper we describe the design and implementation of its signal processing components, which provide a GUI-based environment to perform signal processing research and education.

An overview of a speech recognition system is shown in Figure 1. The tool described here deals with the block known as the Acoustic Front-end, which encapsulates most of the signal processing portions of a recognition system. Signal processing tools extract feature vectors from speech data, and play a critical role in the development of speech recognition systems. Many signal processing toolkits are currently available including popular commercial products such as MATLAB [2]. Such toolkits provide powerful computation and analysis capabilities, and sophisticated graphical interfaces. Nonetheless, they also contain serious deficiencies that limit their usefulness in a research environment. For example, run-time efficiency and

file I/O are two common issues with such high-level tools.

Adding new algorithms to such toolkits requires modifying the base code of the existing system, a potentially time-consuming and costly undertaking that can significantly impede many research efforts. Special problems for speech recognition front ends, such as synchronization and buffering of data along a data flow graph, are difficult to handle in a simple and easy to understand framework. Of even greater importance and difficulty, data preparation for algorithms that require multiple frames of data, such as windows and differentiation, can be problematic.

To address these issues, we have developed a modular, flexible environment for signal processing. The key differentiating characteristics of our system include:

- Competitive technology with maximum flexibility;
- Well-documented and simple APIs;
- An object-oriented software design in C++.

In this paper, we present our software design rationale and approach for maximizing modularity and usability.

2. SOFTWARE DESIGN

Research in the area of speech recognition requires the development of large applications in a relatively short period of time. To address these needs, we designed a large, hierarchical software environment to support advanced research in all areas of speech recognition, including signal processing. This environment contains ISIP foundation classes (IFCs) that provide features ranging from complex

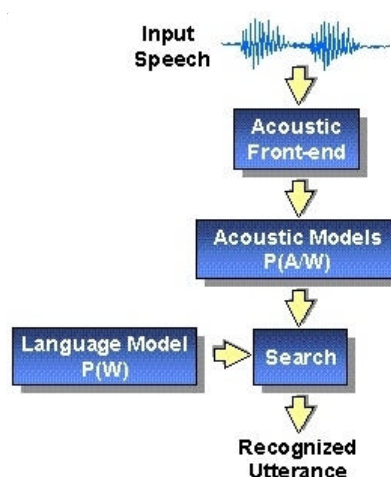


Figure 1: An overview of a speech recognition system.

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data structures to an abstract file I/O interface. IFCs are implemented as a set of C++ classes, organized as libraries in a hierarchical structure. Some key features include:

- Unicode support for multilingual applications;
- memory management and tracking;
- system and I/O libraries that abstract users from details of the operating system;
- math classes that provide basic linear algebra and efficient matrix manipulations;
- data structures that include generic implementations of essential tools for speech recognition code.

We developed our signal processing toolkit to stringently adhere to the IFC design philosophy and framework. The software described in this paper involves primarily the Algorithm library [3] in the IFC class hierarchy. At the outset, it was clear that the tools must not only allow a wide selection of algorithms, but also the ability to vary every parameter of each algorithm easily and finally, provide users an efficient environment for evaluating new research ideas. Thus, the design requirements for these tools included:

- a library of standard algorithms to provide basic digital signal processing (DSP) functions;
- an ability to easily add new algorithm classes and functions without modifying existing classes;
- a block diagram approach to describing algorithms to realize rapid prototyping without programming.

Fulfilling the first requirement enabled users to directly realize a single algorithm such as a window with simple programming by building an algorithm object, and calling its functions. This ability allows users to easily plug our algorithms into their own codes. Meeting the second requirement allowed users to enhance system capabilities according to their new requirements. This is described in Section 2.1. To meet the third requirement, we developed a signal processing control tool and a signal processing configuration tool. By separating the design data flow block diagram using our signal configure tool written in Java GUI with the processing the signal data flow using signal processing tool written in C++, users do not need to program but just draw block diagram. These are described in Section 2.2.

2.1. Algorithm Library

The algorithm library contains a collection of signal processing algorithms implemented as a hierarchy of C++ classes. The implementation of this hierarchy using an abstract base class, `AlgorithmBase`, and virtual functions or methods that comprise the interface contract, is the single most important feature, since it makes the library extensible. All algorithm classes are derived from this base class. However, since it is an abstract class, no objects are ever directly instantiated from it. Instead, it defines the interface

contract, specifying virtual functions that all Algorithm classes must provide, and centralizes useful protected data

Processing:

```
virtual boolean init();
virtual boolean apply();
```

Configuration:

```
virtual const String& className() const;
virtual long getLeadingPad() const;
virtual long getTrailingPad() const;
virtual CMODE getOutputMode() const;
virtual float getOutputSampleFrequency() const;
virtual boolean setParser();
```

Debugging:

```
boolean displayStart();
boolean displayFinish();
boolean displayChannel();
boolean display();
```

Table 1: An overview of the interface contract for the Algorithm classes.

common to all algorithms, such as sample frequency and frame duration. The interface contract is summarized in Table 1:

Note that the key computational steps of any algorithm are limited to two functions: *init* and *apply*. The remaining functions simply facilitate configuration and debugging. The configuration functions deal with retrieving information from the specific algorithm that is needed to coordinate I/O processing. For example, *getLeadingPad* and *getTrailingPad* are used to determine the amount of delay a specific algorithm introduces so buffers can be adjusted accordingly. The debugging methods were introduced to allow users to insert debugging statements within the signal flow graph, and to see intermediate output as the data is being processed.

Expanding the collection of algorithms supported in our Algorithm library is the subject of on-going research. Our current offerings can be sorted into two categories: basic DSP and support. The basic DSP components include commonly used algorithms, such as windows, filter, Fourier transform, covariance, correlation and energy. Support components allow high-level manipulation of data flow through block diagrams, such as math, statistics, mask, connections and mask. Together, they provide a unique and powerful set of signal processing capabilities, some of which include: multi-pass processing of a signal; automatic handling of arbitrary amounts of prior and future data when a recipe is created; processing of a signal, saving a constant derived from that signal to a file, and reloading the constant.

2.2. Signal Processing Configuration Tool

The procedure by which users employ the tools and libraries can be described as follows: First, the signal

processing configuration tool is used to graphically specify the sequence of algorithms and their configuration using a block diagram. This is saved to a file containing a description of the block diagram. This description uses a graph data structure containing components, each of which has its own configuration. Second, a control tool accepts the speech data file and the recipe files produced in the first step as input. It then parses the recipe file using functions provided by the signal processing library to obtain the necessary information for each algorithm. Finally, it applies the corresponding algorithm functions to process the input speech data by calling the correct method in the algorithm library.

We developed a Java GUI tool, shown in Figure 2, to provide users a block diagram approach to designing front ends. We chose the Java language to allow the tool to run across a wide range of platforms and to give the tool an industry-standard look and feel. This tool allows users to select algorithms from an inventory of predefined components, and to connect and configure these components using standard graph drawing tools. Each component represents one algorithm for which the user can specify how to process data; each arc represents a data flow from one algorithm to another. To create a block diagram, the user selects the desired algorithm from the component menu, connects each component using directed arcs, configures each component in the diagram, and saves the configuration or “recipe” into a file. The control tool, described in Section 2.3, uses this file to complete the signal transformation process.

The example block diagram shown in Figure 2 illustrates many unique capabilities of the configuration tool. All synchronization and buffering of data between components within a single data flow and across data flows is automatically handled by these tools. The functional interface described in the previous section enables this. The

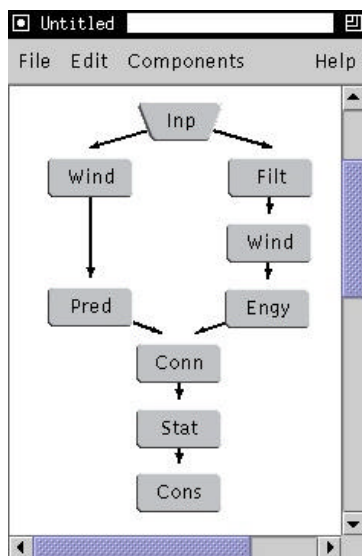


Figure 2: A configuration tool that allows users to create new front ends by drawing signal flow graphs.

user need only draw the block diagram to indicate how the signal should be processed, without concern for data synchronization or buffering. In addition, the example illustrates the support provided for multi-pass processing. The constant saved to a file can be easily reloaded as input to the same data flow diagram or differing diagrams.

Debugging information can be added to the flow graph using a number of mechanisms. A display block can be inserted on any arc to display data to the console (*stdout* or a file) as the data propagates through the graph. Each block can also output debugging information through the use of a debug mode that is part of the algorithm interface contract. Through standardization of such capabilities through an interface contract, we are able to provide a uniform interface for debugging across all utilities and GUIs in the toolkit.

Finally, to increase the extensibility of the tool, algorithms are presented in the interface through the components menu, populated from a resource file. All algorithms appearing in this menu are read from the resource file. Adding a new algorithm requires simply including a description into the resource file according to its format. No modifications to the source code of the signal processing control tool itself are required. This has allowed this tool to be used as interfaces for a number of related applications provided in our toolkit.

2.3 Signal Processing Library and Control Tool

The signal processing library is a collection of specially designed modules, implemented as C++ classes, which serve as an interface between the block diagrams, created by the GUI configuration tool, and the computation algorithms, described in Section 2.1. It should be noted that the work of the signal processing library is hidden from the user by default. Its functions include: parsing the file containing the recipe created by the user with the configuration tool; synchronizing different paths along the block flow diagram contained in this file; preparing input/output data buffers for each algorithm, particularly for those requiring multiple frames of data, such as windows or calculus; scheduling the sequences of required signal processing operations; processing data through the flow defined by the recipe; and finally, managing conversational data.

By combining our algorithm library, signal processing library and signal processing configuration tool, we fulfilled our design goal to provide a tool for users who do not need know how to program but they still can finish complex digital signal processing procedure as long as they understand the signal data flow and build block diagram.

Another important attribute of the signal processing control tool concerns its compatibility with other components of our software. The control functions are embedded in the recognizer so that both tools use exactly the same code base. This further enhances the usability of

the toolkit, enabling researchers to more easily achieve consistency in experimental results by using the same data in recognition as that used in feature extraction. Also, the fact that the front end is embedded in the recognizer allows live input demos to be easily created, again by supporting use of the same data from feature extraction.

3. EXPERIMENTAL RESULTS

We tested the quality of our toolkit along two dimensions, correctness and usability. To verify the correctness of the computation results, we have successfully built several complex front ends, including an industry standard front end based on Mel-frequency cepstrum coefficients (MFCCs) [4]. Our testing procedure entailed first comparing the data generated from the general purpose tools described in this paper to similar data generated from a prior version of our software that has been publicly available for several years, but contained less general implementations of many algorithms. Since there are subtle differences in the way the components are implemented, byte by byte comparisons of the data are not always possible or desirable.

Hence, in addition to directly comparing values in the feature vector streams, we also ran several recognition experiments, including Switchboard (SWB) [5] and Wall Street Journal (WSJ). As one example, we compared the correctness of our results against comparable baseline systems used in the Aurora evaluations on the 5K WSJ0 task [6]. Front ends created using our general purpose tools matched performance on these tasks achieved using the older versions of the software. For example, we achieved an 8.3% WER using our MFCC front end recipe, and this matches the results reported in [4].

Next, we assessed and enhanced the usability of our tools through extensive user testing conducted over the course of many workshops [6]. As part of this testing, we gave the users the description of the signal processing data flow, users who did not have any programming experience can easily create the block diagram and correctly save to recipe files according to the description. Then the user can correctly get the results by using giving commands. As another aspect of this testing, we administered a user survey derived from the Questionnaire for User Interaction Satisfaction (QUIS), a measurement tool designed for assessing user subjective satisfaction with the human-computer interface [7]. Several features of the interface were modified or enhanced as a result of these tests. General examples include reductions in the number of menus, number of menu options, changes in wording of menu options, and modifications to the behavior of the drawing tool itself.

4. CONCLUSIONS

This paper has presented the signal processing component of our public domain speech recognition toolkit.

This component was designed and developed in adherence to our philosophy of providing a flexible, extensible software environment for speech recognition researchers. Our goal was to enable researchers to explore ideas freely, unencumbered by low-level programming issues. To achieve this goal, we implemented several critical features in our signal processing software tools, including a library of standard algorithms for basic DSP functions, the ability to add new algorithms to this library easily, and a GUI-based configuration tool for creating block diagrams to describe algorithms, allowing rapid prototyping without programming.

We have tested and verified this tool for both correctness and usability. It empowers researchers to easily build state-of-the-art front end systems for speech recognition. We continue to monitor feedback from our user community in order to maintain the highest quality of the tool. This tool has been one of the most popular components of our toolkit, and is suitable for teaching basic concepts in digital signal processing.

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