Speech and Text-Image Processing in Documents

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ABSTRACT

Two themes have evolved in speech and text image processing work at Xerox PARC that expand and redefine the role of recognition technology in document-oriented applications. One is the development of systems that provide functionality similar to that of text processors but operate directly on audio and scanned image data. A second, related theme is the use of speech and text-image recognition to retrieve arbitrary, user-specified information from documents with signal content. This paper discusses three research initiatives at PARC that exemplify these themes: a text-image editor[1], a wordspotter for voice editing and indexing[12], and a decoding framework for scanned-document content retrieval[4].

1. INTRODUCTION

Research on application of spoken language processing to document creation and information retrieval has focused on the use of speech as an interface to systems that operate primarily on text-based material. Products of such work include commercially available voice transcription devices, as well as DARPA-sponsored systems developed to support speech recognition and database interaction tasks (e.g., the Resource Management[9] and Air Travel Information Systems[8] tasks, respectively). Similarly, work on text image processing has focused primarily on optical character recognition (OCR) as a means of transforming paper-based documents into manipulable (i.e., ASCII-based) electronic form. In both cases, the paradigm is one of format conversion, in which only enough signal analysis is performed to accomplish the user's goal. The systems are intended to facilitate authoring and editing of documents for which input and output medium are the same. Examples include authoring of voice mail and editing of facsimile-based document images.

A second, and related, research theme is the use of speech and text-image recognition to retrieve arbitrary, user-specified information from documents with signal content. The focus is again on partial document models that are defined in only enough detail to satisfy task requirements. Depending upon application, format conversion may or may not be a desired goal. For example, in retrieving relevant portions of a lengthy audio document via keyword spotting, a simple time index is sufficient. On the other hand, extracting numerical data from tabular images to facilitate on-line calculations requires at least partial transcription into symbolic form.

This paper discusses three research initiatives at PARC that exemplify these themes: a text-image editor[1], a wordspotter for voice editing and indexing[12], and a decoding framework for scanned-document content retrieval[4]. Overviews of the three systems are provided in Sections 2 through 4, respectively; concluding comments are contained in Section 5. The discussion focuses on key concepts embodied in the research that enable novel signal-based document processing functionality.

2. TEXT IMAGE EDITING: IMAGE EMACS

Image EMACS is an editor for scanned documents in which the inputs and outputs are binary text images[1]. The primary document representation in Image EMACS is a set of image elements extracted from scanned text through simple geometrical analysis. These elements consist of groupings of connected components (i.e., connected regions of black pixels)[2] that correspond roughly to character images. Editing is performed via operations on the connected components, using editing commands patterned after the text editor EMACS[11].

Image EMACS supports two classes of editing operations.

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1 Thanks to S. Bagley, P. Chou, G. Kopec and L. Wilcox for agreeing to have their work discussed here. This work represents a sample, rather than a full survey, of relevant speech and text-image research at PARC.

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The first class is based on viewing text as a linear sequence of characters, defined in terms of connected components. Traditional "cut and paste" functionality is enabled by a selection of insertion and deletion commands (e.g., delete-character, kill-line, yank region). As with text, editing is typically performed in the vicinity of a cursor, and operations to adjust cursor position are provided (e.g., forward-word, end-of-buffer). Characters can also be inserted by normal typing. This is accomplished by binding keyboard keys to character bitmaps of the character image bound to a given key against successive connected-component groupings allows for image-based character search.

The second class of operations supported by Image EMACS is based on viewing text as a two-dimensional arrangement of glyphs on an image plane. These operations provide typographic functionality, such as horizontal and vertical character placement, interword spacing, vertical line spacing, indentation, centering and line justification. Placement of adjacent characters is accomplished using font metrics estimated for each character directly from the image. These metrics allow for typographically acceptable character spacing, including character overlap where appropriate.

Taken together, Image EMACS commands are intended to convey the impression that the user is editing a text-based document. In actuality, the system is manipulating image components rather than character codes. Moreover, while the user is free to assign character labels to specific image components, editing, including both insertion and search, is accomplished without explicit knowledge of character identity. This approach enables interactive text-image editing and reproduction, independent of font or writing system.

Figures 1 and 2 show an example of a scanned multilingual document before and after editing with Image EMACS. The example demonstrates the results of image-based insertion, deletion, substitution and justification, as well as intermingling of text in several writing systems and languages (paragraphs 4 through 7). Such capabilities are potentially achievable using a format-conversion paradigm; however, this would require more sophisticated OCR functionality than currently exists, as well as access to fonts and stylistic information used in rendering the original document.

3. AUDIO EDITING AND INDEXING

A second example of signal-based document processing is provided by a wordspotter developed to support editing and indexing of documents which originate and are intended to remain in audio form. Examples include voice mail, dictated instructions and pre-recorded radio broadcasts or commentaries. The wordspotter can also be used to retrieve relevant portions of less structured audio, such as recorded lectures or telephone messages. In contrast with most previous wordspotting applications (e.g., [15, 10]), unconstrained keyword vocabularies are critical to such editing and indexing tasks.

The wordspotter is similar to Image Emacs in at least three ways: 1) it is based on partial modeling of signal content; 2) it requires user specification of keyword models; and 3) it makes no explicit use of linguistic knowledge during recognition, though users are free to assign interpretations to keywords. The wordspotter is also speaker or, more accurately, sound-source dependent. These constraints allow for vocabulary and language independence, as well as for the spotting of non-speech audio sounds.

The wordspotter is based on hidden Markov models (HMM's) and is trained in two stages. The first is a static stage, in which a short segment of the user's speech (typically 1 minute or less) is used to create a background, or non-keyword, HMM. The second stage of training is dynamic, in that keyword models are created while the system is in use. Model specification requires only a single repetition of a keyword and, thus, to the system user, is indistinguishable from keyword spotting. Spotting is performed using a HMM network consisting of a parallel connection of the background model and the appropriate keyword model. A forward-backward search is used to identify keyword start and end times, both of which are required to enable editing operations such as keyword insertion and deletion.

The audio editor is implemented on a Sun Microsystems Sparcstation and makes use of its standard audio hardware. A videotape demonstrating its use in several multilingual application scenarios is available from SIGGRAPH.

4. DOCUMENT IMAGE DECODING

Document image decoding (DID) is a framework for scanned document recognition which extends hidden Markov modeling concepts to two-dimensional image data. In analogy with the HMM approach to speech recognition, the decoding framework assumes a communication theory model based on three elements: an image generator, a noisy channel and an image decoder. The image generator consists of a message source, which generates a symbol string containing the information to be communicated, and an imager, which formats or encodes the message into an ideal bitmap. The channel transforms this ideal image into a noisy observed image by introducing distortions associated with printing and scanning. The decoder estimates the message from the observed image using maximum a posteriori decoding and a Viterbi-like decoding algorithm.
A key objective being pursued within the DID framework is the automatic generation of optimized decoders from explicit models of message source, imager and channel[3]. The goal is to enable application-oriented users to specify such models without sophisticated knowledge of image analysis and recognition techniques. It is intended that both the format and type of information returned by the document decoder be under the user's control. The basic approach is to support declarative specification of a priori document information (e.g., page layout, font metrics) and task constraints via formal stochastic grammars.

Figures 3 through 5 illustrate the application of DID to the extraction of subject headings, listing types, business names and telephone numbers from scanned yellow pages[7]. A slightly reduced version of a sample scanned yellow page column is shown in Figure 3 and a finite-state top-level source model in Figure 4. The yellow page column includes a subject heading and examples of several different listing types. These, in turn, are associated with branches of the source model. The full model contains more than 6000 branches and 1600 nodes. Figure 5 shows the result of using a decoder generated from the model to extract the desired information from the yellow page column. Automatically generated decoders have also been used to recognize a variety of other document types, including dictionary entries, musical notation and baseball box scores.

5. SUMMARY
The speech and text-image recognition initiatives discussed in the preceding sections illustrate two research themes at Xerox PARC which expand and redefine the role of recognition technology in document-oriented applications. These include the development of editors which operate directly on audio and scanned image data, and the use of speech and text-image recognition to retrieve arbitrary information from documents with signal content. Key concepts embodied in these research efforts include partial document models, task-oriented document recognition, user specification and interpretation of recognition models, and automatic generation of recognizers from declarative models. These concepts enable the realization of a broad range of signal-based document processing operations, including font, vocabulary and language-independent editing and retrieval.

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• The introduction of the new Area Codes...
  -- will not increase the cost of your calls,
  -- will not change your regular seven-digit telephone number.

• We are telling you now so that stationery purchases and reprogramming of equipment can be planned.

Para información en español sobre los cambios efectuados a los códigos de área 415/510 y 213/310, favor de llamar gratis a: Servicios Comerciales—811-2733, Servicios Residenciales Zona Norte—811-7730, Servicios Residenciales Zona Sur—811-5855

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Figure 1: Original scanned document image.

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• is running out of telephone profits in both the San Francisco Bay Area and the Los Angeles Area. So, we require all telephones in these areas to be replaced to offset the decline in our economic growth and the decreased demand for our telecommunications services.

• Also, you can anticipate that we...
  -- will increase the cost of your calls,
  -- will change your regular seven-digit telephone number.

• We are telling you now so that new telephone purchases and replacement of equipment can be planned.

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Figure 2: Scanned document image after Image EMACS editing.
Figure 3: Scanned yellow page column, slightly reduced.