Automated Metadata Extraction Services

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(Collaborators: Mohammad Zubair

and Steven Zeil)

Outline

- Problem and Challenges
- Approach
- Target Collections
- Architecture and Implementation
- Validation
- Conclusion

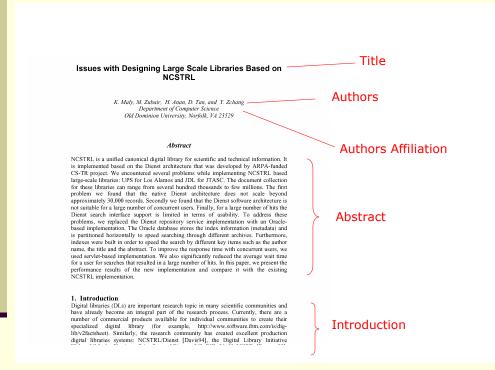
<u>Automated Metadata Extraction</u> for Large, <u>Diverse</u> and <u>Evolving</u> Document Collections

Automation?

The traditional method of creating metadata for large collections by librarians is prohibitively expensive and time consuming. According to one estimate, it would take about 60 employee-years to create metadata for 1 million documents.

Challenges:

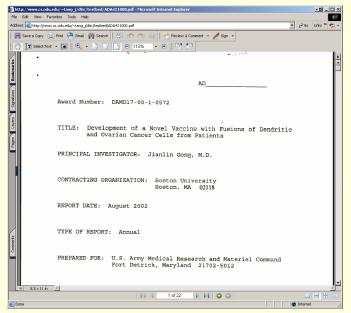
- Extracting metadata programmatically
- Diverse and evolving nature of a collection complicates metadata extraction

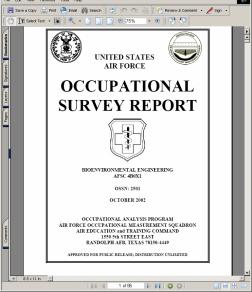


The visual cues in the formatting of the document along with the accumulated knowledge and intelligence makes it easy for a human to identify a string as a title, author, etc. !!

What about writing a computer program to do this task automatically?

State-of-the-art approaches for automated metadata extraction are restricted to documents with a common layout and structure. It is relatively easy to define set of rules or train machines for a homogeneous collection.





A number of federal organizations such as DTIC, GPO, and NASA manage heterogeneous collections consisting of documents with **diverse** layout and structure, where these programs do not work well.

Evolution:

Let us assume we have developed an approach that addresses diversity present in a given collection.

What happen to the approach when collection changes with time?

What happens to the approach when the collection changes over time?

- new types of documents added
- new layouts encountered

Robust:

A commercially viable process for metadata extraction must remain robust in the presence of external sources of error as well as in the face of the uncertainty that accompanies any attempts to automate "intelligent" behavior

Example source of error: Scanned documents with text obscured by smudges, signatures, or poor template

Approach

Existing automated metadata extraction approaches:

- Learning systems such as SVM, and HMM
 - Restricted to homogeneous collection
 - Problem with evolution: inertia to change until a significant number of examples of the new characteristics have been encountered.
- Rule-Based Systems
 - Heterogeneity can result in complex rule sets whose creation and testing can be very time-consuming. Complexity grows much more linearly in the number of rules.

Approach: Meeting the Challenges

Heterogeneity

- A new document is classified, assigning it to a group of documents of similar layout – reducing the problem to multiple homogeneous collections
- Associated with each class of document layouts is a template, a scripted description of how to associate blocks of text in the layout with metadata fields.

Approach: Meeting the Challenges

Evolution

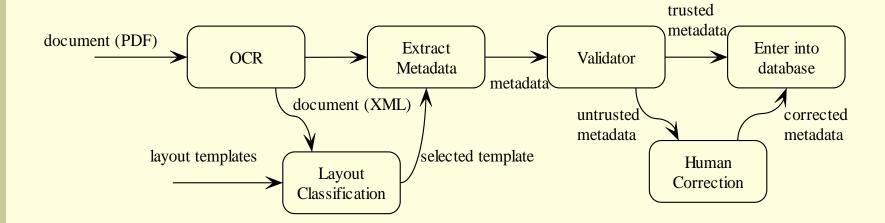
- New classes of documents accommodated by writing a new template
 - templates are comparatively simple, no lengthy retraining required
 - potentially rapid response to changes in collection

Approach: Meeting the Challenges

Robustness

 Use of Validation techniques to detect extraction problems and selection of templates

Approach



Target Collections

DTIC (Defense Technical Information Center) and NASA (National Aeronautics and Space Administration) Collections, Government Printing Office (GPO)

- Millions of Documents.
- Tens of thousands of new documents each year.
- Diverse: scientific articles, slides from presentations, PhD theses, (entire) conference proceedings, promotional brochures, public laws, and acts of Congress.
- Contributors: several organizations with their own in-house standards for layout and format.

Around 50% of the documents contain a Report Document Page (RDP)

 a standardized form that is inserted into the document when the document is added to the collection

Sample GPO Document – With Report Document Page

Extracted Metadata

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						multizone approach. The goal of the P tom hole sensitivity for 10" diameter t					
					compensation procedures. This report documents the results for 5", 10", and 14" diameter billets using calibration standards in a laboratory setting.						
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Sample DTIC Document – Without Report Document Page

Extracted Metadata

AU/ACSC/138/2000-04

AIR COMMAND AND STAFF COLLEGE

AIR UNIVERSITY

IDENTIFYING AND MITIGATING THE RISKS OF COCKPIT AUTOMATION

by

Wesley A. Olson, Major, USAF

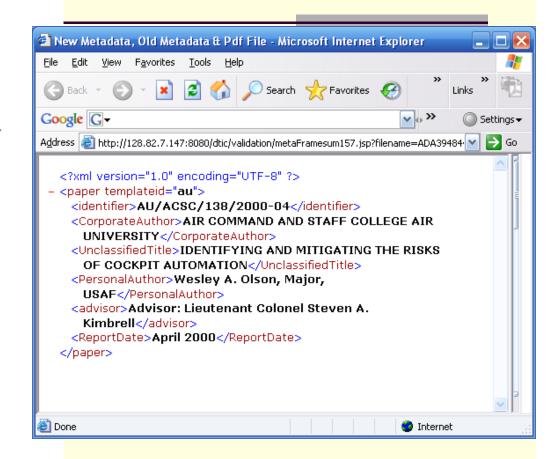
A Research Report Submitted to the Faculty

In Partial Fulfillment of the Graduation Requirements

Advisor: Lieutenant Colonel Steven A. Kimbrell

Maxwell Air Force Base, Alabama

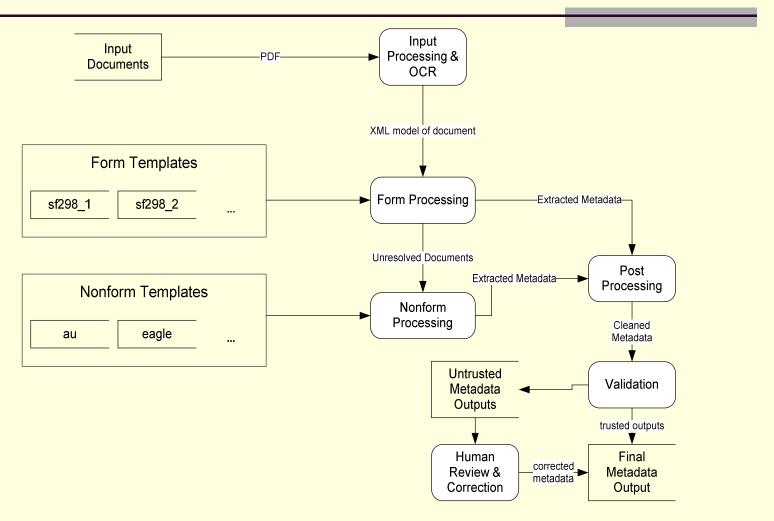
April 2000



November 23, 2010

Service Computation 2010 - Keynote - Lisbon, Portugal

Architecture & Implementation



Input Processing & OCR

- Select pages of interest from image PDF documents
 - First or last five pages of a document using the pdf toolkit (pdftk)
- Apply Off-The-Shelf OCR software
 - OmniPage Professional
- Convert OmniPage OCR output to Independent Document Model (IDM), a XML based format

Independent Document Model (IDM)

- Platform independent Document Model
- Motivation
 - Dramatic XML Schema Change between Omnipage 14 and 15
 - Tie the template engine to stable specification
 - Protects from linking directly to specific OCR product
 - Allows us to include statistics for enhanced feature usage
 - Statistics (i.e. avgDocFontSize, avgPageFontSize, wordCount, avgDocWordCount, etc..)

Documents in IDM

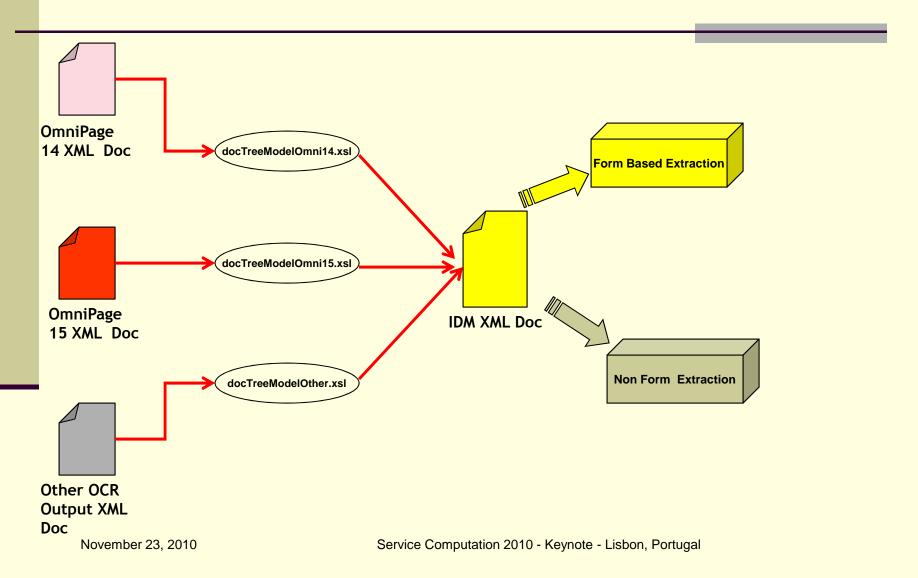
- A document consists of pages
- pages are divided into regions
- regions may be divided into
 - blocks of vertical whitespace
 - paragraphs
 - tables
 - images
- paragraphs are divided into lines
- lines are divided into words

All of these carry standard attributes for size, position, font, etc.

Generating IDM

- Use XSLT 2.0 stylesheets to transform
 - Supporting new OCR schema only requires generation of new XSLT stylesheet. -- Engine does not change

IDM Usage



Form Processing

- Scan document to identify one of six possible forms in the DTIC collection
 - Select form template
- Form extraction engine uses the template to extract metadata from IDM based document

If the form processor fails to match any template the document moves into the non-form extraction process

Sample Form-based template fragment

The (*line*) elements in the (*field*) elements define string matching criteria.

The (rule) elements defined for each (metadata) element defines the geometric placement.

Sample Form

REPORT D		Form Approved OMB No. 0704-0188		
The public reporting burden for this collection of igathering and maintaining the data needed, and co of information, including suggestions for reducting 10704-0188, 1215 Jefferson Davis Highway. Sin subject to any penalty for felling to comply with a PLEASE DO NOT RETURN YOUR FOR	Information is astimated to average 1 hour per responsing the properties of information. Sets the burden, to Department of Detense, Washington, VA 22202-4302. Respondents should be not information if it does not deplay a currently M TO THE ABOVE ADDRESS.	e, including the ti and comments region. Headquarters uild be aware that valid OMB contro	the for reviewing instructions, searching existing data source arding this burden estimate or any other aspect of this collection Services, Directorate for Information Operations and Report instrumental provision of law, no person shall be inumber.	
1. REPORT DATE (DD-MM-YYYY)	2. REPORT TYPE		3. DATES COVERED (From • To)	
18-09-2003	Final Report		1 April 1996 - 31 August 2003	
4. TITLE AND SUBTITLE		5a. CO	NTRACT NUMBER	
		F19628-96-C-0039		
VALIDATION OF IONOSPI	HERIC MODELS	5b. GRANT NUMBER		
		5c. PROGRAM ELEMENT NUMBER		
	Å	61102F		
6. AUTHOR(S)		5d. PROJECT NUMBER		
			1010	
Patricia H. Doherty		5e. TASK NUMBER		
Leo F. McNamara		IM		
Susan H. Delay		2117		
Neil J. Grossbard		5f. WORK UNIT NUMBER		
			AC	
7. PERFORMING ORGANIZATION NAT	ME(S) AND ADDRESS(ES)		8. PERFORMING ORGANIZATION	
Boston College / Institute for S	Scientific Research		REPORT NUMBER	
140 Commonwealth Avenue				
Chestnut Hill, MA 02467-386	2			
9. SPONSORING/MONITORING AGEN	CY NAME(S) AND ADDRESS(ES)		10. SPONSOR/MONITOR'S ACRONYM(S)	

Sample Form (cont.)

9. SPONSORING/MONITORING AGENCY NAME(S) AND ADDRESS(ES)

Air Force Research Laboratory
29 Randolph Road
Hanscom AFB, MA 01731-3010

11. SPONSOR/MONITOR'S REPORT
NUMBER(S)

AFRL-VS-TR-2003-1610

12. DISTRIBUTION/AVAILABILITY STATEMENT
Approved for public release; distribution unlimited.

13. SUPPLEMENTARY NOTES

14. ABSTRACT

This document represents the final report for work performed under the Boston College contract F19628-96C-0039. This contract was entitled Validation of Ionospheric Models. The objective of this contract was to obtain satellite and ground-based ionospheric measurements from a wide range of geographic locations and to utilize the resulting databases to validate the theoretical ionospheric models that are the basis of the Parameterized Real-time Ionospheric Specification Model (PRISM) and the Ionospheric Forecast Model (IFM). Thus our various efforts can be categorized as either observational databases or modeling studies.

15. SUBJECT TERMS

Ionosphere, Total Electron Content (TEC), Scintillation, Electron density, Parameterized Real-time Ionospheric Specification Model (PRISM), Ionospheric Forecast Model (IFM), Paramaterized Ionosphere Model (PIM), Global Positioning System (GPS)

16. SECURITY CLASSIFICATION OF:				18. NUMBER	19a. NAME OF RESPONSIBLE PERSON	
a. REPORT	b. ABSTRACT	c. THIS PAGE	ABSTRACT	PAGES	John Retterer	
7.7	77	**	SAR	PAGES	19b. TELEPHONE NUMBER (Include area code)	
· U	U	U		1	781-377-3891	

Standard Form 298 (Rev. 8/98) Prescribed by ANSI Std. 239.18

Metadata Extracted from Sample RDP (1/3)

- <metadata templateName="sf298_2">
- <ReportDate>18-09-2003</ReportDate>
- <DescriptiveNote>Final Report/DescriptiveNote>
- <DescriptiveNote>1 April 1996 31 August 2003
- <UnclassifiedTitle>VALIDATION OF IONOSPHERIC MODELS
- <ContractNumber>F19628-96-C-0039</ContractNumber>
 - <ContractNumber></ContractNumber>
- <ProgramElementNumber>61102F</ProgramElementNumber>
- <PersonalAuthor>Patricia H. Doherty Leo F. McNamara
 - Susan H. Delay Neil J. Grossbard</PersonalAuthor>
- <ProjectNumber>1010</ProjectNumber>
- <TaskNumber>IM</TaskNumber>
- <WorkUnitNumber>AC</WorkUnitNumber>
- <CorporateAuthor>Boston College / Institute for Scientific Research 140 Commonwealth Avenue Chestnut Hill, MA 02467-3862

Metadata Extracted from Sample RDP (2/3)

- <ReportNumber></ReportNumber>
 - <MonitorNameAndAddress>Air Force Research Laboratory 29 Randolph Road Hanscom AFB, MA 01731-3010
 - <MonitorAcronym>VSBP</MonitorAcronym>
 - <MonitorSeries>AFRL-VS-TR-2003-1610
- <DistributionStatement>Approved for public release; distribution
 unlimited.</DistributionStatement>
- <Abstract>This document represents the final report for work performed under the Boston College contract F I9628-96C-0039. This contract was entitled Validation of Ionospheric Models. The objective of this contract was to obtain satellite and ground-based ionospheric measurements from a wide range of geographic locations and to utilize the resulting databases to validate the theoretical ionospheric models that are the basis of the Parameterized Real-time Ionospheric Specification Model (PRISM) and the Ionospheric Forecast Model (IFM). Thus our various efforts can be categorized as either observational databases or modeling studies.

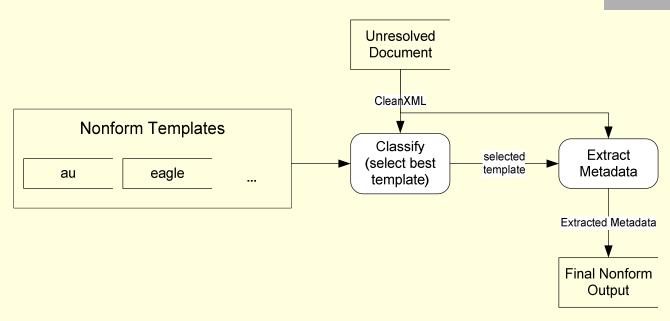
Metadata Extracted from Sample RDP (3/3)

- <Identifier>Ionosphere, Total Electron Content (TEC), Scintillation, Electron density, Parameterized Real-time Ionospheric Specification Model (PRISM), Ionospheric Forecast Model (IFM), Paramaterized Ionosphere Model (PIM), Global Positioning System (GPS)
 - <ResponsiblePerson>John Retterer</ResponsiblePerson>
 - <Phone>781-377-3891</Phone>
- <ReportClassification>U</ReportClassification>
 <AbstractClassification>U</AbstractClassification>
- <AbstractLimitaion>SAR</AbstractLimitaion>
- </metadata>

Non-Form Processing

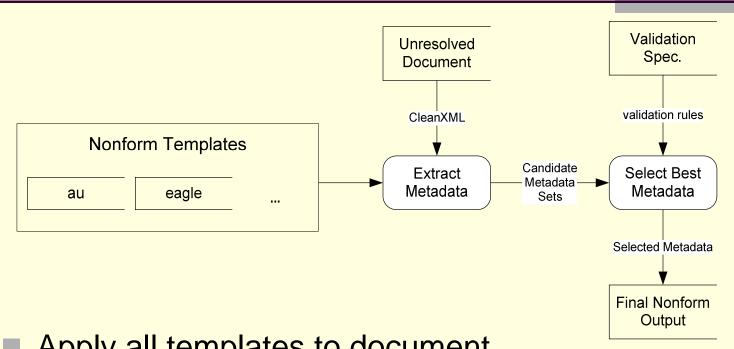
- Classification compare document against known document layouts
 - Select template written for closest matching layout
- Apply non-form extraction engine to document and template

Classification (a priori)



- Previously, we had attempted various schemes for a priori classification
 - x-y trees, bin classification
- Still investigating some
 - image-based recognition

Post-Hoc Classification



- Apply all templates to document
 - results in multiple candidate sets of metadata
- Score each candidate using the validator
 - Select the best-scoring set

Statistical Validation of Extracted Metadata

APPROACH

Measure a relevant property, such as length of the title, and compare it to values known to be typical of documents already in that collection.

The comparison results are quantified and normalized into a confidence value for the value extracted for that metadata field.

Field	Avg.	Std. Dev.	
UnclassifiedTitle	9.9	4.8	
Abstract	114	58	
PersonalAuthor	2.8	0.5	
CorporateAuthor	7	2.3	

Field Length (in words) for DTIC collection

Approach - Reference Models

We need to build models for different metadata fields to support our statistical based validation approach

- From previously extracted metadata
 - specific to document collection
- Phrase dictionaries constructed for fields with specialized vocabularies
 - e.g., author, organization
- Statistics collected
 - mean and standard deviation
 - permits detection of outputs that are significantly different from collection norms

Statistics Collected

- Field length statistics
 - title, abstract, author,...
- Phrase recurrence rates for fields with specialized vocabularies
 - author and organization
- Dictionary detection rates for words in natural language fields
 - abstract, title,..

Sample statistics collected for authors and organization names (corporate authors)

Statistics based on 800,000 human extracted metadata records

Phrase dictionaries are constructed for all phrases of length 1-4 words over a randomly selected set of 600,000 of metadata records. The remaining 200,000 are used to compute the average and standard deviation of the percentage of phrases in each field that are recorded in the phrase dictionary

Field	Phrase	Avg.	Std. Dev.	
	length			
	1	97%	11%	
Personal author	2	83%	32%	
	3	71%	45%	
	1	100%	2%	
CorporateAuthor	2	99%	6%	
	3	99%	10%	
	4	98%	13%	

Phrase Dictionary Hit Percentage, DTIC collection

Validation Process

- Extracted outputs for fields are subjected to a variety of tests
 - Test results are *normalized* to obtain confidence value in range 0.0-1.0
- Test results for same field are combined to form field confidence
- Field confidences are combined to form overall confidence

Validation Tests

- Deterministic
 - Regular patterns such as date, report numbers
- Probabilistic
 - Length: if value of metadata is close to average -> high score
 - Vocabulary: recurrence rate according to field's phrase dictionary
 - Dictionary: detection rate of words in English dictionary

Combining results

- Validation specification describes
 - which tests to apply to which fields
 - how to combine field tests into field confidence
 - how to combine field confidences into overall confidence

Fragment for Validation Specification for DTIC Collection

```
<?xml version="1.0"?>
<val:validate collection="dtic" xmlns:val="jelly:edu.odu.cs...">
 <val:average>
  <val:field name="UnclassifiedTitle">
    <val:average>
      <val:dictionary/>
      <val:length/>
    </val:average>
  </val:field>
  <val:field name="PersonalAuthor">
    <val:min>
     <val:length/>
      <val:max>
       <val:phrases length="1"/> <val:phrases</pre>
                                                   length="2"/> <val:phrases length="3"/>
     </val:max>
    </val:min>
  </val:field>
```

Experimental Design

- How effective is post-hoc classification?
- Selected several hundred documents recently added to DTIC collection
 - Visually classified by humans,
 - comparing to 4 most common layouts from studies of earlier documents
 - discarded documents not in one of those classes
 - 167 documents remained
- Applied all templates, validated extracted metadata, selected highest confidence as the validator's choice
- Compared validator's preferred layout to human choices

Automatic vs. Human Classifications

Manually Assigned Classes	Validator au	Validator eagle	Validator rand	Validator title	Total Manual
au	86	0	0	0	86
eagle	0	8	33	4	45
rand	0	0	8	4	12
title	0	0	1	23	24

- Post-hoc classifier agreed with human on 74% of cases
- Most disagreements were due to "extra" words in extracted metadata (e.g., military ranks in author names) - highlights need for post-processing to clean up metadata
- In our simulated study of post-processing, the agreement between post-hoc classifier and human classification rose to 99%

Extracting Metadata Using Non-Form Extraction Engine

- Transform IDM based document into another XML format called CleanML, which encodes the paragraphs and lines and their corresponding features into an XML structure (this is to support our current simple non-form engine implementation)
- The non-form extraction engine also uses rule-based template extraction to locate and extract metadata. Each template contains a set of rules designed to extract metadata from a single class of similar documents.

Non-Form Template Fragment

- Each desired metadata item is described by a rule set designating the beginning and the end of the metadata.
- The rules are limited by features detectable at the line level resolution. We hope to address this deficiency in future versions.

Non-Form Template Fragment

Non-Form Sample (1/2)

AU/ACSC/012/1999-04

AIR COMMAND AND STAFF COLLEGE

AIR UNIVERSITY

INTEGRATING COMMERCIAL ELECTRONIC EQUIPMENT
TO IMPROVE MILITARY CAPABILITIES

by

Non-Form Sample (2/2)

by

Jeffrey A. Bohler LCDR, USN

A Research Report Submitted to the Faculty

In Partial Fulfillment of the Graduation Requirements

Advisor: CDR Albert L. St.Clair

Maxwell Air Force Base, Alabama

April 1999

Metadata Extracted From the Title Page of the Sample Document

```
<paper templateid="au">
    <identifier>AU/ACSC/012/1999-04</identifier>
    <CorporateAuthor>AIR COMMAND AND STAFF COLLEGE
    AIR UNIVERSITY</CorporateAuthor>
    <UnclassifiedTitle>INTEGRATING COMMERCIAL
    ELECTRONIC EQUIPMENT TO IMPROVE
    MILITARY CAPABILITIES
    </UnclassifiedTitle>
    <PersonalAuthor>Jeffrey A. Bohler LCDR, USN</PersonalAuthor>
    <advisor>Advisor: CDR Albert L. St.Clair</advisor>
    <ReportDate>April 1999</ReportDate>
```

Experimental Results

- DTIC & NASA Testbed: Downloaded 9825 documents from the DTIC collection and 728 from the NASA collection.
- The internal distribution between forms and non-form documents for the collections are 94% RDP forms for DTIC and 21% RDP for NASA.

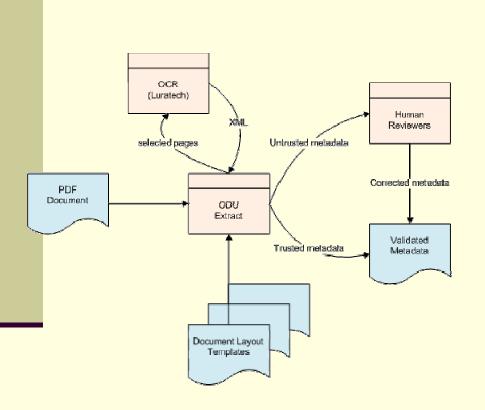
Experimental Results

- Form based extraction
 - Wrote six form based templates
 - The overall accuracy for the for the Form based extraction was close to 99%
- Non-Form based extraction
 - Wrote 11 templates
 - The overall accuracy was 66% for DTIC and 64% for NASA. (Lower values is mostly due to the limited number of templates we were using. Assuming that we write all the necessary templates, we expect accuracy in the 85% range.

Status

- Completely Automated Software for:
 - Drop in pdf file (either text or image(scanned))
 - No OCR necessary for text pdf
 - Process and produce output metadata in XML format
 - Hooks for human intervention and correction
- Completed software for documents with RDP
- Default set of templates for:
 - RDP containing documents (about 10 templates, accuracy 94%)
 - Non-form documents (about 100 templates, cover 50 % of documents, accuracy 80%)
- Statistical models of DTIC collection(800,000 documents)
 - Phrase dictionaries: personal authors, corporate authors
 - Length and English word presence for title and abstract
 - Structure of dates, report numbers

Summary – the ODU Extract System



- Documents (PDF) as input
- Invokes OCR as required
- Applies templates to find and extract metadata
- Validates its own outputs
 - Untrusted metadata outputs are diverted for human review

Summary: Elements of technology

- Partitioning the documents into classes based on visual similarity of metadata-containing pages
- Developing templates for each class, i.e., mechanisms to describe the location of metadata and how to extract them
- Develop a validation process that validates the extracted metadata against a statistical model of the collection and scores the output against the best fitting class
- Develop an automated dataflow model that starts with edocuments and ends with citations in XML or other libraryspecific formats; it should handle text pdfs directly and image pdf through first OCRing them
- Develop a human intervention process that will validate system suggestion of potential problems and correct them.

Summary: Advantages of technology

- Reduces time spent by humans for creating citation
 - Very significantly (greater than 90%) for documents with RDP
 - Very significantly (greater than 90%) for successful, correct extraction
 - Significantly (from 10%-90%) for successful extraction with warnings of low score
 - Significantly (from 10%-90%) for successful, partially incorrect extraction
 - None for unsuccessful extraction

Special Features of ODU System

Post-processing

- Author extraction conforms to DTIC cataloguing standards
 - e.g., "Capt. John James Smith, ASAF" is extracted as "Smith, John J"
- Report dates and Descriptive notes converted as per DTIC cataloguing standards
 - will resolve ambiguous dates where possible
- Distribution and Classification statements standardized
- Special handling for long abstracts, metadata across multiple pages

Dynamic Validation

ODU Extract evaluates its own output, scores it, and flags suspicious outputs for later human inspection,

- catches many instances of software errors, OCR errors, and human data entry errors.
- Statistical models of DTIC collection(800,000 documents)
 - Phrase dictionaries: personal authors, corporate authors
 - Length and English word presence for title and abstract
- Structure of dates, report numbers

Conclusions

- Automated metadata extraction can be performed effectively on a wide variety of documents
 - Coping with heterogeneous collections is a major challenge
- Much attention must be paid to "support" issues
 - validation, post-processing, etc.

Conclusions

- Creating statistical model of existing metadata is very useful tool to validate extracted metadata from new documents
- Validation can be used to classify documents and select the right template for the automated extraction process