

A3-01 2002

#### **Séminaire CNRT:**

#### Text Indexing for multimedia Seminar

|   | ,                             |  |
|---|-------------------------------|--|
| 9:00-9:15   | Welcome                       | Philippe Letellier - Thomson multimedia Catherine Pierre-Radenac – CNRT                                |
|   | 9:15 a                        | m - 12:15 pm   |
| 9:15 am   | Context and needs for text    | ual indexing<br>Louis Chevallier, Nour-Eddine Tazine<br>Thomson multimedia                             |
| The si  | de of Content Owners          | ISTI<br>BIBLIOTECA<br>AR MILLO   |
| 10:00 am  | The value of text indexing i  | n the BBC Colloc A3-01 Chris Wilkie - BBC  |
| Text cl   | assification                  |  |
| 11:15 am  | Automated text categorizat    | ion: tools, techniques and applications<br>Fabrizio Sebastiani<br>Italian National Council of Research |
| 12:15 am  | Lunch                         |  |
| · All The Control of | 2:0                           | 0-5:30 pm ———————————————————————————————————  |
| 2:00 pm   | Automatic topic characteriz   | ation and detection for text classification Pascale Sébillot – <i>Irisa</i>                            |
| Linguis   | tic resources construction    |  |
| 2:45 pm   | Information extraction: fron  | n unstructured texts to knowledge database<br>Thierry Poibeau - <i>Thales</i>                          |
| Comme   | rcial tools                   |  |
| 3:45 pm   | Automatic text categorization | on tools in practice<br>Peter de Bie - <i>Inxight</i>  |
| 1:30 pm   | Linguistic knowledge and d    | ocument processing   |

Claude de Loupy - Sinequa

Centre National de Recherche Technologique
Text Indexing Seminar
Rennes, France – April 3, 2002

Automated Text Categorization: Tools, Techniques and Applications

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secret treasure. There was no personal or world problem whose eloquent solution did not exist in some hexagon. All men felt themselves to be the masters of an intact and books, the first impression was one of extravagant happiness. When it was proclaimed that the Library contained all

 $(\dots)$ 

were inaccessible, seemed almost intolerable. hexagon held precious books and that these precious books excessive depression. The certitude that some shelf in some As was natural, this inordinate hope was followed by an

[Jorge Luis Borges, The Library of Babel, 1941]

# Tackling Information Overload by Text Categorization

Nowadays, there are two main paradigms for tackling information overload

- 1. build high-quality tools for searching an unstructured document base, such as the Web. This is the "standard" answer from text search
- 2. build high-quality tools for structuring the document base, e.g. into a digital library. This is the answer from automated text categorization (ATC)

It is the latter approach that we will concentrate on in this talk.

### Overview of this talk

- 1. A definition of the ATC task
- 2. Learning and evaluating text classifiers
- 3. Applications of ATC
- 4. Applications of ATC at IEI-CNR
- (1) Automated indexing of scientific articles under hierarchical classification schemes
- (2) Personalized information delivery to digital library users
- (3) Automated construction of thematic lexicons
- (4) Automated survey coding
- 5. Conclusion

## 1. A definition of the ATC task

called the classifier, such that  $\Psi$  and  $\Phi$  "coincide as much as possible". Here  $\Psi: \mathcal{D} \times \mathcal{C} \to \{T, F\}$  by means of a function  $\Phi: \mathcal{D} \times \mathcal{C} \to \{T, F\}$ ATC is the task of approximating the unknown | target function

- $\mathcal{C} = \{c_1, \dots, c_{|\mathcal{C}|}\}$  is a fixed set of pre-defined | categories |
- $\mathcal{D}$  is a domain of documents.

Depending on the application, classification may be

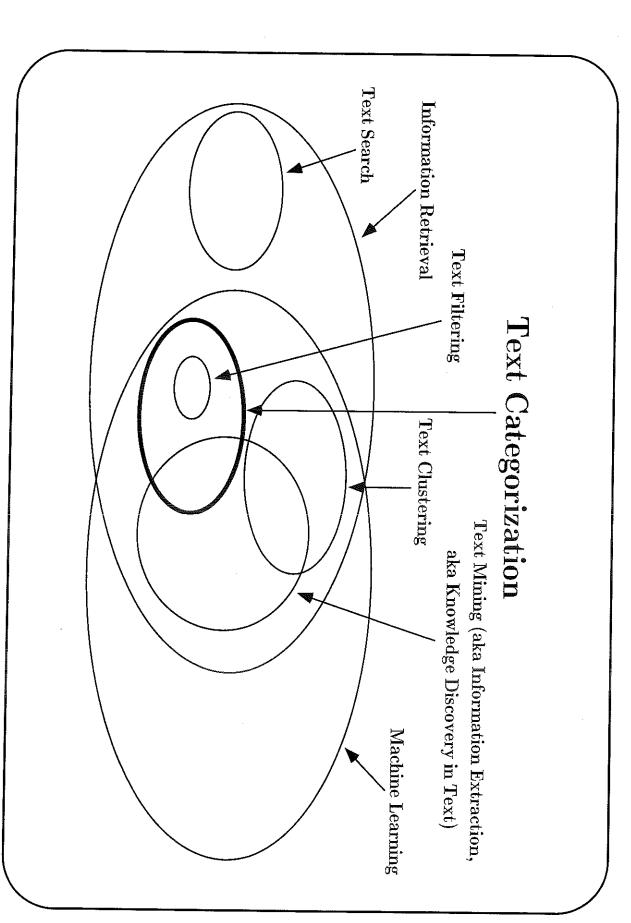
- document. A special case is when  $|\mathcal{C}| = 2$  (the | binary | case). single-label: exactly one category must be assigned to each
- document multi-label: any number of categories can be assigned to each

In general it is the case that:

- the categories are just symbolic labels. No additional knowledge of their "meaning" is available to help in building the classifier; in particular, the "text" constituting the label is not significant;
- the attribution of documents to categories should be realized on metadata that may be available from an external source. the basis of the *content* of the documents, and not on the basis of

a document in a category, cannot be decided deterministically. means that the fundamental notion of ATC, that of | membership | of Given that the content of a document is a subjective notion, this

be verified, and one uses whatever source of knowledge is available. In an operational environment the two assumptions above may not



# 2. Learning and evaluating text classifiers

A binary classifier for class c, i.e. one that decides between c and  $(\mathbf{not}\ c)$ 

- is built automatically, by | machine learning | techniques, from a "training" set of documents preclassified under c and  $(\mathbf{not}\ c)$
- $(\mathbf{not}\ c)$ is tested by comparing its decisions with the human decisions encoded in a "test" set of documents preclassified under c and

Learning techniques often used in ATC are

- "legacy" techniques: decision trees, decision rules, probabilistic (Bayes) classifiers, neural networks, etc.
- "emerging" techniques: boosting, support vector machines.

least once in the training set. where the length of the vector is the number of terms that occur at Usually, a document is represented as a (sparse) vector of weights,

contributes to the semantics of the document). In this latter case, the document), or non-binary (indicating how much the term Weights may be binary (indicating presence or absence of the term in weighting | functions from IR (such as tf \* idf) are used

also be performed in order to improve the efficiency of the system. performed. Dimensionality reduction (aka | feature selection |) may Previous to weighting, stop word removal and stemming are often

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automatic) classifier must be measured by a combination of classifiers are error-prone. The effectiveness of a (human or Classification is a subjective task, and both human and automatic

recall: "How many documents truly belonging to the category have been deemed as such"?

category truly belong to it"? precision: "How many documents deemed to belong to the

human classifiers. The effectiveness of automatically built classifiers now rivals that of

## ATC research at IEI-CNR. Methods

including We have investigated a number of methods and issues in ATC,

- hypertext classification through "blurb indexing" [Attardi et al., THAI'99];
- novel "boosting" algorithms for learning text classifiers [Sebastiani et al., ACM CIKM'00];
- novel "feature selection" techniques for optimizing document representations for ATC [Galavotti et al., ECDL'00];
- "bigram indexing" for ATC [Caropreso et al., TD'01].

### 3. Applications of ATC

Historically, the most important applications of ATC are:

- "descriptors", ...) from e.g. i.e. classifying documents with categories (or "subject codes", or automated document indexing with controlled vocabularies
- the Library of Congress Cataloging Scheme;
- Dewey Decimal System.

This is a form of automated metadata generation.

whether the document belongs to  $User_i$  or  $(\mathbf{not}\ User_i)$ documents to the interested users only by deciding, for  $User_i$ , personalized information delivery |, i.e. routing a stream of

### Other applications of ATC have been:

- filing patents under predefined patent categories
- filing "classified ads" into classes (e.g. deciding whether a given
- filing Web sites, or organizing search results, under Yahoo!-like hierarchical Web directories ad should be printed under Cars for Sale, or Real Estate, ...)
- filtering unsuitable content (e.g. deciding between Pornography and (not Pornography)) or junk mail (Spam vs. (not Spam))
- detecting authorship of documents of disputed paternity (e.g. Shakespeare vs. (not Shakespeare))
- classifying images through the analysis of textual captions
- automatically identifying text genre (e.g. For Kids  $_{
  m VS.}$  (not ForKids))

# ATC research at IEI-CNR. Applications

We have tackled a number of applications of ATC, including

- classification schemes Automated indexing of scientific articles under hierarchical
- Personalized information delivery to digital library users
- Automated construction of thematic lexicons
- Automated survey coding

Applications we intend to tackle in the near future are

- Automated authorship attribution
- Spoken text segmentation and categorization, with topic detection and tracking

#### (1) Automated indexing of scientific articles under hierarchical classification schemes 4. ATC applications at IEI-CNR

characteristics are: building digital libraries in the computer science domain. Its main and concerned with generating a classifier of scientific articles for CompCat is an internally funded project now starting at IEI-CNR the categories are the ones from the ACM Classification Scheme

- (version of 1998)
- the training and test sets are years from 1998 onwards of the ACM Digital Library
- the classifier will be interactive, i.e. will suggest to the user a appropriateness for the document list of categories ranked according to their estimated

solution is problematic since it is c, train a binary classifier that decides between c and (**not** c). This "Trivial" solution to the problem: for each (internal of leaf) category

- this means invoking thousands of classifiers for each document inefficient |: since a document may belong to multiple categories,
- documents in the "leaf" categories thousands of categories, and there may thus be very few training structure of the category set is not exploited. There are ineffective: the information provided by the hierarchical

insights into hierarchical text categorization well-known from the In this work (joint with H. Avancini and A. Rauber) we exploit literature:

- we bring to bear the hierarchical structure of the classifier by "children" of a given node), thus enhancing efficiency; "soft pruning" (i.e. we consider only the k most promising
- we make term statistics more robust by "shrinkage", thus enhancing effectiveness for categories at the lower levels in the hierarchy.

To these intuitions, we add a novel idea, i.e.

- discarding the naturally occurring hierarchy
- generating an artificial one by means of a hierarchical agglomerative clustering technique
- using this latter instead of the naturally occurring one for learning the final classifiers

statistics-based algorithm such as shrinkage can profit more from the former are indeed advantageous for human understanding, a latter than the former. result of machine (statistics-based) conceptualization. While the conceptualization, while the ones generated by clustering are the naturally occurring hierarchical structures are a result of human This apparently counterintuitive idea relies on the observation that

### (2) Personalized information delivery to DL users ATC applications at IEI-CNR

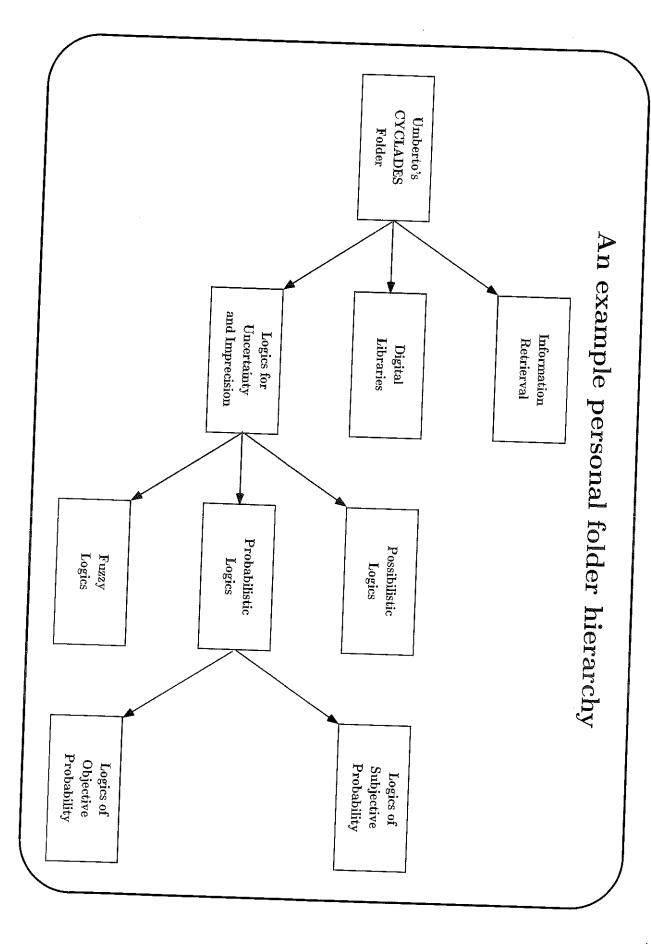
with the Open Archives initiative (OAi) standard. services for users of a distributed DL of grey literature compliant 2001 and coordinated by IEI-CNR, aiming to provide a layer of CYCLADES is a CEC-funded project (IST-2000-25456), started Feb

and U. Straccia) by means of ATC techniques. personalized way. We are building this (joint work with E. Renda allowing a user to interact with the system in a flexible and highly Key to Cyclades is the Personalization Service (PS), responsible for

interaction with the user must take into account the user's interests. Personalization is viewed as a | content-based notion |; i.e. the

modality) from the DL; A CYCLADES user will obtain documents (in either "push" or "pull"

- each document will be | classified | into one or more of a the user's subjective view of a topic which is of interest to her hierarchically structured set of folders, each of which represents
- each user action (e.g. moving, removing, printing a document) the folders, and used to revise the classifiers. will be be interpreted as "implicit feedback" on the meaning of



In addition we intend will be exploited also here (e.g. "shrinkage", "soft pruning", ...). sparseness), which means that some of the solutions adopted there thousands of categories, and "leaf" categories may suffer from data one (e.g. the category set is hierarchically structured; there are This application has meany features in common with the previous

to bring to bear implicit user feedback by adopting an effectiveness for data-sparse categories by exploiting the dynamic "incremental" classifier (BALANCEDWINNOW), thus enhancing

character of this application;

• to bring to bear not only document content, but also | ratings content-based) recommendation. current user. This yields a form of hybrid (i.e. ratings- and given to the document by other users deemed "similar" to the

### (3) Automated construction of thematic lexicons ATC applications at IEI-CNR

thematic lexicons. task, i.e. | term categorization | , for the automated construction of This work (joint with A. Lavelli and B. Magnini) proposes a novel

of classifying terms (into zero, one, or several categories belonging to a predefined set) represented as vectors in a space of documents. vectors in a space of terms, the purpose of term categorization is that purpose of ATC is that of classifying documents represented as The main idea is to apply ATC techniques in a dual way: while the

preclassified documents. Our technique needs instead Previous techniques for learning thematic lexicons need a set of

- discrimination purposes (e.g. ZOOLOGY, SPORT, BASKETBALL) a small set of preclassified terms. For this, we have used the most popular labels used in dictionaries for sense term has been labelled with one or more from a set of 238 among WordNetDomains, an extension of WordNet in which each
- a corpus of unlabelled documents. For this we have used various subsets of the Reuters Corpus Volume I (RCVI).

terms and new terms. from the documents the associations between previously contained Our technique extends the existing thematic lexicons by learning

### ATC applications at IEI-CNR (4) Automated survey coding

This work (joint with D. Giorgetti and I. Prodanof) proposes that automated survey coding (ASC) be viewed as an ATC task. ASC

- is the task of analyzing the answers that a person has given to an on the answers. open-ended questionnaire (e.g. a social survey) and "classifying" this person into one among a predefined set of categories, based
- has many important applications, such as filing resumes into categories, classifying respondents to social surveys, etc.
- had previously been viewed in terms of similarity matching enough, and is often not available. This is unsuitable, since this description is usually not useful between the answer and a textual description of the category.

Research Center (NORC): Example from a social survey conducted by the US National Opinion

way what the situation was? describe in a couple of sentences what made you feel that you felt really angry, irritated or annoyed. Could you Question: Within the past month, think about the last time

### Applicable Categories:

- ANGRYWRK: Situation involved work;
- ANGRYFAM: Situation involved family;
- ANGRYGVT: Situation involved government or government officials;
- :
- OTHER: Situation did not fit the above categories.

#### Conclusion

### Nowadays ATC is considered

- essential for either tackling or supporting any application involving automated indexing with controlled vocabulary or
- essential for tackling any application involving timely personalized information delivery

automated metadata generation

to the categorization of documents expressed in slightly different media; for instance: its methods and techniques to neighbouring fields of application. Techniques typical of ATC have already been extended successfully The success story of ATC is also going to encourage an extension of

- very noisy text resulting from optical character recognition.
- speech transcripts.

# And for those of you interested in ATC ...

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- "Machine Learning in Automated Text Categorization", slides from http://faure.iei.pi.cnr.it/~fabrizio/Slides/ATCslides.pdf the ECDL'01 (and probably ECDL'02) tutorial, at
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