Intelligent Information Filtering

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Outline

INTRODUCTION

- The Information Overloading Problem
- Personalization on the Web
- INFORMATION SEEKING STRATEGIES
- **3** INFORMATION FILTERING
 - Collaborative Filtering
 - Content-based Filtering & User Profiling
- **4** IDEAS FOR INTELLIGENT INFORMATION FILTERING
- SCONCLUDING REMARKS
- 6 LIVE DEMO

People across the world...

Chat

- Exchange e-mail, sms, pictures (mms)
- Buy products and services online
- Use search engines to find information their work and day-to-day life
- Exploit the Web for obtaining informat conventional sources like books, maga libraries



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- Exploit the Web for obtaining information instead of conventional sources like books, magazines and libraries



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- Chat
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- Buy products and services online
- Use search engines to find information useful in their work and day-to-day life
- Exploit the Web for obtaining information instead of



Web	Images	Groups	<u>News</u>	<u>Froogle</u>	Local New!	more	<u>))</u>
Fo	Advanced Search Preferences						
	Google Search		h)	I'm Feeling Lucky			Language Tools

- Chat
- Exchange e-mail, sms, pictures
- Buy products and services online
- Use search engines to find infor their work and day-to-day life
- Exploit the Web (instead of conventional sources like books, magazines and libraries) for obtaining information



Information Overload

Problems...

- Explosion of irrelevant, unclear, inaccurate information
- Users overloaded with a large amount of information impossible to absorb



...and consequences

- Searching is time consuming
- Need for intelligent solutions able to support users in finding documents according to their interests



How Much Information?

- Print, film, magnetic and optical storage media produced between 3.4 and 5.4 exabytes of unique information in 2002
 - ✓ 92% stored on magnetic media, mostly hard disks
 - ✓ 500-800 MB per person each year
- Information flow through electronic channels telephone, radio, TV, and the Internet – contained almost 18 exabytes of new information in 2002 (3 ½ more than recorded in storage media), 281 exabytes in 2007 (45 GB per person)

Medium	Terabytes	Medium	Terabytes
Radio	3,488	Surface Web	167
Television	68,955	Hidden Web	91,850
Telephone	17,300,000	E-mail	440,606
Internet	532,897	Inst. Messaging	274
TOTAL	17,905,340	TOTAL	532,897

Lyman, Peter and Hal R. Varian, *How Much Information*, School of Information Management and Systems, University of California at Berkeley, 2003. URL: http://www.sims.berkeley.edu/how-much-info-2003. Last access: May 23rd, 2007.

How to defend oneself from Information Overloading

Se si vuole trovare una <u>metafora</u> del <u>rapporto fra l'uomo e i mezzi di</u> <u>comunicazione</u>, Umberto Eco suggerisce quella dell'<u>automobilista</u>:

la tecnologia ha messo a disposizione vetture sempre più sofisticate, potenti e veloci; che vengano usate per portare una persona all'ospedale o per fare le gare di velocità sulle strade, dipende da chi è seduto al posto di guida.

Lo stesso si può dire di quella che ormai è diventata una delle relazioni fondamentali della nostra vita quotidiana, cioè il nostro modo di interagire con i mass media, dalla televisione al telefonino, da Internet alla radio, dai libri ai cd e dvd (ebbene sì, anch'essi sono media), alla posta elettronica: dipende dalla cultura e dalla volontà di ciascuno di noi, educato soprattutto dalla scuola e dalla famiglia, mettere a punto una "<u>dieta mediatica</u>"

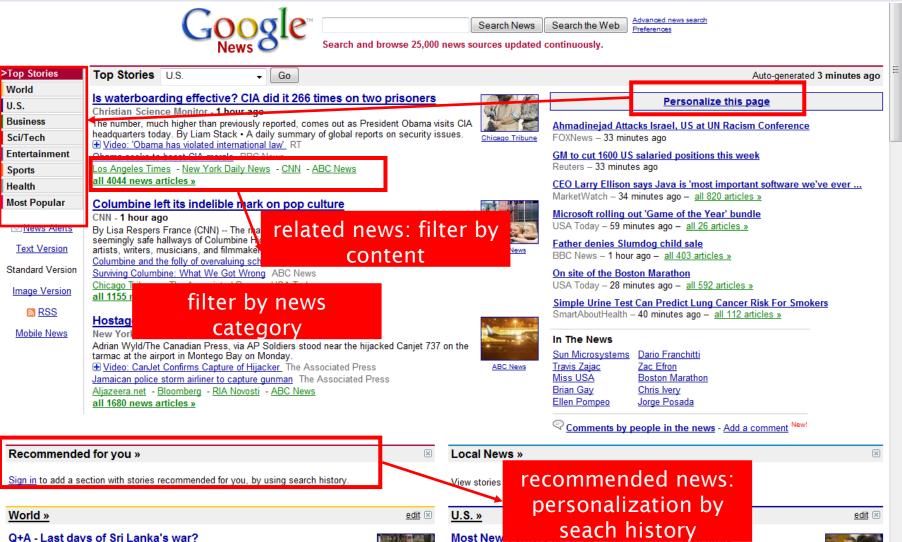
– suggerisce Gianfranco Bettetini – che non provochi né obesità né anoressia.

da: "Mettete a dieta i mass-media"

INTERVISTA A DUE VOCI CON UMBERTO ECO E GIANFRANCO BETTETINI

di Paolo Perazzolo, Famiglia Cristiana n.20 del 20-05-2007 (http://www.sanpaolo.org/fc/0720fc/0720fc54.htm)

How to defend oneself from Information Overload: My... Web - Google News



Q+A - Last days of Sri Lanka's war?

Reuters - 1 hour ago

U.S.

By C. Bryson Hull (Reuters) - Sri Lanka on Monday gave the Tamil Tiger separatists a 24-hour deadline to surrender or face annihilation inside a tiny coastal strip, signalling the guardian.co.uk final fight in Acia's longest running war

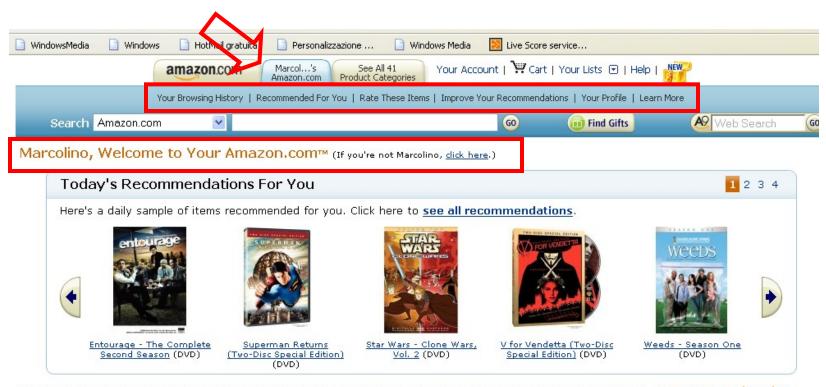
Buffalo News - 9 minutes ago

By Tom Precious ALBANY &mdash: A majority of New Yorkers favor a measure pending in the State Senate to legalize gay marriages in the state, a new poll has found. Ciana Dalle Datassan Ma. Cau Maniana Van Manu Varle Daile M



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How to defend oneself from Information Overload: My... Web - Personalized Stores



User's personalized "view" of Amazon Store: It's a filter which selects potentially interesting items! Your Recent Shopping Recently Viewed Items (0) Your Shopping Cart (2) Open & Recently Shipped Orders

Your Lists Your Wish List Your Gift List Your Shopping List 12/54

Outline

The Information Overloading Problem

Personalization on the Web

INFORMATION SEEKING STRATEGIES

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Popular Information Seeking strategies

Information Retrieval [Baeza-Yates and Ribeiro-Neto 1999]

- "Information Retrieval (IR) deals with the representation, storage, organization of, and access to information items"
- "...the user must first translate this information need into a *query* which can be processed by a search engine (or IR system)".
- "Given the user query, the key goal of an IR system is to retrieve information which might be useful or relevant to the user. The emphasis is on the retrieval of *information* as opposed to the retrieval of *data*".

Information Filtering [Hanani et al. 2001]

"The aim of Information Filtering (IF) is to expose users to only the information that is relevant to them. Some examples of filtering applications are: filters for search results on the internet,... e-mail filters based on *personal profiles*, ... filters for e-commerce applications that address products and promotions to potential customers only..."

Comparing IR and IF

Parameters	Information Retrieval	Information Filtering	
representation of Information Needs	Queries	User Profiles	
Goal	selection of items relevant for query	filtering out irrelevant items or collecting items	
Frequency of use	ad hoc use	repetitive use	
	one time users	long term users	
Type of Users	Not known to the system	"Profiled"	
Database	(relatively) static	very large dynamic	

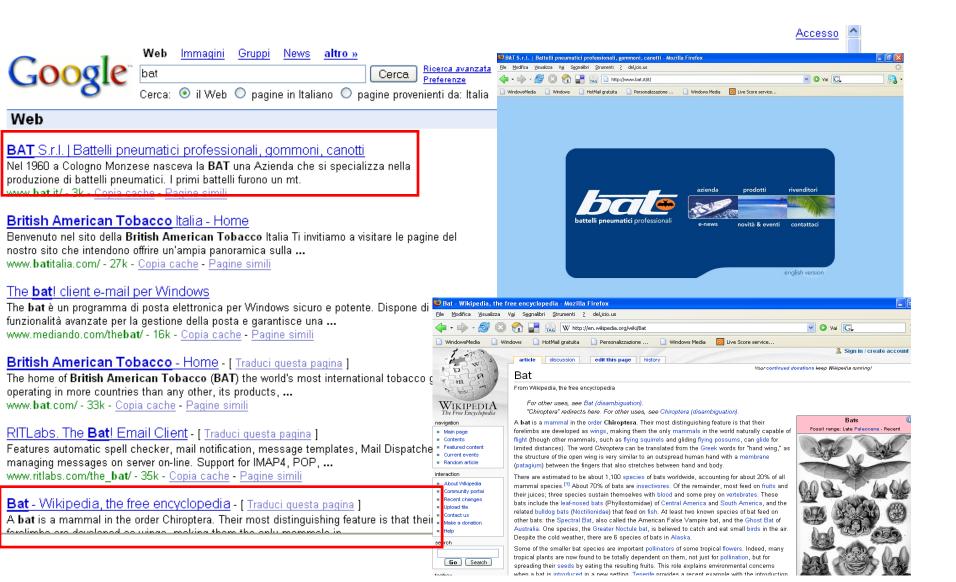
Table adapted from [Hanani et al. 2001]

Comparing IR and IF

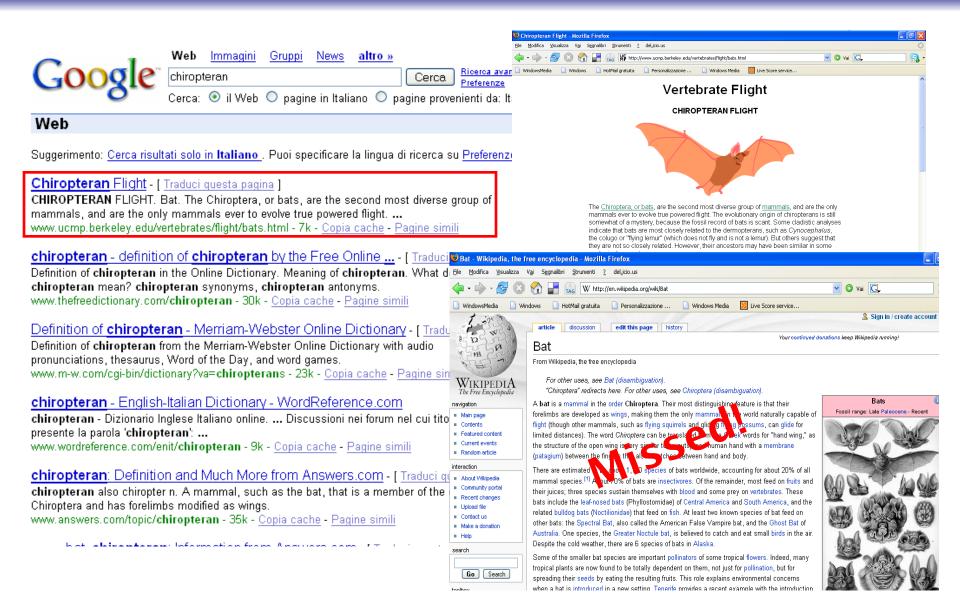
Common Mechanisms

- Representation: Both the user's information need query or profile – and the document set must be represented for comparison
- Comparison: String matching? Concept matching?
- Feedback: To improve the performance of the IR/IF system, a feedback mechanism is usually incorporated.

Some Problems in IR systems... Polysemy



Some Problems in IR systems... Synonymy



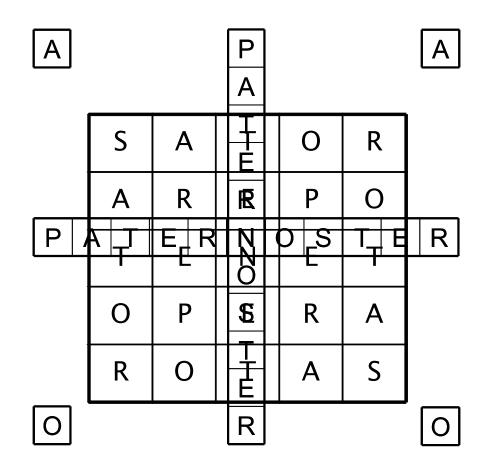
Some Problems in IF systems...

- IF systems perform the filtering task on the basis of user profiles
 - Structured model of the user interests
 - User profiles compared against item descriptions to provide recommendations
- Problems: keywords not appropriate for representing content, due to polysemy, synonymy, multi-word concepts (homography, homophony) – "Sator arepo eccetera" (Eco, 2007)

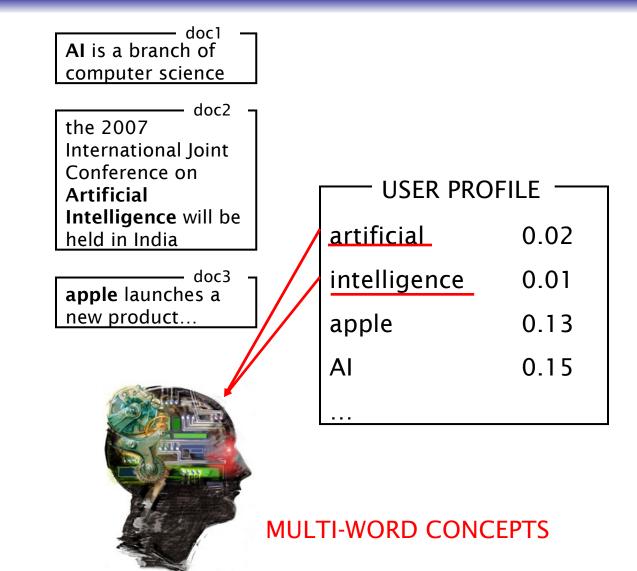


Some Problems in IF systems...

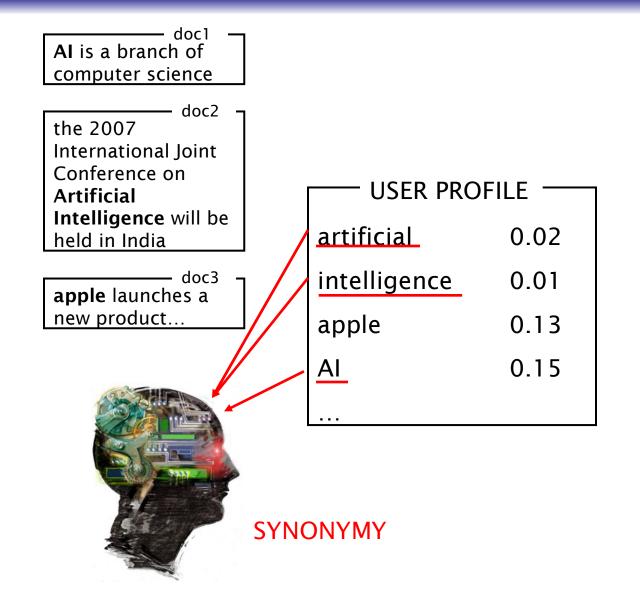
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Keyword-based Profiles

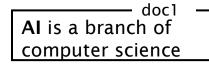


Keyword-based Profiles



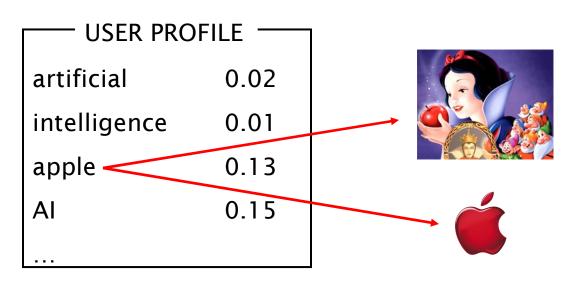
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Keyword-based Profiles



doc2 the 2007 International Joint Conference on Artificial Intelligence will be held in India

apple launches a new product...



POLYSEMY

Tackling IR/IF problems

Putting Intelligence into Search/Filtering Tasks =

1. Semantics: concept identification in documents through advanced NLP techniques \rightarrow "beyond keywords"

╋

1. Personalization: representation of the user information needs in an effective way \rightarrow "(high-accuracy) *user profiles*"

Information Seeking Support Systems (ISSSs):

tools that aid people in managing, analyzing, and sharing sets of retrieved information.

ISSSs provide search solutions that empower users to go beyond single-session lookup tasks.

G. Marchionini and R.W. White. Information-Seeking Support Systems. *IEEE Computer* 42(3):30-32, March 2009.

Meno's Paradox: Search is not so simple as it might seem

Meno: And how will you enquire, Socrates, into that which you do not know? What will you put forth as the subject of enquiry? And if you find what you want, how will you ever know that this is the thing which you did not know?

Socrates: I know, Meno, what you mean; but just see what a tiresome dispute you are introducing. You argue that man cannot enquire either about that which he knows, or about that which he does not know; for if he knows, he has no need to enquire; and if not, he cannot; for he does not know the very subject about which he is to enquire.

Plato Meno 80d-81a
http://www.gutenberg.org/etext/1643

Meno's question in our time

with modern ISSSs:

"How to discover the *magic words* that will connect you to the information you seek?"

The focus of many studies has been to point out methods for text processing to enable a system to make connections between the terminology of the user request (implicit/explicit) and related terminology in the information needed

Need for semantics for text interpretation

Beyond (Keyword) Search: Semantics by way of Knowledge Infusion

- Humans typically have the *linguistic* and *cultural* experience to comprehend the meaning of a text
 - Y How to realize this *capability* into machines?
- In NLP tasks, computers require access to vast amounts of common-sense and domain-specific world knowledge
 - ✓ Infusing lexical knowledge → Dictionaries (e.g. WordNet)
 - \checkmark Infusing cultural knowledge \rightarrow Wikipedia

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Collaborative / Social Filtering

- Word of Mouth / "Wisdom of the Masses"
- Makes use of a database of user preferences in order to:
 - find users with *similar* interests
 - predict whether an *unseen* information item is likely to be of interest for a user based on how *other* users rated that item

Widely adopted in recommender systems [Resnick and Varian 1997, Linden et al. 2003]

> Recommender Systems have the effect of guiding the user in a personalized way to interesting or useful objects in a large space of possible options [Burke, 2002]

[Burke 2002] R. Burke. Hybrid Recommender Systems: Survey and Experiments. *User Modeling and User-Adapted Interaction* 12(4):331-370, 2002.

[Resnick and Varian 1997] Resnick, P. and H. Varian. Recommender Systems. *Communications of the ACM* 40(3):56-58, 1997.

[Linden et al. 2003] Linden, G., B. Smith, and J. York. Amazon.com Recommendations: Item-to-Item Collaborative Filtering. *IEEE Internet Computing* 7(1), 76-80, 2003.

Collaborative / Social Filtering

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Recommender Systems provide personalized suggestions about items that the user might find interesting, by matching items to user profiles or groups [Kangas, 2001]

[Kangas 2001] Sonja Kangas. *Collaborative Filtering and Recommendation Systems*. Technical Report TTE4-2001-35. VTT Information Technology, Espoo, Finland, 2001.

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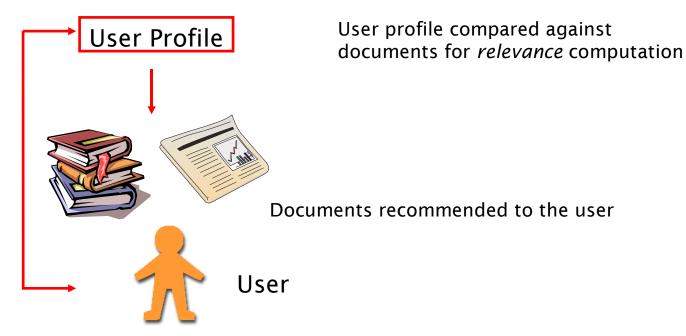
Recommendation process

Given a large set of items and a description of the user's needs, the goal is to present a small set of the items that are suited to the user needs

Content-based Filtering



Information Source



Content-based Filtering

- Each user is assumed to operate independently
- Items are represented by some features
 - Movies: actors, directors, plot, ...
- In the second second
 - Relevance judgment on items, e.g. ratings
 - Training on rated items \rightarrow user profile
- 4 Filtering based on the comparison between the *content* of the items and the user preferences as defined in the user profile
 - ✓ Keyword-based representation for content and profiles → string matching

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Intelligent Content-based Filtering

Semantics

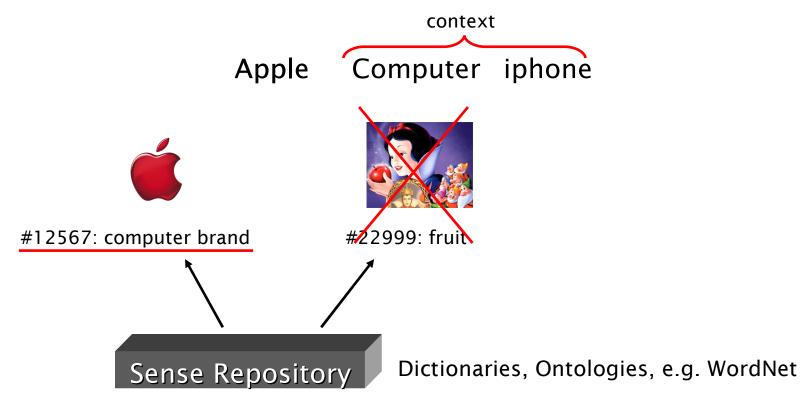
 Novel strategies to represent items and user profiles
 e.g. based on knowledge infusion and NLP methods

Personalization

- Novel strategies to build user profiles
 e.g. taking into account that information seeking
 can be seen no longer as a *solitary* activity
- recent interest in collaborative search and social bookmarking

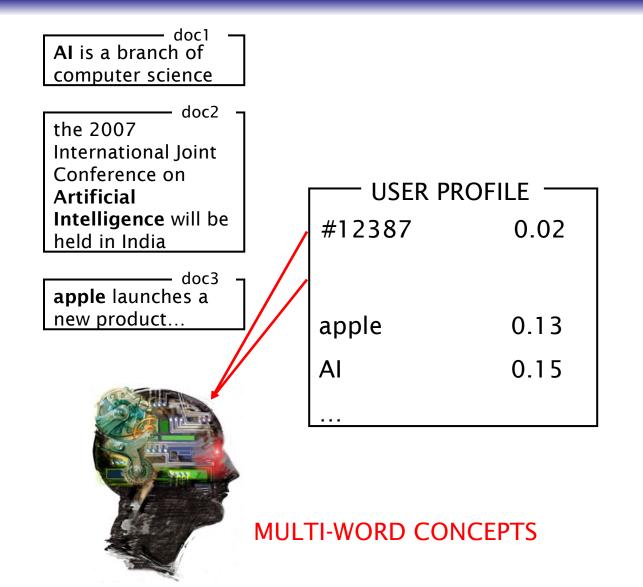
Word Sense Disambiguation (WSD): from words to meanings

WSD selects the proper meaning (sense) for a word in a text by taking into account the context in which that word occurs



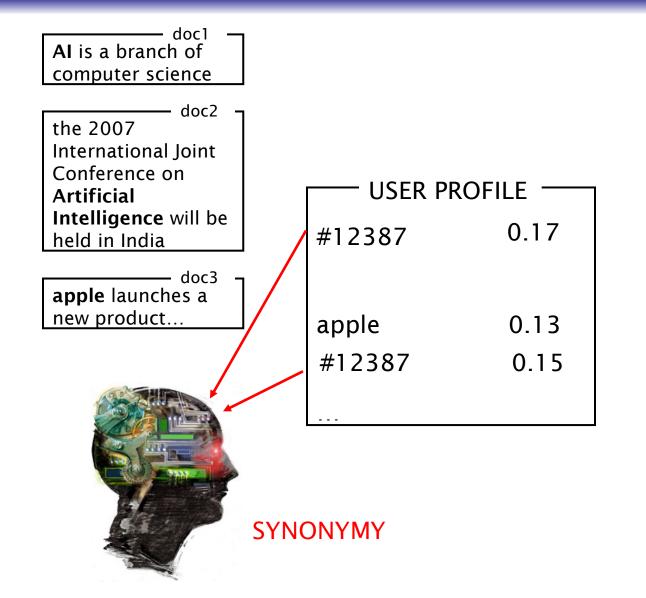
P. Basile, M. Degemmis, A. Gentile, P. Lops, and G. Semeraro. UNIBA: JIGSAW algorithm for Word Sense Disambiguation. In *Proceedings of the 4th ACL 2007 International Workshop on Semantic Evaluations (SemEval-2007), Prague, Czech Republic*, pages 398-401. Association for Computational Linguistics, June 23-24, 2007.

ITR (ITem Recommender) Sense-based Profiles



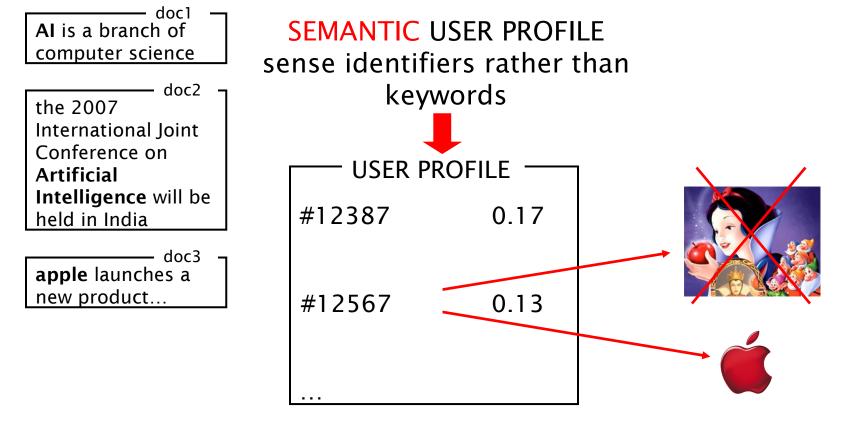
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ITR (ITem Recommender) Sense-based Profiles



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ITR (ITem Recommender) Sense-based Profiles



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M. Degemmis, P. Lops, and G. Semeraro. A Content-collaborative Recommender that Exploits WordNet-based User Profiles for Neighborhood Formation. *User Modeling and User-Adapted Interaction: The Journal of Personalization Research (UMUAI)*, 17(3):217-255, Springer Science + Business Media B.V., 2007.

G. Semeraro, M. Degemmis, P. Lops, and P. Basile. Combining Learning and Word Sense Disambiguation for Intelligent User Profiling. In M. M. Veloso, editor, *IJCAI 2007, Proceedings of the 20th International Joint Conference on Artificial Intelligence, Hyderabad, India, January 6-12, 2007*, pages 2856–2861. Morgan Kaufmann, 2007.

Intelligent Content-based Filtering

Semantics

 Novel strategies to represent items and user profiles
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Web 2.0 & User-Generated Content (UGC)

Linked in









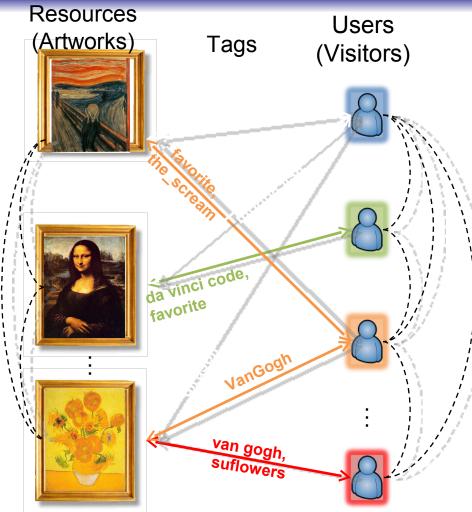






Social Tagging & Folksonomies

- Users annotate resources of interests with free keywords, called *tags*
- Social tagging activity builds a bottom-up classification schema, called a *folksonomy*
 - Folksonomy: "Folks" + "Taxonomy"
- 3 How to exploit folksonomies for advanced user profiling in content-based filtering?



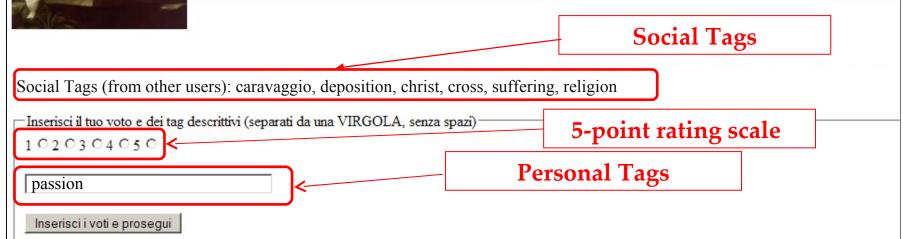
FIRSt (Folksonomy-based Item Recommender syStem) Learning from Ratings & Tags

27) Caravaggio - Deposition from the Cross



-Descrizione dell'opera

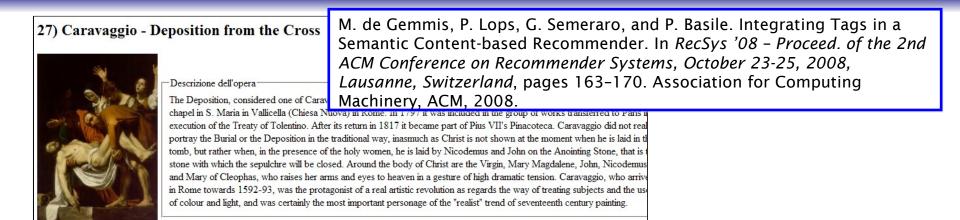
The Deposition, considered one of Caravaggio's greatest masterpieces, was commissioned by Girolamo Vittrice for his family chapel in S. Maria in Vallicella (Chiesa Nuova) in Rome. In 1797 it was included in the group of works transferred to Paris is execution of the Treaty of Tolentino. After its return in 1817 it became part of Pius VII's Pinacoteca. Caravaggio did not real portray the Burial or the Deposition in the traditional way, inasmuch as Christ is not shown at the moment when he is laid in the tomb, but rather when, in the presence of the holy women, he is laid by Nicodemus and John on the Anointing Stone, that is to stone with which the sepulchre will be closed. Around the body of Christ are the Virgin, Mary Magdalene, John, Nicodemus and Mary of Cleophas, who raises her arms and eyes to heaven in a gesture of high dramatic tension. Caravaggio, who arrive in Rome towards 1592-93, was the protagonist of a real artistic revolution as regards the way of treating subjects and the us of colour and light, and was certainly the most important personage of the "realist" trend of seventeenth century painting.

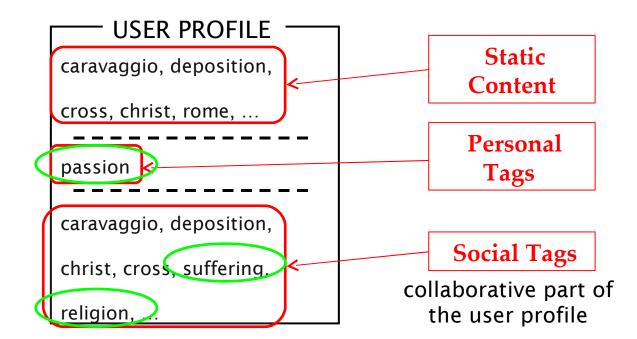


Textual description of

items (static content)

FIRSt (Folksonomy-based Item Recommender syStem) Tags within User Profiles





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Concluding Remarks and Future Work

- Need for *intelligent* solutions and tools for Information Access in the information overload era
- 2 New strategies for Information Filtering & Retrieval
 - Semantics: to capture the meaning of content and user needs
 - Personalization & User Profiling for adapting results to user information needs

Future Work

- Serendipity within Recommender System (avoiding the homophily trap [Zuckerman 2008])
- How to make serendipity operational? An initial attempt in [laquinta et al. 2008]

E. Zuckerman. Homophily, serendipity, xenophilia. April 25, 2008. www.ethanzuckerman.com/blog/2008/04/25/homophily-serendipity-xenophilia/

L. Iaquinta, M. de Gemmis, P. Lops, G. Semeraro, M. Filannino, and P. Molino. Introducing Serendipity in a Content-based Recommender System. In F. Xhafa, F. Herrera, A. Abraham, M. Koppen & J. M. Benitez, editors, *Proceed.of the 8th Int. Conf. on Hybrid Intelligent Systems HIS-2008*, 168-173. IEEE Computer Society Press, Los Alamitos, California, 2008.

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Intelligent Information Filtering & Retrieval @ SWAP group: the **SENSEmare** suite

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МЕГА	‰ ultilanguag & T ext <mark>A</mark> nalyzer
JIGSAW	Word Sense Disambiguator
IT	99 em Recommender
FIRS t	牙olksonomy-based 🗹 tem
	Recommender sy Stem
SENSE	SC mantic <u>N-level</u> Search <u>Cng</u> ine
OJJHO	On the J ip of my J HO ught



Sensemaking

- Many information search tasks are part of a broader class of tasks called *sensemaking*
- Sensemaking involves finding and collecting information from large collections, organizing and <u>understanding</u> that information and producing some product
 - understanding a health problem to make a medical decision
 - deciding which laptop to buy

Sensemaking

- Searching and filtering are the basic steps of the sensemaking overall process
- Reading and extracting information from retrieved documents are essential steps for decision making and other complex mental activities
 - Retrieval/Filtering are necessary but not sufficient

Peter Pirolli. Powers of 10: Modeling Complex Information-Seeking Systems at Multiple Scales. *IEEE Computer* 42(3):33-40, March 2009.

Examples of challenging sensemaking processes

• Just for fun:

OTTHO (On the Tip of my THOught) [Semeraro et al. 2009]

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- La Ghigliottina: very popular linguistic game
- Umberto Eco's article "La nostra ghigliottina quotidiana" on "L'Espresso"

Over the serious of the series of

U. Eco. La nostra ghigliottina quotidiana. April 24, 2008. http://espresso.repubblica.it/dettaglio/La-nostra-ghigliottina-quotidiana/2011914/18

G. Semeraro, P. Lops, P. Basile, and M. Degemmis. On the Tip of my THOught: Playing the Guillotine Game. In C. Boutilier, editor, *IJCAI 2009, Proceedings of the 21th International Joint Conference on Artificial Intelligence, Pasadena, California, July 11-17, 2009*, Morgan Kaufmann, 2009 (in press).

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... for your attention...



...Questions?