

Intelligent Interfaces for Information Retrieval: A Review

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May 6, 1996

Abstract

An intelligent interface for an information retrieval system has the aims of controlling an underlying information retrieval system, directly interacting with the user, and allowing him to retrieve relevant information without the support of a human intermediary. Developing intelligent interfaces for information retrieval is a difficult activity, and no well established models of the functions that such systems should possess are available. Despite of this difficulty, many intelligent interfaces for information retrieval have been implemented in the past years. This paper surveys these systems, with two aims: to stand as a useful entry point for the existing literature, and to sketch an analysis of the functionalities that an intelligent interface for information retrieval has to possess.

1 Introduction

Information Retrieval (IR) [64, 75] is the discipline devoted to build software systems (IR systems) for the storing, management and retrieving of large amounts of unstructured data (typically, natural language texts, images, sounds, and so on). The task at hand is difficult: the standard ‘syntactical’ solutions used in database theory [25], based on the structured nature of the data, are not suited, and a new approach, at a more ‘semantic’ level, is needed. Moreover, often the user has to face some problems, for instance to

express a need regarding an unknown field [37]. For overcoming this difficulties, the user of an IR system is usually supported by a human expert, an *intermediary*, that acts as an interface between the user and the IR system.

This solution is effective, but the growing diffusion of databases (CD-ROMs, Internet, and so on) creates a new scenario: the human intermediary is not always available and the user has to interact directly with the system. Of course, the system has to be modified: an *artificial intermediary* between the IR system and the user is needed, and this brings into the scenario the discipline of *human-computer interaction* [26, 55]. Moreover, for designing and implementing such artificial intermediary, *artificial intelligence* techniques [63, 34] seem to be promising, as the problems at hand are ill-defined (it is difficult, if not impossible, to define a formal, correct and complete relation between input and output): the obtained system is named *Intelligent Interface* (II) to an IR system, or II for IR.

The field of IIs for IR is thus the intersection of three fields: IR, human-computer interaction and artificial intelligence. All these three disciplines are difficult ones on their own, and the resulting area is obviously a complex one. As a matter of fact, no well established theoretical model of the functionalities that an II for IR has to provide is available, though some proposals have appeared (MONSTRAT and MEDIATOR models [37]), and some studies have been published (for instance [6]). Notwithstanding all these obstacles, many experimental IIs for IR have been developed in the last decades: Table 1 reports, in temporal order, the IIs for IR described in literature. In each column, the table shows: the name of the system; the date of its realization, derived from the bibliographic references; and its main bibliographic references.

This paper presents a survey of IIs for IR, and has two aims: to stand as a useful entry point to the large literature on IIs for IR and to sketch the functionalities that the ideal II for IR has to possess. The description style is very schematic, in order to have the maximum synthesis and clearness. In the next section I select and briefly describe some of the most representative IIs for IR (the sixteen ones in boldface in Table 1). These systems are analyzed more in depth in Section 3. The last section concludes the paper.

2 Selected Intelligent Interfaces for Information Retrieval

In this section some of the most significative IIs for IR are selected and briefly described. The selection is made on the basis of the functionalities provided by the systems: it is by no means intended to classify the goodness of the systems.

Table 1: The IIs for IR described in literature. The entries labelled with (*) do not have a system name; I have used the name of the author.

Name	Year	Refs.
THOMAS	1977	[56]
RITA	1978	[79]
EUREKA	1979	[13]
CITE	1979	[27]
GRUNDY	1979-83	[61, 62]
TTIRS	1980	[81]
CSIN	1981	[36]
CONIT	1981	[41, 42]
Shoval (*)	1981-85	[66]
IIDA	1982	[48, 49]
OL'SAM	1982	[73]
OASIS	1982-85	[82, 83]
IRUS	1983	[5]
IR-NLI	1983-86	[8, 33]
OKAPI	1983-95	[28]
POISE	1984	[22]
NP-X	1984	[68]
CIRCE	1985	[3]
RABBIT	1985	[19]
RUBRIC	1985-87	[45, 74]
EP-X	1985-89	[69, 70]
FIRSTUSER	1986	[20]
FRED	1986	[38]
RESEARCHER	1986	[40]
CANSEARCH	1986-87	[57, 58]
Eurisko	1987	[4]
IRNLI-II	1987	[9, 10]
IOTA	1987	[15, 16]
GRANT	1987	[18]
CODER	1987	[29]
COALSORT	1987	[52]
IMIS	1987	[80]
PROBIB-2	1987	[80]
PLEXUS	1987-88	[77, 78]
I3R	1987-89	[23, 72]
TOPIC	1988	[17]
OAKDEC	1988	[46]
ODA	1988	[53]
MenUSE	1988	[59]
KIRA	1988	[65]
Tome	1988-89	[76]
EUROMATH	1989	[44]
OFFICER	1989	[21]
KIWI	1989	[39]
BOOKHOUSE	1989	[43]
OAK	1989	[47]
MOSS	1989	[54]
ESOCKS	1989	[84]
IANI	1989	[85]
Gauch (*)	1989-90	[30, 31, 32]
ISIR	1990	[60]
SIMPR	1990	[67]
RADA	1990	[71]
AI-STARS	1990-93	[1, 2]
INQUERY	1992	[14]
LYBERWORLD	1994	[35]
FIRE	1995-96	[11, 12]

The selected systems, in temporal order, are:

CITE allows the user to express his need as a natural language request. The system translates the request into a query for MEDLINE (a medical database) and returns a ranked set of documents. It also provides query expansion through relevance feedback: the user can judge the relevance of the retrieved documents and the MeSH (Medical Subject Heading, the controlled vocabulary of MEDLINE) terms contained in the relevant ones are automatically added to the query.

GRUNDY interacts with the user, asking him questions with the aim of building, updating, refining, and revising a model of the user. The model is then used for suggesting to the user a novel interesting for him. The user models are implemented via stereotypes, hierarchically organized: a stereotype, containing features like age, sex, perseverance, independence, political preferences, and so on, is assigned to the user. To each feature a numerical value, a confidence rating, and one or more justifications are associated. On the basis of the features of the user, the system selects the most adequate novel; if the user does not accept it, GRUNDY tries to find which feature(s) has to be modified, eventually asks other questions, and proposes something different.

CONIT interacts with the user through menus and a simple command language. The user expresses his need as a set of terms, and the system autonomously chooses to which of the available databases (DIALOG, ORBIT, MEDLINE) to send the query.

Shoval has the aim of reformulating an input query through spreading activation in a thesaurus. The spreading activation starts from the query terms, is governed by a production rules knowledge base, and the steps taken can be explained to the user.

RUBRIC takes in input a rule-based description of user's need. For instance, the two rules

$$\begin{aligned} & \text{"information" AND "retrieval"} \\ & \Rightarrow \text{"information retrieval"} \quad (0.6) \end{aligned}$$
$$\begin{aligned} & \text{"information" ADJACENT "retrieval"} \\ & \Rightarrow \text{"information retrieval"} \quad (0.9) \end{aligned}$$

mean that a document containing both the terms "information" and "retrieval" has a 0.6 probability of being about "information retrieval" and that this probability is 0.9 if the two terms are adjacent. The rule representation is then translated into a boolean query. This system is interesting and original: the user is the source of the domain knowledge,

and the information need description may be more precise and flexible than through the usual boolean operators. The problems arise when the user has an ill-defined information need, and thus does not know the terminology of the field in which he is doing the search, or when he has not enough time for completely specify his need.

EP-X tries to work at a semantic level: its domain knowledge is represented as a hierarchy of frames (thus it is richer than a classical thesaurus), and this knowledge is used to understand user's query (a set of terms) and to reformulate it. The reformulation activity is not merely interactive query expansion: also linguistic ambiguities are resolved.

CANSEARCH is a menu-based interface that helps the user in formulating a query for MEDLINE using terms from MeSH. MeSH is hierarchically organized, with broader-term and narrower-term relations among the terms; the system proposes a series of menus on the basis of such hierarchy.

IOTA is an expert system prototype with the aim of experimenting the effectiveness of artificial intelligence techniques in IR. The system takes in input a natural language request and, using computational linguistic techniques (parsing, learning of new terms from the context), and linguistic knowledge (two vocabularies, one thesaurus), translates it into a boolean query. IOTA does not search in a real database, but in a unique document indexed by keywords. It also performs a limited automatic query expansion, inserting broader terms from the thesaurus and modifying the logical structure of the boolean query.

PLEXUS asks questions to the user for building a user model and getting a natural language description of user's need. The request is translated into boolean form by semantic analysis (the system has a dictionary containing the terms of the domain, gardening, and a description of their meaning) and, using a stop-word list and stemming algorithms, sent to an underlying IR system. If not items are found, the automatic reformulation phase starts: some of the tactics contained in a knowledge base (tactics based both on the query structure and on the meaning of terms) are applied for modifying the query in order to retrieve at least one item.

I3R is one of the more complete IIs for IR. It is based on a blackboard architecture, with seven cooperating expert systems ('User Model Builder', 'Request Model Builder', 'Domain Knowledge Expert', and so on) coordinated by a 'Control Expert'. Initially, the system asks questions to the user, with the aim of building the user model (via stereotypes) for

choosing the better interaction style. On this basis, the system tries to build the model of the information need, and then searches in the database. If the results are not satisfying, the reformulation starts: the user can browse an integrated, filtered by the system, hypertext-like network of concepts and documents, evaluate documents, and insert new terms in the query.

EUROMATH tries to model the need of the users in a more complete way than the usual one. It asks to the users (mathematicians), besides the classical topic aspect of user's need [11, 12, 50, 51], the type of the need (verificative, similar to a previous one, conscious topical, ill-defined). This information, that is continuously verified during the interaction on the basis of user's behavior, and eventually updated (by either the user or the system), is used for choosing the best kind of interaction with the user. Also the preferred type of documents is asked to the user and used in the retrieving phase. Thus, EUROMATH builds and maintains a primitive form of user model, and uses it both for the interaction and the search.

BOOKHOUSE supports, through an icon-based graphical interface, unexperienced users and librarians for the retrieving and indexing of documents, respectively. This system is supportive: it does not simulate a human intermediary, but supports the user that works autonomously.

Gauch supports the user during the search of relevant passages in a full-text database. The system accepts a boolean query and the number of desired passages, performs a search and, if the results are not satisfying (*i.e.* not matching the desired number of hits), starts an automatic reformulation phase in which the knowledge bases of the system (domain knowledge, *i.e.* terms from a thesaurus and stemming, and expert knowledge, *i.e.* tactics [7]) are used for modifying the query and retrieve the wanted number of items. The query can be modified in three ways: via addition or replacement of related terms, via relaxation or reinforcement of the proximity operators, and via structural modification of the boolean operators.

ISIR provides a graphical interface that allows the user to search in different commercial databases, connected via TELNET, and to browse the thesaura of the databases (INSPEC, MeSH, ERIC), with a unique syntax.

AI-STARS uses linguistic knowledge (morphological analysis, stemming, thesaurus) for improving the indexing of documents, the interpretation of the query and the query expansion. AI-STARS takes in input a

query in natural language, translates it into a boolean representation and presents it to the user in a way similar to a spreadsheet, helping the user to understand the meaning of the boolean operators (AND and OR) and allowing him to directly manipulate the terms, dragging them with the mouse for changing the logic of the search. The system can also add or substitute terms in the query, thus providing help for reformulation.

FIRE has the main goal of emulating some of the functions of a human intermediary by interacting directly with end-users and by supporting them during query reformulation. These capabilities are accomplished by FIRE through explicit representations of knowledge about the intermediary skills (tactics [7] and plans, namely predefined sequences of tactics, that a human intermediary executes for reformulating the query) and about the subject domain (thesauri and morphological knowledge). FIRE allows the user to enter a boolean query, to retrieve relevant documents, and to read and classify them. If the user is not satisfied with the obtained results, he can start a query reformulation process, in which FIRE selects and controls a general strategy for a semi-automatic reformulation: starting from the initial representation of the information need, it proposes to the user a set of alternative or additional terms, among which the user can choose the ones that better describe his need.

3 Analysis of Intelligent Interfaces for Information Retrieval

In this section, four tables are presented, in order to analyze the main features of the selected IIs for IR. The tables report features about: the knowledge used in the systems, the interaction system-user, the representation of the need, and the user model built by the system, respectively.

In Table 2 the *knowledge* types used in the IIs for IR and the sources from where such knowledge was obtained are described. The values in the cells of this and the following tables are: ● for present, ○ for partial, a void cell for absent and ? for unknown. The columns have the following meaning:

Area groups the columns representing the types of knowledge possessed by the system. The letters in the column headings stand respectively for knowledge about:

I: (Information) the information retrieval activity;

S: (System) the information retrieval system(s) underlying the interface;

Knowledge	Area					Source				
	I	S	D	P	U	U	D	L	P	E
CITE	•	•	•				?	•	?	?
GRUNDY	•		•	•	•		•			
CONIT		•								
Shoval	•		•				•	?		
RUBRIC	•	•	•			•	•	?		
EP-X	•	•	•				•	•	•	
CANSEARCH		•	•				?	?		?
IOTA	•	•	•	•	•	•	•	•		•
PLEXUS	•	•	•	•	○	○	•	•	•	•
I3R	•	•	•	•	•	•	•	•		
EUROMATH	•	•	•	•	•		•	•		
BOOKHOUSE	•	•	•	•				•	•	
Gauch	•	•	•				•	•		
ISIR	•	•	•				•	•		
AI-STARS	•	•	•				?	?	?	?
FIRE	•	•	•				•	•		•

Table 2: Types and sources of knowledge used in the systems.

D: (Domain) the database domain, often limited to terminological knowledge represented as a thesaurus;

P: (Population) the user population, *i.e.* the features that the users of the particular system possess;

U: (User) the specific user interacting with the system.

Source groups the possible sources used for obtaining the knowledge:

U: Users;

D: Designers;

L: Literature;

P: Protocol analysis;

E: Experts (*i.e.* human intermediaries).

In Table 3 some aspects of the *interaction* between the user and the system are summarized. The meaning of the columns are:

Modality groups the columns regarding the modality of the interaction:

Interaction	Modality						Ctrl	General		
	C	N	M	W	I	G		A	F	I
CITE		•					S			S
GRUNDY	•		•				S	•	◦	A
CONIT	•	◦	•				U		◦	?
Shoval	•						S			S
RUBRIC				•			U	•		A
EP-X		•					S			S
CANSEARCH			•				U		•	A
IOTA		•					S	•	•	S
PLEXUS		•					S			S
I3R		•	•	•		•	M	•	◦	A
EUROMATH			•	•		•	M	•	•	S
BOOKHOUSE					•	•	U	•	•	A
Gauch	•						S			S
ISIR			•	•			U		•	S
AI-STARS		•				•	M		•	A
FIRE			•	•		◦	M		•	S

Table 3: Interaction user-system.

C: Command line;

N: Natural language;

M: Menus;

W: Windows;

I: Icons;

G: Graphic.

Control indicates who (the system or the user) has the initiative of the interaction. It can assume as value:

S: System;

U: User;

M: Mixed.

General groups the columns regarding general aspects of the interaction:

A: (Adaptable) describes if the systems adapts his behavior to the specific user;

Need	Topic				Task			Tools				
	B	F	S	L	O	C	T	E	F	C	D	Cl
CITE				•				•	•			
GRUNDY			•									
CONIT			•									
Shoval			•					•				
RUBRIC	•					◦						
EP-X		•	•		?	•		•				
CANSEARCH			•					◦				
IOTA	•			•				•		◦		
PLEXUS	•			•				E				
I3R	•		•	•	•	•		•	•	•	•	•
EUROMATH	•	•			•	•	•	•	•	•		
BOOKHOUSE	•				•	•	•	•		◦	•	•
Gauch	•	•			◦	•		•				
ISIR			•					◦		•		
AI-STARS	•	•		•				•				
FIRE	•	•			•	•		•	•	•		•

Table 4: Expression of the user's information need.

F: (Friendly) describes if the system is easy to use for an unexperienced user;

I: (Intermediary) describes the approach chosen in the design of the system. It can assume as value:

S for a system that Simulates the intermediary;

A for a system that follows an Alternative approach.

In Table 4 some aspects of the representation of the user's *information need* are presented. The meaning of the columns is:

Topic groups the columns describing how the topic of the information need is communicated from the user to the system:

B: Boolean;

F: Facets;

S: Single term;

L: natural Language.

User Model	LT	ST	Can.	Ind.	Expl.	Ded.	Stat.	Dyn.
GRUNDY	•		•	•	•	•		•
IOTA		•	•			•	•	
PLEXUS		•	•		•	?	•	
I3R	•		•	•	•		•	◦
EUROMATH		•	•		•		•	
BOOKHOUSE			•				•	
FIRE	•		•	•	•	•		•

Table 5: User model types.

Task groups the columns describing if the task and context [11, 12, 50, 51] of the search are communicated from the user to the system:

- O:** search Objectives (for instance, high recall or high precision);
- C:** Constraints on the search (for instance, the desired number of retrieved documents);
- T:** Type of the need (for instance, verifacative *vs.* conscious topical *vs.* muddled [37]).

Tools groups the columns describing which tools support the user’s expression of his need. Each column describes if the system possesses tools for:

- E:** query Expansion (adding —or substituting— to the terms in the query other related terms);
- F:** relevance Feedback (adding to the query terms taken from relevant documents);
- C:** Concept browsing (allow the user to browse the domain knowledge of the system);
- D:** Document browsing (allow the user to browse the documents);
- Cl:** Classification of documents (allow the user to classify the retrieved documents as relevant, not relevant useful, not useful, etc.).

Table 5 reports the features of the system’s *user model* [24] (only for the systems that have a user model, see Table 2; also FIRE is included, though its user model capabilities are not yet integrated in the system, but only foreseen). Each column describes if the user model:

LT and ST: (Long Term and Short Term) is stored or not stored at the end of the session for being re-utilized;

Can. and Ind. (Canonical and Individual) represents a canonical user or the actual user;

Expl. and Ded. (Explicit requests and Deduced) is built asking questions to the user or deducing information from user behavior;

Stat. and Dyn. (Static and Dynamic) changes during a session.

4 Conclusions

Some of the most representative IIs for IR have been presented and analyzed. On the one side, this work is useful for a high level view of the existing IIs for IR. On the other side, it has some interesting future developments: on the basis of the analysis of the systems, it is possible to individuate a set of functions that an ideal system has to possess, and to provide a detailed specification of such a system. This specification, being based on already implemented functions, is more likely to be really implementable than more theoretical models like MONSTRAT and MEDIATOR: only integration problems will occur.

Acknowledgements

This paper could not have been written without Francesco Gri and Carlo Tasso. I would like to thank them for their work on the classification of the systems.

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