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**Contextual Search:  
A Computational  
Framework**

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# Contextual Search: A Computational Framework

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*Published, sold and distributed by:*

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PO Box 1024  
Hanover, MA 02339  
USA  
Tel. +1-781-985-4510  
[www.nowpublishers.com](http://www.nowpublishers.com)  
[sales@nowpublishers.com](mailto:sales@nowpublishers.com)

*Outside North America:*

now Publishers Inc.  
PO Box 179  
2600 AD Delft  
The Netherlands  
Tel. +31-6-51115274

The preferred citation for this publication is M. Melucci, Contextual Search: A Computational Framework, *Foundations and Trends<sup>®</sup> in Information Retrieval*, vol 6, nos 4–5, pp 257–405, 2012

ISBN: 978-1-60198-600-9

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Information Retrieval**  
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Foundations and Trends<sup>®</sup> in Information Retrieval, 2012, Volume 6, 5 issues. ISSN paper version 1554-0669. ISSN online version 1554-0677. Also available as a combined paper and online subscription.

Foundations and Trends<sup>®</sup> in  
Information Retrieval  
Vol. 6, Nos. 4–5 (2012) 257–405  
© 2012 M. Melucci  
DOI: 10.1561/15000000023



# Contextual Search: A Computational Framework

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## Abstract

The growing availability of data in electronic form, the expansion of the World Wide Web (WWW) and the accessibility of computational methods for large-scale data processing have allowed researchers in Information Retrieval (IR) to design systems which can effectively and efficiently constrain search within the boundaries given by context, thus transforming classical search into contextual search. Because of the constraints imposed by context, contextual search better focuses on the user's relevance and improves retrieval performance, since the out-of-context aspects of the search carried out by users that are likely linked to irrelevant documents are left apart.

This survey introduces contextual search within a computational framework based on contextual variables, contextual factors and statistical models. The framework adopted in this survey considers the data observable from the real world entities participating in contextual search and classifies them as what we call contextual variables. The contextual variables considered are content, geotemporal, interaction, and social variables. Moreover, we distinguish between

contextual variables and contextual factor: the former is what can be observed, the latter is what cannot be observed, yet this is the factor affecting the user's relevance assessment. Therefore, in this survey, we describe how statistical models can process contextual variables to infer the contextual factors underlying the current search context.

In this survey we provide a background to the subject by: placing it among other surveys on relevance, interaction, context, and behavior; providing the description of the contextual variables used for implementing the statistical models which represent and predict relevance and contextual factors; citing and surveying useful publications to the reader for further examination; providing an overview of the evaluation methodologies and findings relevant to this subject; and briefly describing some implementations of contextual search tools.

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# 1

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## Introduction

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Context: from Latin *contextus*, where *con* stands for “together” and *texere* stands for “to weave”.  
Oxford Dictionary

### 1.1 Motivation of this Survey

Many researchers with various backgrounds believe that context can enhance the user’s experience and improve the system’s effectiveness of search. In so doing, they frame Information Retrieval (IR) within the more general notion of contextual search, although from differing viewpoints. The different perspectives at which context has been viewed have led to definitions of context with different potential of implementation.

At one extreme, context can be defined as the circumstances that form the setting for an event, statement, or idea, and in terms of which it can be fully understood and assessed. From this perspective, some publications relevant to contextual search are being written mostly from an information seeking and retrieval point of view. Although such a point of view is rooted in strategically important disciplines like user behavior, cognition or human interaction, it cannot fully help see how to

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proceed with what should be observed and computed for implementing context within an IR system. At the other extreme, context can be viewed as the parts of something written or spoken that immediately precede and follow a word or passage and clarify its meaning.

These two extreme perspectives clearly differ in their potential of implementation. While the definition of context in terms of events and settings cannot obviously be utilized for designing algorithms and data structures, the definition of context as a text window around a word is easier to implement and is strictly related to the nature of text and is part of common sense. However, this view is quite reductive and considers only one of the ways context occurs.

From one extreme to the other, computational approaches to contextual search followed one another, ranging from sophisticated and computationally expensive approaches to more simple and efficient ones, but each one of them has been useful for writing this survey.

### 1.2 Definitions and Scope of the Survey

A *variable* is any observable value that is liable to change. Variables can be: qualitative or quantitative; ordinal or not; if ordinal, cardinal or not; if cardinal, integer or not; and so on. When the variables are random, they change according to a probability distribution in such a way that its observation value occurs with a given probability. Such a characterization allows inference to be made on estimation and prediction of potential relationships between variables in such a way that the variation of some independent variables determines the variation of some dependent variables.

A *contextual factor* is any unobservable circumstance or fact of search such as query intent, personal interest and document quality, which affects relevance. We concentrate on three contextual factors: query intent, personal interest, and document quality. (The contextual factors are illustrated in Sections 2, 3, and 4.)

Query intent refers to the objectives of the user who issued the query. In this situation, a query is viewed as a means to accomplish a task such as “dissertation writing,” “finding a resource,” “bibliography compilation” and a query intent is an objective to be achieved in order

to accomplish the task. Intent is a property of a query and is not necessarily tied to a user in the way a personal interest is.

Personal interest in general refers to the user's state of wanting to know or learn about a thing, a person or an event. An interest is an information need that has the quality of sparking curiosity or holding the user's attention, and may be viewed as a property of an information need that makes the information need crucial to the user.

Document quality refers to the property of a document that is able to be trusted as being up-to-date, authoritative, exhaustive, accurate, reliable, and clear. A high-quality document is considered to be the best of its kind and unlikely to be improved upon.

A *contextual variable* is any set of variables dependent on contextual factors. This survey classifies the contextual variables observed by the real world entities participating in contextual search as: content variables, interaction variables, geographical variables, and social variables. These contextual variables are introduced in Section 1.3.

A *statistical model* is a set of computable mathematical rules defined over a set of variables or factors for example height and weight are related in a way that they can be plotted as points along a straight line, or the frequency of a term within a document and relevance assessment are related in a way that the higher the frequency, the more likely the document is relevant. In this survey, the rules of a statistical model express computable relationships between contextual variables and contextual factors.

The denotation of context in this survey is thus essentially computational and allows us to introduce a computational framework of contextual search summarized by the following

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**Definition 1.1.** *Context* is represented as a set of contextual variables and contextual factors weaved together by statistical models of estimation and prediction.

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An objective of this survey is to inform the reader which contextual variables, contextual factors, and statistical models have been utilized in the literature to represent context by means of a

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computer system and therefore which of these provide some useful hints about what to use to extend a traditional IR system toward context.

The definitions of contextual variable and contextual factor point out two main aspects: observability and dependence. Observability is a necessary condition of the meaning of context in this survey, that is, context can be operationalized as variables and can therefore be exploited in search only if some variables can be defined and observed; context implementations not referred to as variables are not considered in this survey. A basic example is standard IR: index term occurrence is a contextual variable observed from the document content, in contrast, aboutness is a contextual factor affecting relevance and cannot directly be observed.

Dependence is between contextual factors, relevance, and variables. Research in IR often assumes that a variation of contextual factors reflects upon a variation of the contextual variables. The relationship posited between personal interest and term frequency is an example of the relationship between a contextual variable and a contextual factor; term frequency may increase in a document if this document becomes interesting for a user. Therefore, if some variations of the contextual variables are observed, a variation of the contextual factor is likely to have occurred; for example, if term frequency is higher in a document than in another document, the former is more likely interesting than the latter.

In the computational framework presented in this survey, attention is also paid to discovering the contextual factors that affect relevance assessments. Thus, a variation of relevance assessment is due to a variation of the contextual factors; for example, if query intent is viewed as contextual factor and term frequency is a contextual variable, a variation of frequency may result from a variation of query intent which in turn affects relevance.

Another feature of this survey is the attention paid to statistical models. The statistical models mentioned in this survey provide some advantages in illustrating context. They allow researchers to implement context because these models are suitable for estimating and predicting context starting from the variables observed in objects. Another

advantage is of a computational nature. Most of the statistical models scale up when the size of data available from the user's environment, social network, and personal dimension increases by hundred-fold, thus keeping high levels of effectiveness and efficiency. Everyone prefers computationally efficient approaches to search in context; however, computational efficiency is not a feature of every one of the approaches (e.g., those arising from Artificial Intelligence), despite the computational potential that makes contextual variable implementation more unbridged than in the past.

A computational framework for contextual search like that described in this survey may resemble classical modeling in noncontextual search in which the "best" model is selected for optimizing effectiveness. We think that this approach is not constrictive since in the past there have definitely been statistical models for contextual search that resulted in significant improvements in IR systems. We do not claim that a computational approach is the only approach to explaining contextual search, but we do claim that it is the best approach for making contextual variables useable; relevance feedback is an example of a computational approach to contextual search thoroughly investigated in the past.

Although (or maybe precisely because) investigated and employed for a long time, relevance feedback is still crucial in contextual search, since it mainly relies on content variables and in particular on document content. Indeed, relevance feedback has recently been reevaluated and experimented with huge test collections and very short noisy queries through initiatives such as the relevance feedback track of TREC. However, despite it being relevant, explicit or pseudo-relevance feedback is not addressed in this survey because our focus is on recent developments of contextual search while there are already surveys of relevance feedback and query expansion.

In contrast, implicit relevance feedback is the backbone of the incorporation of behavior in contextual search. The research conducted within implicit relevance feedback has aimed to use the contextual information generated during the interaction between the user and information as implicit evidence of relevance. Hence, a key question is whether implicit relevance feedback can effectively be used in

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contextual search systems in comparison with traditional content-based ranking functions or more advanced yet well experimented methods such as anchor text or link analysis algorithms.

As it happens, contextual search is not relevant only to IR, of course, but to other research areas too with which they interact yet they seem rather distant from IR. Examples are Psychology, Mobile Communication, Electronic Commerce, Nomadic Computing, Human Computer Interaction. All these subjects are relevant to this survey although they cannot be looked at thoroughly because the topic of this survey is already vast enough. A few things that are on the side of the context of a document and are not the primary focus are: temporal context (e.g., two e-mail messages sent right after the same event); storage context (e.g., two documents found in the same file system folder); conversational context (e.g., one e-mail message is a reply to another).

### **1.3 Contextual Variables**

This survey considers four types of contextual variables: content, geographical, interaction, and social variables.

Content variables refer to the informative content and relationships of queries and documents. The data are content features observed from text, image, video, audio; link anchors; layout; genre; lexical properties (e.g., part-of-speech tags); user's tags (e.g., image tags or file names); category labels (e.g., Wikipedia category labels); demographic labels (e.g., authorship) and anything used to describe informative contents or to enrich information need representations.

Geographical variables are any variable with the state of existing within or having some relationship with space location. Examples are geographical names added to documents or queries, digital photographs tagged with geographical coordinates, typically the latitude and longitude of the space location perhaps associated to a user.

Interaction variables are observed over time during the interaction between users and IR systems. (Geographical variables are not necessarily referred to a user.) These variables are for example: click-through data; data about queries or search sessions; user



judgments or assessments; user behavior data (e.g., document retention, display time, eye or mouse movements).

Social variables refer to user communities or groups and are observed for example from: “tweets”; social connections (e.g., friendship); hyperlinks (e.g., a link between two WWW pages).

### 1.3.1 Content Variables

Content is a contextual variable exploited in contextual search to decide whether an additional or special action that is different from time to time should be performed by an IR system when the user is interacting with the informative content managed by the system for meeting his information needs. Content variables can be observed from the documents of a collection, search engine result pages, queries, or from parts of them such as windows, fragments, and passages.

The main *medium* addressed in the literature of this survey is text. It is perhaps the richest source of evidence for predicting context since text is an expression of natural language, that is, the main means used by humans to communicate information and needs. Text can easily be managed because words or terms can be suggested to the user who in turn can understand them by leveraging common cognitive abilities and feed data back into the system: positive words represent what items the user would like to retrieve; negative words indicate what the user does not want; neutral words are not good indicators of her information needs. We are not dealing with multi-lingual text IR; the literature utilized in this survey refers to the English language only.

### 1.3.2 Geographical Variables

In our view, geographical variables are observed and are instrumental for detecting contextual factors such as query intent and personal interests. Geographical variables differ from other variables due to the intents underlying the queries referring to geographical information. However, they raise issues similar to the issues raised by natural language processing. In particular, when geographical variables are names, the issues are: name or reference detection,

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name disambiguation, name clustering, linking or association, name weighting, and document ranking.

Our use of geographical variables complements the view of geography as a relevance factor; for example, Raper [143] defines geographical information needs based on cognitive and geographic criteria and argues that geographic relevance is best defined as a spatio-temporally extended relation between geographic information needs and geographic documents.

### **1.3.3 Interaction Variables**

Whenever users have difficulty in expressing their information needs, contextual variables based on interaction are precious because an IR system can be enabled to automatically deduce a user's interest based on the data generated during the interactions with the system. Indeed, the data observed over time during the interaction of the user with a contextual search system form an interaction history where history also means "finding out." Thus, the value of interaction variables is not only the individual pieces of content, but their organization within a coherent stream of data — it is the observation of these pieces together which makes history valuable; for example, if the user has requested some documents recently, it is likely that the user is in a given context, and the retrieved documents can form the basis of supervised learning for the user's preferences because of recency and not only because of the amount of data.

When the interaction data employed for estimation are very close in time to the user's actions, the estimated models are more closely related to the user than the models that would be estimated with the farthest data in time. The data employed for estimation are very close in time to the user's actions when, for instance, it is of interest to the contextual search system to recognize the correct query sense or intent of the user. On the other hand, implicit relevance feedback information collected over a long period of time is less likely to be very useful for predicting the individual's interests than the immediate search context and feedback information; they may be useful for predicting the interests of a group.

#### 1.3.4 Social Variables

What a user either directly or indirectly learns from or teaches to the communities is a crucial contextual variable because human relationships are at the basis of many conditions. Hence, it is not surprising that quite an appreciable proportion of the literature on contextual search addresses the issues of the social dimensions of the users. Community is meant in a broad sense and is not confined to social networks or similar user organizations; the same algorithms can make the social variables observed from any user community a useable source of evidence for contextual search.

In this survey, we consider some cases of community behaviors where members participate in a collective activity and unwittingly collaborate to build collective knowledge. These kinds of community behaviors differ in the degree to which a member is aware of belonging to a community. For example, the users tagging resources are often aware of their membership to a community (e.g., they log into a system) whereas the users clicking ads are not aware that their clicks are collected and exploited for boosting ad ranking. What the various kinds of community behaviors have in common is that they leverage the (large) size of the communities involved in a way that the large quantity of observed data can be exploited to estimate parameters and discover patterns useful for implementing contextual search.

In IR and related disciplines there have already been research works that to a certain extent investigate how members of a community interact, perhaps indirectly, to building knowledge that is further exploited by the community (the link analysis methods addressed in this section are an example and the earlier bibliometrics is another notable example).

Numerous papers addressing social variables as contextual variables are based on link analysis algorithms since a graph is a natural way to represent a community; nodes are members and edges are relationships between members. However, in this survey, we not only address link analysis but also address other statistical models suited to mining useful information from community contextual variables. To this end, we are drawing the reader's attention to a couple of social variables,

that is, “tweets” and tags, which can efficiently provide useful information about the context on a large scale.

Another research area is known as digital annotation systems. Annotations affixed to digital documents is a little more recent than bibliometrics because the use and production of the digital documents has grown since the 1980s at the earliest; Agosti et al. [5] introduced digital annotation systems. However, since their advent their use is still limited, thus making the exploitation of these data for contextual search through statistical methods difficult. We focus on two types of annotation (ESP games and “tweets”) that in contrast to “traditional” annotation affixed to digital documents stimulate the implementation of large scale statistical methods for contextual search.

## **1.4 Historical Background**

In this section we provide a background to contextual search by: placing the subject among other surveys on relevance, interaction, context, and behavior; citing and surveying useful publications to the reader for further examination.

Before the relatively renowned and growing interest in contextual search viewed in the recent literature of IR, context had been on the scene, or perhaps better stated behind the scenes, for many years (perhaps for decades) as the IR literature since the 1970s shows. As the literature is by now quite vast, we can distill only some aspects and issues and cannot be more exhaustive than the publications already available on this topic.

This section is then devoted to providing a summary of and the references to the publications in which contextual search has been thoroughly considered. These publications may provide the reader with complementary information, and give a background to this survey. In particular, this section draws the reader’s attention to the papers by Belkin et al. [18]; Ingwersen and Järvelin [78]; Mizzaro [131]; Ruthven [149]; Saracevic [152]; Spink [159].

### **1.4.1 Relevance**

Because users of an IR system assess whether a document is relevant in a context, context has been a crucial aspect of relevance for decades.

Hence, relevance is intrinsically dependent on context. Due to the complexity of context and relevance, the most common IR models are a mere simplification of the reality in which users are called on to assess the relevance of documents to their information needs.

If the items of a context are gathered together, a sort of relation is obtained; actually, a mathematical relation as it is intended by a DBMS. Saracevic [153, p. 1918] suggested an understanding of relevance as a relation. According to this understanding, relevance is a relation over information objects and contexts which include information needs, tasks, and other elements. In Saracevic's review, context is an element of relevance ("Relevance has a context") and it is viewed as a complex, dynamic "interaction between a number of external and internal aspects, from a physical situation to cognitive and affective states, to motivations and beliefs, to situations, and back to feedback and resolution." Context is "ambiguous, even amorphous" and at most "context is a plural."

In the review of relevance authored by Mizzaro [131], context "includes everything not pertaining to topic and task, but however affecting the way the search takes place and the evaluation of results." This definition suggests the view that the user has some context that is not stated in the query but which we could nonetheless model. Mizzaro's paper also cites literature relevant to context introduced as a factor, component or container of the content, user, task, and so on.

#### 1.4.2 Anomalous State of Knowledge

The Anomalous State of Knowledge (ASK) by Belkin et al. is another useful element for understanding contextual search. The first part of the paper reported by Belkin et al. [18] introduces the ASK hypothesis stating "that an information need arises from a recognized anomaly in the user's state of knowledge concerning some topic or situation and that, in general, the user is unable to specify precisely what is needed to resolve that anomaly" [18, p. 62]; the second part reported by Belkin et al. [19] describes an experiment. The information need of the ASK hypothesis stems from a "topic or situation" which might better be named as problematic situation or task. In the words of Belkin et al., "the user, faced with a problem, recognizes that her/his state of

knowledge is inadequate for resolving that problem, and decides that obtaining information about the problem area and its circumstances is an appropriate means toward its resolution.”

When the ASK hypothesis is valid, the user is unable to make his information need explicit because what he would be asked to say is precisely what he does not know. A consequence of this impossibility which is relevant to this survey is that, to address the ASK, an IR system should be interactive and iterative, thus calling into play various contextual factors such as query intents, personal interests, and document qualities. Sometimes, the combination of different contextual variables leads to concept networks. Belkin et al. [18, p. 68] defined concept networks as networks of inter-related documents and named them as “formal context.” Such a network becomes a description of context and at the same time a source of evidence from which data can be observed to represent context. Networks of concepts have been further elaborated in Agosti et al.[4] within the most naturally interactive system, that is, hypermedia systems.

### **1.4.3 Interactive Information Retrieval**

Ingwersen and Järvelin [78] introduced the Integrated Cognitive Research Framework for IR. The components of this framework are: information objects (e.g., documents); the IT component (e.g., search engines); the interface (e.g., WWW clients); the cognitive actor (e.g., the user); the socio-cultural and organizational context (e.g., the workplace or the community). Between the components, which are depicted in Figure 1.1, there are influence or exchange relationships depicted as unidirectional and bidirectional arrows, respectively, and there are solid or dashed unidirectional arrows corresponding to influence and influence over time, respectively. Within the Integrated Cognitive Research Framework for IR, the definition of context suggested in Ingwersen and Järvelin [78] becomes: “in information seeking and retrieval actors and objects [are] associated with each component of the cognitive information seeking and retrieval framework function as context for their own elementary cognitive structures (intra-object context), as context to one another (inter-object context), and in context of

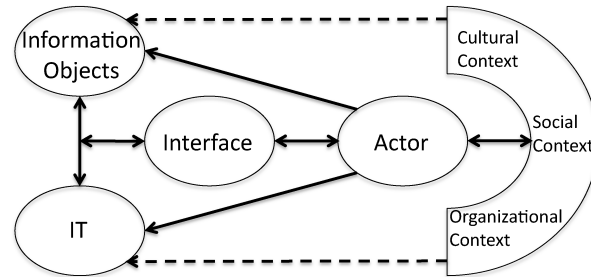


Fig. 1.1 Ingwersen and Järvelin [78]’s Integrated Cognitive Research Framework for IR.

the interaction processes between framework components, which themselves are contextual to each other. In the latter case one may talk about social/organization/cultural as well as systemic contexts. The context of interactive IR processes ranges from algorithmic IR processes in context of interactive IR as well as information seeking processes to information behavior. All information seeking and retrieval components and activities are in context of common social, physical and technological infra-structures as well as their history over time.”

As it happens, circumscribing a notion like that of context to something simpler and perhaps simplistic makes its implementation easier or more understandable than general or perhaps vague definitions. An example is a user interface-oriented notion of context which would help visualize the different components of Ingwersen and Järvelin’s framework. Ruthven [148] gives a user interface-oriented notion and states that “Our ideas on context (from both a soft and hard laboratory perspective) often manifest themselves at the interface.” Lalmas [106] adopted this definition.

Ruthven [149] some years later provides another definition: context is “a complex set of variables describing our intentions, our personal characteristics, the data and systems available for searching, and our physical, social and organizational environments” or it is also thought as the fact that “personal [context] information can cover any information that we have experienced (such as webpages we have visited), information that we have received (such as email) or information that we have created (such as documents or images). [...] [T]he range of contextual factors that might be important is vast ranging from age, physical

and cognitive ability (which may require altering the presentation of search results as well as the selection of results), learning styles, education level, or mood of searcher. The most common personal context investigated so far is the searchers topical search interests, particularly through applications of information filtering.”

To personal context, we may add social context, which is somehow related yet independent of other contexts since it is about “how people use systems and for what purposes. We can mine this information — the context of use — for many purposes including filtering information to obtain better search results” according to Ruthven [149]. From this point of view task is the information problem, for example, finding a holiday destination, writing an essay, giving a lecture, which is the reason why the user expresses his information need through queries, browsing, clicking, etc. Thus, task context covers any information that describes the user’s problem and that makes relevance, usefulness or authoritativeness of documents dependent on the task, with all the other variables being equal.

Space–time reality is perhaps the most intuitive and common setting where we experience context. Thus, it is quite straightforward to define physical context as the container of important data for providing situationally relevant information (e.g., GPS coordinates or time). Similarly, environmental context relates to any information about the type of location where the user’s search takes place (e.g., whether the user is in a public place, the weather is nice, the roads are congested) according to Ruthven [149].

Contextual search can barely be separated not only from IR and information seeking and retrieval but also from the notion of human information behavior defined by Spink [159] as follows: human information behavior “refers to a wide range of processes which people employ when engaged with information and to related cognitive and social states and effects”. In a sense, human information behavior studies are orthogonal to ASK, IR and information seeking and retrieval since they aim at understanding how and why the users interact with information when this information is contained in documents or queries. Spink in particular is interested in the user’s behavior during the formulation of the ASK. She defines information seeking and retrieval



as “one sub-process within human information behavior that includes the purposive seeking of information in relation to” an ASK because information seeking and retrieval starts when an ASK has been recognized, continues when relevance assessments have been observed and ends when the ASK has been solved. From this description it is then clear that information seeking and retrieval is as highly dependent on context as human information behavior is. The remarks made by Spink [159] about human information behavior within communities and the personal dimension are relevant to this survey.

## 1.5 Concluding Remarks and Suggestions

The computational framework underlying contextual factors, contextual variables, and statistical models is the main conceptual contribution of this survey. Other researchers are allowed to place other contextual factors, contextual variables and statistical models in this framework, thus preserving the overall consistency of the illustration of contextual search proposed in this survey. Some results illustrated in the remaining sections may well be placed in more than one contextual factor (e.g., understanding the intent of a given user may be placed in Section 2, in Section 3, or both). However, these decisions are a matter for the researchers implementing this framework. Appendix A briefly illustrates some prototypes of contextual search tools.

We conclude by giving some bibliographic references relevant to the computational framework introduced in this survey and to the general notion of contextual search. Alpaydin [7] describes support vector machines. Alpaydin [7] is a reference on machine learning. Azzopardi [12] gives a thorough study that starts from theoretical issues, investigates whether and how language models can be an efficient and effective theoretical framework for contextual search, and ends with experiments. Bai et al. [14, 15] are examples of text window-based context papers with co-occurrence analysis, an interesting modeling of contextual factors based on language models and an analysis of domain knowledge and language model combination. Bartholomew et al. [17] provide a perspective of the factorial models that are relevant to the notion of computational

framework used in this survey. Bian et al. [23] are worth reading as for the Expectation-Maximization algorithm. Blei et al. [24]'s is the original publication on latent Dirichlet allocations. The notion of geographical variable is discussed by, for example, Cai [33]. The remarks made by Chakrabarti et al. [41] on how to build an effective model and avoid bias, overfitting, etc. are useful to a newcomer to machine learning because they explain basic issues in a realistic scenario. Croft and Lafferty [47] survey language models for IR. The study by Efthimiadis [57] describes query expansion whereas the paper by Carpineto and Romano [38] is an up-to-date survey of this topic. Feller [60]; Levinson et al. [113]; Rabiner and Juang [140] are some reference publications on Bayes' rule, Markov chains and hidden Markov models. Halmos [68] explains Singular Value Decomposition and in general vector spaces. The paper written by Hu et al. [75] is easy to read and has a computational flavor. As for interaction variables, the reader may want to spend some time reading Inmon [79, 80] who introduced the notion of time-variancy, since click-through datasets may be viewed as an instance of data warehouses. The special journal publication edited by Jones and Purves [90] is a useful reference on the issues of geographical variables. The papers on implicit relevance feedback by Kelly and Belkin [95, 96]; Kelly and Fu [97]; Kelly et al. [98]; Kelly [92, 93, 94] are definitely worth reading. The survey by Lalmas and Ruthven [107] provides a precise, recent and exhaustive account of relevance feedback. Lau et al. [108] address context at difference abstraction levels, from the conceptual, to the logical up to the statistical level. Lau et al. [109] present an interesting application of their theoretical framework and show that the vector space model is still a good baseline for search in context. Metzler and Croft [130] illustrate conditional random fields. Ponte and Croft [138] introduce language models for IR. The notion of geographical variable is also discussed by Reichenbacher [144]; Reichenbacher and De Sabbata [145]. The paper written by Shannon [155] is the reference for entropy. The papers on exploratory search by White et al. [172]; White and Kelly [173]; White et al. [170, 171]; White and Roth [174]; White [169] are also useful reading.

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