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1



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ABSTRACT

This is a thorough analysis of two techniques applied to Geographic Information Retrieval 23 (GIR). Previous studies have researched the application of query expansion to improve the 24 25 selection process of information retrieval systems. This paper emphasizes the effectiveness of the filtering of relevant documents applied to a GIR system, instead of query expansion. 26 Based on the CLEF (Cross Language Evaluation Forum) framework available, several exper-27 28 iments have been run. Some based on query expansion, some on the filtering of relevant documents. The results show that filtering works better in a GIR environment, because rel-29 30 evant documents are not reordered in the final list.

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34 1. Introduction

Geographical information is recorded in a wide variety of media and document types. Over the past few decades, information technology for accessing geographical information has focused on the combination of digital maps and databases that characterize the majority of Geographic Information Systems (GIS) (Bolstad, 2005; Chang, 2007; Wise, 2002). It is only in recent years that much attention has been paid to the development of computer systems to retrieve geographically specific information from the relatively unstructured documents that compose the Web (Jones & Purves, 2008; Larson, 1996; McCurley, 2001).

Much current research in Geographic Information Retrieval (GIR) can be regarded as an extension of the field of Information Retrieval (IR) (Baeza-Yates & Ribeiro-Neto, 1999), and indeed has primarily been undertaken by researchers from the IR rather than the GIS research community. It includes all of the research areas that have traditionally made up the core of research in Information Retrieval, except that it has an emphasis on spatial and geographic indexing and retrieval (Larson, 1996). GIR is therefore concerned with improving the quality of geographically specific information retrieval with a focus on access to unstructured documents (Jones & Purves, 2008).

This paper aims at a brief description of GÎR and a thorough analysis of two techniques, namely query expansion and filtering of relevant documents.

Because a GIR system can be seen as an IR system from a functional point of view, GIR systems can work with the same document collections as IR systems, accepting that the queries usually include location entities (something that has a dis-

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2

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M.Á. García-Cumbreras et al. / Information Processing and Management xxx (2009) xxx-xxx

tinct, separate existence, concerning a location) and spatial awareness (a well thought-out awareness of things in the space 51 52 around us). For instance,

53 • "Wine regions around rivers in Europe."

- "Cities within 100 km of Frankfurt." 54
- "Whisky making in the Scottish Islands." 55

56 • "Events at St. Paul's Cathedral." 57

Evaluation campaigns such as the **Cross Language** Evaluation Forum¹ (CLEF) have tasks to evaluate geographical IR sys-58 tems, all grouped under the name "GeoCLEF"². The aim of GeoCLEF is to provide the necessary framework in which GIR systems 59 are evaluated for search tasks involving spatial and multilingual aspects. We have participated in the last three editions of GeoC-60 LEF (Perea-Ortega, Cumbreras, Vega, & Lpez, 2007, 2008; Vega, Cumbreras, Lpez, & Perea-Ortega, 2006).

There are several approaches to solve the GIR task, including simple IR systems that do not use geographical terms or 62 spatial references. More complex systems rely on Natural Language Processing (NLP) methods to detect locations and spatial 63 references. Some techniques that are often applied are "geographic entities extraction", "semantic analysis", "geographical data 64 bases" (thesauri, gazetteers), "query expansion" methods and "geographical disambiguation". 65

Most of the systems presented in GeoCLEF, for the English monolingual task between 2005 and 2008, preprocess the col-66 67 lections and the queries applying a *stop-word* list (Salton, 1971) (to delete the most frequent words in each language) and a 68 stemmer (Porter, 1980) (to extract the stem of each non-stop-word).

They also use a Named Entity Recognizer (NER), a module that detects and recognizes named entities. Some groups have 69 implemented their own NER module using geographical databases and thesauri (FerrTs & Rodrfguez, 2007; Larson, 2007), 70 71 but most groups use an external one, like LingPipe³ (Andogah & Bouma, 2007; Buscaldi & Rosso, 2007; Hu & Ge, 2006).

72 Although GIR is not typically multilingual, in our GIR system a translation module is necessary because it works with mul-73 tilingual queries.

74 The paper is organized as follows: Section 2 describes the complete system and the operations; Section 3 presents the resources used in this empirical comparison and the experimental results; and in Section 4 the conclusions of this work 75 are written. 76

77 2. System overview

The SINAI GIR system is made up of five main subsystems: Translator, Entity Extraction, Geo-Information Extraction, Infor-78 79 mation Retrieval and Filtering and Re-ranking. They are introduced in the following section. For query expansion experiments, 80 we have added a Query Expansion subsystem. In addition, we make use of the Geonames⁴ gazetteer as geographic knowledge base for the whole system, and Lemur⁵ as IR index-search engine. 81

First of all, each translated query is preprocessed and analyzed by the Geo-Information Extraction subsystem, identifying 82 83 their geo-entities and spatial relationships. Each query will be run later against the IR subsystem. On the other hand, the document collection is preprocessed and its entities are extracted by the Entity Extraction module. Finally, the documents 84 recovered by the IR subsystem are filtered and re-ranked by means of the Filtering and Re-ranking subsystem, making use 85 of all geo-information extracted from document collection and queries. 86

- 2.1. System design and operations 87
- As we can see in the Fig. 1, the main modules of our system are: 88
- Translator subsystem. We have developed and used the SINTRAM⁶ translator (GarcíaCumbreras, Ureña-López, Martínez-89 Santiago, & PereaOrtega, 2006). This own subsystem translates different queries into English and implements some heuristics 90 in order to combine various translations of the same query. A comprehensive evaluation showed that Systran⁷ worked best 91 for German and Portuguese (Vega et al., 2006). 92
- **Entity Extraction subsystem.** GATE⁸ is an infrastructure for developing and deploying software components that process 93 94 human language. It includes a NER module to detect and to recognize entities.

³ http://www.alias-i.com/lingpipe.

⁴ http://www.geonames.org/. Geonames is a geographic database which contains over eight million geographical names and consists of 6.3 million unique features whereof 2.2 million populated places and 1.8 million alternate names.

- http://www.lemurproject.org/.
- ⁶ SINai TRAnslation Module.
- 7 http://www.systransoft.com.
- ⁸ http://gate.ac.uk/.

¹ http://www.clef-campaign.org/.

² http://ir.shef.ac.uk/geoclef/.

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M.Á. García-Cumbreras et al./Information Processing and Management xxx (2009) xxx-xxx

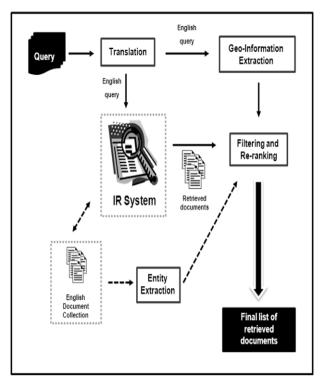


Fig. 1. Architecture of the SINAI GIR system.

- Geo-Information Extraction subsystem. The aim of this module is to recognize spatial relationships and locations in the queries:
- As in the *Entity Extraction* subsystem, we have used GATE in order to detect the geo-entities. All locations detected are also verified using the GeoNames gazetteer.
- In order to find spatial relationships, we have used manual rules. A spatial relationship is a constraint that appears in a spatial query to select spatial objects (Clementini, Sharma, & Egenhofer, 1994). They may include such simple selections as, "Lakes *in* the state of Maine", or more complex ones like "Find the *shortest path* from Boston to Bangor". Spatial relationships are divided into three main categories (Clementini & DiFelice, 2000): topological (e.g., *museums in Rome*), metric (e.g., *Rome is not far from L'Aquila*) and projective (e.g., *the shops on the right of the road*). Our system detects and recognizes topological spatial relationships. Some examples of them are: *in*, *of*, *near*, *north of*, *next to*, *in or around*, *in the west of*, etc.
- Information Retrieval subsystem. We have used Lemur as IR index-search engine.
- Filtering and Re-ranking subsystem. It is intended to filter the list of relevant documents recovered by the IR subsystem, establishing what of them are valid, depending on the locations and the spatial relationships detected in the query. Another important function is to establish the final ranking of documents, based on manual rules and the initial position of the document in the ranking. This process is explained in more detail in the next section.
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113 2.2. Relevant documents filtering

This process filters the list of relevant documents recovered by the IR system, validating them and obtaining a new ranking. This validation uses two important indexes:

Documents Index. It stores the preprocessed information of documents collection. We have preprocessed the collection with linguistic tools to remove *stop-words* and to mark stems. This index contains all the stems from the collection, including the original entities. We have used Lemur to build the documents index.

- **Geographical Index**. It stores all the locations detected in the collection by the *Entity Extraction* subsystem. All entities typified as *LOC* (location) by the NER are checked using the GeoNames gazetteer. Organizations are not included.
- 122 On the other hand, before the filtering process, the *Geo-Information Extraction* subsystem determines the type of each 123 location detected in the query. We have considered four main types of locations:

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| | 4 M.Á. García-C | Cumbreras et al./Information Processing and Management xx. | x (2009) xxx-xxx |
| 104 | Continent Coollands used the | CONT and for the continuet true It was an | |
| 124 125 | | <i>CONT</i> code for the continent type. It recognize th America, Oceania and Antarctica. | es seven entities as continents: Europe |
| 125 | | dministrative divisions of a country as country | type CeoNames uses several codes in |
| 120 | ~ - | ADM2, ADM3, ADM4, ADMD, etc. We have also | |
| 127 | as country type. | | considered pointear entities (rel code |
| 129 | | GeoNames to identify a city or a village: PPL fo | r populated place <i>PPLC</i> for capitals of a |
| 130 | | ted localities, <i>PPLS</i> for populated places, etc. We | |
| 131 | entity as city type. | | |
| 132 | 0 0 0 1 | ot been considered as some type before. | |
| 133 | | 51 | |
| 134 | Fig. 2 describes the basic archited | ture for the filtering process. | |
| 135 | In the beginning, the rules applie | ed in the filtering process were very restrictive | e, and few heuristics were used. For in- |
| 136 | | s valid if it contained only one location which a | |
| 137 | tering process did not make a re-rai | nking with the valid documents. They are inclu | uded with the score assigned by the IF |
| 138 | system. | | |
| 139 | In the last experiments carried of | out in 2008, the filtering process includes new | w heuristics and a re-ranking process |
| 140 | Depending on the location type and t | he spatial relationship detected in the query, the | e documents recovered by the IR system |
| 141 | are filtered and re-ranked. Some exa | mples of these filtering rules are: | |
| 142 | • If the entity type is <i>country</i> and it h | nas associated "in the north of" as spatial relation | nship, the system obtains the maximum |

- If the entity type is *country* and it has associated "*in the north of*" as spatial relationship, the system obtains the maximum and minimum latitudes of all locations which belong to that country, using again the GeoNames gazetteer. In order to estimate the *mid-latitude* of a country, the system subtracts the maximum and minimal latitudes. Any location which is above of this *mid-latitude* will be considered in the northern part of the country and, therefore, the document will be valid. Some approaches take a more sophisticated point of view of partitioning space or regions according to direction (van Kreveld & Reinbacher, 2004).
- If the entity type is *city* and its spatial relation associated is "*near*", there is a difficult problem to solve. Spatial reasoning is a complex activity that involves at least two levels of representation and reasoning; a geometric level where metric, topo-
- logical, and projective properties are handled (Herskovits, 1986); and a functional level where the normal function of an
- 151 entity affects the spatial relationships attributed to it in the context (for example, the meaning of "*near*" for a bomb is

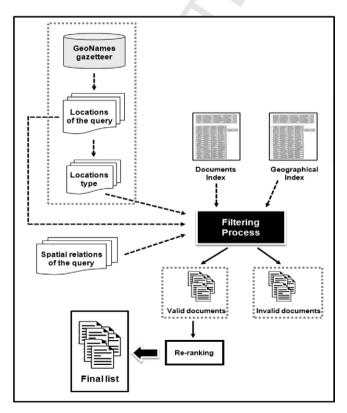


Fig. 2. Basic architecture of the filtering process.

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M.Á. García-Cumbreras et al. / Information Processing and Management xxx (2009) xxx-xxx

quite different from the meaning of "*near*" for other objects of the same size (Coventry, 1998; Garrod, Ferrier, & Campbell, 153 1999)). Without much further on this issue, only for *point-to-point* distances, we have decided to consider that a location is 154 *near* to another when their distance is less than 50 km. This value has been determined based on experiments carried out 155 in the framework proposed in this paper. In order to measure the *point-to-point* distance in kilometers between two loca-156 tions, we have used the *Great-Circle* formula Sinnott (1984):

$$D = \arccos((\sin a)(\sin b) + (\cos a)(\cos b)(\cos P))$$

159 where

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160 *D* is the distance, in kilometers;

- 161 a is the latitude of point A;
- 162 b is the latitude of point **B**;
 - *P* is the longitudinal difference between points A and **B**.
- If the entity type is *continent* or *country* and its spatial relationship associated is "*in*", "*of*", "*at*", "*on*", "*from*" or "*along*", the system will accept the document wether it has at least one location that belongs to that continent or country.
- After the filtering process, in order to make the final ranking, the system handles two documents lists:
- The list of valid documents. In this list there are only documents recovered by the IR system which satisfy at least one of the filtering rules.
- The list of invalid documents. This list includes the documents which have not satisfied any filtering rule.

The valid documents appear in the final list, whereas the invalid ones are deleted. The initial score of each document is provided by the IR system. For the re-ranking of the valid documents, the system increases their initial scores depending on the filtering rules which comply and the position of the document in the ranking:

$$RSV_{new} = \log(RSV_{old} + 1) + w_{rule}$$

where w_{rule} is the weight associated to the filtering rule that has been applied to the document. These weights have been optimized and established empirically using the framework proposed in this work.

180 2.3. Query expansion

We have developed a *Query Expansion* architecture in order to carry out the query expansion approach. The aim is to expand queries with many relevant words as possible. This applies not only to geographical terms, like countries, towns and cities, but also to synonyms, terms from the thesauri collections and any word in the same semantic domain.

184 The *Query Expansion* subsystem is made up of three main modules:

- Named Entity Recognition. It detects and recognizes the location entities in the queries in order to expand the topics
 with geographical data. Examples of location terms are: towns, cities, capitals, countries and even continents.
- Ceographical Information. The purpose of this module is to expand the locations detected in the previous module, using
 geographical information of Geonames. To do this, it is also necessary to detect and recognize spatial relationships in the
 query.

• **Thesaurus Expansion**. This module has its own thesauri collection, generated from the GeoCLEF training corpus according to a very high word co-location rate (Croft & Yufeng, 1994; Jones, 1971). This module finds the best thesaurus terms and adds them to the query. All terms are considered, including the geographic entities.

193

194 Fig. 3 describes the architecture for the query expansion.

195 **3. Experiments and results**

In this section the experiments carried out during the years 2006–2008 are described. We have used the evaluation framework provided by GeoCLEF (Gey et al., 2006; Mandl et al., 2007). Its aim is to evaluate GIR systems for search tasks, involving spatial and multilingual aspects. This framework provides an English document collection and different textual queries in several languages in order to test each GIR system:

The document collection. It consists in 169,477 documents, composed of stories and newswires from the British news-paper *Glasgow Herald* (1995) and the American newspaper *Los Angeles Times* (1994) (Braschler & Peters, 2004). This collection contains stories covering international and national news, therefore representing a wide variety of geographical regions and places. The documents have a common structure: newspaper-specific information like date, page, issue, special filing numbers and usually one or more titles, a byline and the actual text. They were not geographically tagged and contained no semantic location-specific information (Mandl et al., 2007).

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M.Á. García-Cumbreras et al./Information Processing and Management xxx (2009) xxx-xxx

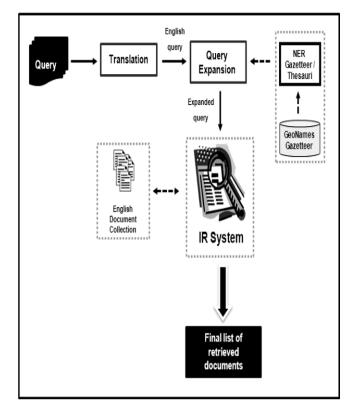


Fig. 3. Architecture of the query expansion.

- The textual queries. They are textual descriptions with three fields: *title* (T), *description* (D) and *narrative* (N). In some experiments we have used all fields (TDN), and other only TD labels. The format of these queries are showed in the Fig. 4.
- 209
- 210 3.1. Experimental method
- 211 The experiment framework has been set as follows:
- (1) The document collection and the queries are preprocessed using the English *stop-words* and the Porter stemmer algorithm. Geo-entities are also extracted.
- (2) The non-English queries are translated by means of the *Translator* subsystem.
- (3) In order to generate the *Documents Index* (see Section 2.2), we have used Lemur. It is also used to retrieve the relevant documents for each query, applying traditional weighting methods such as TF · IDF, Okapi (Robertson & Walker, 1999) and the use of Pseudo-Relevant Feedback (PRF) (Salton & Buckley, 1990). We have used the *Okapi* + PRF weighting
- 218 function because it always offers the best performance in all experiments.

```
<top>
<num>10.2452/58-GC</num>
<title>Travel problems at major airports near to London</title>
<desc>To be relevant, documents must describe travel problems
at one of the major airports close to London.</desc>
<narr>Major airports to be listed include Heathrow, Gatwick,
Luton, Stanstead and London City airport.</narr>
</top>
```

Fig. 4. Example of a query.

M.Á. García-Cumbreras et al./Information Processing and Management xxx (2009) xxx-xxx

(4) For the query expansion experiments, we have used the GeoNames gazetteer in order to expand with geographic
 terms, and a thesaurus generated from the collection in order to expand with terms treated like synonyms. Both
 approaches are described in Section 2.3.

223 The final list of documents retrieved is evaluated using the relevance judgements provided by GeoCLEF (Gev et al., 2006: 224 Mandl et al., 2007) and the TREC evaluation method. The evaluation has been accomplished by using the Mean Average Pre-225 cision (MAP) (Harman, 1994) and the R-Precision. The MAP measure computes the average precision over all queries. The 226 average precision is defined as "the mean of the precision scores obtained after each relevant document is retrieved, using zero 227 as the precision for relevant documents that are not retrieved" (Salton & Buckley, 1990). R-precision is the precision at the number of relevant documents in the collection for the query. If R-precision is 1.0, it means a perfect relevance ranking and a 228 perfect recall. Recall is a measure of the extent to which relevant documents are found or retrieved. Recall is 1.0 when every 229 230 relevant document is retrieved.

Table 1 shows all results per experiment using the GeoCLEF 2006 topics. Table 2 shows the same experiments but using the GeoCLEF 2007 topics. We have summarized them, showing only the best results. They have been achieved using all topic labels (TDN). The results obtained using TD labels are worse in general.

3.2. Experiments with query expansion

In this section we study the impact of the query expansion approach. We considered as baseline case for this approach not
 use any expansion method. We compared two types of query expansion:

- **Using geographic data**. We have used several heuristics or manual rules in order to expand locations using geographic data. This geo-information is extracted from the GeoNames gazetteer. Previously, the spatial relationship and the location included in the query have been detected. Examples of these heuristics are:
- If the location type is *continent* and it has associated *in*, *of* or *from* as spatial relationships, we expand the query with the
 highest population countries of that continent. For example, if the query is "*Wine regions around rivers in Europe*", we
 expand it with *France, Germany, Russia* and *United Kingdom* as geo-terms, and therefore the preprocessed query would
 be "wine region river europe france german russia united kingdom".
- If the location type is *country* and it has associated *in*, *of* or *from* as spatial relationships, we expand the query with the highest population cities of that country. For instance, if the query is "*Eta in France*", we expand it with *Paris*, *Marseille* and *Lyon* as geo-terms, and therefore the preprocessed query would be "*eta france paris marseille lyon*".

| Language | Experiment | R-Precision | MAP |
|----------|---------------------------------|-------------|--------|
| English | Baseline | 0.2934 | 0.3223 |
| - | Expansion using geographic data | 0.2028 | 0.2295 |
| | Expansion using thesaurus | 0.2261 | 0.2610 |
| | Filtering | 0.2934 | 0.3225 |
| German | Baseline | 0.1818 | 0.1965 |
| | Expansion using geographic data | 0.1506 | 0.1838 |
| | Expansion using thesaurus | 0.1556 | 0.1864 |
| | Filtering | 0.1774 | 0.2113 |
| Spanish | Baseline | 0.2335 | 0.2572 |
| | Expansion using geographic data | 0.2341 | 0.2604 |
| | Expansion using thesaurus | 0.2334 | 0.2565 |
| | Filtering | 0.2260 | 0.2634 |

Table 1 Summary of results using the GeoCLEF 2006 topics.

Table 2

Summary of results using the GeoCLEF 2007 topics.

| Language | Experiment | R-Precision | MAP |
|----------|---------------------------------|-------------|--------|
| English | Baseline | 0.2578 | 0.2619 |
| | Expansion using geographic data | 0.2376 | 0.2380 |
| | Filtering | 0.2716 | 0.2661 |
| German | Baseline | 0.0666 | 0.0652 |
| | Expansion using geographic data | 0.0819 | 0.0842 |
| | Filtering | 0.0827 | 0.0763 |
| Spanish | Baseline | 0.2200 | 0.2361 |
| | Expansion using geographic data | 0.2060 | 0.2076 |
| | Filtering | 0.2369 | 0.2441 |

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30 April 2009 Disk Used

M.Á. García-Cumbreras et al./Information Processing and Management xxx (2009) xxx-xxx

- If the location type is *city* and it has associated *near* as spatial relationship, we expand the query with the highest population cities closest to that location. In order to measure the *point-to-point* distance in kilometers between two locations, we have used the *Great-Circle* formula (see Section 2.2). For example, if the query is "*Travel problems at major airports near to London*", we expand it with *Luton, Reading* and *Maidstone* as geo-terms. The preprocessed query would be "travel problem major airport london luton reading maidstone".
- **Using thesaurus**. We have generated an own thesaurus from the GeoCLEF document collection, attending to a very high word co-location rate. We have used all the preprocessed words from queries, including the geographic terms. We expand the queries with the synonyms that have a similarity higher than a predefined threshold. The best results were obtained with a threshold value of 0.7. An example of query expansion using thesaurus is showed in the Fig. 5.
- The analysis of results using the GeoCLEF 2006 topics shows that the use of geographical and thesaurus information does not improve the retrieval. The expansion method obtained poor results compared to the simple model in which we only use an IR system (baseline experiment). The use of the text from all topic labels improves the retrieval process, as opposed to consider the text from one or two topic labels. The query expansion with words from the thesaurus included noise but, in general, it worked better than the expansion using geographic knowledge. Two main reasons can explain the worse results obtained with the expansion of topics:
- For the information retrieval process, the inclusion of geographic knowledge as location entities introduced noise in the queries because the expansion heuristics were not correct.
- Sometimes, the NER module did not work well. For several queries the NER did not recognize some compound entities and locations (*New England*, *Middle East*, *Eastern Bloc*, etc.) and therefore, for these topics, the system was unable to realize the query expansion using geographical information.
- This query expansion method has also been tested using the 2007 GeoCLEF topics. For these experiments we have expanded the queries using geographical information only and all the labels in the IR process. Table 2 shows these results. Figs. 6 and 7 show a graphic comparison of these experiments based on MAP and R-Precision measurements respectively.
- 273 3.3. Experiments with relevant documents filtering

In this section we study the impact of the relevant documents filtering. In 2007, we completely changed the approach of our system, because previous experiments demonstrate that the query expansion heuristics applied do not improve the re-

> Original topic: Malaria in the tropics Expanded topic: malaria tropic plasmodium heimlich neurosyphili timpone vivax

> > Fig. 5. Example of a query expansion using thesaurus.

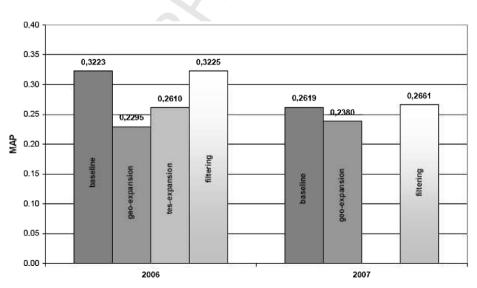


Fig. 6. Graphic comparison based on MAP measure using English topics.

M.Á. García-Cumbreras et al./Information Processing and Management xxx (2009) xxx-xxx

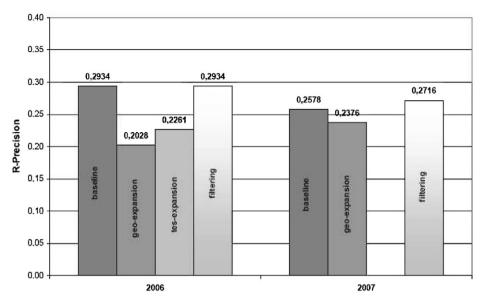


Fig. 7. Graphic comparison based on R-Precision measure using English topics.

sults. We introduced a very restrictive system: we removed those documents recovered by the IR system that did not satisfy certain filtering rules. Some of these manual rules have been explained in the Section 2.2.

For the filtering experiments we used all topic labels in the retrieval process. As baseline experiment we considered a simple information retrieval without applying any filtering process. We tried several weighting functions in the IR process (TF · IDF, Okapi and PRF), but in Tables 1 and 2 we only show the best results, using *Qkapi + PRF*.

The main conclusion of this approach is that the filtering of the documents retrieved by the IR system works better than the expansion of queries. In addition, the results obtained by the filtering process improved the baseline cases (all experiments and all languages). Instead, the difference between the use of query expansion approach and the use of document filtering is important.

As we can see in Fig. 6, using 2006 English topics, document filtering has an improvement of 9.3% with regard to use query expansion with geographic information. This difference is also appreciated using 2006 German topics (2.7%) and using 2007 Spanish topics (3.6%). In other experiments, this difference is not very important, although document filtering gets always better results than query expansion, except using 2007 German topics, where the *Translator* subsystem did not work very well.

In some filtering experiments, the mean average precisions obtained with this approach and the baseline experiments are similar. The main reason is that the documents filtered are valid but their positions in the final ranking are similar to the original ones in the baseline result. The *Filtering and Re-ranking* subsystem should work better for these experiments in the future, maybe doing a thorough query analysis, identifying more accurately the geographic part in the topic. This would allow a more optimal re-ranking process for each relevant document.

295 4. Conclusions and future work

In this work we present a comparison between two approaches applied to the Geographic Information Retrieval task: query expansion vs. relevant documents filtering. We have used the GeoCLEF framework in order to carry out the experimentation. The results show that relevant documents filtering works better than query expansion in a basic GIR system, so the use of a filtering module is an important method for GIR systems.

However, the query expansion is an interesting method for some geographic queries. For instance, for those topics that contain a subregion of the world not clearly defined (*Northern Africa, Sub-Saharan Africa, Caribbean, Patagonia, Iberian Peninsula*, etc.). As future work, it would be interesting to consider the best way to make the geographical expansion of a topic, depending on the query type.

In order to improve the final results, we think that GIR systems require the combination between relevant documents retrieved by an IR system and the application of a geographical module. Usually, general approaches expand the queries with geographical information but in a wrong way. Experiments reported in this work show that a filtering process, instead of the query expansion, produces better results. In this case, an improvement could be the use of a higher number of documents recovered by the IR subsystem, so the probability of valid documents returned by the filtering and re-ranking process is higher.

On the other hand, the application of Pseudo-Relevant Feedback (PRF) and the traditional Okapi weighting method works
 well and good results are obtained by the IR subsystem.

30 April 2009 Disk Used

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M.Á. García-Cumbreras et al. / Information Processing and Management xxx (2009) xxx-xxx

312 Our future work will include an improvement of the filtering and re-ranking process when the relevant documents ob-313 tained from the IR phase report a low score. The filtering method does not introduce new relevant documents that are not returned by the IR system, so we will focus on improving those queries that report a low IR score. Another query expansion 314 315 module will be developed for these queries. In addition, since the Web contains valid geo-referenced information, we are interested in the development of a new module that validates the final relevant documents using the Web. 316

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