

EPIMINING: Using Web News for Influenza Surveillance

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Abstract. Epidemiological surveillance is an important issue of public health policy. In this paper, we describe a method based on knowledge extraction from news and news classification to understand the epidemic evolution. Descriptive studies are useful for gathering information on the incidence and characteristics of an epidemic. New approaches, based on new modes of mass publication through the web, are developed: based on the analysis of user queries or on the echo that an epidemic may have in the media. In this study, we focus on a particular media: web news. We propose the EPIMINING approach, which allows the extraction of information from web news (based on pattern research) and a fine classification of these news into various classes (new cases, deaths, and so forth). The experiments conducted on a real corpora (AFP news) showed a precision greater than 94% and an F-measure above 85%.

1 Introduction

In the context of epidemiological surveillance, the analysis of relevant information is crucial to the decision-making process when an expert has to decide to trigger or not an alarm. The question we tackle in this article is the following: Can the flow of information exchanged on the Web be used to improve the daily monitoring of the epidemiological reality that epidemiologists themselves sometimes have difficulty to establish?

Health professionals can use news as new resources of information. However, they have to deal with the abundance of information. How to sort efficiently this pool of resources, to keep only the relevant information according to the specific issue?

The work presented in this paper is based on a collaboration between the Nevantropic company and the LIRMM laboratory. The company focuses on the development of operational monitoring of the environment at local and regional scales. In this collaboration framework, we are particularly interested in the automatic tracking of the evolution of H1N1 from heterogeneous resources of the Web. Our goal is to extract knowledge from news to provide new indicators for health authorities in order to assist them in the decision-making process. In this paper, we present a method

for automatic detection of weak signals (task of epidemiological surveillance) from a news series. This method is based on pattern research to extract information from a news corpus and on the classification of these annotated sentences of news into topics (e.g. news cases, death...).

Our contribution comes in treefold: (1) to annotate the news according to a set of concepts; (2) to classify the news into categories, (2) to identify, count and locate the cases associated with an epidemic thanks to this classification. A brief state-of-the-art is presented in Section 2. In Section 3, we present the EPIMINING approach. The conducted experiments are described in Section 4, and discussed in Section 5. Finally in Section 6, we conclude with future work.

2 Background

Agencies managing the traditional systems of epidemiological surveillance (e.g. Institut National de Veille Sanitaire in France, European Influenza Surveillance Schema, US CDC Center for Disease Control and Prevention⁴) generally use virologic data, clinical information from medical reports or pharmacies in order to monitor an epidemic. For example, in France, one of the objectives of the Sentinel Network⁵ composed of physicians and pharmacists is to monitor, according to the medical consultations, various diseases (e.g. asthma, diarrhea, influenza-like illness...). Even if these approaches are very effective, the proposed analyzes only focus on the events of the previous weeks and only few approaches are able to monitor outbreak in real-time [1].

Recently, Yahoo and Google have proposed systems which take advantage of the mass of information now available online for epidemiological surveillance. In 2008, [2] have examined the relationship between searches for influenza and actual influenza occurrences, using search queries from the Yahoo! search engine. The principle is based on the assumption that when a person has disease symptoms, he tends to query the Web like: "What are the symptoms of this disease?", "Which web sites deal with this disease?". Using the keywords chosen by the web users and their location, it is possible to define what are the trends of the users and consequently to predict potential outbreaks. [3] made a similar proposal by using the Google search engine to predict in advance the H1N1 epidemic peaks. The results of these two experiments showed that these approaches predicted an increase of the epidemic up to 5 weeks in advance from the US CDC. Even if these approaches are very effective, they require to access to the content of the user's requests and also to have a sufficient number of users to define a prediction model.

⁴ <http://www.invs.sante.fr/>,
<http://www.ecdc.europa.eu/en/activities/surveillance/EISN/Pages/home.aspx>,
<http://www.cdc.gov/>

⁵ <http://websenti.b3e.jussieu.fr/sentiweb/>

Other authors [4–6] use the information available in Web documents (news, reports, and so forth). The principle generally used is the following one: From a large volume of Web documents, they extract features such as numbers and location. The collected numbers are often used to display with different colors (more or less dark) information that may be located on a map. For example, systems such as MedISys, Argus, EpiSpider, HealthMap, or BioCaster⁶ support the global and real-time monitoring of a disease for a country. These systems are not intended to replace the traditional collaborative systems based on the exchange of official data, but allow a trigger a pandemic alert by integrating data from regions or countries for which official sources are limited or unavailable. However, these approaches suffer from some drawbacks. Because of the aggregate view, it is difficult to monitor an epidemic with a low granularity (time and space). For example, it could be interesting for the epidemiologist to identify which city or village develop new cases instead of having the information for a country. Moreover, most of the systems rarely support a fine result classification (e.g. difference between new cases or deaths). Knowing that in a country there are occurrences of the H1N1 virus is relevant but classifying the information retrieved in the news into new cases or new deaths is also informative. Finally, these methods return documents but not relevant segments in these documents. The epidemiologists have to read all the documents to find a section of interest.

Our objective in this paper is to address the limitations of the previous approaches. We are interested in the echo that may have an epidemic in the media through news that we classify automatically according to their content into categories. For this, we first use an extraction method to annotate the news based on pattern recognition and a classification algorithm that takes into account the number of patterns retrieved in the news.

The classification is not done at the level of the document but at the level of the segments in the documents. Finally, in order to assist the decision-maker, the epidemiologist, we provide different visualizations of the results either as graphical statistics (histogram, pie chart), or as geographical representations of events using GoogleMap.

3 The EPIMINING approach

In this section, we present the overall EPIMINING approach given detailed in the Figure 1.

⁶ <http://medusa.jrc.it>,
<http://biodefense.georgetown.edu/projects/argus.aspx>,
<http://www.epispider.org>,
<http://www.healthmap.org>,
<http://www.biocaster.org>

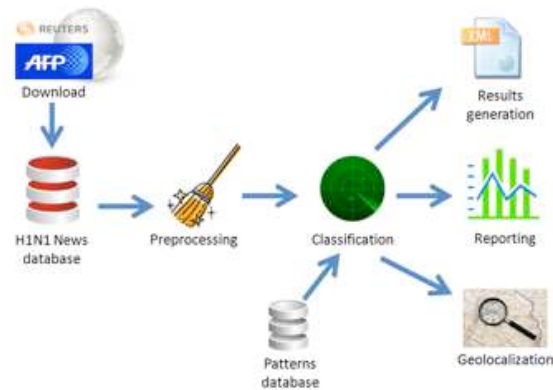


Fig. 1. EPIMINING approach

Ten dead as H1N1 flu returns to Britain

Recommander Une personne recommande ça. Soyez le premier de vos amis.

LONDON | Sat Dec 11, 2010 11:48am EST

(Reuters) - The H1N1 swine flu virus which swept the globe last year has returned to Britain with 10 people dying in the last six weeks, health officials said Saturday.

Britain's Health Protection Agency said the 10 deaths had occurred in adults all under the age of 65, most of whom had underlying health issues.

"Over the last few weeks, we have seen a rise in the number of cases of seasonal flu both H1N1 (2009) and flu B in the community," Professor John Watson, head of the HPA's respiratory diseases department, said in a statement.

Factbox
 Factbox: Tobacco - One of the world's biggest health threats
 Thu, Nov 25 2010

Related News
 Doctors encouraged pregnant women to get flu shot
 Thu, Dec 2 2010
 Haiti's cholera part of old pandemic:
 000

Fig. 2. An example of H1N1 news.

3.1 Acquisition and pre-processing of the corpus

To feed the News database, we queried sites such as Reuters or the French equivalent AFP. We used keywords associated with the disease

(e.g. swine flu, H1N1, influenza, and so forth). We tokenize and tag words that appear in the retrieved news with the TreeTagger tool [7]. For example, let us consider the subpart of the second sentence of the news presented in figure 2:

"10 deaths had occurred in adults all under the age of 65 in England"

The associated lemmatized sentence, composed of the original form of each word (i.e. first element), the grammatical category (i.e. second element), and the lemma (i.e. third element) is:

"10/CD/Card deaths/NNS/death had/VHD/have occurred/VVN/occur in/IN/in adults/NNS/adult all/RB/all under/IN/under the/DT/the age/NN/age of/IN/of 65/CD/Card

3.2 Annotation of the news

Pre-treated news are automatically annotated thanks to a Pattern Database which enables to identify the relevant concepts. We apply an approach similar to the one described in [8] who details different Information Extraction (IE) tools in order to find specific information in free texts. Like our method, the developed tools used patterns associated to part-of-speech knowledge. Note that the EPIMINING system described in this paper is more specific to the epidemiology domain. To recognize patterns in documents, we rely on their linguistic characteristics and other syntactic rules of their arrangement. Specifically, the tagged documents are parsed in order to detect patterns. The analysis started by applying a set of syntactic rules to locate all the patterns present in different sections of the document. A filter is then applied to favour the longest pattern among several patterns sharing the same lemmatised words. For example, in the sentence of the Figure 2, we identify the concept PERSON thanks to the presence of the lemmatized word "adult". Similarly, the concept YEARS_OLD is retrieved via the pattern series: <PERSON> followed by the expression "under the age of" followed by the number 65.

The Pattern Database is composed of patterns specified by an expert. These patterns were identified after a textual analysis of the news content. They take into account the specificities of the news regarding the other types of text documents. Applying this method, we identify in the previous example the concepts: NUMBER, DEATH, PERSON, YEARS_OLD, CITY. To refine the information about the location in the news, we use a database of geographic information (Geolocalisation database). After this step, documents are labelled for an easier classification: When this was possible, each sentence is associated with a number of sick and dead people, a location, a date... Finally, we obtain the following annotations:

"<NUMBER>10</NUMBER><PERSON><DEAD>death</DEAD><AGE> under 65</AGE></PERSON> <CITY> London </CITY>".

3.3 Classification

A news can contain information, which can be classified into various categories. For example, we can find in the same news information about sick and dead patients. Consequently, the news classification at the document level is often not relevant. To obtain a fine classification, we decide not to classify the news but sentences of these news. The classification is performed as follows: Each class is associated with a set of patterns. If patterns of a defined class are retrieved in a news, the one is associated to this class. For instance, the news of the Figure 2 is associated to the class Death because we have found the news with the concept DEAD. For each association between a sentence and a class, we calculate the EPIMINING score according to the following heuristic. The score equals 1 if all the elements that are expected are found in the sentence (e.g. for the association between a sequence containing a date, a number of death, a geographical location and class Death). The score is based on the reliability of extracted information. For example, if the location is not in the sentence, the search is expanded in the nearest sentence to find the missing information and the score is decreased.

4 Experiments

In order to evaluate the performance of our approach, two data sets in French were used for experiments: a database of 510 AFP news over the period of September 2009 to February 2010 and a database of 353 Reuters news over the period of January 2009 to February 2010. To analyze the quality of the returned results, 477 AFP news, and 7147 sentences have been manually annotated. The objective was to evaluate the news classification into four categories. The first two ones depends on when the cases mentioned in the news are listed: "New cases" corresponds to the description of information about new patient at a given time and "Report" corresponds to older cases. The last two categories correspond to the categorization of the patient: "Dead patient" and "Sick patient". Two types of evaluations have been conducted (1) by considering the documents as objects to be classified and (2) by considering the sentences. To evaluate the results of these two classifications, we measure the precision (ratio of relevant documents found on the total number of selected documents), recall (ratio of relevant documents found in the total number of relevant documents), and the F-measure (harmonic average between precision and recall).

In Table 1, results of the tests conducted on the news classification are reported. The best results are obtained for classes "Report" and "Dead patient". This is justified by the fact that the distinction between illness and death is not always present in the news and by the fact that the concept of novelty is more difficult to detect. Even when the analysis is conducted by an expert, the difference between the two classes is not necessarily obvious to capture.

Table 2 presents the experiments conducted on the classification of the sentences. We worked with different EPIMINING score values corresponding to the search for patterns in different sentences close to the evaluated segment. With a high confidence score (i.e. [50..100]), we obtain the best precision (83.6%).

Finally, Tables 1 and 2 show that the EPIMINING approach focuses on precision. Indeed, the patterns are often quite restrictive to return relevant elements. To increase the recall, we can consider the sentences with a large EPIMINING confidence as shown in Table 2 .

| Classes | Retrieved and relevant | Retrieved | Relevant | Precision | Recall | F-Measure |
|---------|------------------------|-----------|----------|-----------|--------|-----------|
| Dead | 100 | 106 | 128 | 94.3% | 78.1% | 85.5% |
| Ill | 43 | 55 | 65 | 78.2% | 66.2% | 71.7% |
| Report | 88 | 103 | 114 | 85.4% | 77.2% | 81.1% |
| New | 48 | 59 | 78 | 81.4% | 61.5% | 70.1% |

Table 1. News classification

| Confidence | Retrieved and relevant | Retrieved | Relevant | Precision | Recall | F-Measure |
|------------|------------------------|-----------|----------|-----------|--------|-----------|
| [0..25[| 20 | 46 | 280 | 43,5% | 7,1% | 12,3% |
| [25..50[| 58 | 97 | 280 | 59,8% | 20,7% | 30,8% |
| [50..100] | 112 | 134 | 280 | 83,6% | 40,0% | 54,1% |
| [0..100] | 190 | 277 | 280 | 68,6% | 67,9% | 68,2% |

Table 2. Sentence classification

5 Discussion

A prototype dedicated to healthcare professionals was set up by the Nevantropic company. Figure 3 shows an interface of this tool that presents various indicators that can be used for decision-making.

On the left, the evolution of the number of cases of sick and died people identified through the news dealing with H1N1 are presented over several months or years. On the right, the cases are located on a GoogleMaps at a given time. Of course, only the indicators derived from the news classification are presented on this image but of course, it is the combination of several indicators that make sense for healthcare professionals

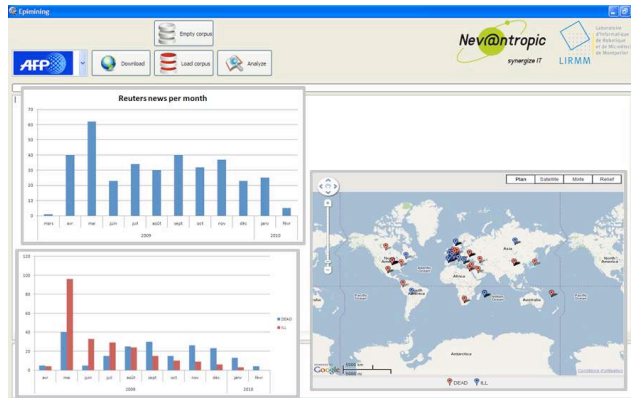


Fig. 3. EPIMINING tool

who must take a decision. For example, the tool can be used by epidemiologists who should or should not trigger an alarm or by physicians to guide their diagnosis during the visit of a patient in suspected cases of epidemic in the country where he travel back. The proposed architecture for the monitoring of H1N1 is of course adaptable to other types of epidemics.

The type of approach presented in this paper, based on the exploitation of massive data published on the Web, like the approaches proposed by Yahoo and Google, are relevant because they help to early alert health authorities. The results of these methods must be considered as new and indispensable sources of information that have to be crossed with more traditional sources of information provided by the agencies managing the traditional sources of epidemiological surveillance, either to confirm, disprove or in most cases to clarify. These methods are especially useful in geographic areas that do not have a conventional surveillance infrastructure but where the deployment of the Internet is already well advanced.

6 Conclusion and Future Work

In this paper, we have proposed a new approach, called EPIMINING , to monitor epidemics, based on automatic knowledge extraction and news classification. EPIMINING have been illustrated by a prototype for monitoring indicators on the H1N1 epidemic. The advantage of our approach, by measuring the echo of an epidemic in the media, is to be complementary to traditional surveillance networks and user's queries analysis proposed by Yahoo and Google systems for instance. The perspectives associated to our proposal are numerous. We can easily improve the classification with learning methods in order to automatically extract the representative patterns of a class. In addition, we plan to extend our

approach to other types of textual datasets (e.g. weblogs). We also plan to combine this method with the ones based on other types of datasets (air transport, meteorological, entomological data...). Finally, to answer to our initial question, we can say that the data issued from the web seem to be relevant variables, which can be included into the models of epidemics to better anticipate and predict their dynamics.

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