Automatic ICD-9-CM Classification for textual medical reports

Margherita Lazzarini, Massimiliano Fasi¹

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¹{margherita.lazzarini, massimiliano.fasi}@studio.unibo.it

Abstract

An algorithm for automatic **ICD-9-CM** classification is presented, as well as a working though experimental implementation.

The interface returns to the user a set of possible classification for a given textual description of the diagnosis. The user has to select the most suitable classification. Multiple classifications for a same session are allowed.

Tools

We used the following software frameworks:

- Server side
 - Java servlet within Tomcat container on golem.cs.unibo.it
 - jdbc Java-based data access technology
 - MySQL database on golem.cs.unibo.it
 - Python

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 - Python
- Client side
 - HTML and CSS3
 - AJAX technologies
 - JQuery framework

Database structure

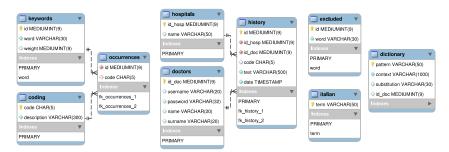


Figure: General ER Schema of the database

ICD-9-CM core tables

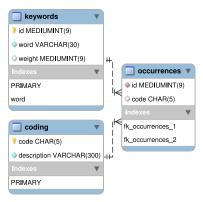


Figure: Detail of the ER Schema

- coding contains the ICD-9-CM diseases classification
- keywords contains every significant word of the classification with the corresponding weight
- occurrences links each keyword to the codes whose description contains it

ICD-9-CM history management tables

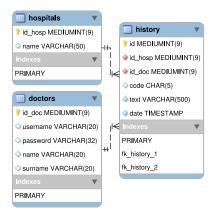


Figure: Detail of the ER Schema

- doctors contains the personal and login information of each authorized user
- hospitals contains the information about each hospital using the service
- history links the doctor with the textual diagnosis, the diagnosis time and the chosen ICD-9-CM class

Text parsing and mining tables



Figure: Detail of the ER Schema

- excluded contains the list of stopwords to be replaced during the normalization step
- italian is a complete list of Italian terms produced expanding a context-free grammar
- dictionary contains shortenings widely used in medical context, along with the corresponding expansions

Database's population

The initial state

Ab ante, the database is populated by *Python* scripts that parse comma separated files (.csv) containing the classification, the Italian dictionary and a given set of past years diagnoses.

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The state's update

During the execution, many of the tables are modified by the *Java* servlets:

- dictionary is updated both when new diagnoses are inserted and when new expansions of the shortenings are manually (by dedicated servlet) inserted
- history is updated just when a new association between diagnosis and classification is sent to the system

Algorithm's outline



The string inserted by the end user is split into **tokens**, and each one of them, if considered significant, is turned into an object of the class WeightedWord, that is closed under a structured **sum** operation.

By successive sums, a single object of that class is created, and the output is sent back to the client through an *XML* string, **parsed** by an appropriate *JavaScript* function.

Tokenization and Normalization



The medical report, sent through a *GET* request to the servlet, is divided into single words, that, if not recognized as *stopwords*, are normalized as follows:

- everything is lower cased,
- accents, if any, are removed,
- tokens that matches one element of the PATTERN column of the DICTIONARY table, are replaced accordingly to the corresponding SUBSTITUTION.

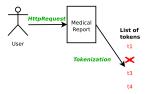
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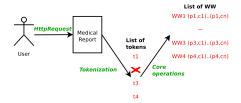
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Inverted indexes queries



For each token longer than three characters, accessing the core tables of the database, a <code>WeightedWord</code> object is created. These ones associate a string (the word) to a list of couples (<code><code></code>, <code><weight></code>). The sum object is then calculated, and the list is ordered by weight.

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Presentation of the results



The resultant WeightedWord object is sent to the client after being converted in an XML string. On the client side a JavaScript function parses it, showing a list of selectable codes with descriptions. Once the user has chosen one of the presented options, he is given the possibility of repeating the whole process, adding other diagnoses.

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Example of the XML string

```
<response>
    <item>
      <code>060</code>
3
      <description>Febbre gialla</description>
    </item>
5
    <item>
      <code>0600</code>
7
      <description>Febbre gialla silvestre</description>
    </item>
9
    <item>
11
      <code>0601</code>
      <description>Febbre gialla urbana</description>
13
    </item>
  </response>
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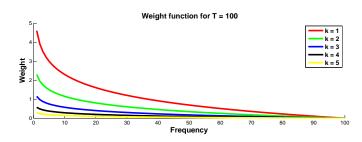
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After some trials with completely customized algorithms, we found an adaptation of the *inverse document frequency* approach (*idf*) very suitable for our purposes.

We obtained a good receipt by replacing the *term frequency* related factor (tf) with another metric, concerned with the position of the word in the string, as described into details below.



The formula we devised to calculate the weight of a term i is

$$w_i^k = \frac{1}{2^k} \log \left(\frac{T}{f_i} \right)$$

where

- k is the relative position of the word i in the typed string,
- f_i is the frequency of the word in the ICD-9-CM specification,
- T is the number of tuples of the **coding** table.

Addition of two WeightedWord

Example

By a logical viewpoint, a WeightedWord object has this structure: $\{gialla, \{(060,3.45), (0600,3.45), (0601,3.45), (0609,3.45), (9793,3.45)\}\}$

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The plus() static method of the WeightedWord class performs the logical union of two sets of WeightedCode objects, adding the weights associated to the same ICD-9-CM code.

Adding repeatedly, we obtain, from the medical report, a single WeightedWord, with the list of codes ordered by the weight field. If two codes have the same weight, the one with the shortest code string is preferred, since more generic.

History management

Every time a code and its description are chosen by the user, a new row is inserted in the HISTORY table. By clicking on the link <u>Visualizza storico</u> at the end of the main page, results of last successful searches can be viewed, organized as follows:

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The DOCTORS and HOSPITALS tables are provided but not displayed yet, since we had no data to populate them.

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The collection of words to be substituted has been found working on a set of past diagnosis (from 2006 to 2011), selecting terms not in the Italian dictionary.

Searching for fibrosi the results are:

- Fibrosi polmonare postinfiammatoria
- Fibrosi tubercolare del polmone
- Tiroidite fibrosa cronica
- Fibrosi cistica
- Fibrosi endomiocardica

Searching for fibrosi cistica the results are:

- Fibrosi cistica
- Fibrosi cistica senza ileo da meconio
- Fibrosi cistica con ileo da meconio
- Fibrosi polmonare postinfiammatoria
- Fibrosi tubercolare del polmone

Searching for febbre the results are:

- Febbre tifoide e paratifoide
- Febbre da morso di ratto
- Febbre gialla
- Febbre emorragica da artropodi
- Febbre ricorrente

Searching for febbre gialla the results are:

- Febbre gialla
- Febbre gialla silvestre
- Febbre gialla urbana
- Febbre gialla non specificata
- Avvelenamento da vaccino contro la febbre gialla