

Topic categorization in Portuguese news articles

André F. Santos

CRACS & INESC-Porto LA

Faculty of Sciences, University of Porto

Porto, Portugal

afs@inesctec.pt

Abstract—This document is a model and instructions for L^AT_EX. This and the IEEEtran.cls file define the components of your paper [title, text, heads, etc.]. *CRITICAL: Do Not Use Symbols, Special Characters, Footnotes, or Math in Paper Title or Abstract.

Index Terms—topic categorization, machine learning, text mining

I. INTRODUCTION

The last decades have witnessed a fast increase on the rate of publication of digital text documents. Traditional document types, such as news articles, scientific papers or books are now published online together with new formats, such as blog posts or tweets, each having thousands or millions of new documents published each day.

And it was not only the publication step which has moved to the digital world; in fact, most often nowadays the whole document lifecycle happens digitally, with virtual tools available for preparing, writing, styling, publishing and sharing.

Having the entire workflow happening within the digital world presents some opportunities when compared to the traditional process. In particular, due to today's processing power, tasks related to the manipulation of the information contained within these documents (searching, compiling, annotating, sharing, ...) can now be performed automatically and targeting a large amount of articles.

In addition to the document content (for example, in a news article, the *title*, *lead* and *body*), its metadata is also important: author(s), date of publication, source, topic, mentioned entities and their relations, etc. Some of this metadata might be filled in and stored along with the document (e.g. *author* and *date of publication*); other is usually extracted from the document content (e.g. mentioned entities).

Online news articles first appeared as reprints from traditional newspapers; nowadays, however, they represent now the primary source of news for some segments of the population, both in developed and developing countries (whether consumed directly in the newspaper website, or indirectly (e.g. through a social media application or a feed catcher).

Unofficially known as the *fourth branch of government*, the press plays a vital role within our society, keeping us informed regarding the current state of affairs (at a local and global scale) and acting as a watchdog for the other three powers (legislative, executive and judicial). The (lack of) freedom of

press and access to the news in a given country is even often considered an indicator of a lack of democracy[1].

As such, improving the ways citizens can access the information (view it, query it and search it) contained in news articles has the potential to contribute for a more informed society and, ultimately, a better society.

An example of a feature which improves information access is the categorization of news articles by the topic (or topics) of its content. The presence of such a categorization may influence the way the information is stored, organized, displayed and queried.

The simplest way of achieving this categorization is to have the author of the article to manually introduce it (e.g. the journalist typing it on the news article authoring framework); however, this solution presents some challenges:

- it increases the amount of work the author has to do
- the author might not be sure which categories are available
- the author might not be sure which category is the best (e.g. *Economics* vs *Finance*)
- it does not scale – e.g. if the goal is to categorize an existing (large) corpus

Thus, an automated way of categorizing news articles could solve some of these problems and decrease the burden of this task.

The challenges of document classification have been well studied within the machine learning research field of study. Given a corpus of already classified documents, several algorithms might be applied to train a classifier capable of determining the category of additional articles.

II. METHODS

A. The dataset

We gathered a dataset of news articles published in *Observador*¹, one of the main Portuguese newspapers, and which stands out from the others for being fairly young (it was created in May 2014) and for existing exclusively online.

We gathered all the categories used by Observador, and ordered them from the most common to the least common. We selected the ones which had more than 1.000 articles in our dataset, and reduced our original dataset to include only articles from these categories, which were²:

¹<http://observador.pt>

²Category names translated to English by the authors.

- 1) International
- 2) National
- 3) Economics
- 4) Politics
- 5) Sports
- 6) Science
- 7) Culture
- 8) Music

We then randomly selected, from each category, 700 articles to be used to train the classifiers, and 200 to be used to evaluate their performance.

For each article, we had available its contents (title, lead, body) and several metadata fields (publication date, category, tags, etc). A truncated JSON representation of an article can be found in Listing 1.

Listing 1. Example of JSON representation of an article

```
{
  "Type" : "sapo.obj.creativework.article",
  "Source" : {
    "Name" : "Observador"
  },
  "Pretitle" : "Benfica",
  "Title" : "Ruben Amorim com rotura total do
    ligamento cruzado",
  "Author" : {
    "Name" : "Observador"
  },
  "Tags" : [
    "benfica",
    "desporto",
    "futebol",
    "ruben amorim"
  ],
  "PublishDate" : ISODate("2014-08-25T18:33:00Z"),
  "Lead" : "Depois de Fejsa, mais uma baixa. O
    internacional português [...]",
  "Body" : "<p>O pior cenário confirmou-se. O
    Benfica informou esta segunda-feira [...]",
  "URL" : "http://observador.pt/2014/08/25/ruben-
    amorim-com-rotura-total-ligamento-cruzado/",
  "CategoryPaths" : [
    "Desporto"
  ],
  "Domain" : "observador.pt",
  "Language" : "pt_PT",
  ...
}
```

B. Preprocessing the articles

C. Classification algorithms

- 1) *Decision tree*:
- 2) *k-nearest neighbors*:
- 3) *Naive Bayes*:
- 4) *Neural network*:
- 5) *Support vector machine*:

TABLE I
CATEGORIES CONFUSION MATRIX (K-NEAREST NEIGHBORS)

	ciencia	cultura	desporto	economia	mundo	musica	pais	po
ciencia	64	18	25	29	21	37	4	
cultura	4	37	16	9	19	107	3	
desporto	1	4	130	8	19	35	2	
economia	0	6	9	127	12	32	1	
mundo	9	8	25	31	82	32	1	
musica	0	6	1	1	0	180	0	
pais	8	15	24	42	17	64	14	
politica	1	8	7	36	19	24	3	

D. Most informative terms

III. RESULTS

A. Evaluation

B. Limited lexicon

IV. DISCUSSION

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- [1] Goode, L.: Social news, citizen journalism and democracy. *New media & society* 11(8), 1287–1305 (2009)