

Text mining a Portuguese book on Freemasonry: Disclosing network communities' features

Conceição Rocha^{1,3}; **Alípio Mário Jorge**^{2,3}; **Márcia Oliveira**^{1,3};
Paula Brito^{1,3}; **João Gama**^{1,3}; **Carlos Pimenta**^{1,4}

¹Faculdade de Economia da Universidade do Porto

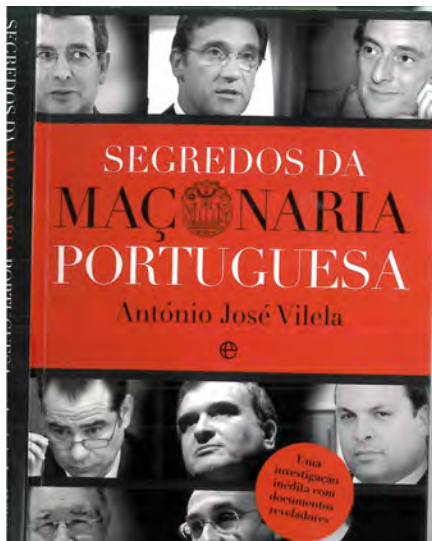
²Faculdade de Ciências da Universidade do Porto

³LIAAD/INESC TEC

⁴OBEGEF

INForum 2014

Objective



Extract named entities from a Portuguese book on Freemasonry and explore network communities based on their co-occurrences in the same sentences

Difficulties

Some difficulties:

- 1 Pre-processing the text
 - the structure of the scanned book text - page breaks
 - 'junk' like page numbers
 - mistakes/limitations of Optical Character Recognition (OCR)
- 2 Named entities extraction
 - limitations of free software on Portuguese language
 - different designations used for the same entity

Software

Program developed in R

packages: **tm**, **gdata**, **stringr**, **cwhmisc**, **openNLP** and **Hmisc**

Social network analysis - Gephi software

Methodology

Process main steps:

- Phase 1

- ① remove page numbers and empty lines
- ② remove 'junk' based on their patterns
- ③ extract the named entities using regular expressions (capital letters and lower — e.g. *presidente, câmara, deputado*)

- Phase 2

- ① tag terms list as part-of-speech
- ② remove all the terms that do not have at least one tag 'prop'
- ③ remove the first word from terms starting by 'pron-det'
- ④ remove some stop words
- ⑤ identify the named entities

Validation

12650 events corresponding to 5502 unique terms in the book

To evaluate the term extraction:

125 book' pages with 3866 named entities have been manually labeled
 (1/3 of the text book)

	Phase 1	Phase 1 + Phase2
extracted terms	5089	3075
named entities	3205	2982
<i>recall</i>	0.84	0.78
<i>precision</i>	0.63	0.97
<i>F – measure</i>	0.72	0.865

Network characteristics

4730 nodes and 24997 edges
 undirected and weighted graph

Table : Statistics attributes

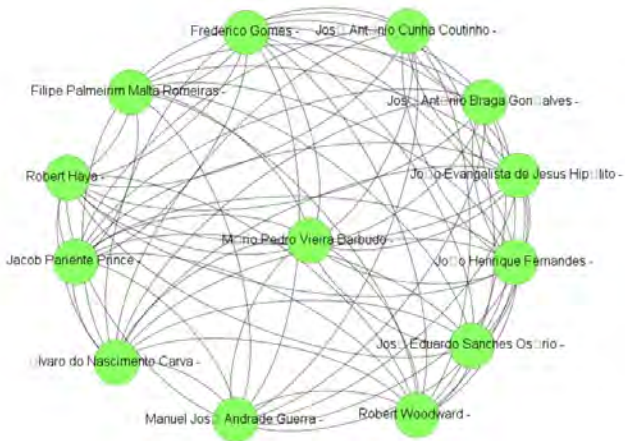
Network	average clustering coefficient	0.851
	average path length	3.445
	average degree	10.57
	average weighted degree	11.75
	network diameter	12
	network radius	1
	graph density	0.002
	modularity	0.682
	N^c of communities	268
N^c weakly connected components	238	

Main community

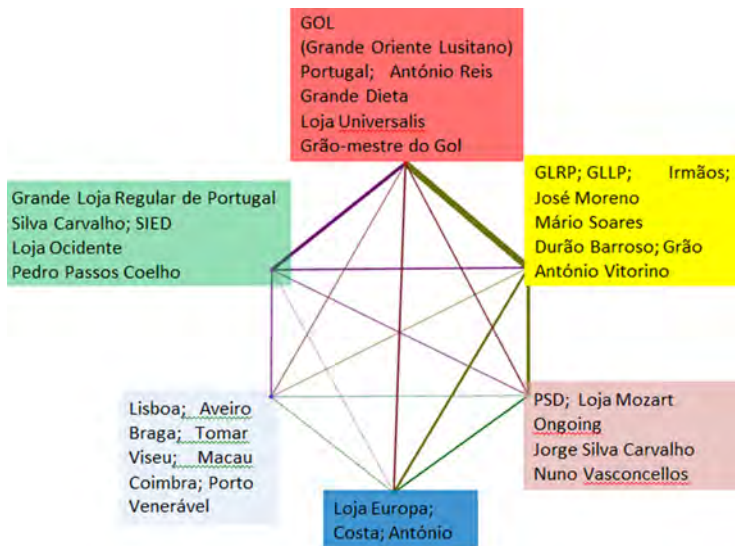
comprises the majority of nodes
heterogeneous
exhibits a rich internal structure



A component and a clique of the network



Higher-level representation of the Social network



Remarks and Conclusions

- ⇒ Inclusion of the second phase on the process improves the quality
- ⇒ F – *measure* increases from 0.72 to 0.865

Considering that:

- the text mining procedure to extract entity names is not finished
 - the relation between entities is given by their co-occurrence in the same sentence
- ⇒ The results are quite meaningful and we can see relevant connections in terms of some political organizations, politicians and other public figures

Further Work

- including an entity synonymy step and a disambiguation step
- adjusting the network model so that links between entities are based on the verbs

The results obtained so far may also be considered a step towards the creation of a text intelligence system to be used in the study of the social context of possible economic and financial offenses.

Acknowledgments

This work is partially funded by FCT/MEC through PIDDAC and ERDF/ON2 within project NORTE-07-0124-FEDER-000059 and through the COMPETE Programme (operational programme for competitiveness) and by National Funds through the FCT - Fundação para a Ciência e a Tecnologia (Portuguese Foundation for Science and Technology) within project FCOMP-01-0124-FEDER-037281. Márcia Oliveira acknowledges funding from FCT, through PhD grant SFRH/BD/81339/2011.