

# Automatically Enriching a Thesaurus with Information from Dictionaries

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- 1 Introduction
- 2 Proposed approach
- 3 Enriching TeP with synonymy in PAPEL
- 4 Evaluation
- 5 Concluding remarks



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- Thesaurus, lexical networks, lexical ontologies, ...



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  - ▶ See Princeton WordNet [Fellbaum, 1998]



# Free lexical knowledge bases for Portuguese

- Public domain thesaurus:
  - ▶ TeP [Maziero et al., 2008]
  - ▶ OpenThesaurus.PT<sup>2</sup>

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- More **complementary** than overlapping
- Fruitful to **merge** some of them in a unique broader resource

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- **Integrate synonymy information from dictionaries in a thesaurus**





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  - ① Extraction of synpairs from dictionaries



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- **Integrate synonymy information from dictionaries in a thesaurus**
  - 1 Extraction of synpairs from dictionaries
  - 2 Assigning synpairs to synsets



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- **Integrate synonymy information from dictionaries in a thesaurus**
  - 1 Extraction of synpairs from dictionaries
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  - 3 Clustering remaining pairs
- Apply the procedure in the enrichment of TeP with PAPEL



# Extracting synpairs from dictionaries

- **mente**, n: *cérebro, cabeça, intelecto*  
[**mind**, n: *brain, head, intellect*]
  
- **máquina**, n: **o mesmo que** *computador*  
[**machine**, n: *the same as computer*]



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# Assigning synpairs to synsets

- $p = (w_x, w_y) + S_a = (w_1, w_2, \dots, w_n) \rightarrow S_a = (w_1, w_2, \dots, w_n, w_x, w_y)$





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- Synonymy graph  $G$ 
  - ▶ All the extracted synpairs
  - ▶ Nodes represent words (eg.  $w_x, w_y$ )
  - ▶  $p = (w_x, w_y)$  establishes an edge between  $w_x$  and  $w_y$



## Assigning synpairs to synsets

For each synpair  $p = (w_x, w_y)$

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# Clustering remaining pairs

$G'$  is established by the remaining pairs

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- 5 For each  $S_i : S_i \cup S_j = S_j$  and  $S_i \cap S_j = S_i$ ,  $S_i$  is discarded.



# Coverage of the synpairs by TeP

POS	Synpairs	In TeP	$ C ^4 = 0$	$ C  = 1$	$ C  > 1$	$\overline{ C }$
Nouns	37,452	27.38%	14.98%	12.01%	45.63%	3.86
Verbs	21,465	43.01%	1.34%	4.04%	51.66%	6.64
Adjectives	19,073	37.60%	5.58%	8.22%	48.60%	4.26

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## Results – words

Thesaurus	POS	Words			
		Total	Ambiguous	Avg(senses)	Most ambig.
TeP 2.0	Nouns	17,158	5,805	1.71	20
	Verbs	10,827	4,905	2.08	41
	Adjectives	14,586	3,735	1.46	19
After assignments	Nouns	23,775	10,418	2.09	37
	Verbs	12,818	7,094	2.64	42
	Adjectives	17,158	6,294	1.83	22
Clusters	Nouns	8,546	701	1.15	8
	Verbs	502	8	1.02	3
	Adjectives	1,858	39	1.03	4
Final thesaurus	Nouns	30,369	12,045	1.96	38
	Verbs	13,090	7,221	2.62	42
	Adjectives	18,525	6,550	1.80	23





## Results – synsets

Thesaurus	POS	Synsets				
		Total	Avg(size)	size = 2	size > 25	max(size)
TeP 2.0	Nouns	8,254	3.56	3,079	0	21
	Verbs	3,978	5.67	939	48	53
	Adjectives	6,066	3.50	3,033	19	43
After assignments	Nouns	8,254	6.01	1,930	179	150
	Verbs	3,978	8.50	702	217	148
	Adjectives	6,066	5.17	2,369	120	110
Clusters	Nouns	3,524	2.78	2,247	0	13
	Verbs	220	2.34	174	0	6
	Adjectives	820	2.33	656	0	10
Final thesaurus	Nouns	11,778	5.05	4,177	179	150
	Verbs	4,198	8.18	876	217	148
	Adjectives	6,886	4.84	3,025	120	110

# Assignments evaluation

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- Two judges for each assignment



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POS	Sample	Correct	Incorrect	Agreement
Nouns	100 assigns. × 2	153 (76.50%)	47 (23.50%)	77.00%
Verbs	100 assigns. × 2	142 (71.00%)	58 (29.00%)	74.00%
Adjectives	100 assigns. × 2	151 (75.50%)	49 (24.50%)	75.00%



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POS	Synpair	Synset	Judge 1	Judge 2
Nouns	( <b>escrutínio</b> ,votação)	votação;voto;sufrágio	1	1
	(decisão, <b>desempate</b> )	resolução;objetivação;tenção;intenção	0	1
	(plano, <b>gizamento</b> )	planície;chã;chanura;plano;plano;planura	0	0
Verbs	(venerar, <b>homenagear</b> )	venerar;cultuar;adorar;idolatrar	1	1
	(atacar, <b>combater</b> )	atacar;inciar	0	1
	( <b>obter</b> ,rapar)	depilar;despelar;pelar;raspar;rapar;rascar	0	0
Adjectives	(grandioso, <b>épico</b> )	admirável;fabuloso;grandioso	1	1
	(delicado, <b>requintado</b> )	difícil;complicado;delicado	0	1
	( <b>falido</b> ,queimado)	queimado;incendiado	0	0



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Table: Evaluation of clustering

POS	Sample	Correct	Incorrect	Agreement
Nouns	105 × 2	179 (85.24%)	31 (14.76%)	91.43%
Verbs	105 × 2	193 (91.90%)	17 (8.10%)	87.62%
Adjectives	105 × 2	189 (90.00%)	21 (10.00%)	85.71%



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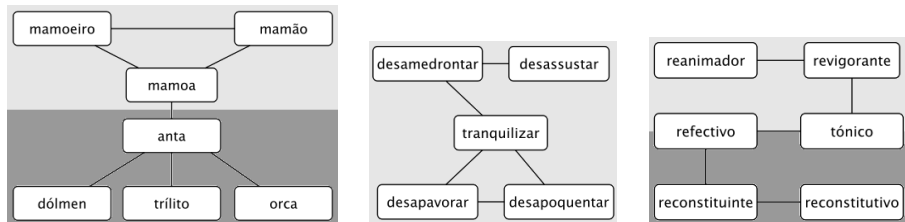


Figure: Examples of connected subgraphs and resulting clusters.





# Update: computing similarity

- Sum the adjacencies

- ▶ One vector per synset:  $[C_j] = \sum_{k=1}^{|C_j|} [w_k] : w_k \in C_j$ ;
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  - ▶ Variable cut point  $\theta$  on similarity



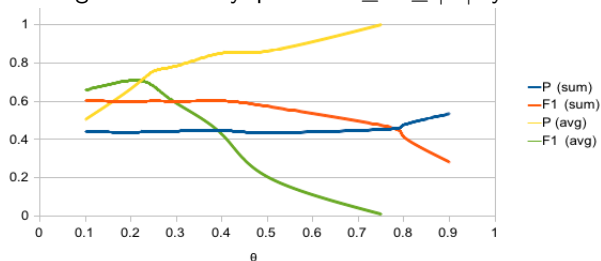
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  - ▶ Automatic creation of a lexical ontology for Portuguese



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  - ▶ Soon freely available!
  - ▶ Check <http://ontopt.dei.uc.pt>



# Thank you!



# References I

- [Fellbaum, 1998] Fellbaum, C., editor (1998).  
*WordNet: An Electronic Lexical Database (Language, Speech, and Communication)*.  
The MIT Press.
- [Gonçalo Oliveira and Gomes, 2010] Gonçalo Oliveira, H. and Gomes, P. (2010).  
Onto.PT: Automatic Construction of a Lexical Ontology for Portuguese.  
In *Proc. 5th European Starting AI Researcher Symposium (STAIRS 2010)*. IOS Press.
- [Gonçalo Oliveira et al., 2010] Gonçalo Oliveira, H., Santos, D., and Gomes, P. (2010).  
Extracção de relações semânticas entre palavras a partir de um dicionário: o PAPEL e sua avaliação.  
*Linguamática*, 2(1):77–93.
- [Maziero et al., 2008] Maziero, E. G., Pardo, T. A. S., Felippo, A. D., and Dias-da-Silva, B. C. (2008).  
A Base de Dados Lexical e a Interface Web do TeP 2.0 - Thesaurus Eletrônico para o Português do Brasil.  
In *VI Workshop em Tecnologia da Informação e da Linguagem Humana (TIL)*, pages 390–392.

