

# Computational Lexical Semantics: Methods and Applications

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# Plan

## 1 Semantic Similarity

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# Similarity Measures

- **Similarity measure** is a numerical measure of the degree the two objects are alike
- **Dissimilarity measure** is a numerical measure of the degree to which the two objects are different.
- Both similarity and dissimilarity scores are scalars in range  $[0; 1]$  or  $[0; \infty]$ .
- Two similar objects  $i$  and  $j$  will have a high similarity score  $s_{ij}$  and a low dissimilarity score  $d_{ij}$ .
- Similarity to dissimilarity and vice versa:
  - if  $d_{ij} \in [0; 1]$ , then  $s_{ij} = 1 - d_{ij}$ , where  $s_{ij} \in [0; 1]$ ;
  - if  $s_{ij} \in [0; 1]$ , then  $d_{ij} = 1 - s_{ij}$ , where  $d_{ij} \in [0; 1]$ ;
  - if  $d_{ij} \in [0; \infty]$ , then  $s_{ij} = 1 - \frac{d_{ij} - \min_{i,j}(d_{ij})}{\max_{i,j}(d_{ij}) - \min_{i,j}(d_{ij})}$ , where  $s_{ij} \in [0; 1]$ ;
  - if  $s_{ij} \in [0; \infty]$ , then  $d_{ij} = 1 - \frac{s_{ij} - \min_{i,j}(s_{ij})}{\max_{i,j}(s_{ij}) - \min_{i,j}(s_{ij})}$ , where  $d_{ij} \in [0; 1]$ .

# Semantic Similarity Measures

## Definition

A semantic similarity measure quantifies semantic relatedness input terms  $c_i, c_j$  with the similarity score  $s_{ij} = \text{sim}(c_i, c_j)$ :

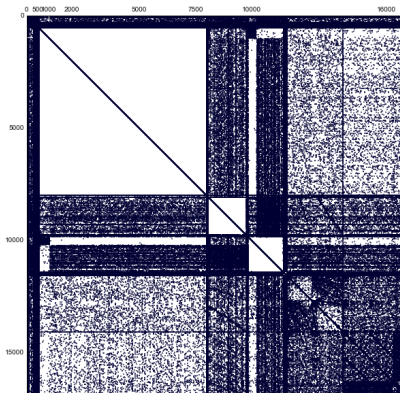
$$s_{ij} = \begin{cases} 1 & \text{if } \langle c_i, c_j \rangle \text{ is a pair of } \textit{syn}, \textit{hyper}, \textit{cohyponym} \\ 0 & \text{otherwise} \end{cases}$$

## Properties

- Nonnegativity:  $0 \leq s_{ij} \leq 1$ ;
- Reflexivity:  $s_{ij} = 1 \Leftrightarrow c_i = c_j$ ;
- Symmetry:  $s_{ij} = s_{ji}$ ;

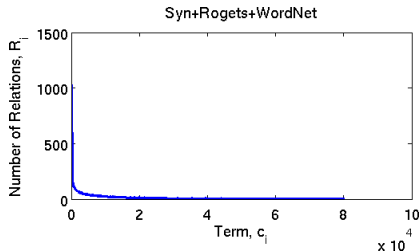
# Word similarity matrix $\mathbf{S}$

- $\mathbf{S}$  – word \* word similarity matrix;
- $s_{ij} \in \mathbf{S}$  – similarity of words  $w_i$  and  $w_j$ ;
- $s_{ij} = \text{sim}(w_i, w_j), s_{ij} \in [0; 1]$ .

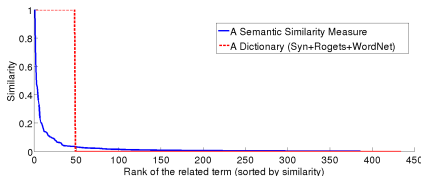


# Semantic Similarity Measures

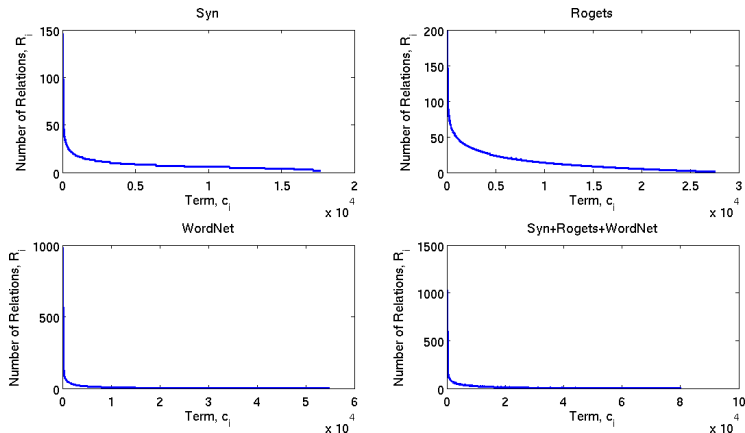
- Many dissimilar pairs, few similar pairs:  $s_{ij} \sim \exp(\lambda)$ :



- Similarity distribution of the term “doctor”:



# Number of relations in semantic resources



**Figure:** Number of relations (synonyms and hyponyms) per term in the dictionaries: a dictionary of synonyms, Roget's thesaurus, WordNet and a union of these three resources.



# Evaluation of Semantic Similarity Measures

- 1 correlations with human judgments (**MC**, **RG**, **WordSim**);
- 2 semantic relation ranking (**BLESS**, **SN**);
- 3 semantic relation extraction;
- 4 using extracted relations in an application:
  - a short text classification system (**iCOP**);
  - a lexico-semantic search engine (**Serelex**).

# Evaluation of Semantic Similarity Measures

## 1 Correlations with human judgments:

- Criterion: Pearson correlation ( $\rho$ )  $\checkmark$  Spearman correlation ( $r$ ).
- Datasets: MC, RG, WordSim.

## 2 Semantic relation ranking:

- Criterion: Precision, Recall, F-measure.
- Dataset: BLESS, SN.

## 3 Semantic relation extraction:

- Criterion: Precision@k.
- Data: annotation and/or dictionaries.

## 4 Application-based evaluation:

- short text classification system (**iCOP**);
- lexico-semantic search engine (**Serelex**).

Panchenko A., **Similarity Measures for Semantic Relation Extraction**. PhD thesis. Université catholique de Louvain. 197 pages, 2013, (Chapter 1).

# Correlations with human judgments

word, $c_i$	word, $c_j$	human, $s$	sim, $s$	human (rank), $r$	sim (rank), $\hat{r}$
tiger	cat	7.35	0.85	1	3
book	paper	7.46	0.95	2	2
computer	keyboard	7.62	0.81	3	1
...	...	...	...	...	...
possibility	girl	1.94	0.25	64	65
sugar	approach	0.88	0.05	65	23

## Datasets:

- WordSim353 – 353 word pairs (Finkelstein, 2002)
- MC – 30 word pairs (Miller Charles, 1991)
- RG – 65 word pairs (Rubenstein Goodenough, 1965)

**Pearson correlation:**  $\rho = \frac{\text{cov}(\mathbf{s}, \hat{\mathbf{s}})}{\sigma(\mathbf{s})\sigma(\hat{\mathbf{s}})}$

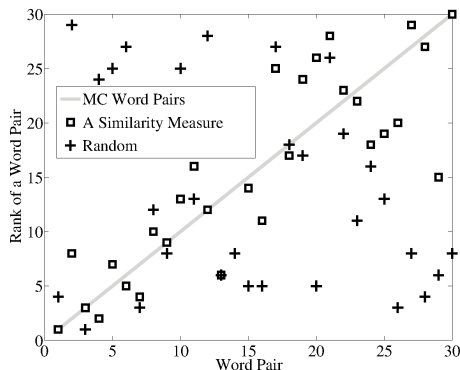
**Spierman correlation::**  $r = \frac{\text{cov}(\mathbf{r}, \hat{\mathbf{r}})}{\sigma(\mathbf{r})\sigma(\hat{\mathbf{r}})}$

# Correlations with human judgments

Table 1.4: Miller-Charles (MC) dataset and scores obtained with a similarity measure.

Word, $c_i$	Word, $c_j$	Human Score, $s_k$	Score, $\hat{s}_k$	Human Rank, $r_k$	Rank, $\hat{r}_k$
automobile	car	3.92	0.884	1	1
journey	voyage	3.84	0.592	2	8
gem	jewel	3.84	0.581	3	3
boy	lad	3.76	0.325	4	2
coast	shore	3.70	0.440	5	7
asylum	madhouse	3.61	0.190	6	5
magician	wizard	3.50	0.556	7	4
midday	noon	3.42	0.692	8	10
furnace	stove	3.11	0.296	9	9
food	fruit	3.08	0.300	10	13
bird	cock	3.05	0.145	11	16
bird	crane	2.97	0.190	12	12
implement	tool	2.95	0.260	13	6
brother	monk	2.82	0.174	14	21
crane	implement	1.68	0.016	15	14
brother	lad	1.66	0.219	16	11
car	journey	1.16	0.124	17	25
monk	oracle	1.10	0.057	18	17
cemetery	woodland	0.95	0.056	19	24
food	rooster	0.89	0.027	20	26
coast	hill	0.87	0.186	21	28
forest	graveyard	0.84	0.069	22	23
shore	woodland	0.63	0.076	23	22
monk	slave	0.55	0.101	24	18
coast	forest	0.42	0.145	25	19
lad	wizard	0.42	0.083	26	20
cord	smile	0.13	0.020	27	29
glass	magician	0.11	0.078	28	27
noon	string	0.08	0.026	29	15
rooster	voyage	0.08	0.005	30	30

# Correlations with human judgments



**Figure:** Spearman correlation on the Miller-Charles (MC) dataset.  $\rho$  of a similarity measure is 0.843 ( $p < 0.001$ ) and  $\rho$  of a random measure is -0.173 ( $p = 0.360$ ).

# Semantic Relation Ranking and Classification

СЛОВО, $c_i$	СЛОВО, $c_j$	ТИП ОТНОШЕНИЯ, $t$
judge	adjudicate	syn
judge	arbitrate	syn
judge	asessor	syn
...	...	...
judge	pc	random
judge	fare	random
judge	lemon	random

## Datasets:

- BLESS (Baroni and Lenci, 2011) – 26554 relations (hyper, coord, mero, event, attri, random)
- SN (Panchenko, 2012) – 14682 relations (syn, random)
- RT, AE, AE2 (Panchenko et al., 2015)

# Semantic Relation Ranking and Classification

- Precision  $P(k = 50) = \frac{1}{7} \approx 0.86$

word, $c_i$	word, $c_j$	relation type	$s_{ij}$
aficionado	enthusiast	syn	0.07197
aficionado	fan	syn	0.05195
aficionado	admirer	syn	0.01964
aficionado	addict	syn	0.01326
aficionado	devotee	syn	0.01163
aficionado	foundling	random	0.00777
aficionado	fanatic	syn	0.00414
aficionado	adherent	syn	0.00353
aficionado	capital	random	0.00232
aficionado	statute	random	0.00029
aficionado	blot	random	0.00025
aficionado	meddler	random	0.00005
aficionado	enlargement	random	0.00003
aficionado	bawdyhouse	random	0.00000

# Semantic Relation Ranking and Classification: BLESS

Target, $c_i$	Relation, $c_j$	Relation Type, $t$	Score, $s_{ij}$	Target, $c_i$	Relation, $c_j$	Relation Type, $t$	Score, $s_{ij}$
hawk	bird	hyper	1	hawk	reverse	random	0.34406
hawk	dove	coltypo	1	hawk	include	random	0.32461
hawk	eagle	coltypo	1	hawk	large	attn	0.29802
hawk	falcon	coltypo	1	hawk	further	random	0.29732
hawk	owl	coltypo	1	hawk	star	random	0.27972
hawk	vulture	coltypo	1	hawk	prison	random	0.2759
hawk	sparrow	coltypo	0.99999	hawk	talon	mero	0.27262
hawk	swoop	event	0.99997	hawk	snip	mero	0.2581
hawk	rapier	hyper	0.99996	hawk	dark	random	0.24776
hawk	swan	coltypo	0.99987	hawk	aim	random	0.24598
hawk	nest	event	0.99976	hawk	lay	event	0.24067
hawk	woodpecker	coltypo	0.99973	hawk	paint	random	0.23446
hawk	feather	mero	0.99964	hawk	old	attn	0.23066
hawk	crow	coltypo	0.99952	hawk	pinon	mero	0.21344
hawk	pigeon	coltypo	0.99948	hawk	experienced	random	0.20552
hawk	fly	event	0.99946	hawk	crate	random	0.20478
hawk	goose	coltypo	0.999	hawk	live	event	0.19195
hawk	pheasant	coltypo	0.99868	hawk	inhabit	event	0.18192
hawk	predator	hyper	0.9985	hawk	christmas	random	0.18135
hawk	wing	mero	0.96836	hawk	conference	random	0.16668
hawk	beak	mero	0.96824	hawk	trim	random	0.15775
hawk	soar	event	0.96733	hawk	limestone	random	0.14231
hawk	robin	coltypo	0.96644	hawk	brother	event	0.13926
hawk	penguin	coltypo	0.96451	hawk	die	event	0.13781
hawk	nostr	event	0.96013	hawk	binding	random	0.11747
hawk	creature	hyper	0.97307	hawk	convict	random	0.11744
hawk	animal	hyper	0.97369	hawk	sign	random	0.11432
hawk	head	mero	0.97016	hawk	kerb	random	0.11242
hawk	claw	mero	0.96907	hawk	nominate	random	0.11112
hawk	gray	attn	0.96	hawk	jacaranda	random	0.10915
hawk	hunt	event	0.95608	hawk	strengthen	random	0.10065
hawk	plumage	mero	0.95465	hawk	everytime	random	0.090728
hawk	hover	event	0.93882	hawk	present	random	0.088429
hawk	catch	event	0.92327	hawk	difficulty	random	0.080612
hawk	vertebrate	hyper	0.91701	hawk	elapse	random	0.08037
hawk	eye	mero	0.88698	hawk	practical	random	0.07723
hawk	wild	attn	0.88417	hawk	no-one	random	0.071844
hawk	grey	attn	0.8643	hawk	educational	random	0.06702
hawk	young	attn	0.79682	hawk	economic	random	0.065984
hawk	foot	mero	0.78725	hawk	judicial	random	0.065793
hawk	plume	mero	0.76141	hawk	concern	random	0.065013
hawk	passenger	random	0.66433	hawk	contextual	random	0.064861
hawk	spot	event	0.65185	hawk	feign	random	0.063987
hawk	sail	event	0.65024	hawk	localise	random	0.063514
hawk	big	attn	0.63627	hawk	neutron	random	0.06142
hawk	circle	event	0.58847	hawk	mem	random	0.060879
hawk	beow	attn	0.55922	hawk	genesis	random	0.060742
hawk	chordate	hyper	0.54453	hawk	improvisation	random	0.059412
hawk	ether	random	0.51446	hawk	employer	random	0.059407
hawk	sea	random	0.51117	hawk	triathlon	random	0.056394
hawk	windmill	random	0.51327	hawk	idealism	random	0.055417
hawk	see	event	0.51181	hawk	fla	random	0.054996
hawk	first	random	0.49451	hawk	co-operative	random	0.047968
hawk	cat	event	0.45854	hawk	ira	random	0.044595
hawk	strong	attn	0.42362	hawk	mindfulness	random	0.042337
hawk	aggressive	attn	0.34521	hawk	shortcoming	random	0.031764



# Semantic Relation Ranking and Classification

- Based on the number of **correctly classified or ranked** semantic relations.
- $R$  – set of non-random relations, e.g. not like  $\langle animal, random, bishop \rangle$
- $\hat{R}(k)$  set of extracted relations at  $k$  nearest neighbours

## Evaluation metrics

- Precision:  $P(k) = \frac{|R \cap \hat{R}(k)|}{|\hat{R}(k)|}$ ,
- Recall:  $R(k) = \frac{|R \cap \hat{R}(k)|}{|R|}$ ,
- F1-measure:  $F(k) = 2 \cdot \frac{P(k) \cdot R(k)}{P(k) + R(k)}$ ,