



Insights & Applications

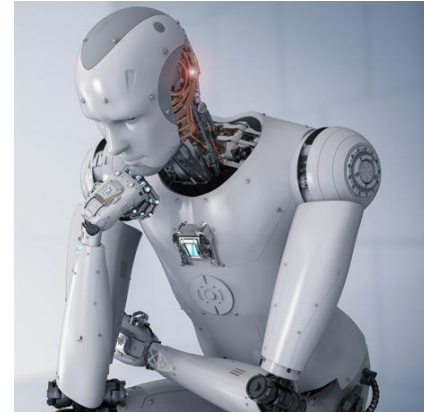
Haim Dubossarsky, h.dubossarsky@qmul.ac.uk

Why change matters?



If you are a linguist

If you are a historian, sociologist
or interested in societal changes



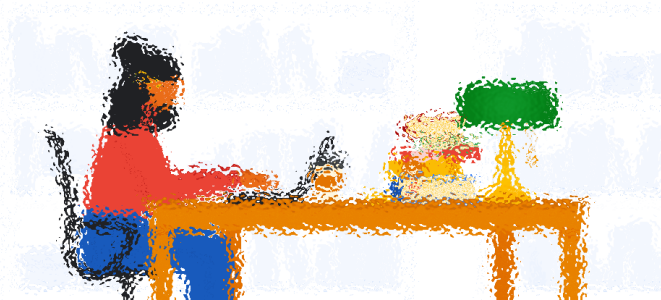
If you work on AI

Why change matters?



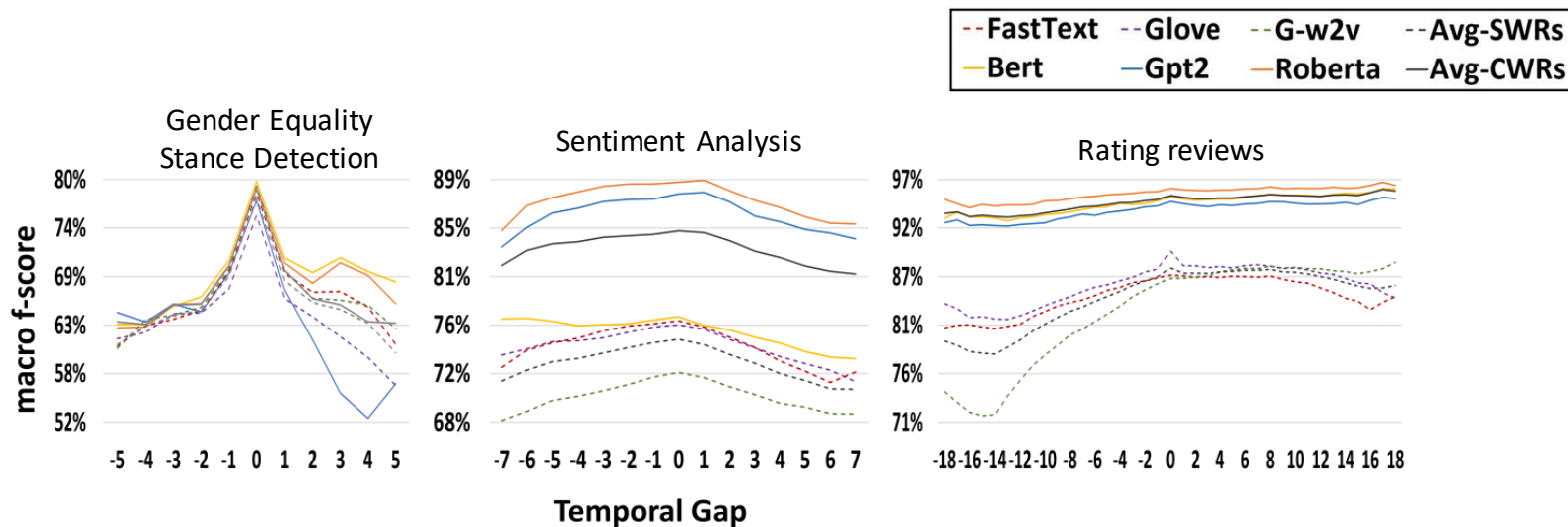
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If you work on AI

Models age: Performance drop with temporal gap



Why change matters?



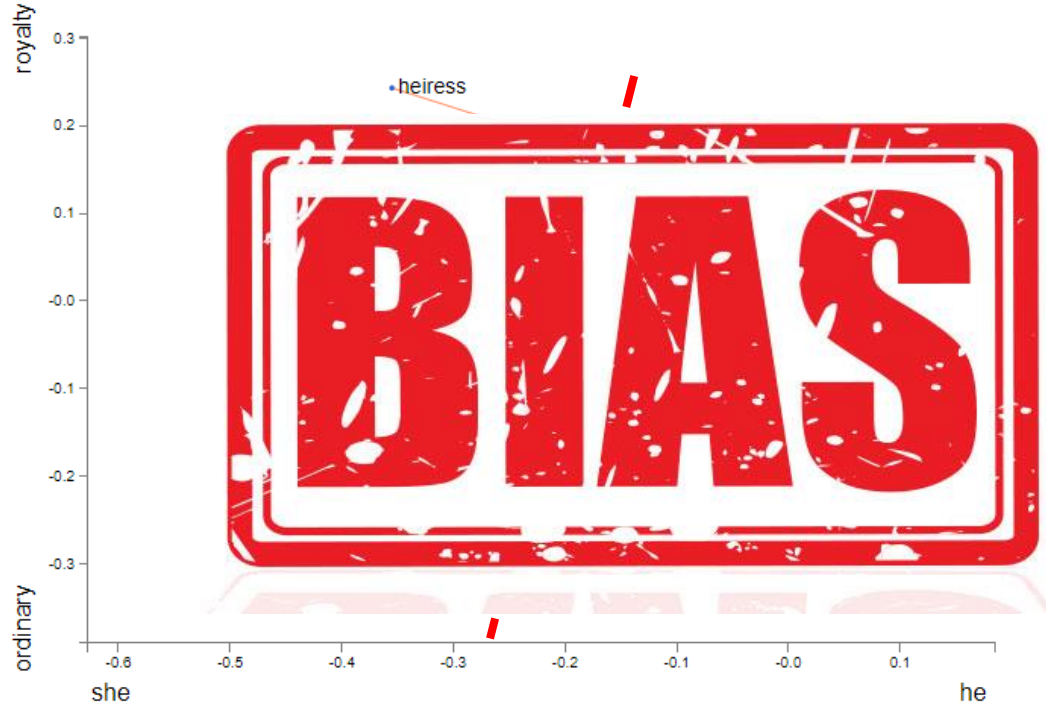
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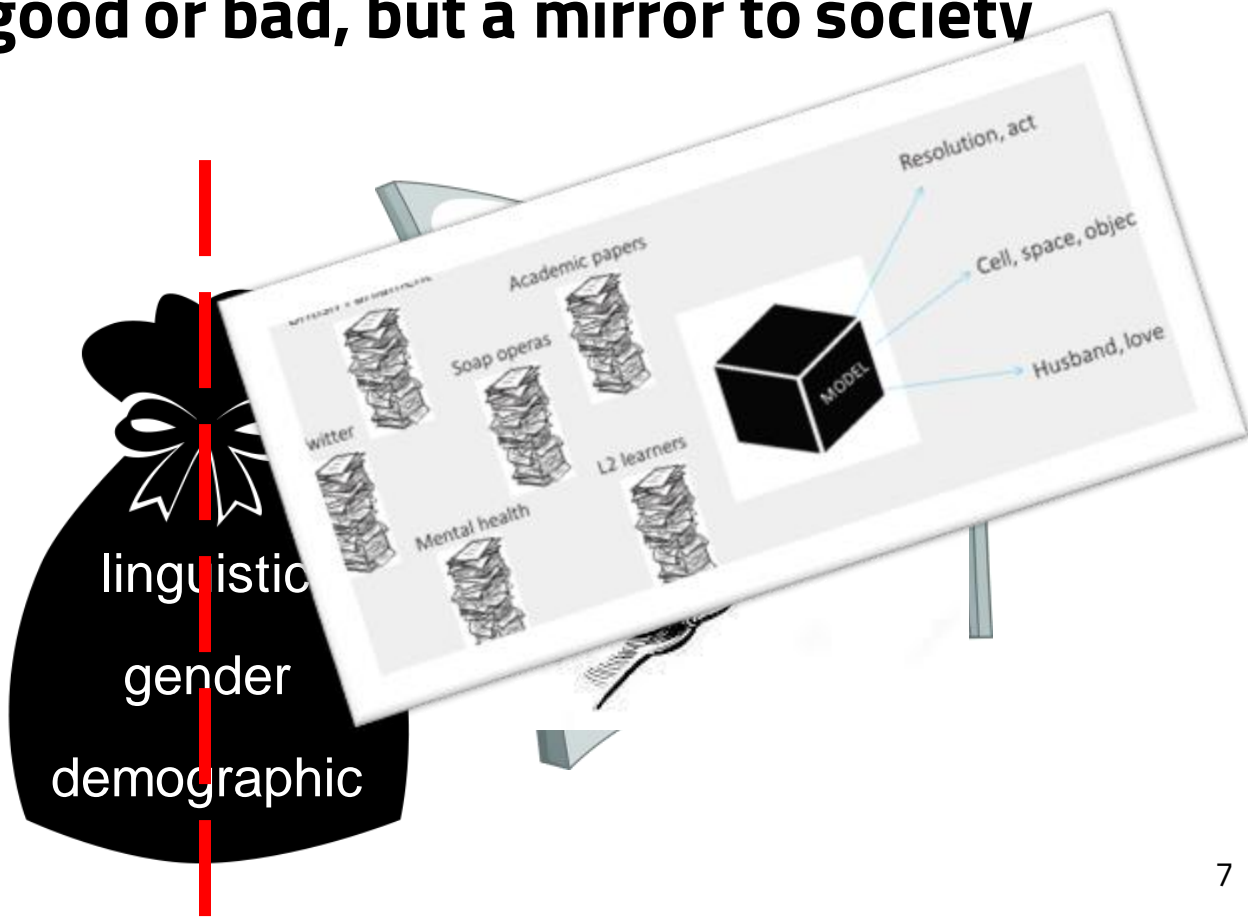
Models are biased



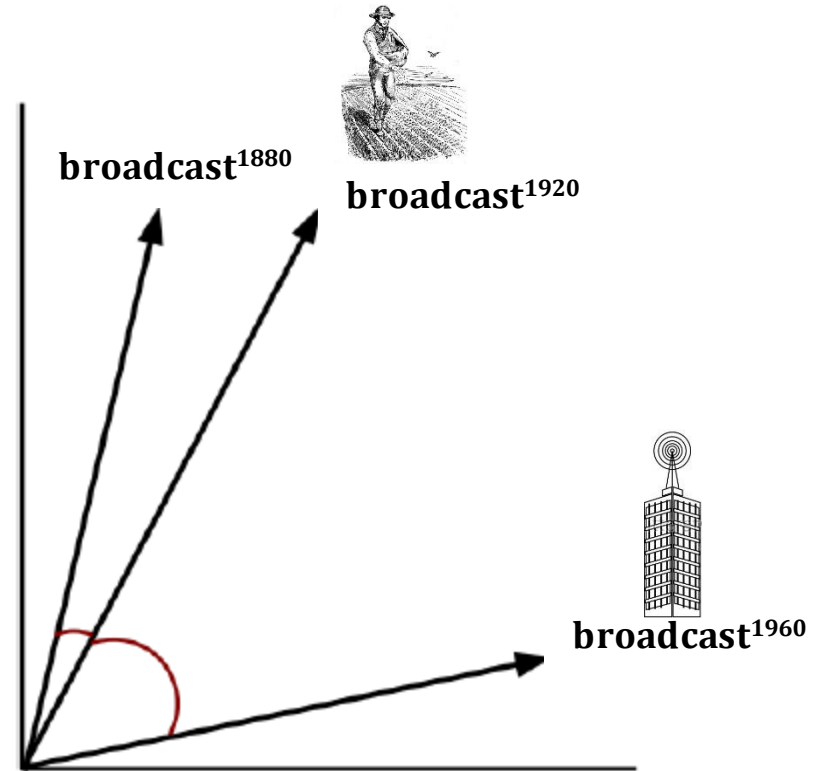
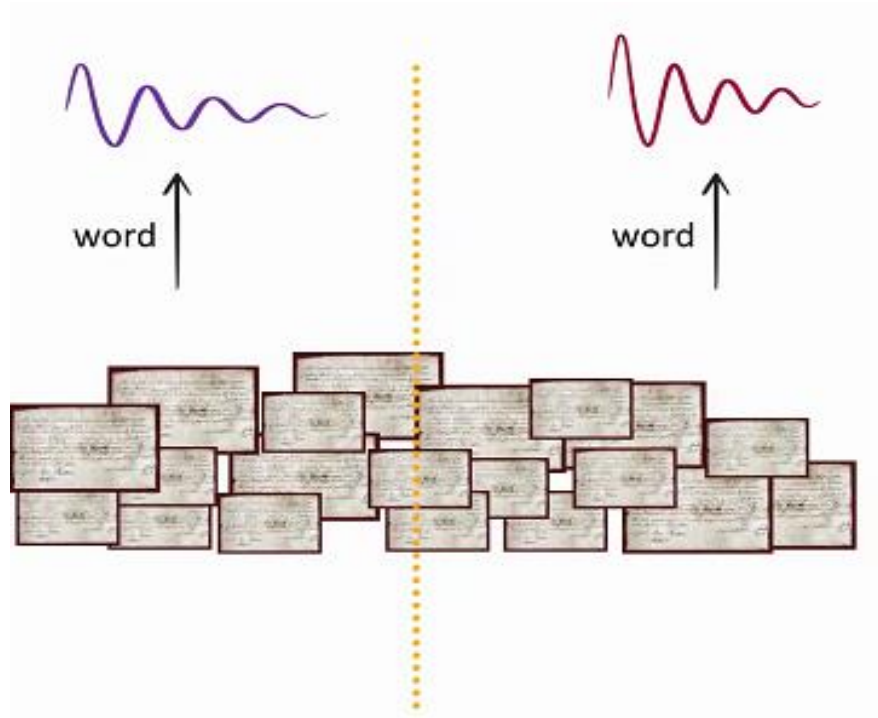
Bias is neither good or bad, but a mirror to society



Democratizing AI



Study how bias changes over time & across domains



In the finance domain

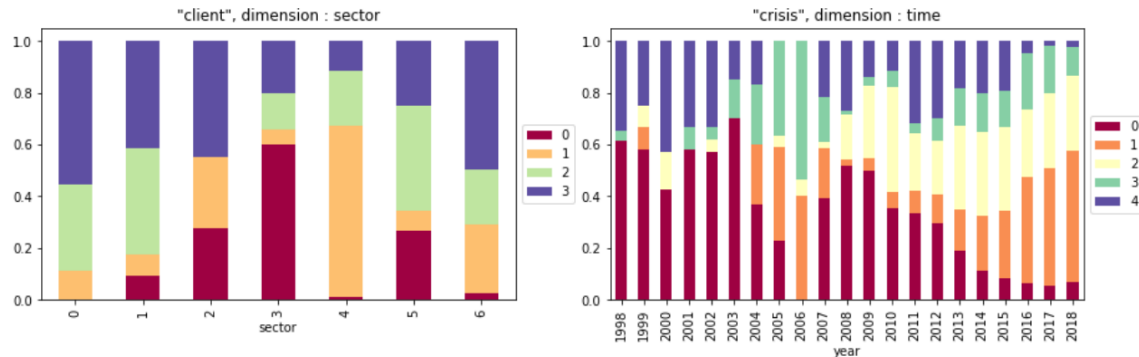
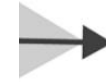
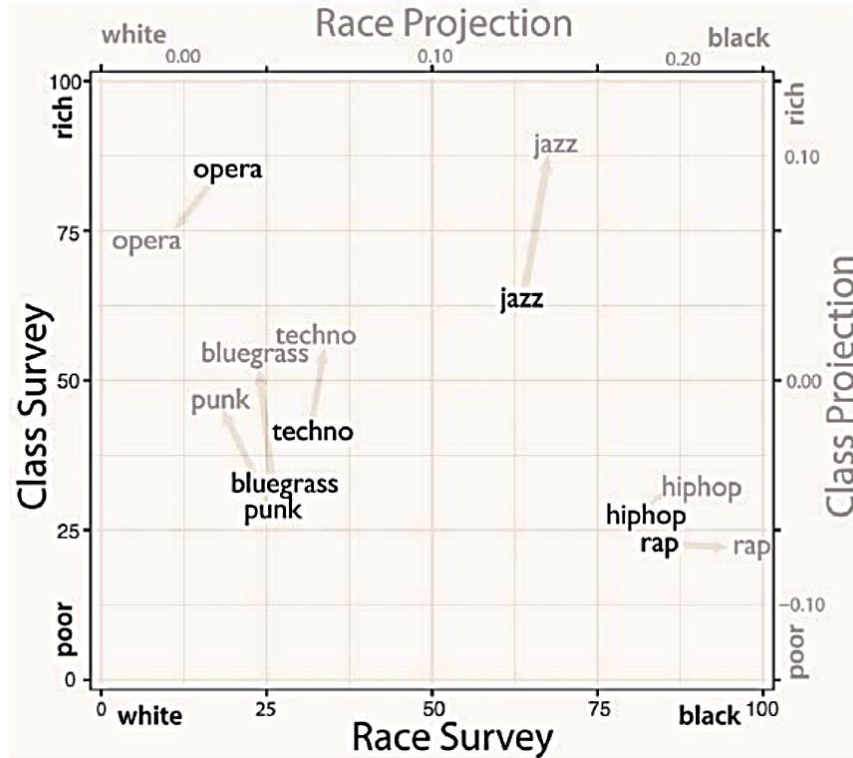
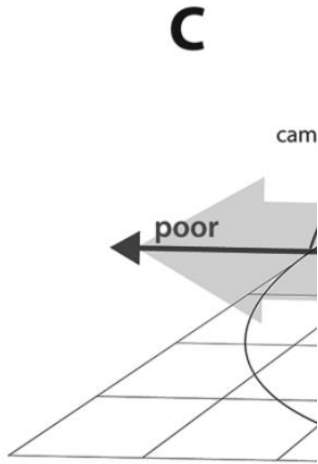


Figure 1: Distribution of clusters per Office for the word *client* (left) and per year for the word *crisis* (right) in the SEC-Edgar corpus. The Offices are described in Table 4

N°	Keyword examples - Word = <i>client</i>	N°	Keyword examples - Word = <i>crisis</i>
0	server, products, data, applications, services, systems	0	liquidity, funding, contingency, cash, collateral, outflows
1	revenue, contract, risk, costs, loss, business, fees	1	marketing, business, management, design, advertising, media
2	assets, funds, cash, interest, balances, investment	2	european, debt, credit, sovereign, countries, eurozone, banks
3	services, business, revenue, growth, management, products	3	financial, accident, capital, regulatory, loss, liquidity, funding
		4	credit, financial, global, markets, debt, european, recession

Table 5: List of clusters and keyword examples for the words *client* (left) and *crisis* (right) in the SEC-Edgar Corpus

In the cultural domain

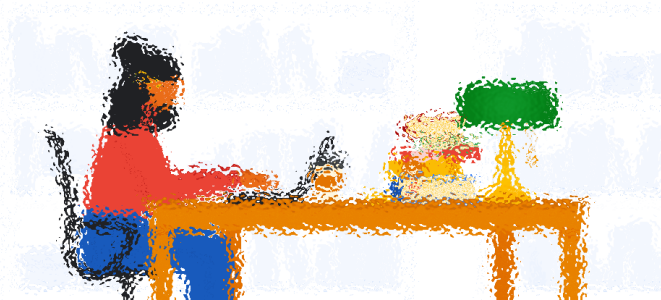


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Exploratory research

\mathcal{H} : Do certain word classes change more than others?



Availability effect (Gentner 1981)

Cognitive Science, 2022

VERB METAPHORIC EXTENSION UNDER SEMANTIC STRAIN

Daniel King & Dedre Gentner

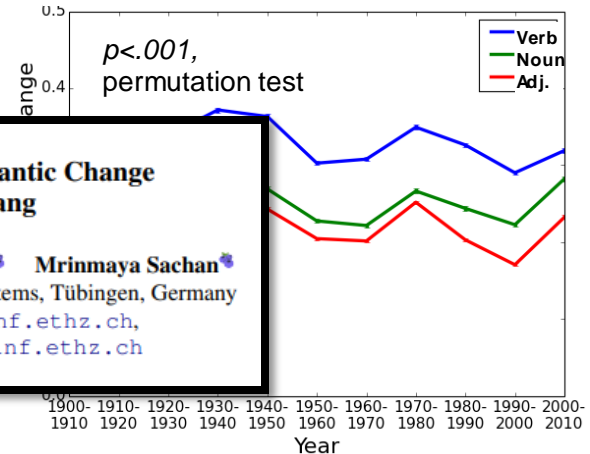
Northwestern University

NOTE: This is an unpublished preprint of

Slangvolution: A Causal Analysis of Semantic Change and Frequency Dynamics in Slang

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Word-class effect (Dubossarsky et al. 2016)



tion between object-reference concepts and relational concepts. Object-reference



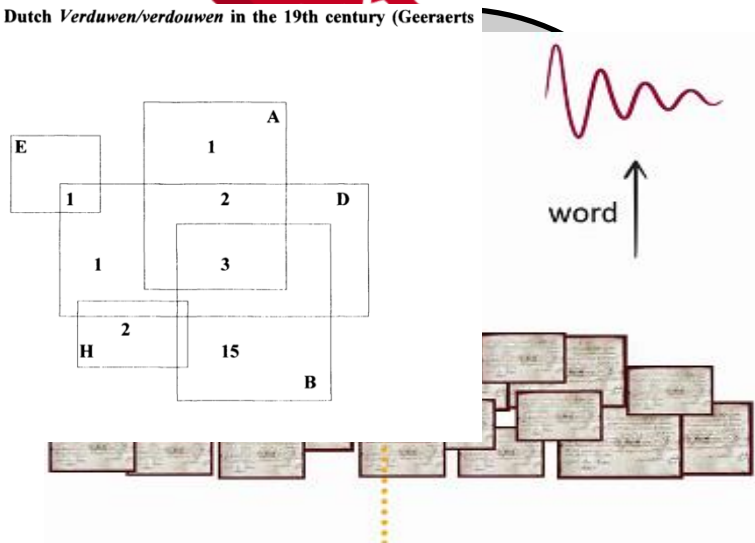
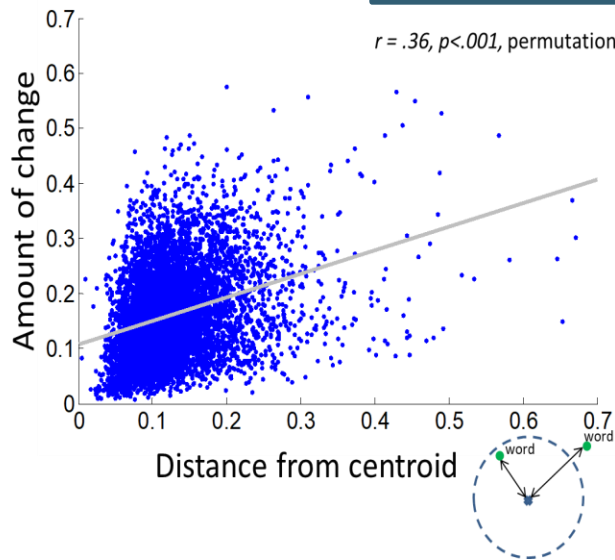
Testing linguistic theories

\mathcal{H} : can category membership explain semantic change?

Law of prototypicality

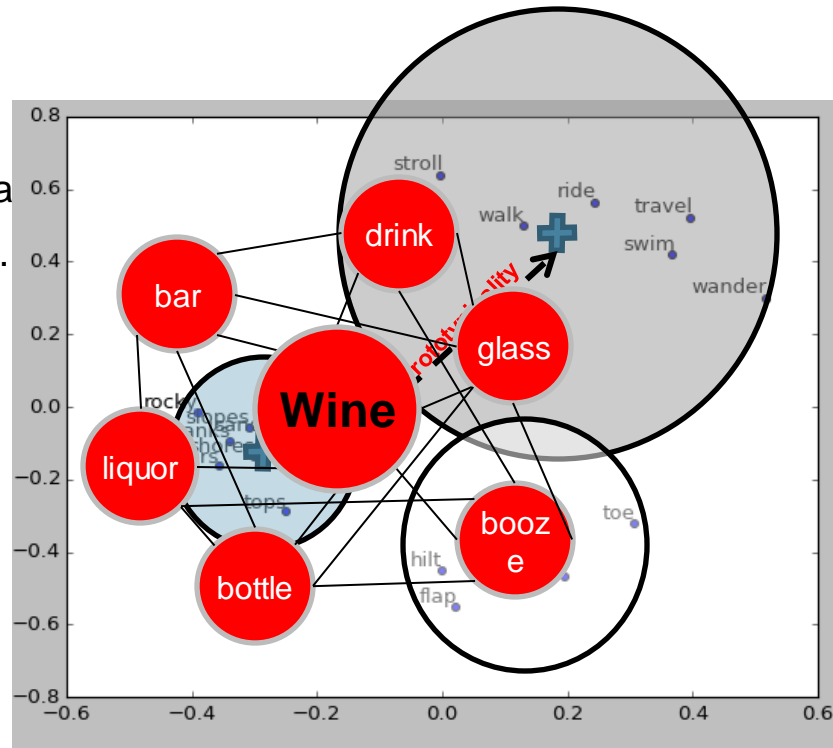
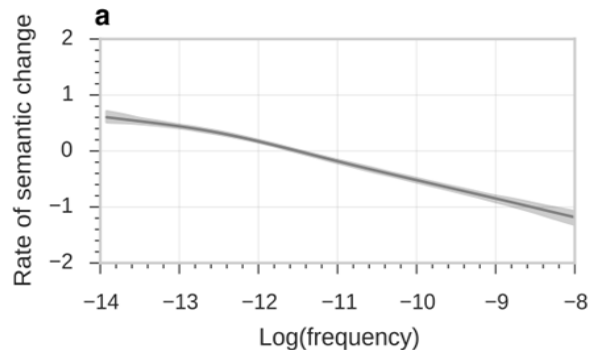
$r = .36, p < .001$, permutation test

Diagram (2.11): Dutch *Verduwen/verdouwen* in the 19th century (Geeraerts 1992:198).



Critical analysis of our models: laws of semantic change

- Law of Prototypicality (Dubossarsky et. al.)
- Law of Innovation (Polysemy, Hamilton et. al.)
- Law of Conformity (Frequency, Hamilton et. al.)



Critical analysis of our models: laws of semantic change

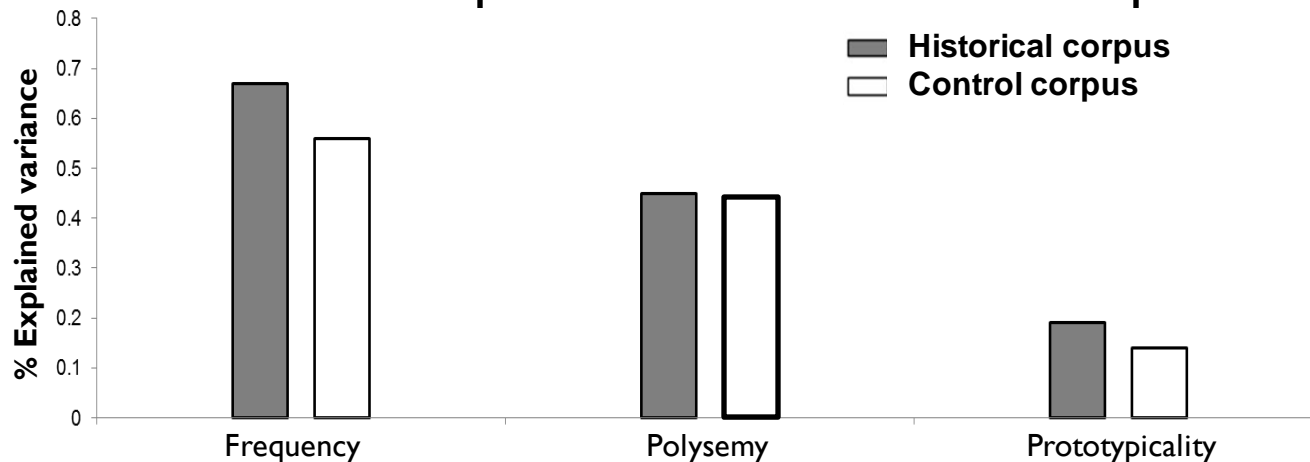
- Law of Prototypicality
- Law of Innovation
- Law of Conformity

Theoretically

In a control condition:

$$\mathbb{E}[x_{t+1}] = \mathbb{E}[x_{t+2}] = \mathbb{E}[x]$$

So, semantic change of a word is,



The search for laws of semantic change continues

Revisiting Statistical Laws of Semantic Shift in Romance Cognates

The return of the “law of conformity”
 Good research is one that steers a discussion

	Coe			
Intercept	0.00	0.03	0.00	1.00
FREQ _{lat}	-0.08	0.04	-1.82	0.07
POLY _{lat}	0.10	0.04	2.28	0.02
LEN _{lat}	-0.21	0.03	-6.29	0.00
FREQ _{rom}	-0.54	0.03	-18.40	0.00
NORM _{rom}	–	–	–	–
EDIT	0.13	0.03	4.07	0.00

Table 3: Results of regression analysis on distance scores of French–Spanish cognate pairs ($N = 794$, $Adj.R^2 = 0.35$). NORM_{rom} was kept out by model selection methods.

	Language pair	Adj. R^2	N
French	French–Italian	0.29	812
	French–Spanish	0.35	794
	Italian–Spanish	0.35	842
Italian	French–Italian	0.29	812
	French–Spanish	0.33	794
	Italian–Spanish	0.38	842
Spanish	French–Italian	0.27	812
	French–Spanish	0.35	794
	Italian–Spanish	0.39	842

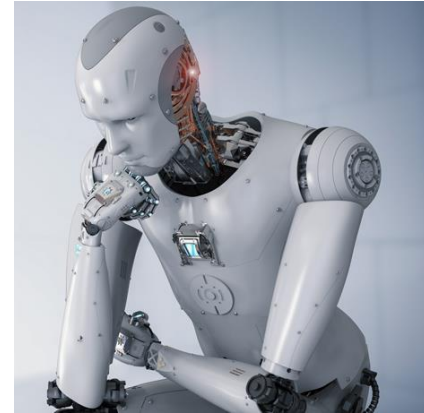
Table 4: Adjusted R-squared for respective language pairs in different embedding spaces.

Summary



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