Word representations and modelling ambiguity: A case study of metaphor

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Polysemy and word senses

The children ran to the store
If you see this man, run!
Service runs all the way to Cranbury
She is running a relief operation in Sudan
the story or argument runs as follows
Does this old car still run well?
Interest rates run from 5 to 10 percent
Who's running for treasurer this year?
They ran the tapes over and over again
These dresses run small

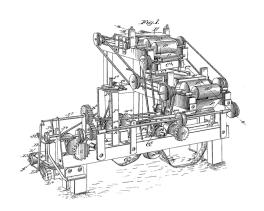
Polysemy

- homonymy: unrelated word senses. bank (raised land) vs bank (financial institution)
- bank (financial institution) vs bank (in a casino): related but distinct senses.
- regular polysemy and sense extension
 - zero-derivation, e.g. tango (N) vs tango (V), or rabbit, turkey, halibut (meat / animal)
 - metaphorical senses, e.g. swallow [food], swallow [information], swallow [anger]
 - metonymy, e.g. he played Bach; he drank his glass.
- vagueness: nurse, lecturer, driver
- cultural stereotypes: nurse, lecturer, driver

No clearcut distinctions.

Dictionaries are not consistent.

What is metaphor?



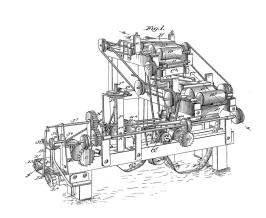
What is metaphor?

"A political machine"

"The wheels of the regime were well oiled and already turning"

"Time to mend our foreign policy"

"20 Steps towards a Modern, Working Democracy"



How does it work?

Conceptual Metaphor Theory (Lakoff and Johnson, 1980)



Metaphorical associations between concepts

POLITICALSYSTEM is a MECHANISM

target

source

Cross-domain knowledge projection and inference

Reasoning about the target domain in terms of the properties of the source

Computational metaphor processing tasks

Learn metaphorical associations from corpora

"POLITICAL SYSTEM is a MECHANISM"

Identify metaphorical language in text

"mend the policy"

Interpret the metaphorical language

"mend the policy" means "improve the policy; address the downsides of the policy"

N: game	N: politics
1170 play	31 dominate
202 win	30 play
99 miss	28 enter
76 watch	16 discuss
66 lose	13 leave
63 start	12 understand
42 enjoy	8 study
22 finish	6 explain
	5 shape
20 dominate	4 influence
18 quit	4 change
17 host	4 analyse
17 follow	
17 control	2 transform

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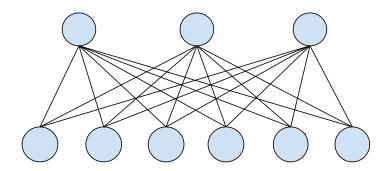
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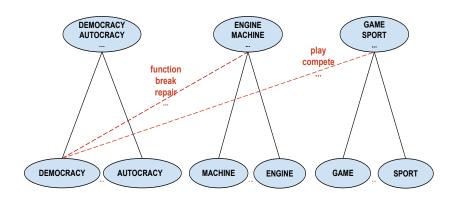
NEED TO FIND A WAY TO PARTITION THE SPACE

Learning metaphorical associations by soft clustering

Unsupervised Metaphor Identification Using Hierarchical Graph Factorization Clustering. Shutova and Sun, 2013. NAACL.



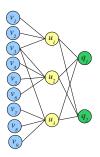
Learning metaphorical associations by soft clustering



Creating the graph

Experimental setup

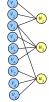
- ALGORITHM: Hierarchical graph factorization clustering (Yu, Yu, Tresp, 2006)
- DATASET: 2000 frequent nouns
- FEATURES: verbs in subject, direct and indirect object relations (Gigaword corpus)
- LEVELS: 10





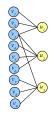
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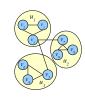
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$$W'$$
: $w'_{ij} = \sum_{p=1}^{m} rac{b_{ip}b_{jp}}{\lambda_p}$

$$\lambda_i = \sum_{i=1}^n b_{ip}$$





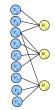


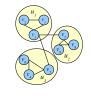
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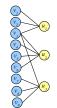


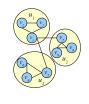
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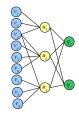
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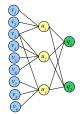
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Identifying metaphorical associations in the graph

- start with the source concept, e.g. "fire"
- output a ranking of potential target concepts



SOURCE: fire

TARGET: sense hatred emotion passion enthusiasm sentiment hope interest **feeling** resentment optimism hostility excitement anger TARGET: coup **violence** fight resistance clash rebellion battle drive fighting

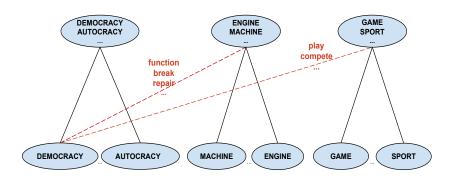
riot revolt war confrontation volcano row revolution struggle

SOURCE: disease

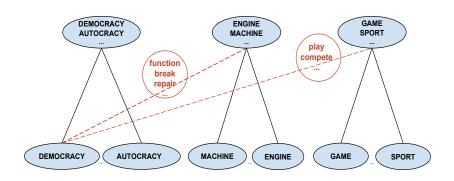
TARGET: fraud outbreak offence connection leak count **crime** violation abuse conspiracy corruption terrorism suicide

TARGET: opponent critic rival

Identifying metaphorical expressions



Identifying metaphorical expressions



Metaphorical expressions retrieved

FEELING IS FIRE

anger blazed passion flared fuel resentment anger crackled etc.

CRIME IS A DISEASE

cure crime abuse transmitted suffer from corruption diagnose abuse etc.

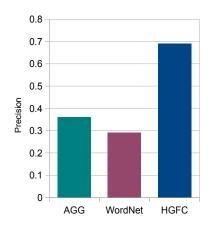
Output sentences from the BNC

EG0 275 In the 1930s the words "means test" was a curse, **fuelling the resistance** against it both among the unemployed and some of its administrators.

HL3 1206 [..] he would strive to **accelerate progress** towards the economic integration of the Caribbean.

HXJ 121 [..] it is likely that some **industries will flourish** in certain countries as the **market widens**.

How well does it work?



Multilingual metaphor processing (Shutova et al. 2017, Computational Linguistics)

Cross-cultural differences

Spanish: poverty metaphors POVERTY IS AN ENEMY, PAIN

English: immigration metaphors IMMIGRATION IS A DISEASE, FIRE

Russian: sporting events / competitions associated with WAR

tions associated with WAIT