Social Media & Text Analysis

lecture 6 - Paraphrase Data Sources

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Website: <u>socialmedia-class.org</u>

Natural Language Processing

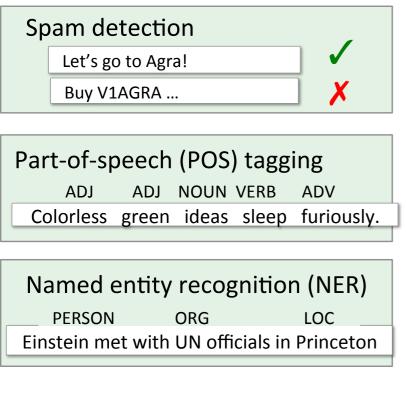
Dan Jurafsky

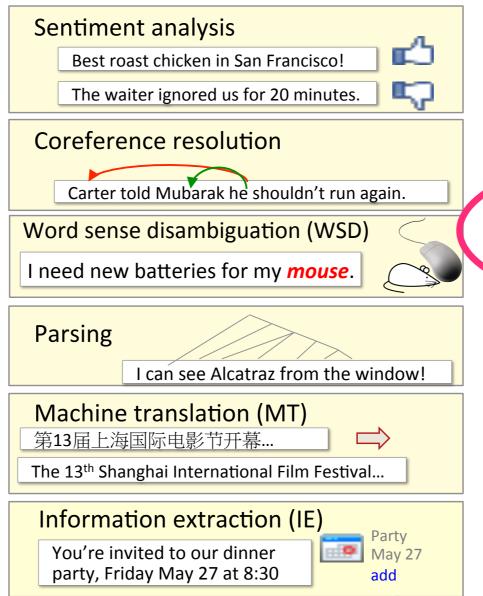


Language Technology

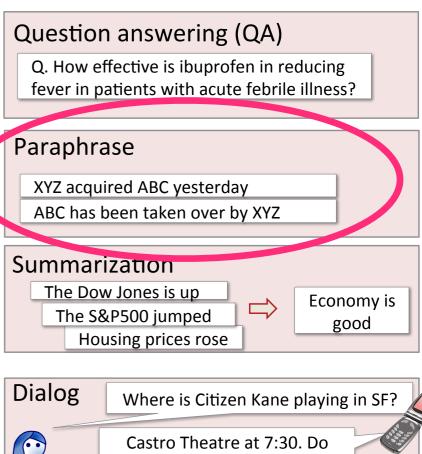
making good progress

mostly solved





still really hard



you want a ticket?

what is Paraphrase?

"sentences or phrases that convey approximately the same meaning using different words" — (Bhagat & Hovy, 2013)

wealthy

word

rich

the king's speech

phrase

His Majesty's address

... the forced resignation of the CEO of Boeing, Harry Stonecipher, for ...

sentence

... after Boeing Co. Chief Executive Harry Stonecipher was ousted from ...

What's good about Paraphrases?

fundamentally useful for a wide range of applications

e.g. Question Answering

Who is the CEO stepping down from Boeing?

... the forced resignation of the CEO of Boeing, Harry Stonecipher, for after Boeing Co. Chief Executive Harry Stonecipher was ousted from ...

What's good about Paraphrases?

fundamentally useful for a wide range of applications

e.g. Question Answering

Who is the CEO stepping down from Boeing?

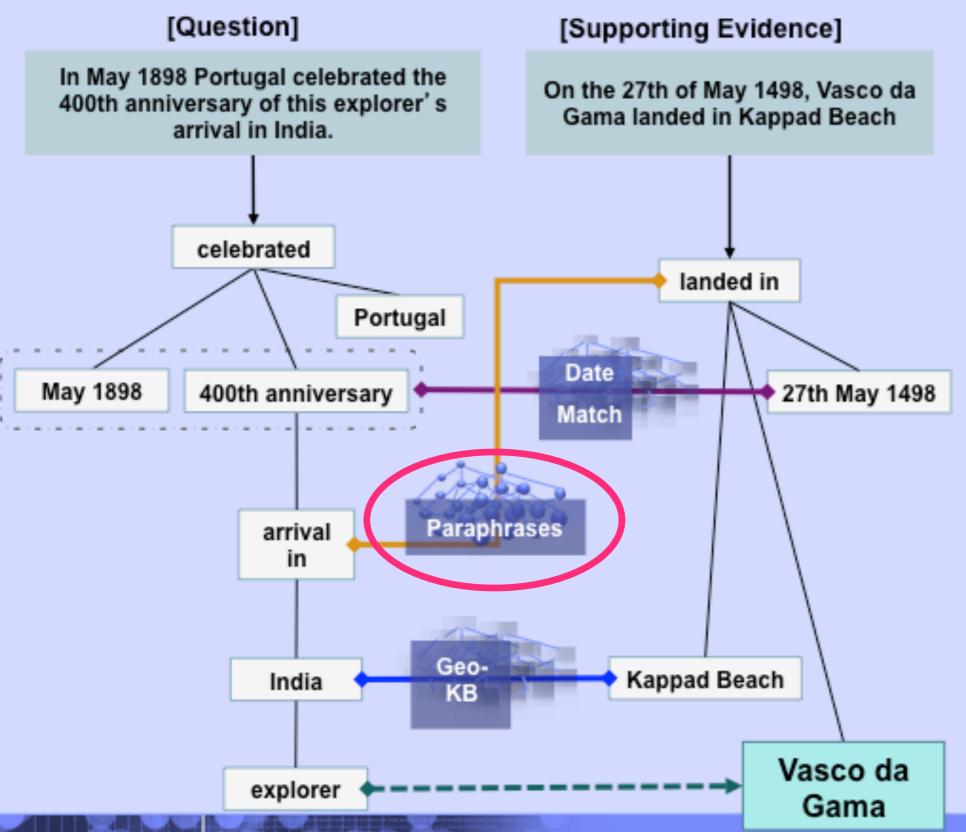


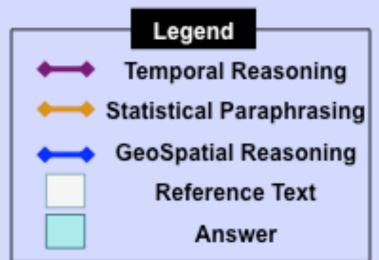
... the forced <u>resignation</u> of the CEO of Boeing, Harry Stonecipher, for after Boeing Co. Chief Executive Harry Stonecipher was ousted from ...





Watson leverages multiple algorithms to perform deeper analysis



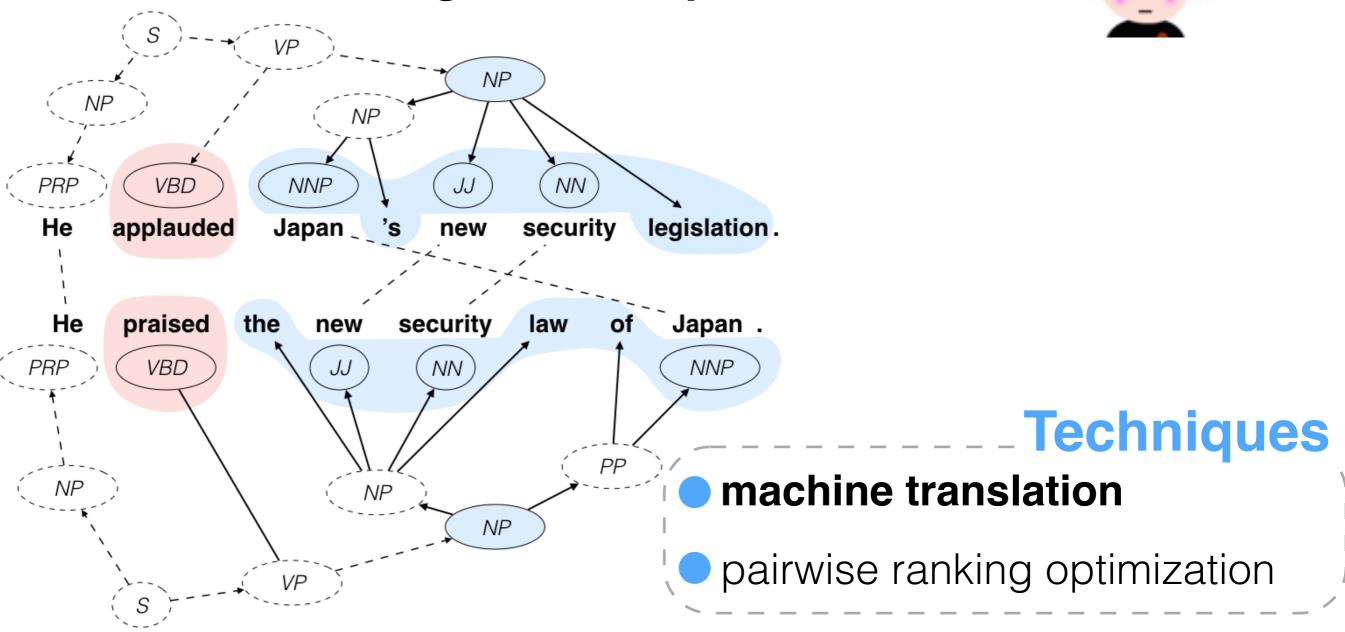


Stronger evidence can be much harder to find and score...

- Search far and wide
- Explore many hypotheses
- Find judge evidence
- Many inference algorithms

Natural Language Generation

e.g. Text Simplification



Wei Xu, Chris Callison-Burch, Courtney Napoles. "Problems in Current Text Simplification Research: New Data Can Help" in TACL (2015) Wei Xu, Courtney Napoles, Ellie Pavlick, Chris Callison-Burch. "Optimizing Statistical Machine Translation for Simplification" in TACL (2016)

Digital Humanities



e.g. Stylistic Rewriting / Poetry Generation



Palpatine:

If you will not be turned, you will be destroyed!

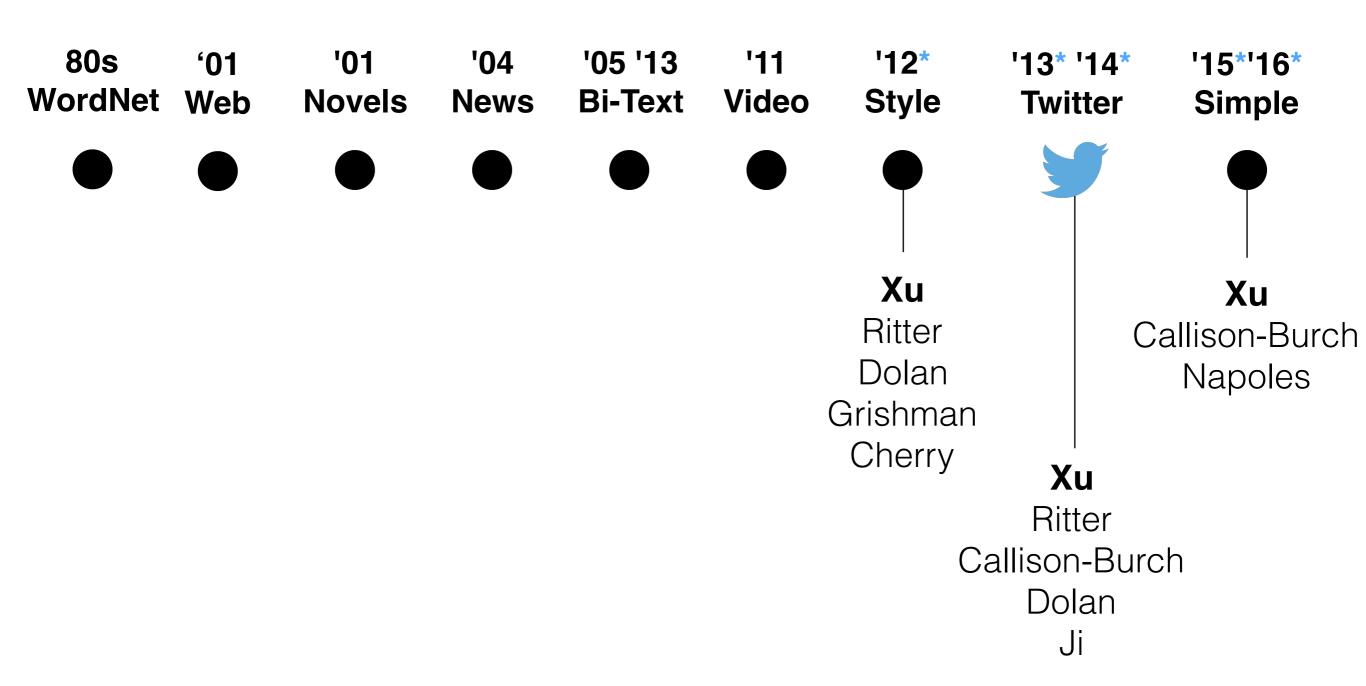
If you will not be turn'd, you will be undone!

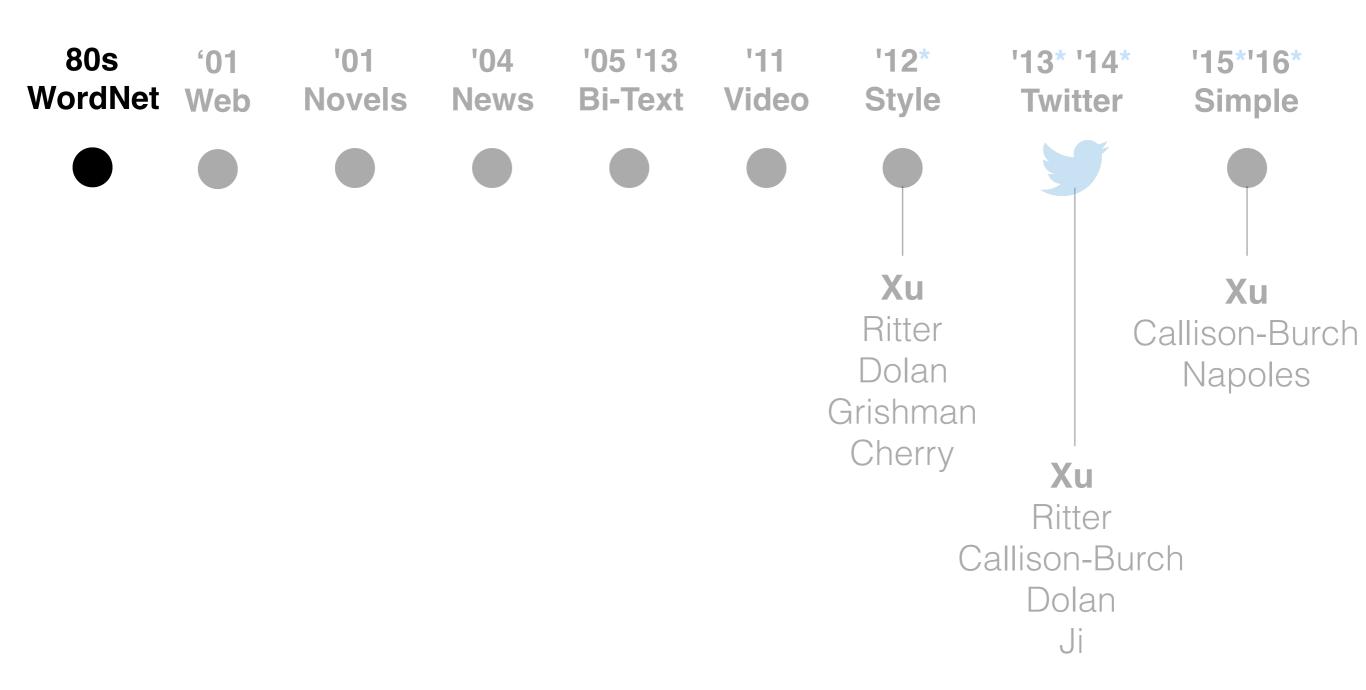
Luke:

Father, please! Help me!

Father, I pray you! Help me!







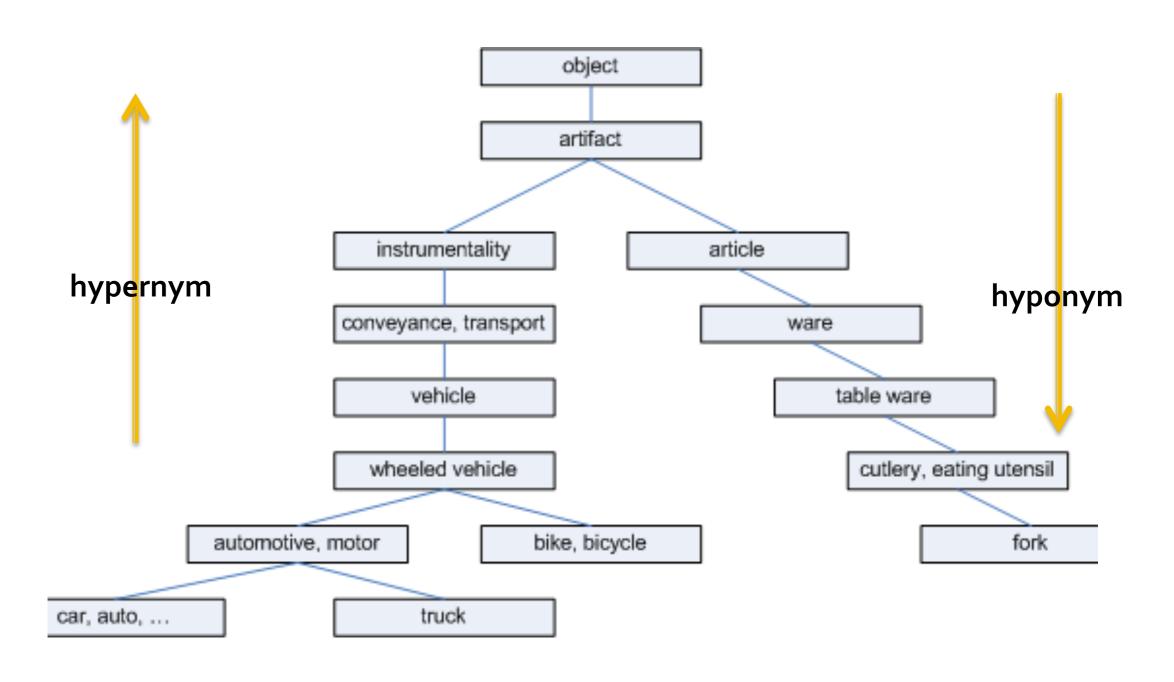
- What is it?
 - a large lexical database of English (155,287 words, latest version in 2005~6)
 - created (from mid-1980s) and maintained by Cognitive Science Lab of Princeton University
 - designed to establish the connections between words

- What is it?
 - a combination of dictionary and thesaurus
 - try it out http://wordnet.princeton.edu/
 - In other languages: http://globalwordnet.org/ wordnets-in-the-world/

Dictionary contains meaning, definition, pronunciation, orthography, and etymology of a word. Thesaurus contains synonyms and antonyms of words.

- 4 types of Parts of Speech (POS)
 - Noun, Verb, Adjective, Adverb
- Synset (synonym set)
 - the smallest unit in WordNet
 - represents a specific meaning of a word
 - S: (n) search (an investigation seeking answers) "a thorough search of the ledgers revealed nothing"; "the outcome justified the search"
 - S: (v) search, seek, look for (try to locate or discover, or try to establish the existence of) "The police are searching for clues"; "They are searching for the missing man in the entire county"

- Synsets are connected to one anther through semantic and lexical relations
- Type of relations (based on POS)
 - hypernyms (kind-of): 'vehicle' is a hypernym of 'car'
 - hyponyms (kind-of): 'car' is a hyponym of 'vehicle'
 - holonym (part-of): 'building' is a holonym of 'window'
 - meronym(part-of): 'window' is a meronym of 'building'
 - similar to: 'smart' is similar to 'intelligent'
 - antonyms: 'smart' is antonym of 'unintelligent'



- Interfaces
 - Unix-style manual
 - Web Interfaces
 - Local Interfaces/APIs (Java, Python, Perl, C# ...)

http://wordnet.princeton.edu/wordnet/related-projects/



wordnet

Q

Articles

About 94,700 results (0.08 sec)

Any time

Since 2017 Since 2016 Since 2013

Custom range...

Sort by relevance

Sort by date

✓ include patents

✓ include citations

Create alert

WordNet: a lexical database for English

GA Miller - Communications of the ACM, 1995 - dl.acm.org

Abstract Because meaningful sentences are composed of meaningful words, any system that hopes to process natural languages as people do must have information about words and their meanings. This information is traditionally provided through dictionaries, and

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[BOOK] WordNet

C Fellbaum - 1998 - Wiley Online Library

Abstract **WordNet** (Miller, Beckwith, Fellbaum, Gross, & Miller 1990; Miller & Fellbaum, 1991; Miller, 1995; Fellbaum, 1998), a lexical database for English, can be thought of as a large electronic dictionary. It contains information about some 155,000 nouns, verbs, adjectives,

☆ 55 Cited by 13461 Related articles All 12 versions
⇒>

Introduction to WordNet: An on-line lexical database

GA Miller, R Beckwith, C Fellbaum... - International journal ..., 1990 - academic.oup.com Abstract **WordNet** is an on-line lexical reference system whose design is inspired by current psycholinguistic theories of human lexical memory. English nouns, verbs, and adjectives are organized into synonym sets, each representing one underlying lexical concept. Different

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WordNet:: Similarity: measuring the relatedness of concepts

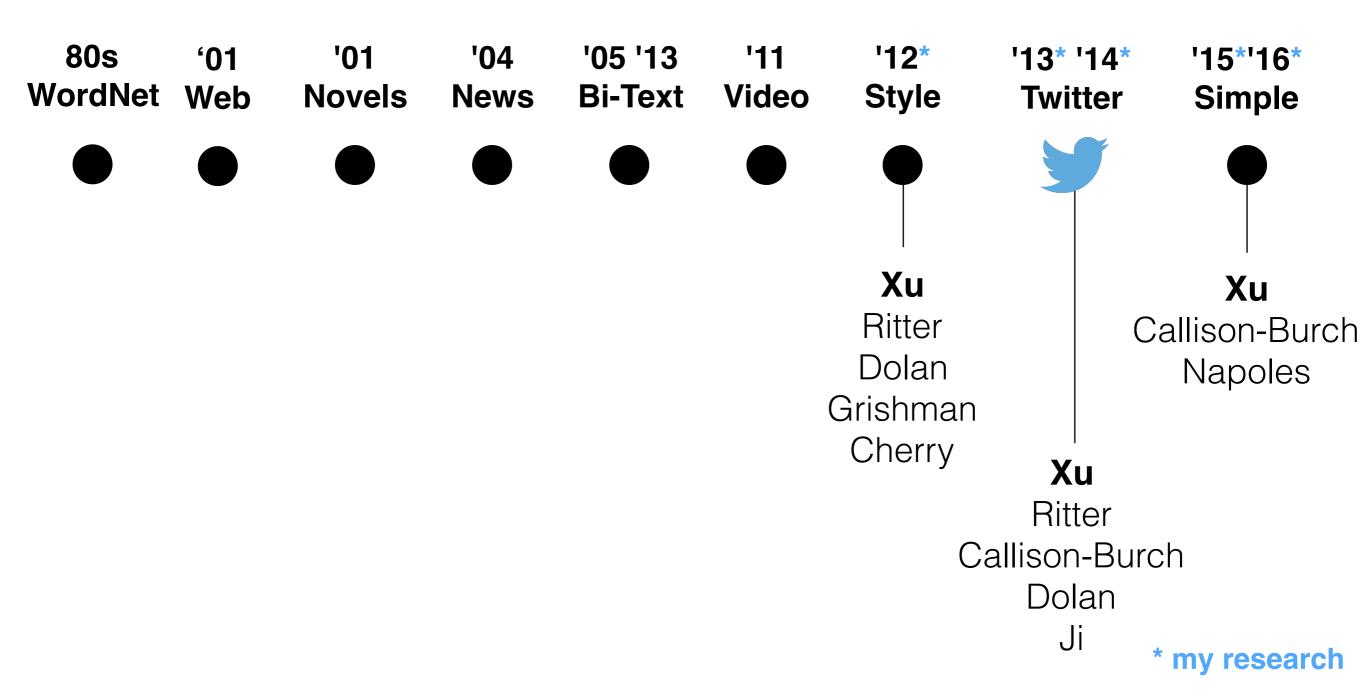
<u>T Pedersen</u>, <u>S Patwardhan</u>, J Michelizzi - Demonstration papers at HLT- ..., 2004 - dl.acm.org Abstract **WordNet**:: Similarity is a freely available software package that makes it possible to measure the semantic similarity and relatedness between a pair of concepts (or synsets). It provides six measures of similarity, and three measures of relatedness, all of which are

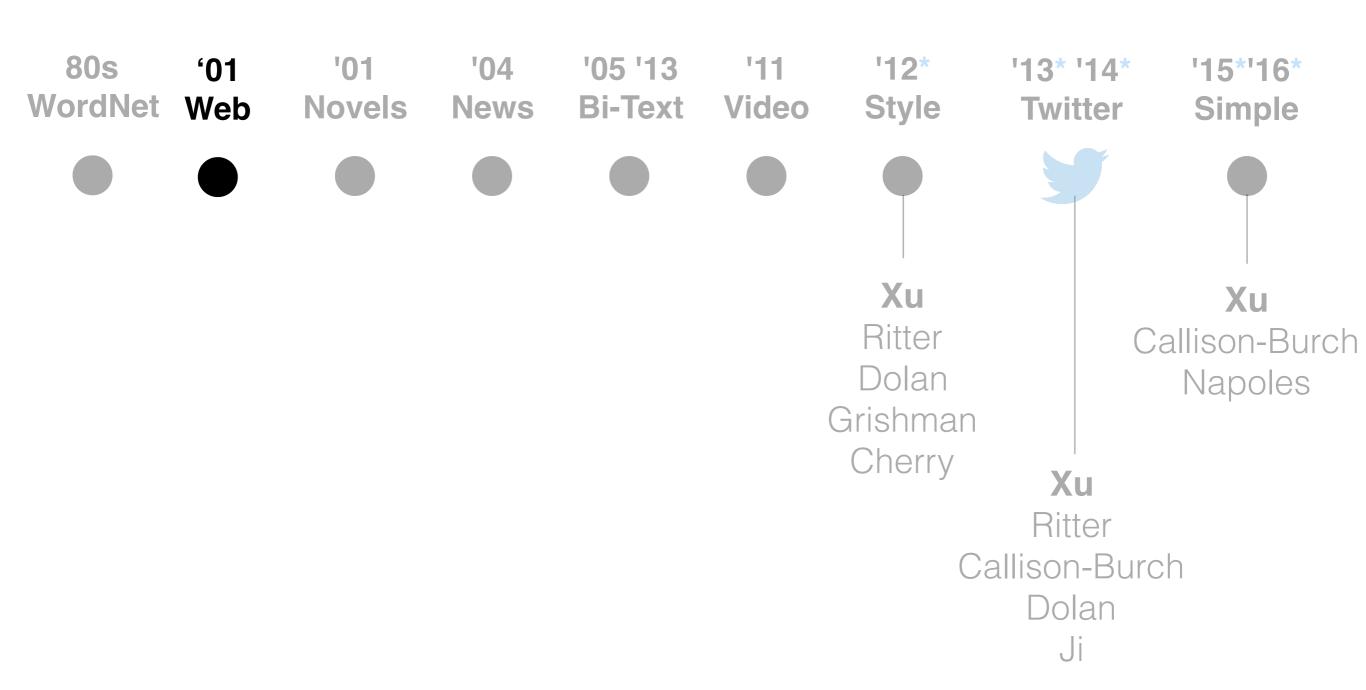
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[PDF] semanticscholar.org

[PDF] academia.edu

[PDF] aaai.org





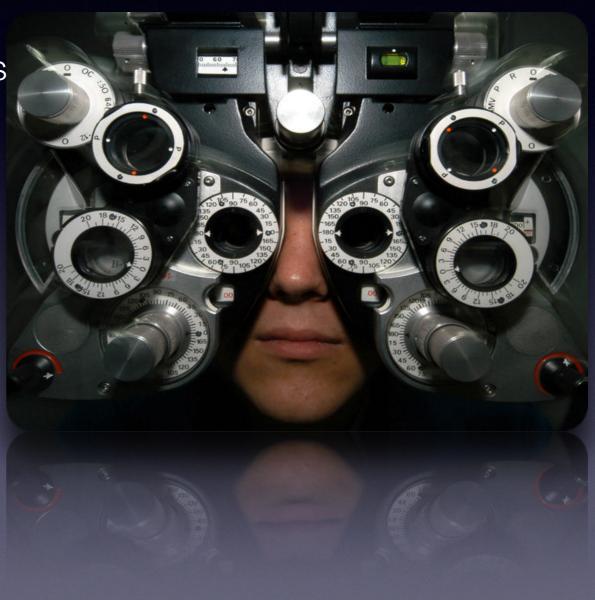
Distributional Hypothesis

If we consider oculist and eye-doctor we find that, as our corpus of utterances grows, these two occur in almost the same environments. In contrast, there are many sentence environments in which oculist occurs but lawyer does not...

It is a question of the relative frequency of such environments, and of what we will obtain if we ask an informant to substitute any word he wishes for **oculist** (not asking what words have the same meaning).

These and similar tests all measure the probability of particular environments occurring with particular elements... If A and B have almost identical environments we say that they are synonyms.

-Zellig Harris (1954)



DIRT

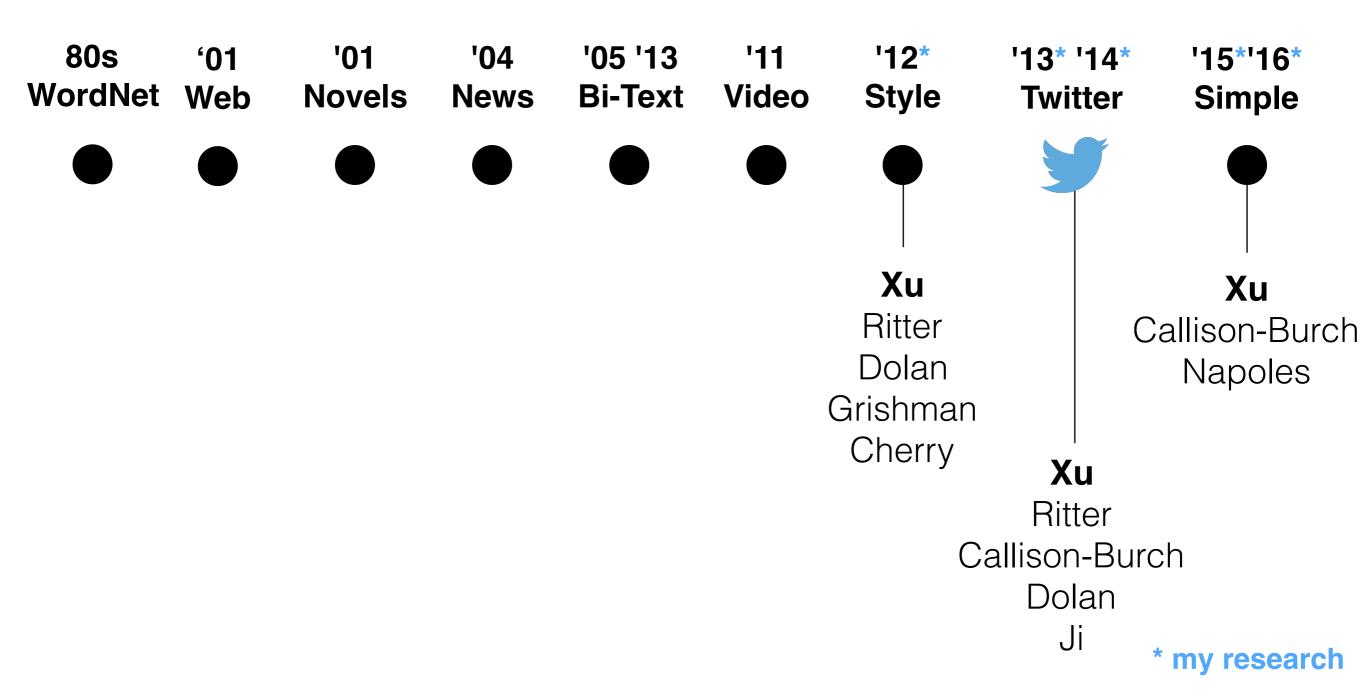
(Discovery of Inference Rules from Text)

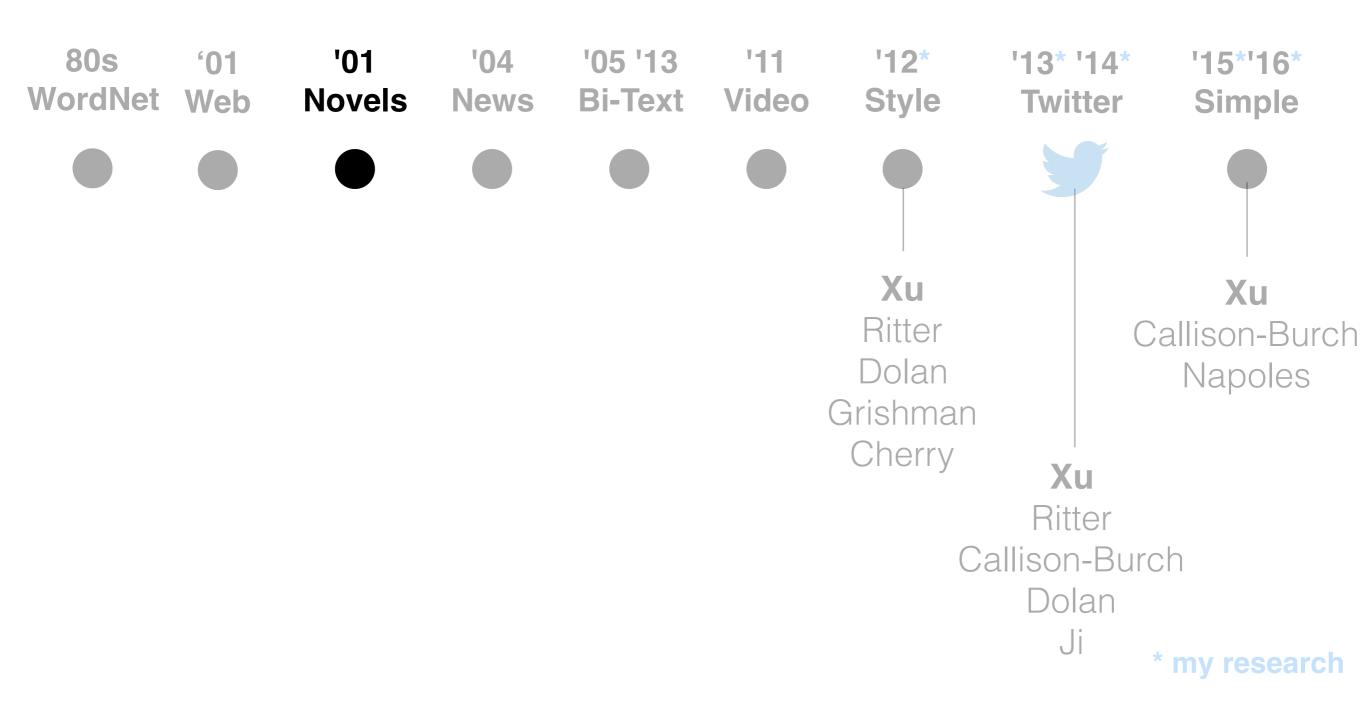
Lin and Panel (2001) operationalize the Distributional Hypothesis using dependency relationships to define similar environments.

Duty and responsibility share a similar set of dependency contexts in large volumes of text:

modified by adjectives	objects of verbs
additional, administrative, assigned, assumed, collective, congressional, constitutional	assert, assign, assume, attend to, avoid, become, breach

Source: Chris Callison-Burch







What a scene! Seized by the tentacle and glued to its suckers, the unfortunate man was swinging in the air at the mercy of this enormous appendage. He gasped, he choked, he yelled: "Help! Help!" I'll hear his harrowing plea the rest of my life!

The poor fellow was done for.

What a scene! The unhappy man, seized by the tentacle and fixed to its suckers, was balanced in the air at the caprice of this enormous trunk. He rattled in his throat, he was stifled, he cried, "Help! help!" That heart-rending cry! I shall hear it all my life.

The unfortunate man was lost.

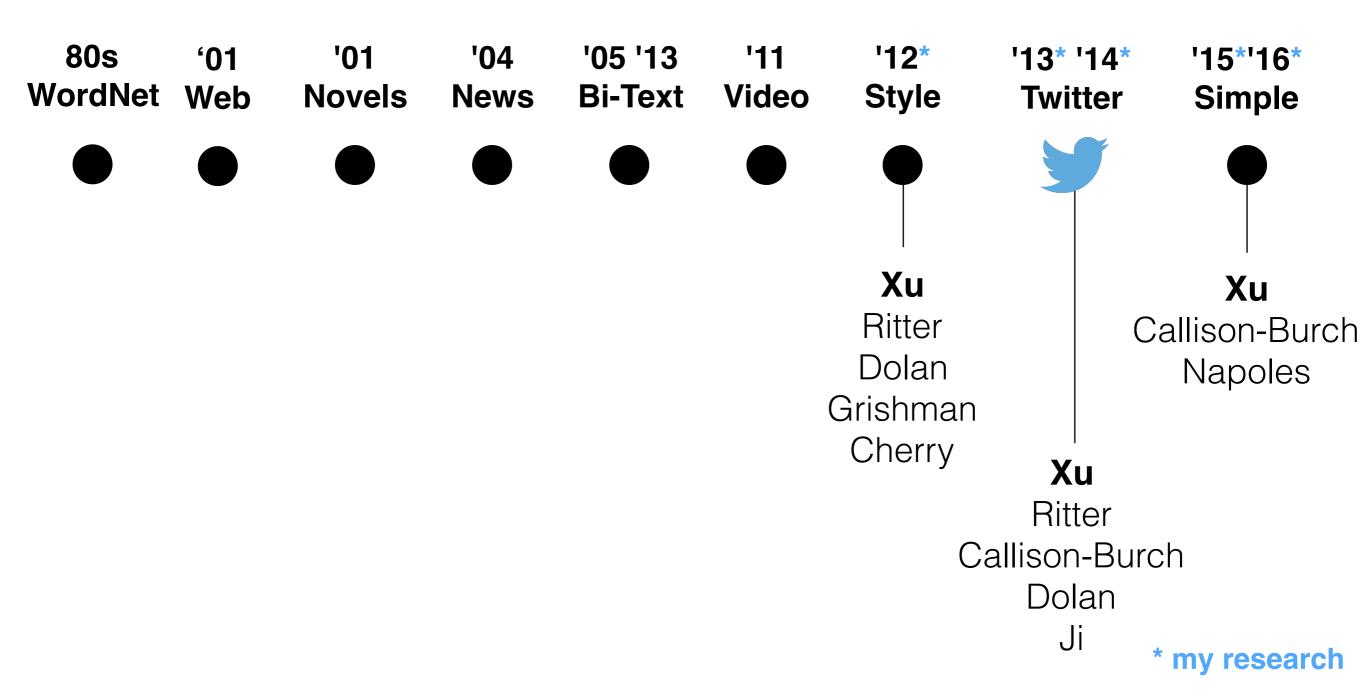
Novels (parallel monolingual data)

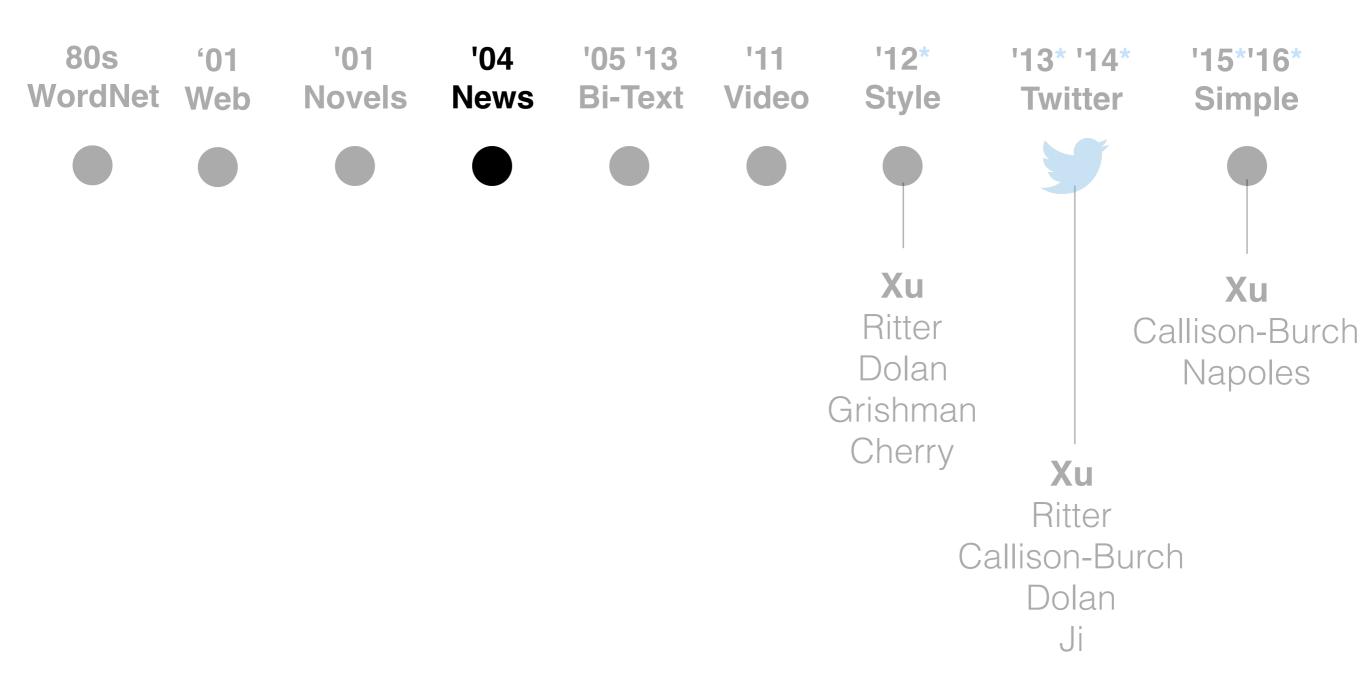
Barzilay and McKeown (2001) identify paraphrases using identical contexts in aligned sentences:

Emma burst into tears and he tried to comfort her, saying things to make her smile.

Emma cried and he tried to console her, adorning his words with puns.

burst into tears = cried and comfort = console





News



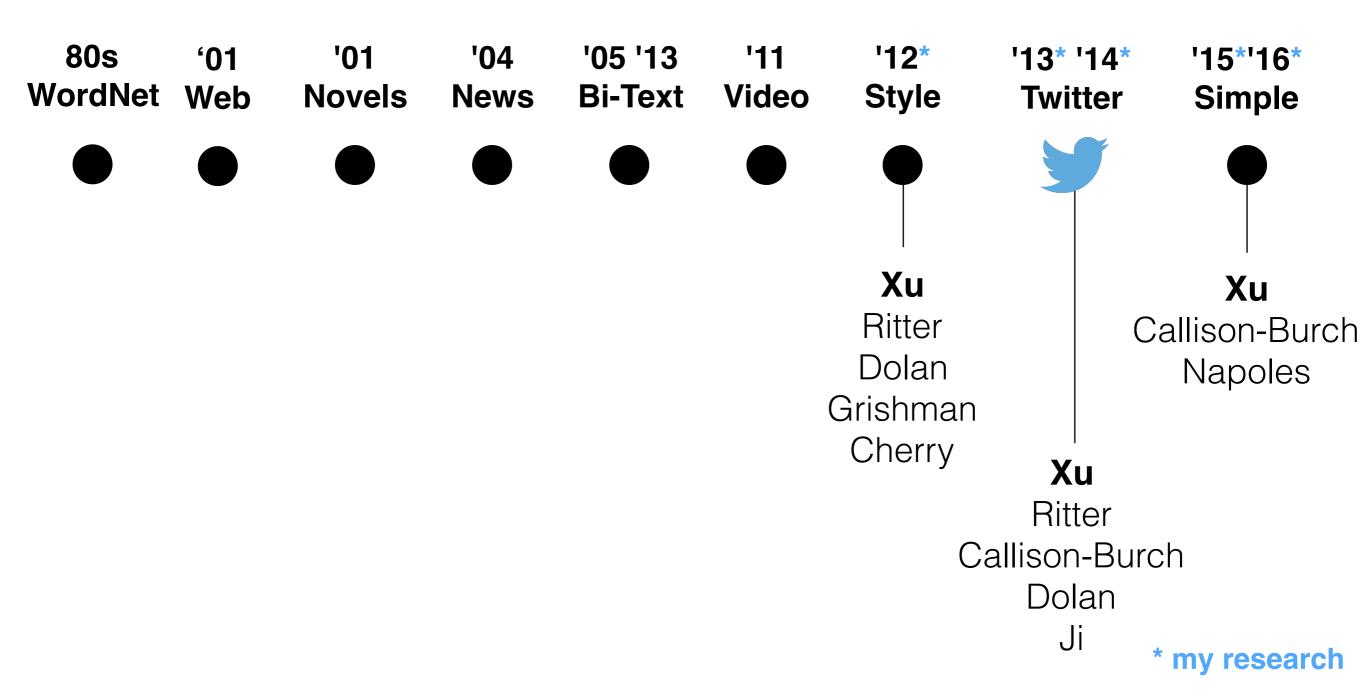
Microsoft Research Paraphrase Corpus

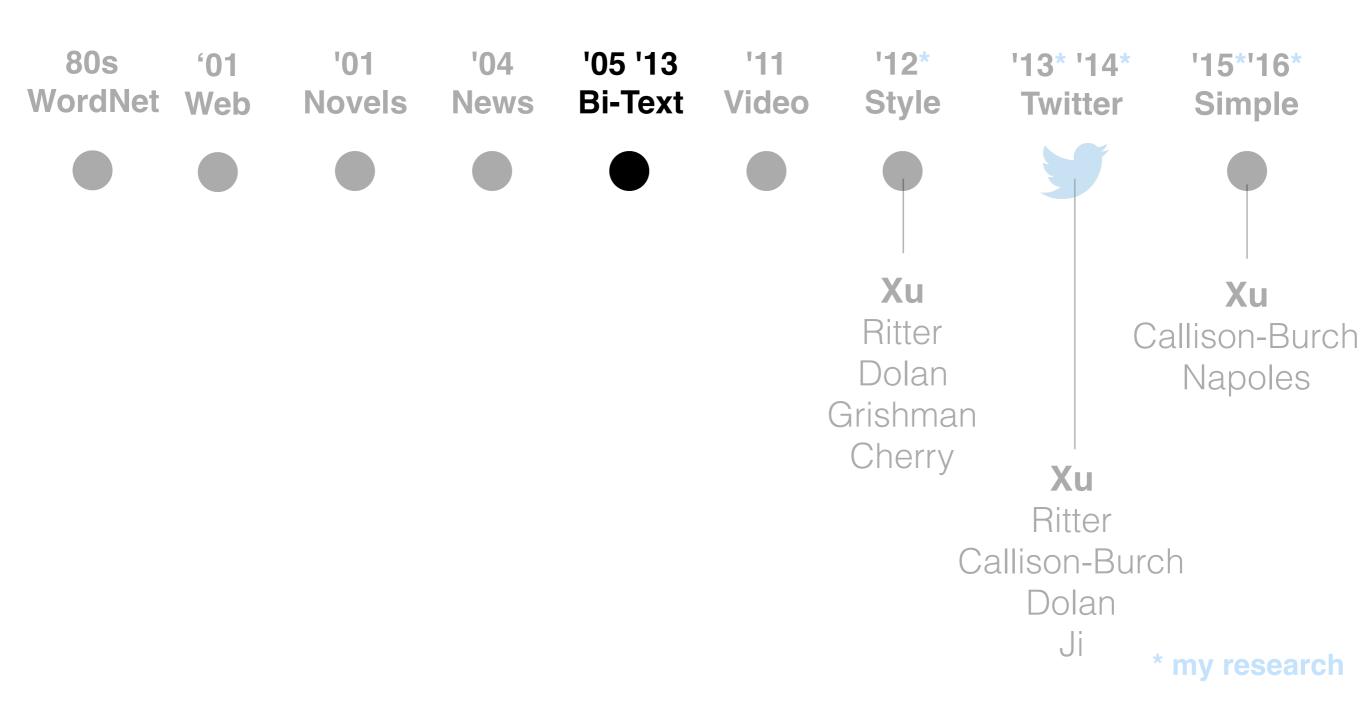
News (comparable texts)

Dolan, Quirk, and Brockett (2004) extract sentential paraphrases from newspaper articles published on the same topic and date:

On its way to an extended mission at Saturn, the Cassini probe on Friday makes its closest rendezvous with Saturn's dark moon Phoebe.

The Cassini **spacecraft**, which is **en route** to Saturn, is about to make a **close pass** of the ringed planet's **mysterious** moon Phoebe.





'01 Novels

Monolingual parallel:

English – English

'01 Novels

Monolingual parallel:

English – English

'01 Web

Plain monolingual:

English

'01 Novels

Monolingual parallel:

English – English

'01 Web

Plain monolingual:

English

'04 News

Monolingual comparable:

English ~ English

Monolingual parallel:

English – English

Plain monolingual:

English

Monolingual comparable:

English ~ English

Monolingual parallel:

English – English

Plain monolingual:

English

Monolingual comparable:

English ~ English

Bilingual parallel:

English – French

Paraphrasing & Translation

Translation is re-writing a text using words in a different language.

Paraphrasing is translation into the same language.

Bilingual Data

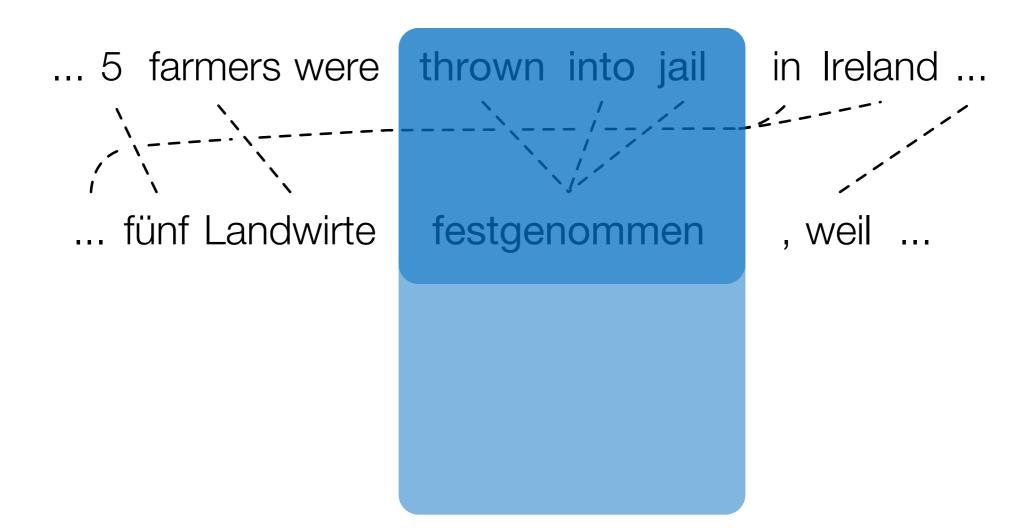
Sentence-aligned parallel corpora in English and any foreign language

Available in large quantities

Strong meaning equivalence signal

... but different languages.

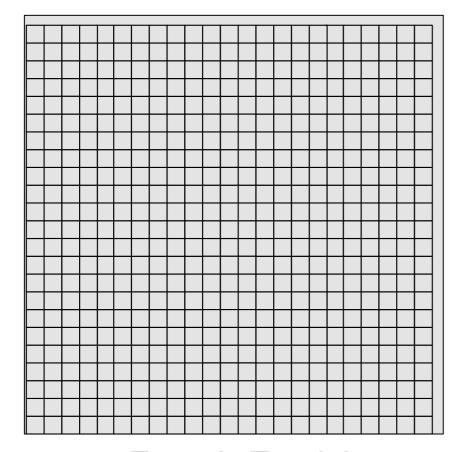
Bilingual Pivoting



Large and diverse

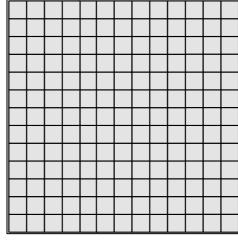
Bilingual Data Sets

1000M



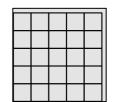
French-English 10^9 word webcrawl

2 languages @ 250M each



DARPA
GALE Program

21 languages @ 50-80M each



European Parliament

Wide range of Paraphrases

thrown into jail

arrested detained imprisoned incarcerated jailed locked up taken into custody thrown into prison who are held in detention

be thrown in prison been thrown into jail being arrested in jail in prison put in prison for were thrown into jail

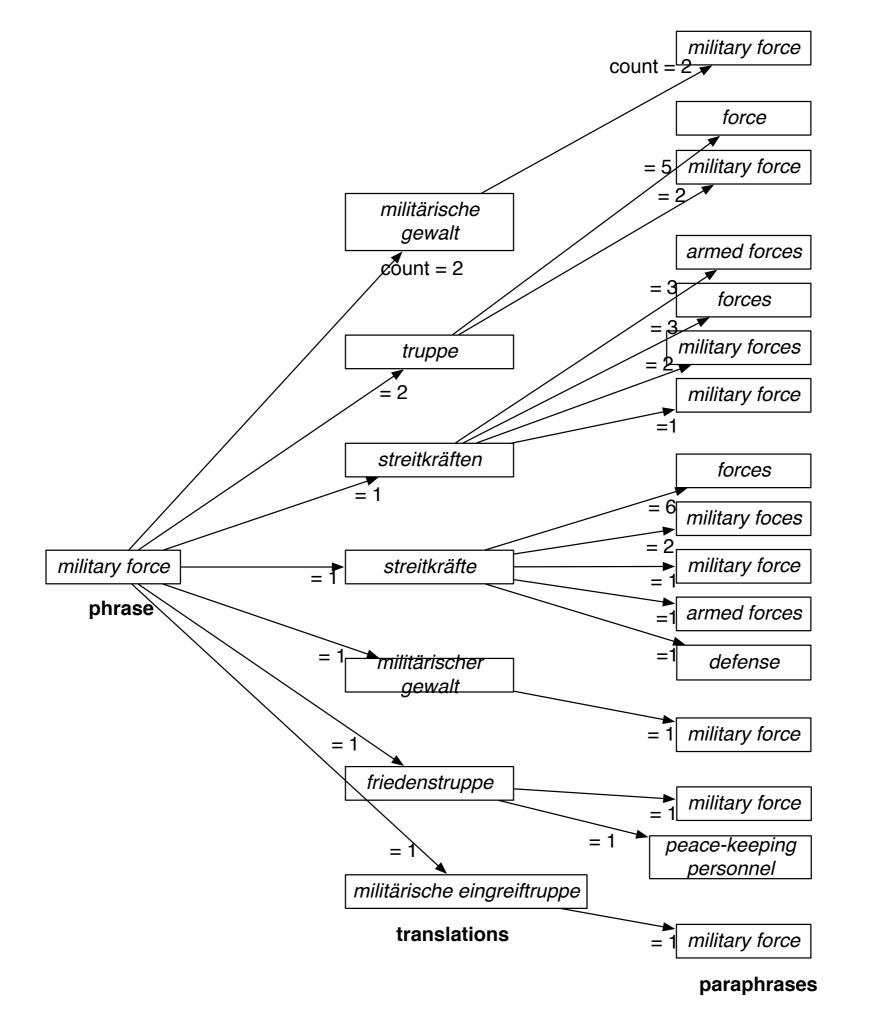
arrest cases custody maltreated owners protection thrown

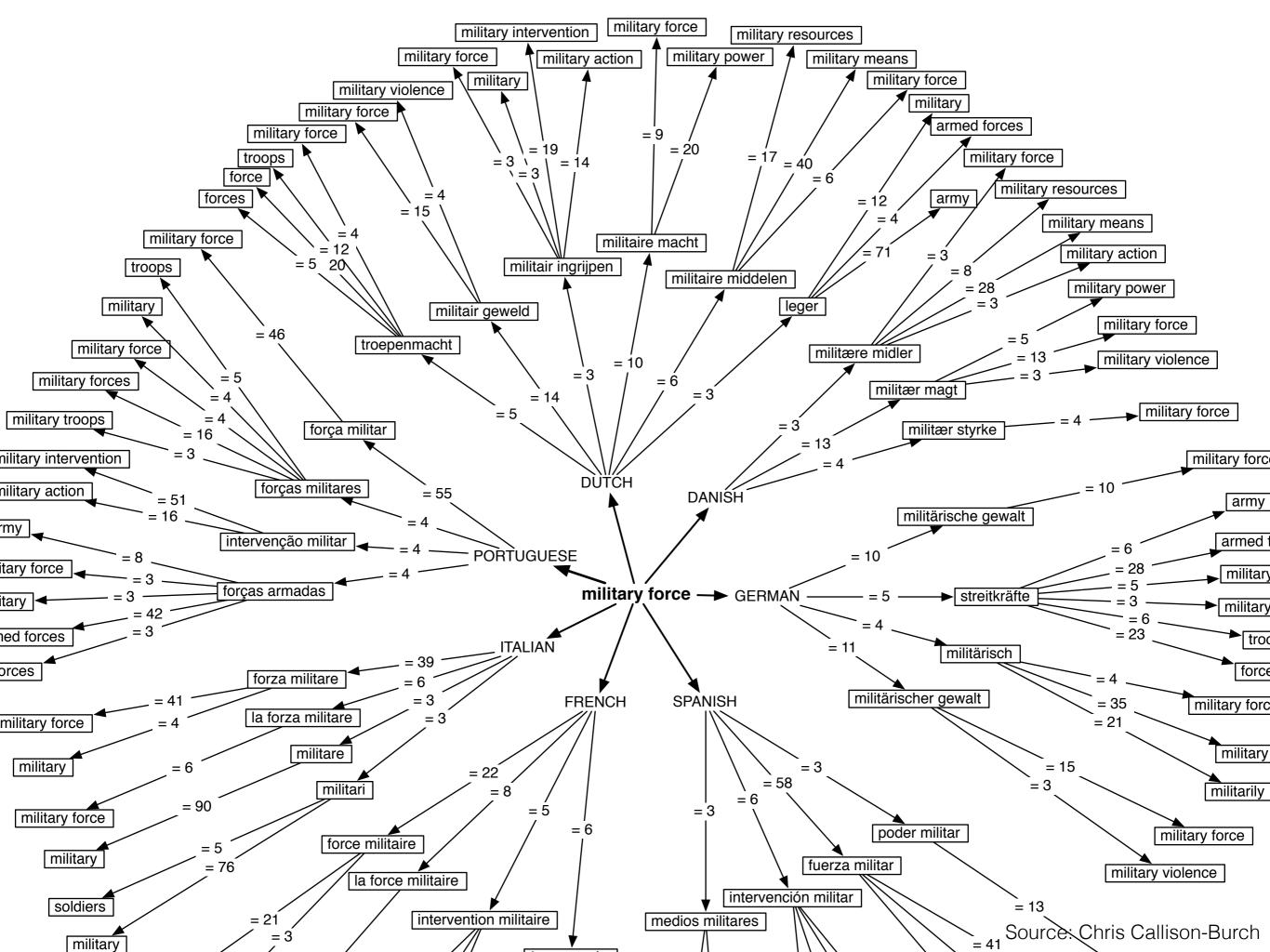
Paraphrase Probability

$$p(e_2|e_1) = \sum_{f} p(e_2, f|e_1)$$

$$= \sum_{f} p(e_2|f, e_1) p(f|e_1)$$

$$\approx \sum_{f} p(e_2|f) p(f|e_1)$$





Syntactic Constraints

thrown into jail

arrested detained imprisoned incarcerated jailed locked up taken into custody

be thrown in prison been thrown into jail being arrested in prison put in prison for were thrown into jail thrown into prison who are held in detention

custody maltreated OWners thrown

Source: Chris Callison-Burch

Distributional Similarity

Idea: similar words occur in similar contexts.

Characterize words by their contexts

Contexts represented by co-occurrence vectors, similarity quantified by cosine

"Are these paraphrases substitutable?"

Similarity

Easy for lexical & phrasal paraphrases

More involved for syntactic paraphrases

..sip from a cup of cocoa..
..a cup of coffee.

cup

 \leftrightarrow

..sip from a mug of cocoa...
..a mug of coffee.

mug

..anxiously awaiting the king's speech..

the king's speech

 \longleftrightarrow

..anxiously awaiting His Majesty's address..

His Majesty's address

?

one JJ instance of NP

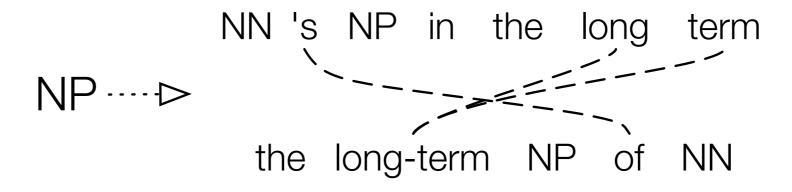


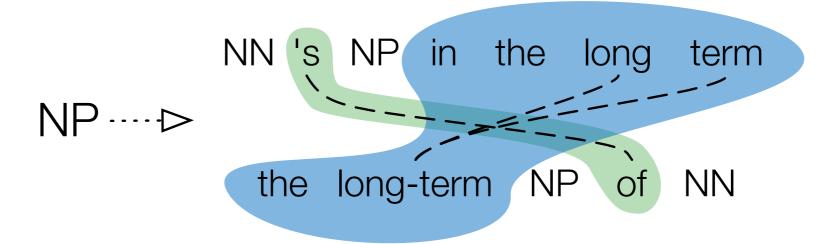
a JJ case of NP

NN 's NP in the long term

NP····⊳

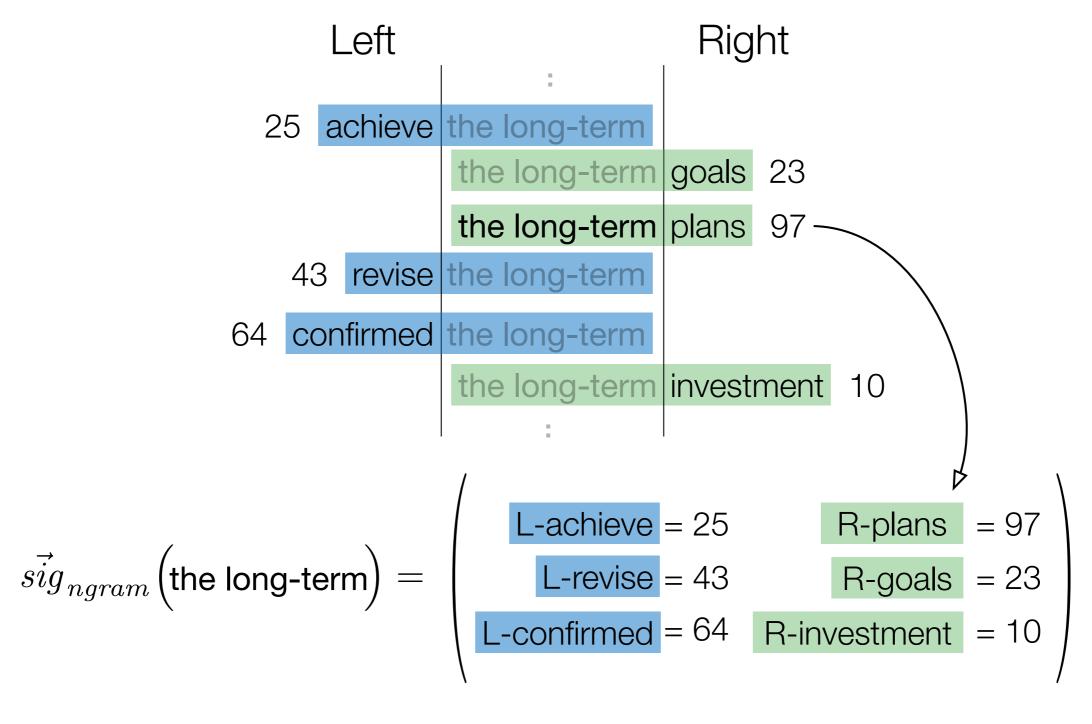
the long-term NP of NN





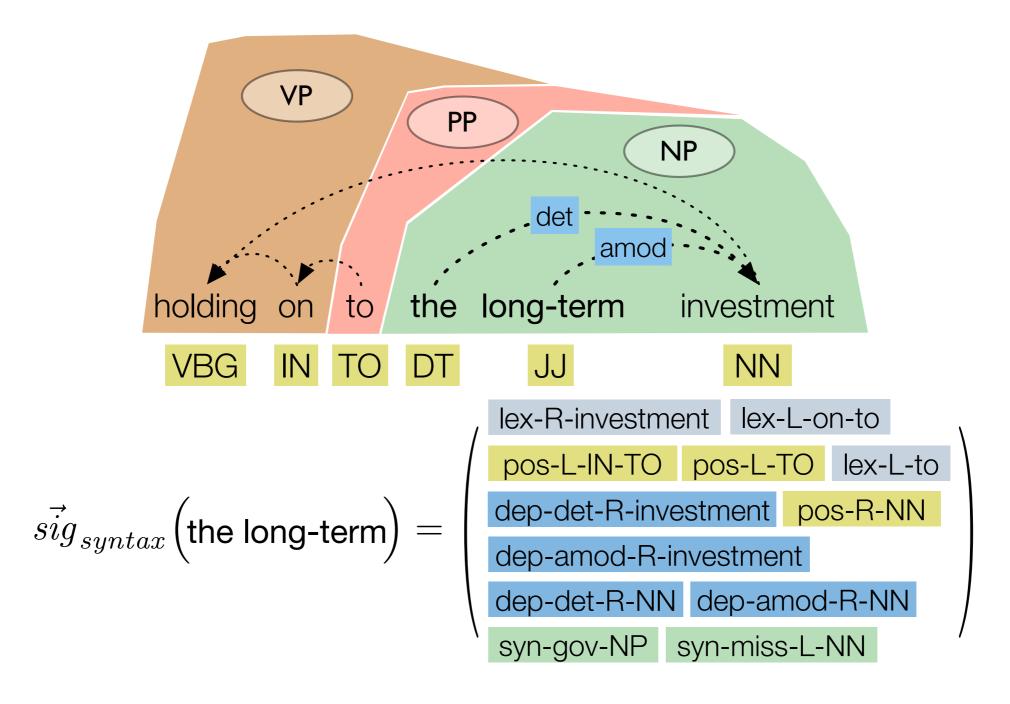
$$sim(\mathbf{r}) = \frac{1}{2} \left(sim \binom{\text{the long-term}}{\text{in the long term}} + sim \binom{\text{'s}}{\text{of}} \right)$$

n-gram Context



Source: Chris Callison-Burch

Syntactic Context



Large Monolingual Data Sets

Google n-grams

Collection of 1 trillion tokens with counts

Based on vast amounts of text

Annotated Gigaword (AKBC-WEKEX '12)

Collection of 4 billion words, parsed and tagged

Source: Chris Callison-Burch

PPDB: The Paraphrase Database

- A huge collection of paraphrases
- Extracted from 106 million sentence pairs,
 2 billion English words, 22 pivot languages

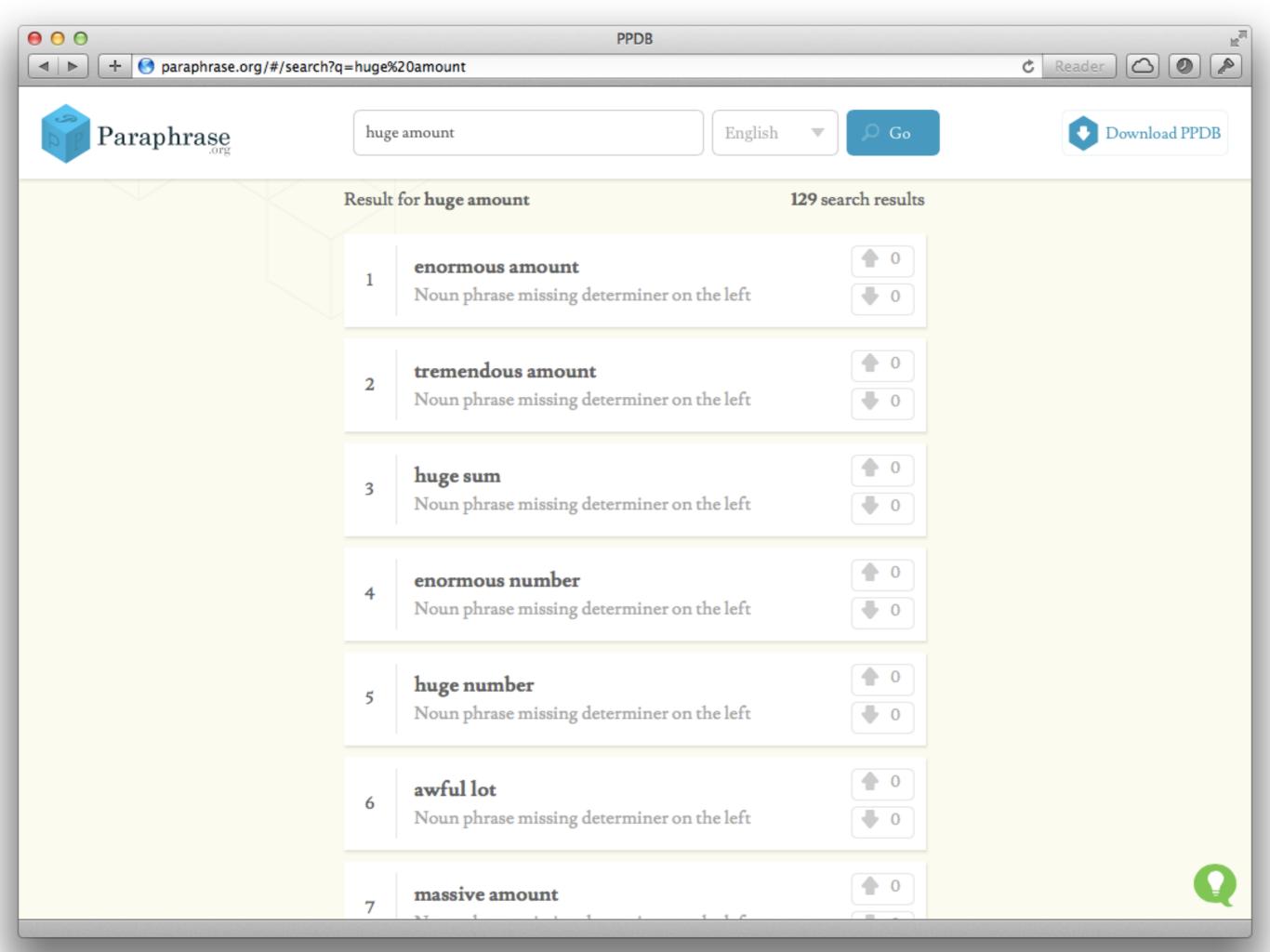
	Paraphrases			
Lexical	7.6 M			
Phrasal	68.4 M			
Syntactic	93.6 M			
Total	169.6 M			

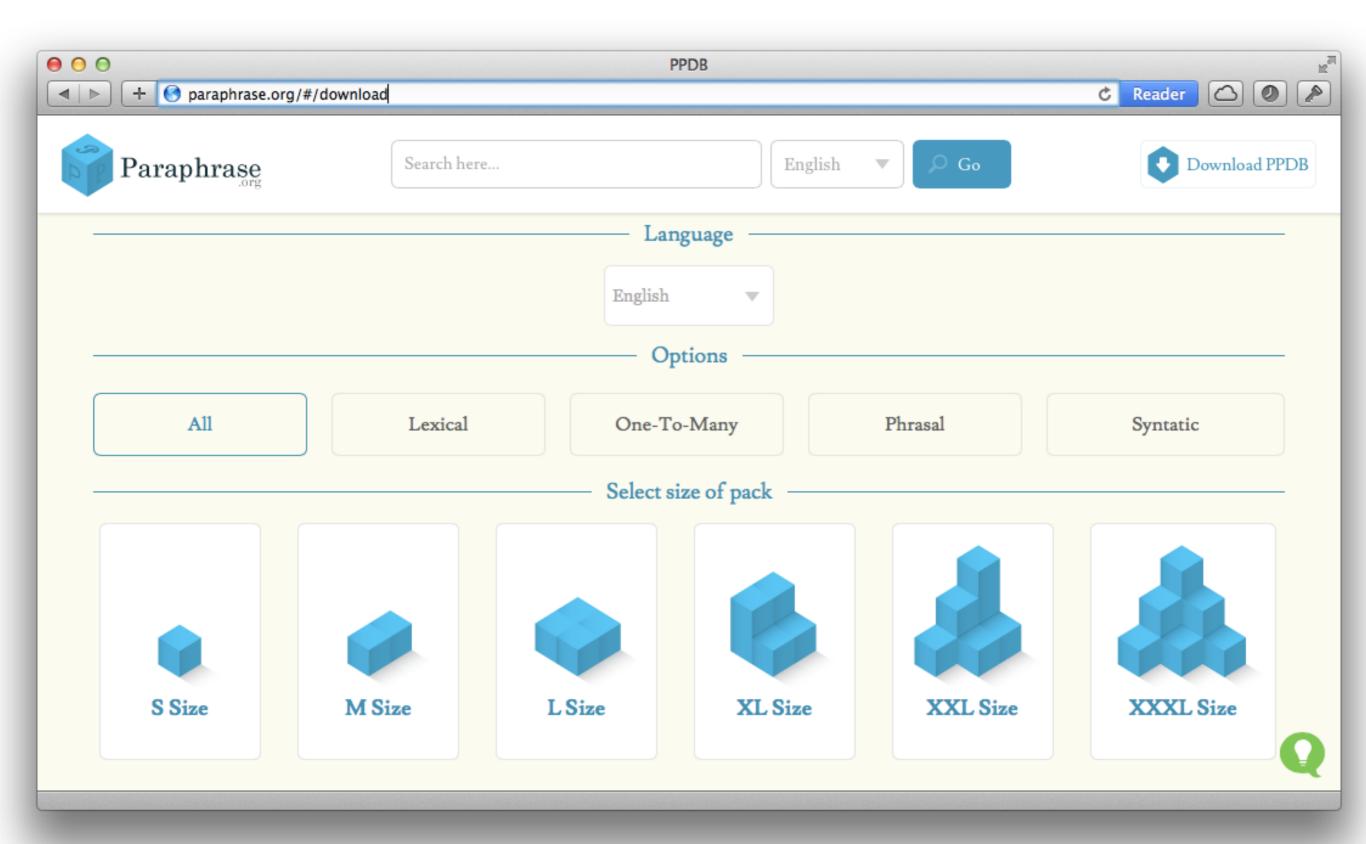
http://paraphrase.org

PPDB: The Paraphrase Database

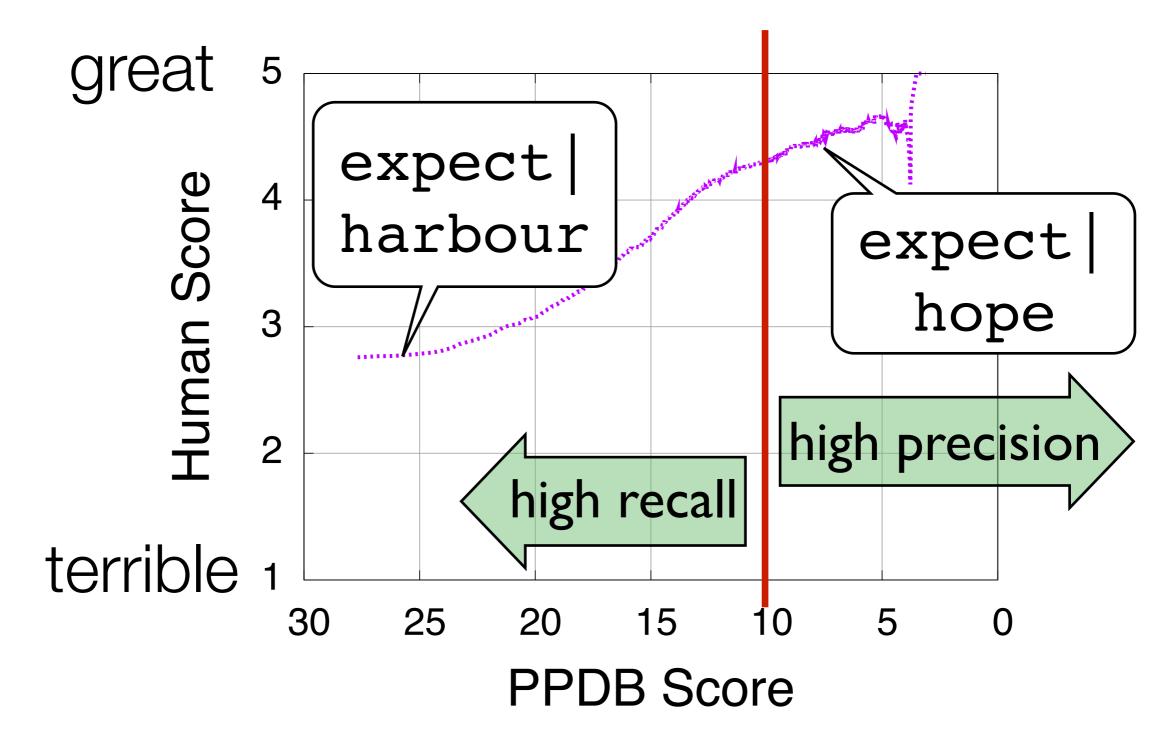
Language	Code	Number of Paraphrases			
		Lexical	Phrasal	Syntactic	Total
Arabic	Ara	119.7M	45.1M	20.1M	185.7M
Bulgarian	Bul	1.3M	1.4M	1.2M	3.9M
Czech	Ces	7.3M	2.7M	2.6	12.1M
German	Deu	7.9M	15.4M	4.9M	28.3M
Greek	Ell	5.4M	9.4M	7.4M	22.3M
Estonian	Est	7.9M	1.0M	0.4M	9.2M
Finnish	Fin	41.4M	4.9M	2.3M	48.6M
French	Fra	78.8M	254.2M	170.5M	503.5M
Hungarian	Hun	3.8M	1.3M	0.2M	5.3M
Italian	Ita	8.2M	17.9M	9.7M	35.8M
Lithuanian	Lit	8.7M	1.5M	0.8M	11.0M
Latvian	Lav	5.5M	1.4M	1.0M	7.9M
Dutch	Nld	6.1M	15.3M	4.5M	25.9M
Polish	Pol	6.5M	2.2M	1.4M	10.1M
Portuguese	Por	7.0M	17.0M	9.0M	33.0M
Romanian	Ron	1.5M	1.8M	1.1M	4.5M
Russian	Rus	81M	46M	16M	144.4M
Slovak	Slk	4.8M	1.8M	1.7M	8.2M
Slovenian	Slv	3.6M	1.6M	1.4M	6.7M
Swedish	Swe	6.2M	10.3M	10.3M	26.8M
Chinese	Zho	52.5M	46.0M	8.9M	107.4M

Source: Chris Callison-Burch





Do the Scores Work?



Fun PPDB Examples

munchies | hungry

PARENTAL am

ADVISORY
sheeit
EXPLICIT CONTENT

abso-fucking-lutely | | indeed

Pivoting w/ Neural MT

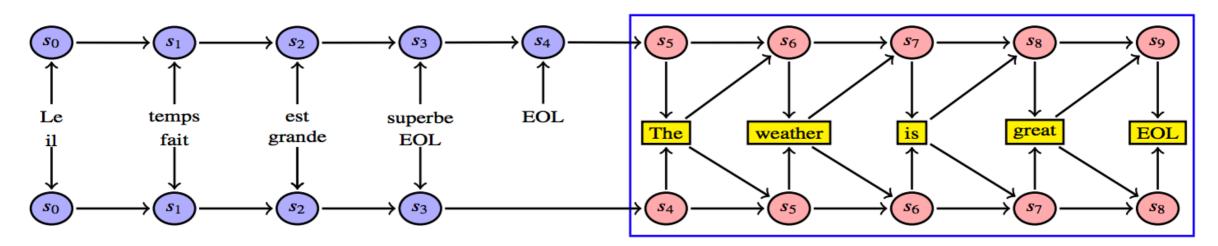


Figure 1: Late-weighted combination: two pivot sentences are simultaneously translated to one target sentence. Blue circles indicate the encoders, which individually encode the two source sentences. After the EOL token is seen, decoding starts (red circles). At each time step the two decoders produce a probability distribution over all words, which are then combined (in the yellow square) using Equation (6). From this combined distribution a word is chosen, which is then given as input to each decoder.

Pivoting w/ Neural MT

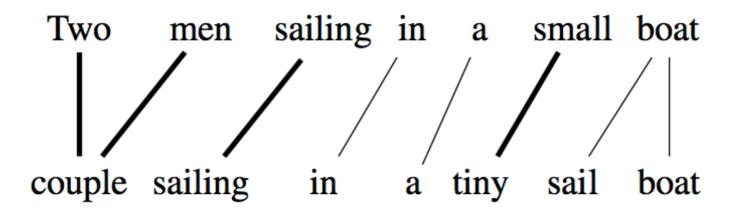


Figure 2: Attention between two sentences. Line thickness indicates the strength of the attention.

$$\alpha(E_2^i, E_1^j, \mathcal{F}) = \sum_{F}^{\mathcal{F}} (P(E_2|E_1, F) \cdot \sum_{m}^{T_F} (\alpha_{i,m}^{E_2, F} \cdot \alpha_{m,j}^{F, E_1}))$$

Improve MT w/ PPDB

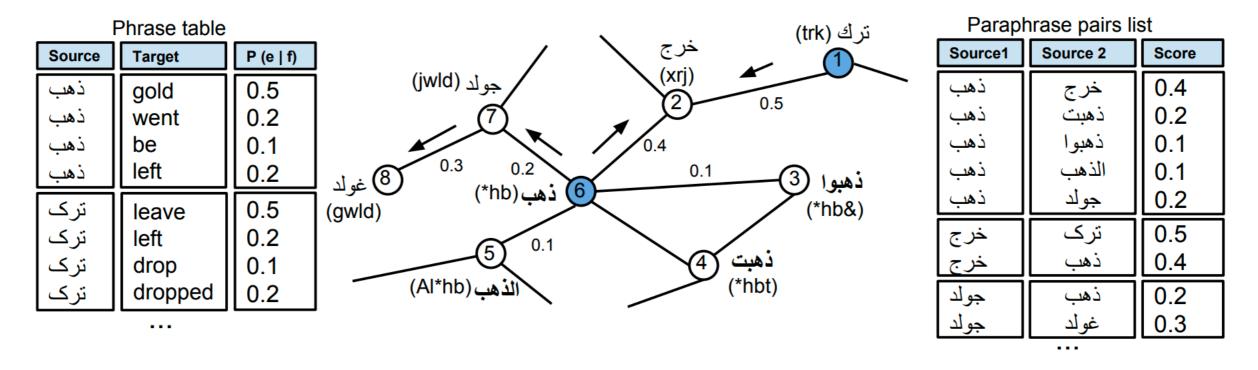


Figure 2: A small sample of the real graph constructed from the Arabic PPDB for Arabic to English translation. Filled nodes (1 and 6) are phrases from the SMT phrase table (unfilled nodes are not). Edge weights are set using a log-linear combination of scores from PPDB. Phrase #6 has different senses ('gold' or 'left'); and it has a paraphrase in phrase #7 for the 'gold' sense and a paraphrase in phrase #2 for the 'left' sense. After propagation, phrase #2 receives translation candidates from phrase #6 and phrase #1 reducing the probability of translation from unrelated senses (like the 'gold' sense). Phrase #8 is a misspelling of phrase #7 and is also captured as a paraphrase. Phrase #6 propagates translation candidates to phrase #8 through phrase #7. Morphological variants of phrase #6 (shown in bold) also receive translation candidates through graph propagation giving translation candidates for morphologically rich OOVs.

Thank You!

