Word Knowledge vs. World Knowledge

Augmenting WordNets with (Un)Common Sense for Robust Web Applications



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The Missing Link: Bridging "Word" and "World" Knowledge

Real Texts (in Real Life) use words and categories in unexpected ways ...



The Rutting Chimpanzee

The Rational Animal

WordNets vs. WorldNets: Lightweight vs. Heavyweight

Dictionaries & WordNets are just one part of a language-processing solution

We must be realistic about what WordNets can and cannot offer the user

WordNets are simultaneously aimed at very different kinds of user

Linguists & language scholars | dictionary users | AI/NLP computer scientists

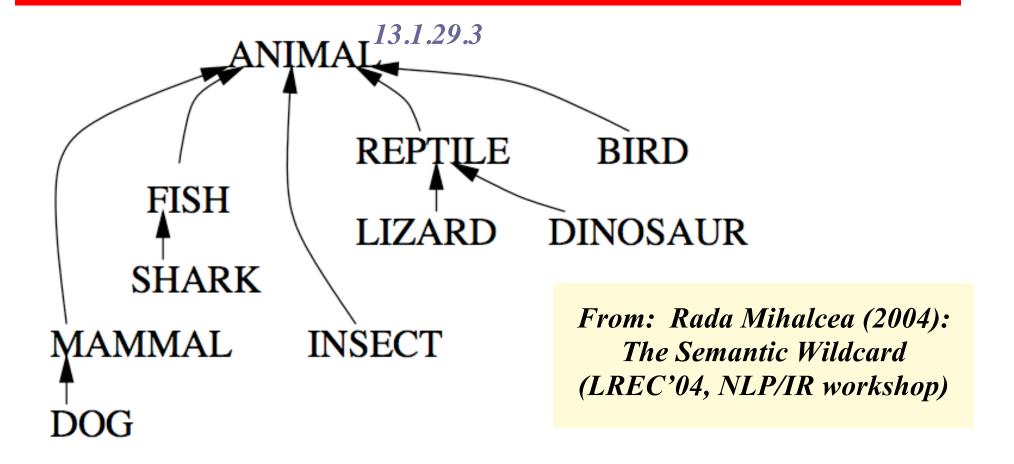
WordNets are lightweight ontologies. WordNets are not WorldNets

We can integrate WNs with sources of encyclopaedic knowledge [Cyc | SUMO]

Natural language processing requires word knowledge and word knowledge

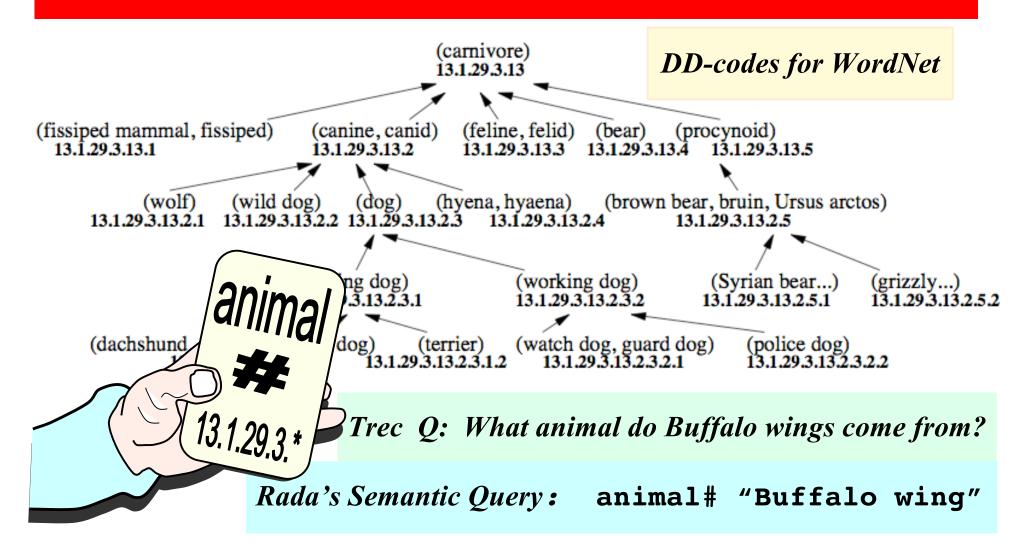
WordNets provide most of the former, some of the latter. But we need more ...

Using WordNets for Semantics: Rada Mihalcea's Semantic Wildcard



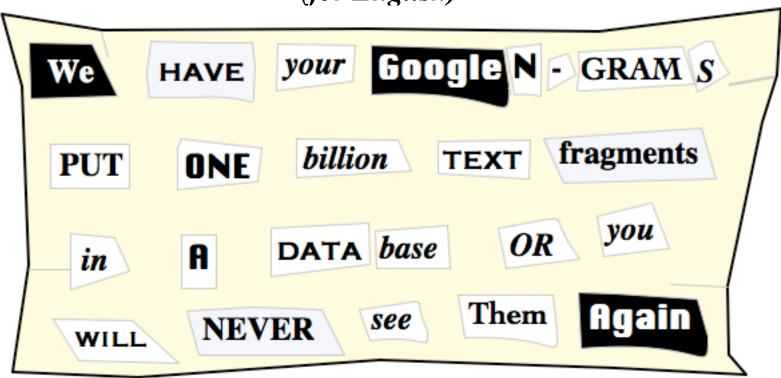
Trec Q: What was the largest dinosaur? A: Diplodocus? Argentinosaurus?

Using WordNet for Answer Retrieval: Semantic Wildcard Matching



A More Fluid View of Semantic Categories: Large Corpora

The Google N-Grams is vast database of recurring text fragments on the Web (for English)



Web n-grams: suited to mining knowledge from recurring small text fragments

Mining Collocations from Corpora: Robust Category Structures

5-grams

Roger Federer, Tiger Woods
Rafael Nadal and Roger Federer
Roger Federer, Andy Roddick
Thierry Henry, Roger Federer
Tiger Woods, Roger Federer
David Beckham, Thierry Henry
Tom Cruise & David Beckham
Tom Cruise and Katie Holmes
Steven Spielberg, Tom Cruise
Tom Hanks / Steven Spielberg
Dan Brown and Tom Hanks
Tiger Woods vs. Thierry Henry

4-grams

tennis and golf players
tennis / squash players
soccer and hockey moms
polo and tennis teams
squash and tennis courts
soccer and rugby fields
tennis and soccer fans
soccer and tennis players
polo and lacrosse teams
soccer vs. golf players
TV and movie stars
radio and TV stars

3-grams

polo and tennis
artists and scientists
apples and oranges
players and fans
coaches and players
golf vs. soccer
terrorism and extremism
soccer versus tennis
Hollywood / Bollywood
radia V
a lirectors

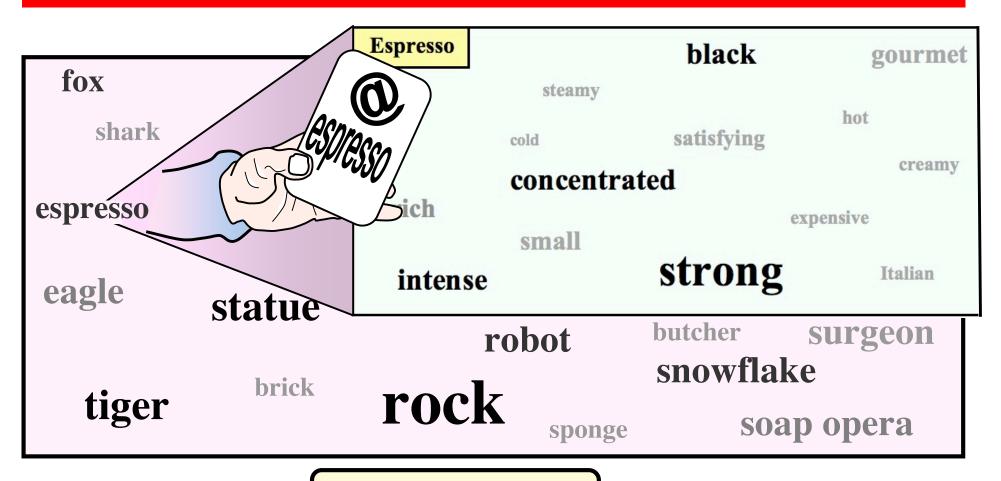
E.g., we use the <u>Google N-Grams 1T Web Corpus</u> $(N \le 5)$

Pragmatic Comparability Versus Semantic Similarity

disaster	<u> </u>	terrorist		beas	t —
tragedy	99	extremist	90	savage	97
catastrophe	99	radical	88	animal	96
calamity	98	anarchist	83	brute	95
destruction	90	subversive	83	wolf	94
famine	89	revolutionary	82	vulture	86
hardship	89	insurgent	72	pet	83
plague	89	separatist	72	plant	73
misfortune	88	guerrilla	71	thief	73
mishap (85	tyrant (71	bird	70
affliction	84	hacker	70	reptile	64
death dicas	Place	rebel ATTO	LZQ	bandit 🚺	63
explosion		liberal Alvilon	ANK.	insect \	63
: \		: \	~ / ·	: \he	asi L
			<u> </u>		MA CANA

Calculate WordNet-based semantic similarity for each coordination

Stereotypical Associations: Mine Simile patterns from the WWW



Use Web query pattern

" as * a | an as * "

to harvest 1000's of similes

Stereotypical Properties co-occur in pragmatic clusters

Adjacency matrix of mutually-reinforcing properties acquired from WWW:

	hot	spicy	humid	fiery	dry	sultry	•••
hot	?	35	39	6	34	11	•••
spicy	ho'		0	15	1	1	•••
humid				0	1	0	•••
fiery	6		0	-	0	0	•••
dry	6		0	0		0	•••
sultry	11	1	0	2	0		•••
• • •	•••	•••	•••	•••	•••	•••	•••

Use the Google query

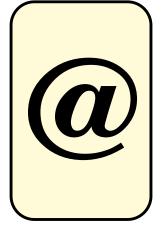
"as * and * as"

to acquire associations

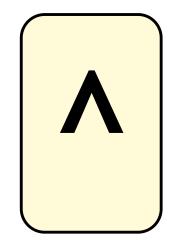
Creative Information Retrieval with Pragmatic Wildcards



Pragmatic Neighbourhood



Stereotype



Named Category





Matches:



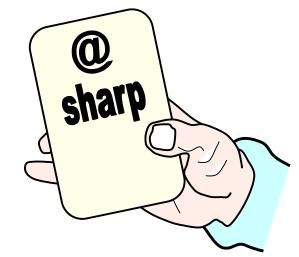
savage, animal, brute... Matches: painful, nasty, depraved...



Matches: brute, barbarian, bully, cannibal, criminal ...

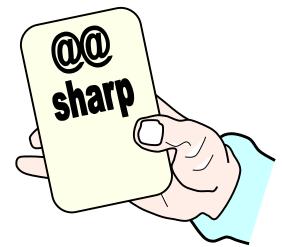


Matches: violent, dirty, vile, embarrassing, humiliating...



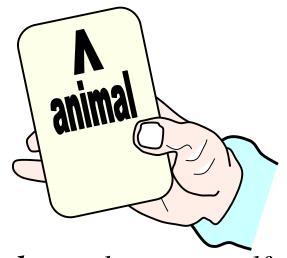


Matches: sword, razor, laser... Matches: strong, ugly, wild...



Matches: stiff, lethal, piercing... Matches: tiger, toad, ape, bear...





Matches: dog, cat, wolf, ape,...



Matches: violent, ugly, vile, muscular, scary, ...

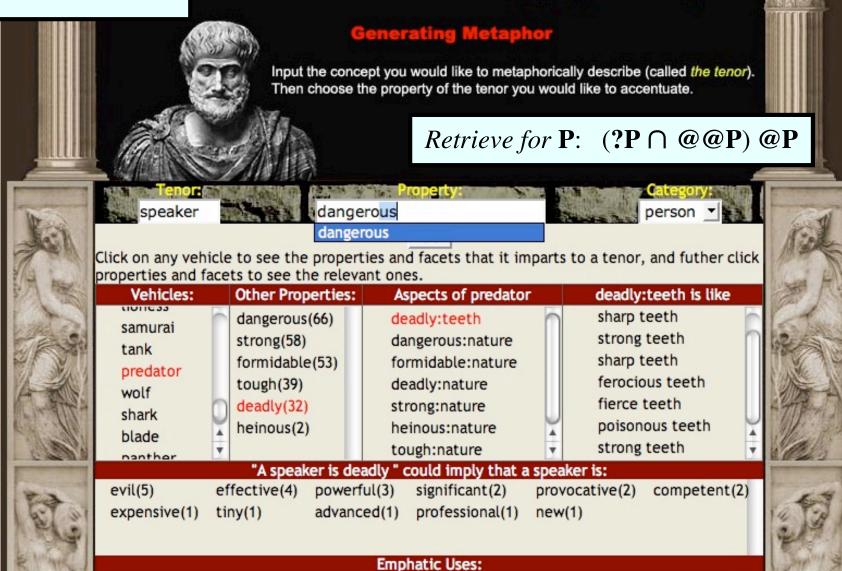


Matches: wild, strong, hairy,...



Matches: bull, bear, gorilla, hog, warthog, dog...

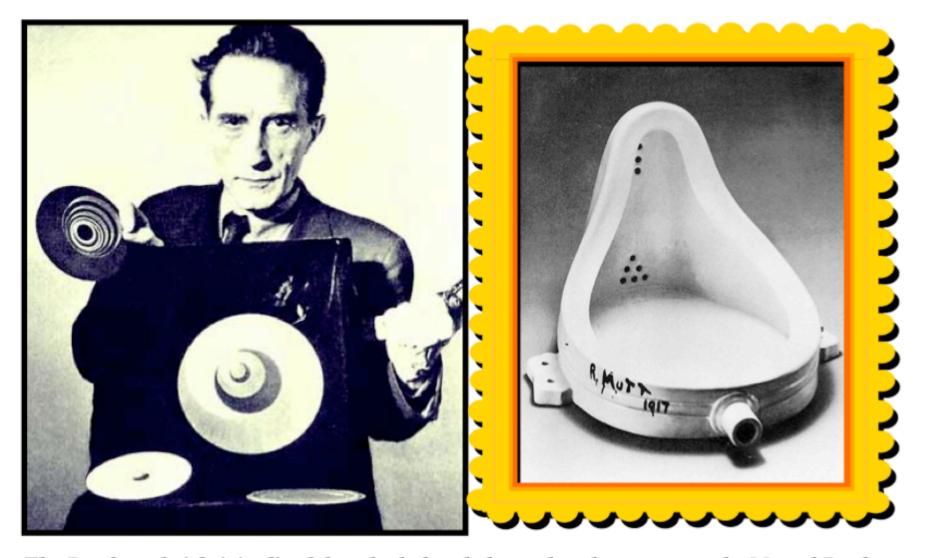
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George Orwell expressed a deep and persuasive distrust of re-use & readymades in language



The Readymade / Originality debate had already been placed centre-stage by Marcel Duchamp His infamous "Fountain" showed that art required neither true originality nor manual craft



Creativity arises as much from intentions as from meanings, and from decisions as much as actions

Everyday objects, wrenched from their conventional contexts of use, can acquire resonant new meanings in new contexts.

And so it is for familiar phrases.

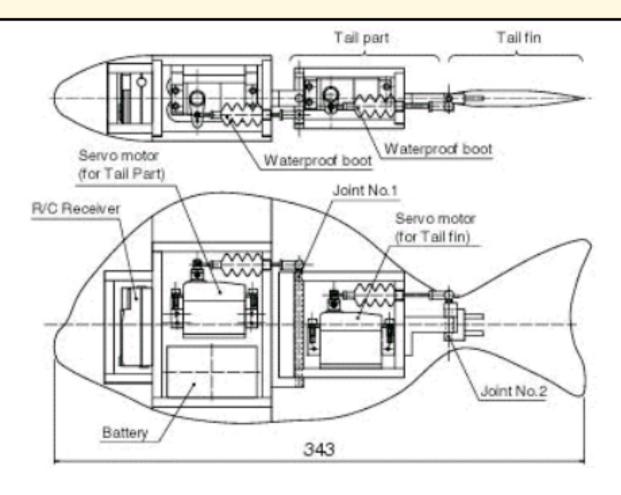
If wrenched from their common contexts of use, they can acquire creative new meanings.

Example: The Google 2-gram

"robot fish"

Stereotype for coldness

Stereotype for coldness



Input an adjectival property

as cold

as Go

Afflatus.UCD.ie/jigsaw

Phrases in blue are computer-generated; all other phrases are automatically mined from large corpora

The Jigsaw

Co-Occurring Properties of 'cold'

cold and slippy

cold and dreary

cold and heartless

cold and motionless

cold and miserable

cold and inorganic

cold and unsympathetic

Peotic Elaborations

the eye of a storm (10365) the eye and power of a storm (10365) the eye and voice of a storm (10365) the eye and air of a storm (10365) the eye and wake of a storm (10365) the power of a storm (2828)

the power and eye of a storm (2828)

the power and voice of a storm (2828) the power and air of a storm (2828)

the power and fury of a storm (2828)

Afflatus Home Aristotle Sardonicus

Simple Elaborations

a wet haddock (6155)

a wet fish (6152)

a wet snow (6142)

a wet January (6118)

a wet storm (6112)

a wet cucumber (6111)

a wet mackerel (6109)

a wet snowball (6106)

a wet snowstorm (6106)

an unfeeling robot (2411)

a heartless robot (2207)

a gray January (2109)

a lifeless corpse (2031)

a lifeless robot (2006)

a bitter storm (1714)

a bitter January (1713)

pale corpse (1610)

a bitter snowstorm (1707)

Retrieve: ?P @P

Complex Elaborations

a fish-eyed storm (10040)

Bard

a glacier with the eye of a fish (10040)

the belly of a fish (10032)

the wake of a storm (10032)

the wall of a cave (10032)

a snow blizzard (10029)

a snowy January (10023)

a fridge with a refrigerator freezer (10023)

a refrigerator freezer (10023)

the flesh of a fish (10022)

the fury of a storm (10020)

a bullet-riddled corpse (10019)

the eyes of a fish (10018)

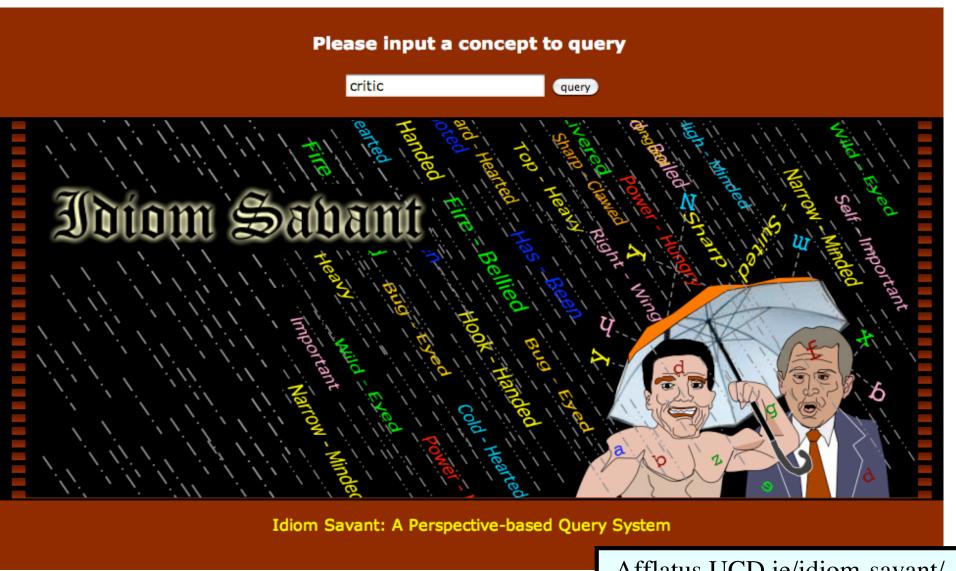
the power of a storm (10018)

a robotic fish (10018)

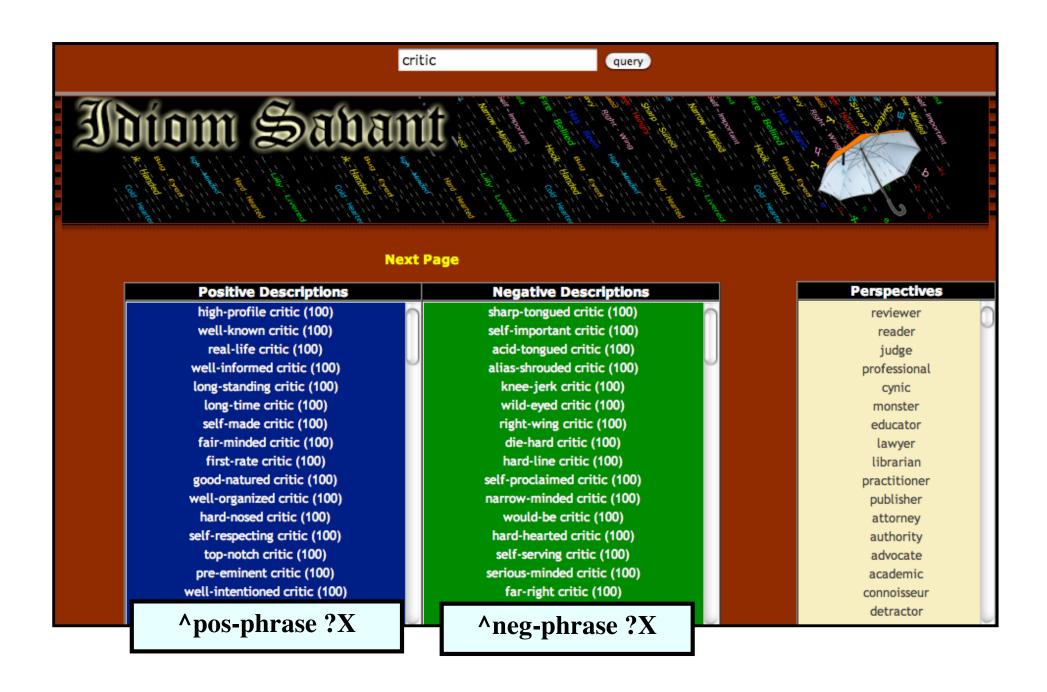
the surface of a steel (10018)

the heart of a killer (10017)

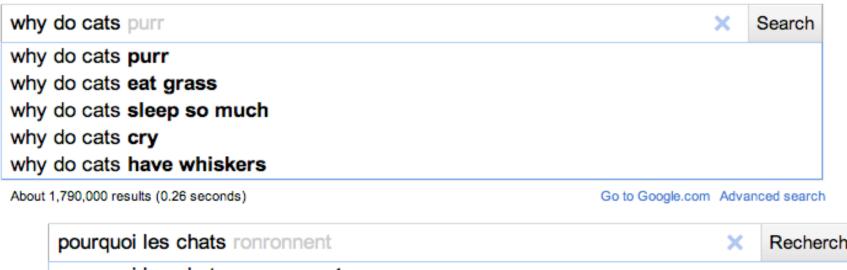
Retrieve: @P @P



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Common Questions On the Web: A Source of World Knowledge

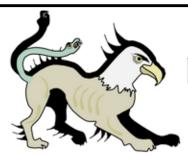




Environ 144 000 000 résultats (0,05 secondes)

Google.com in English Recherche avancée

We "milk" question completions from Google, and parse them into axioms

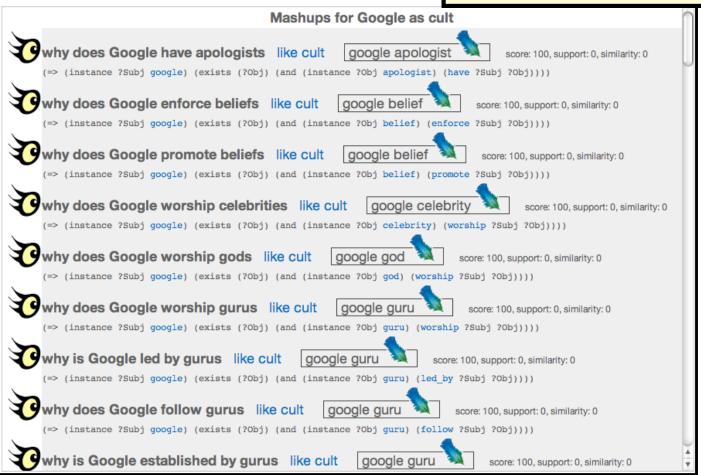


Google is a cult

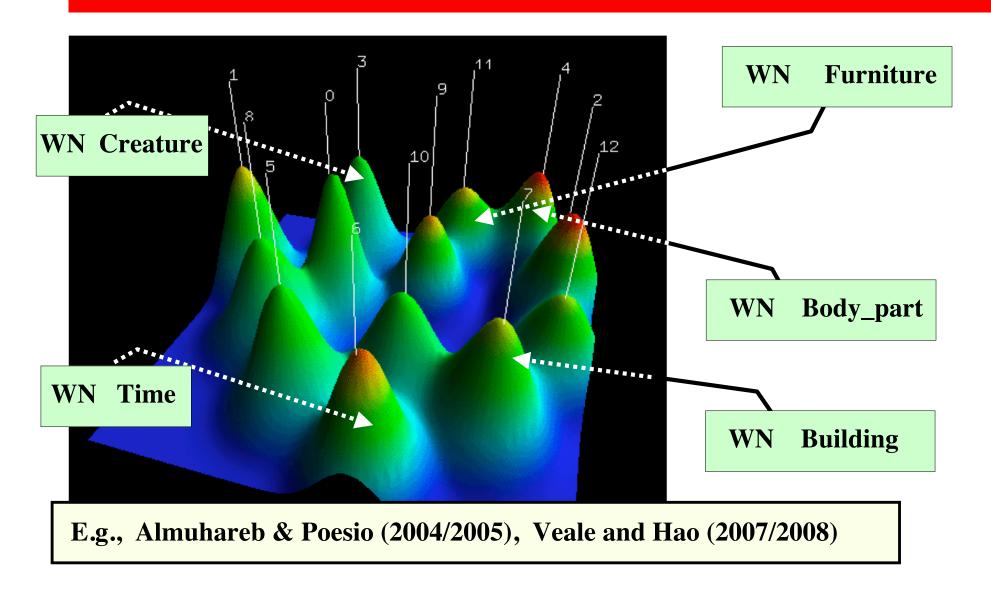
Metaphorize This!

E.g., Scientists as Artists, or, just Scientists

ngrams.UCD.ie/metaphor-eye/

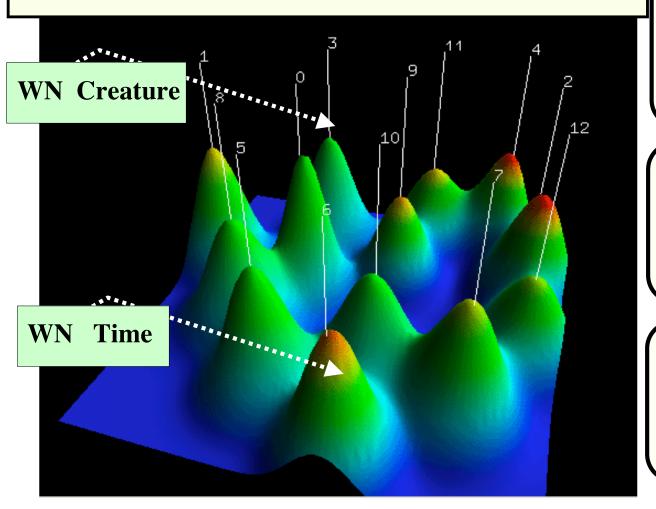


Using WordNet(s) as a Gold-Standard for Pragmatic WorldNet(s)



How Well Does Corpus/Web-based Categorization Do w.r.t. WordNet?

13-way clustering of 214 nouns, compared to WordNet



Almuhareb & Poesio (2004)

Weak text-derived features

~ 60,000 features for 214 nouns

Result: 0.855 cluster purity

Veale & Hao (2007 / 2008)

Strong simile-derived features
~7,300 features for 214 nouns

Result: 0.902 cluster purity

Veale & Li (2009)

Generic Clique-derived features

 \sim 8,300 features for 214 nouns

Result: 0.934 cluster purity

Semantics vs. Pragmatics: Similarity vs. Comparison

WordNets are a good source of word knowledge, lightweight semantics
They must be used as a coherent part of an applied, pragmatic, NLP solution

- Pragmatic knowledge can come from large corpora of real "language use" Comparisons, similes and other tropes are a fluid source of tacit knowledge
- WordNet provides the semantics of similarity (e.g., Budinitsky & Hirst)

 Large corpora / usage data provide the pragmatics of comparability
- A simple framework, instantiated in many ways

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The Aristotle, Jigsaw Bard, Idiom Savant and Metaphorize applications



Questions?