#### Semantic Word Sketches

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Semantic Word Sketches

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#### Outline

- The Sketch Engine
  - Concordances
  - Word Sketches
- Semantic Tagging
  Super Sense Tagger (SST)
  - SST Supersenses
- 3 Semantic Tags in Sketch Engine
  - In the Concordance
  - Semantic Word Sketches
  - $\bullet$  Other Possibilities from  ${}_{\rm SST}$  Output

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4 Comparison to FrameNet



Concordances Word Sketches

#### The Sketch Engine

- concordances, word lists, collocations
- word sketches
  - create and examine syntactic profiles and collocations of words
  - input automatic part-of-speech tags and a bespoke 'sketch grammar'
- automatic thesauruses: which other words have similar profiles?
- sketch differences between words

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#### The Sketch Engine

Semantic Tagging Semantic Tags in Sketch Engine Comparison to FrameNet Conclusions References

Concordances Word Sketches

#### The Sketch Engine

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ordance	Query mouse	49,985 > Random sample 250 (0.16 per million)
List Sketch	Page 1	of 13 Go Next   Last
aurus	site=sange	completed the first phase of reading the mouse 'book of life', reaching its goal on time
h-Diff	site=i-sis	supplements may reverse the damage. Obese mice given dietary supplements were found to
h-Eval	site=sange	identity). Thus, by comparing human and mouse genome sequences, the regions of high similarity
us Info	site=sange	set of BAC clones that covers the entire mouse genome is being sequenced. The BAC data
ge corpus	site=phgu	extraembryonic stem cell lines derived from single mouse blastomeres. Nature advance online publication
bs	site=sange	investigations of the function of genes using the mouse as model genetic system. The principal
	site=birdt	all the group members who were as quiet as mice . We eventually heard Grey-headed Quail-dove
	site=manch	their experiments on monkey kidney cells and mouse skin cells with similar results. Email
	site=littl	very much. To be honest, I'd be fine with mice if they were in a cage. Mice are pretty
ubcorpus	site=blog	said that it wasnt for me it was for my mouse as way of an explanation - she just raised
options	site=i-sis	whereas those from Sertoli cells of immature mice died at an unusually early age. By contrast
c l	site=maryw	the same graphical interaction without a mouse or other pointing device. Alternative text
ence	site=www-e	tissue reactions in normal and immunized mice to a reticulotropic strain of Trypanosoma
	site=theno	open in a new window Open Map ROLL your mouse on and off the EDGE of the maps to reveal
	site=readg	work with Plasmodium chabaudi in laboratory mice (Buckling, Taylor et al. 1997; Taylor, Walliker
t	site=pract	particularly quirky, and involved using the word " mouse " wherever possible inside words, as well

Semantic Word Sketches

The Sketch Engine Semantic Tagging Semantic Tags in Sketch Engine Comparison to FrameNet

References

Concordances Word Sketches

#### The Sketch Engine

Word Sketches: syntactic profiles

Sketch								• Q <u>s</u>	end fee	edback cor	pus: <u>uk</u>	WaC [
Concordance Word List Word Sketch	eat	verb) kWaC fre	q = <u>13</u>	7,789 (88.34 pe	er million	)						
Thesaurus	object	<u>60,826</u>	5.50	subject	<u>16,966</u>	2.20	modifier	<u>22,796</u>	0.70	and/or	<u>9,742</u>	0.70
Sketch-Diff	meat	1,802	8.71	binge	<u>92</u>	7.21	healthily	<u>662</u>	9.82	drink	<u>3,168</u>	10.59
Sketch-Eval	disorder	<u>1,915</u>	8.57	LTY	<u>63</u>	6.81	properly	<u>327</u>	6.84	sleep	<u>681</u>	8.31
Corpus Info	meal	<u>1,812</u>	8.17	RECOMMEND	47	6.47	sensibly	<u>90</u>	6.82	cook	<u>214</u>	7.04
Manage corpus	food	5,233	8.16	ye	<u>89</u>	5.84	enough	<u>212</u>	6.35	shop	<u>73</u>	6.82
My JODS	diet	1,376	8.05	lion	<u>68</u>	5.83	away	437	6.35	socialise	<u>45</u>	6.58
C	fruit	<u>1,298</u>	7.84	vegetarian	<u>40</u>	5.78	regularly	<u>225</u>	6.23	breathe	<u>87</u>	6.41
	lunch	<u>1,070</u>	7.74	flesh	77	5.73	alfresco	<u>47</u>	6.05	swallow	<u>57</u>	6.36
Save	breakfast	<u>910</u>	7.72	slug	<u>37</u>	5.69	outdoors	<u>50</u>	6.01	relax	<u>133</u>	5.97
Change options	habit	<u>751</u>	7.60	caterpillar	<u>34</u>	5.69	happily	<u>66</u>	5.90	smoke	<u>56</u>	5.81
Sorting	bread	<u>694</u>	7.53	cheap	<u>30</u>	5.67	together	<u>455</u>	5.88	chew	<u>31</u>	5.74

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The Sketch Engine

Semantic Tagging Semantic Tags in Sketch Engine Comparison to FrameNet Conclusions References

Concordances Word Sketches

#### Sketch Grammars

Under the hood

- Definitions: define('any\_noun','"N.."')
  - . . .
- Relations

=subject/subject\_of 2:any\_noun rel\_start? adv\_aux\_string\_incl\_be 1:verb\_not\_pp 2:any\_noun rel\_start? adv\_aux\_string\_incl\_be aux\_have adv\_string 1:past\_part

```
1:past_part adv_string [word="by"] long_np
```

Semantic Word Sketches

(a)

Super Sense Tagger (SST) SST Supersenses

#### Semantic Class Tagging

- aim to build word sketches on syntactic and semantic information
- automatic 'superclass' tagging technology
- superclass: a coarse grained semantic class that is applicable to multiple words (e.g. **animal** for *cat*, *fly*, *hare*, *pig* etc...
- allow search and analysis with these classes and
- semantic word sketches: basic semantic frame with semantic preferences for arguments

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Super Sense Tagger (SST) SST Supersenses

#### Semantic Class Tagging

Super Sense Tagger (SST) Ciaramita and Altun (2006) (http://sourceforge.net/projects/supersensetag/)

- semantic tags are WordNet Fellbaum (1998) lexicographer classes
- supervised word sense disambiguation (i.e. it requires hand labelled data for training) using a Hidden Markov Model e.g. labels *mouse* as **animal**, **artifact**)
- SemCor (Landes et al., 1998) used as training data
- Named Entity Recognition e.g. < *RHM Technology Ltd.*> organization
- Multiword tagging using multiwords from WordNet e.g. *couch potato*

Super Sense Tagger (SST) SST Supersenses

#### SST WordNet Noun Classes (25)

. . .

acts or actions act object natural objects (not man-made) animal animals quantities and units of measure quantity artifact man-made objects natural phenomena phenomenon attribute attributes of people and objects plant plants food and drinks food

Semantic Word Sketches

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Super Sense Tagger (SST) SST Supersenses

#### SST WordNet Verb Classes (15)

. . .

body grooming, dressing and bodily care emotion feeling change size, temperature change, intensifying motion walking, flying, swimming thinking, judging, analyzing, doubting cognition seeing, hearing, feeling perception telling, asking, ordering, singing communication buying, selling, owning possession creation sewing, baking, painting, performing

. . .

Semantic Word Sketches

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In the Concordance Semantic Word Sketches Other Possibilities from SST Output

#### Experiments

- just over 25% of the UKWaC Ferraresi et al. (2008)
- SST tagged with
  - part-of-speech tags (Penn TreeBank)
  - supersenses (WordNet labels)
  - Named Entity Labels
  - WordNet multiwords

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#### Semantic Tags in the Concordance

Corpus: UKWaC super sensed Hits: 15772 (42.6 per million)			
First   Previous Page 3	of 789 Go	Next   Last	
#1493049		credibility that surrounds the <mwe><ne> Mickey Mouse /NNP/other.</ne></mwe>	n <m< th=""></m<>
#1496976		hovering , <mwe> seeking out </mwe> a vole or mouse /NN/animal	n . INFORMATION F
#1545720		audio visuals and the <mwe><ne> Soviet Spy Mouse /NNP/other.</ne></mwe>	n Trail
#1561637		even after disruption . Control through the mouse /NN/animal	n Interactive stori
#1561653		keyboard . They are controlled entirely by the mouse /NN/animal	n which moves the
#1561673		of effects . In a very real sense , the mouse /NN/animal	n represents cont
#1561693		importance of access to the controlling device ( mouse /NN/animal	n or keyboard , <n< th=""></n<>
#1561728		, 1993 ) . Seen in this light , the mouse /NN/animal	n might be conside
#1561813		different methods were used when passing the mouse /NN/animal	n to the <ne> next</ne>
#1561826		Sometimes , <mwe> for example </mwe> , when the mouse /NN/animal	n had been left in
#1561862		remaining member of the group moved the mouse /NN/animal	n towards the nex
#1561928		within the group . If simply leaving the mouse /NN/animal	n is seen as unhel
#1562003		members who were not in possession of the mouse /NN/animal	n issued a significa
#1562048		in <mwe> actual possession </mwe> . While the mouse /NN/animal	n gave the undisp
#1562081		directed the same commands to the holder of the mouse /NN/animal	n , adding extra p
#1608486		and dragging the control points using the mouse /NN/animal	.n (left button) or
#1608563		dialog or dragging the points using left <mwe> mouse /NN/artifact</mwe>	.n button
#1664217		with a cenral finger-ball (e.g. marble mouse /NN/animal	n ). Just more ph
#1768071		students attending workshops ( eg. <mwe> mouse /NN/artifact</mwe>	.n mats  , p
#1831217		include the <mwe><ne> Yellow Necked Mouse /NNP/other.</ne></mwe>	n ,
First   Previous Page 3	of 789 Go	Next   Last	

#### Semantic Tags in the Word Sketch (selected)

#### eat (verb) UKWaC super sensed freq = 26329 (71.2 per million)

transframe	<u>1241</u>	6.6	<u>intransframe</u>	<u>1021</u>	2.7
person.n_*consumption.v_food.n	178	11.42	animal.n_*consumption.v	<u>85</u>	10.84
group.n_*consumption.v_food.n	<u>57</u>	10.25	person.n_*consumption.v	382	10.25
person.n_*consumption.v_plant.n	<u>37</u>	9.77	group.n_*consumption.v	<u>145</u>	9.82
person.n_*consumption.v_animal.n	35	9.62	0_*consumption.v	<u>61</u>	9.73
animal.n_*consumption.v_animal.n	<u>30</u>	9.56	state.n_*consumption.v	<u>20</u>	8.89
animal.n_*consumption.v_plant.n	25	9.32	time.n_*consumption.v	<u>19</u>	8.74
animal.n_*consumption.v_food.n	<u>24</u>	9.21	communication.n_*consumption.v	<u>30</u>	8.57
person.n_*consumption.v_person.n	52	8.97	artifact.n_*consumption.v	30	8.51
0_*consumption.v_food.n	20	8.92	food.n_*consumption.v	<u>19</u>	8.4
animal.n_*consumption.v_artifact.n	19	8.79	other.n_*consumption.v	15	8.1

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#### Semantic Tags in the Word Sketch (selected)

### laugh (verb) UKWaC super sensed freq = 6489 (17.5 per million)

V_PP	<u>148</u>	9.1	<u>intransframe</u>	<u>1101</u>	8.9
*body.v_at_cognition.n	14	11.02	person.n_*body.v	556	10.49
*body.v_at_communication.n	<u>12</u>	10.94	group.n_*body.v	<u>143</u>	10.02
*body.v_at_person.n	<u>7</u>	9.77	0_*body.v	<u>102</u>	10.33
*body.v_as_person.n	<u>6</u>	9.78	artifact.n_*body.v	<u>49</u>	9.19
*communication.v_at_location.n	<u>6</u>	7.8	time.n_*body.v	23	8.49
*body.v_in_cognition.n	5	9.58	location.n_*body.v	<u>16</u>	7.88
*communication.v_at_communication.n	5	8.39	event.n_*body.v	<u>9</u>	7.56
*communication.v_at_person.n	5	8.14	other.n_*body.v	<u>8</u>	7.38
-			cognition.n_*body.v	<u>7</u>	7.2
			communication.n_*body.v	7	6.89

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In the Concordance Semantic Word Sketches Other Possibilities from SST Output

#### Semantic Word Sketch Grammar

An example for the intransitive frame

=intransframe \*COLLOC "%(2.sense)\_\*%(1.sense)-x"

2:any\_noun rel\_start? adv\_aux\_string\_incl\_be 1:verb\_not\_pp not\_np\_start

2:any\_noun rel\_start? adv\_aux\_string\_incl\_be aux\_have adv\_string 1:past\_part not\_np\_start

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#### MWEs: detected by ${\rm SST}$

## bread (noun) UKWaC super sensed freq = 8355 (22.6 per million)

mwe	1887	0.8
bread_and_butter_possession.n	203	11.37
loaf_of_bread_food.n	<u>180</u>	11.25
white_bread_food.n	171	11.3
garlic_bread_food.n	<u>74</u>	10.22
brown_bread_food.n	71	10.16
french_bread_food.n	53	9.77
unleavened_bread_food.n	42	9.45
bread_street_location.n	34	9.15
bread_maker_person.n	31	9.03
rye_bread_food.n	30	8.98
bread_knife_artifact.n	<u>21</u>	8.48

### MWEs: Sketch Diff chip (green) vs chips (red)

mwe	325	314	-2.0	-2.0
fish_and_chip_food.n	0	<u>70</u>	0.0	10.4
tortilla_chip_food.n	0	<u>14</u>	0.0	8.3
potato_chip_food.n	<u>10</u>	<u>40</u>	7.7	9.7
memory_chip_artifact.n	25	32	9.0	9.4
gene_chip_artifact.n	<u>11</u>	<u>10</u>	7.9	7.8
poker_chip_artifact.n	<u>14</u>	12	8.3	8.0
silicon_chip_artifact.n	<u>40</u>	27	9.7	9.1
bargaining_chip_attribute.n	<u>29</u>	Z	9.3	7.2
chocolate_chip_food.n	14	0	8.3	0.0
chip_shot_act.n	15	0	8.4	0.0

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#### Portion of Sketch Diff laugh (green) vs cry (red)

V_PP	148	117	9.1	7.0
*communication.v_in_location.n	0	<u>6</u>	0.0	7.3
*stative.v_for_act.n	0	<u>11</u>	0.0	6.6
*communication.v_at_location.n	<u>6</u>	0	7.8	0.0
*communication.v_at_person.n	<u>5</u>	0	8.1	0.0
*communication.v_at_communication.n	<u>5</u>	0	8.4	0.0
*body.v_in_cognition.n	<u>5</u>	0	9.6	0.0
*body.v_at_person.n	Z	0	9.8	0.0
*body.v_as_person.n	<u>6</u>	0	9.8	0.0
*body.v_at_communication.n	<u>12</u>	0	10.9	0.0
*body.v_at_cognition.n	<u>14</u>	0	11.0	0.0

# Semantic Word Lists: CQL + Word Frequency (Communication Verbs)

Frequency limit	0 Set limit						
Pres []							
Page 1	GO Next >						
lemma	Freq						
p/n sav	492025						
p/n ask	141001						
p/n tell	131081						
p/n call	118589						
p/n ensure	105345						
p/n agree	69722						
p/n suggest	67352						
p/n contact	67261						
n describe	66546						
p/n out	65818						
p/n report	65657						
p/n mean	62248						
p/n discuss	58197						
p/n talk	57588						
p/n post	57088						
p/n explain	56597						
p/n speak	54133						
p/n introduce	49905						
p/n write	45315						
p/n teach	44558						
p/n record	42960						
p/n address	41014						
p/n find	40432						
p/n recommend	39143						
p/n claim	38448						
n indicate	36305						
p/n announce	35623						
p/n determine	33633						
p/n state	32990						
p/n allow	32909						
p/n present	31577						
p/n mention	31167						
p/n advise	29322						
	20040	4 E b	4 A 1	A 3 b	A 3 b	-	

#### Semantic Word Lists: FindX (communication verbs)

0.1	122.9	sayv	271	
0.1	119.4	tellv	<u>62</u>	
0.1	112.0	askv	75	
0.2	101.9	outv	53	
0.2	100.0	humour-v	53	
0.2	100.0	critique-v	142	
0.2	100.0	underline-v	2166	
0.2	100.0	stammer-v	<u>50</u>	
0.2	100.0	reintroduce-v	<u>501</u>	
0.2	100.0	re-introduce-v	<u>109</u>	
0.2	100.0	shriek-v	<u>116</u>	
0.2	100.0	exhort-v	88	
0.2	100.0	publicise-v	1244	
0.2	100.0	chide-v	116	
0.3	100.0	interrogate-v	730	
0.3	100.0	fate-v	67	
0.3	100.0	bemoan-v	277	
0.3	100.0	absolve-v	136	
0.3	100.0	signpost-v	160	
0.3	100.0	unrated-v	321	
0.3	100.0	chronicle-v	487	
0.3	100.0	telegraph-v	<u>75</u>	
0.3	100.0	<u>spam-v</u>	218	
0.3	100.0	misquote-v	<u>119</u>	
0.4	100.0	extol-v	87	
0.4	100.0	eschew-v	322	
0.4	100.0	nominates-v	71	
0.4	100.0	evince-v	156	
0.4	100.0	spoof-v	<u>66</u>	
0.4	100.0	<u>rejuvenate-v</u>	205	
0.4	100.0	symbolise-v	292	
0.4	100.0	pardon-v	445	

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#### Comparing to FrameNet (Ruppenhofer et al., 2010)

FrameNet contains lots of useful information e.g. [FRAME employing:

Frame Elements: Employer Employee Position Tasks

Compensation ...

Definition: An Employer *employs* an Employee whose Position entails that the Employee perform certain Tasks in exchange for Compensation

- lots of other information
  - lexical units employ.v commision.v staff.n employment.n
  - precedes frame firing
  - with corpus examples, *I employed him as Chief Gardener for ten years*

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- but manually produced so low coverage
- Semantic word sketches can provide additional information and high coverage

### Summary

- semantic tagging alongside part-of-speech for semantic word sketches
- provide syntactic and semantic profiling for
  - semantic queries and word lists
  - semantic and syntactic profiling in the word sketch
  - comparing words by the profiles

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#### Future Possibilities

- try other semantic tagsets, taggers and tools
- sketch grammar could be developed further
- no identification of semantic roles as yet in contrast to FrameNet (Ruppenhofer et al., 2010), Propbank (Palmer et al., 2005) and VerbNet (Kipper-Schuler, 2005)
- Semantic word sketches could be used to provide selectional preferences and corpus information to such resources

#### Thank You

Query thank, V.\* 14,920 > Positive filter (excluding KWIC) you 5,551 > Shuffle 5,551

Page 1	of 278 Go Next   Last
http://www	see , a little more really means a lot - Thank /communication.
http://www	Jamie and Lynne Reilly 29 June , 2002 `` Thank /communication.
http://www	Team , ( Jack , Billy and young Andy , ) THANK $_{/communication}$
http://www	important member of the family , and we thank /communication.
http://www	B. I would have liked the opportunity to thank /communication.
http://www	obtained some very interesting information . I thank /communication.
http://bee	again for eternity in heaven . I want to thank /communication.
http://www	to cover , but thank you for coming and thank /communication.
http://www	ZENAB () posted : 08.03.2006 message : Thank /communication.
http://www Jo	panne.gowing@pizzaexpress.com Happy donating and thank /communication.
http://www	Platform : GameCube Sent by : Andrew Bernish ( Thank /communication.
http://www	to keep Nicholas in all of your prays and thank /communication.
http://www	top Thank you June 2006 I would like to thank /communication.
http://spo	Peter MacLeod . Hello to you too , sir , and thank /communication.
http://www	, I do . You seem so damn sure . HOMER : Thank /communication.

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