SUPPLEMENTAL MATERIAL FOR "VISUALIZING EXPANDED QUERY RESULTS"

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Here, we present detailed supplemental material to the EuroVis paper "Visualizing Expanded Query Results". Each section number refers to the associated section in the paper.

8.1 TASKS

We selected 10 ambiguous topics from the TREC web tracks 2009 – 2014. For each topic, we performed the given query and presented two different sub-topics to the users. We always chose the main description, as well as one dissimilar sub-topic, which could be resolved by ConceptNet.

```
<topic number="99" type="ambiguous">
  <query>satellite</query>
  <description>
   Find background information about man-made satellites.
  </description>
  <subtopic number="1" type="inf">
   Find background information about man-made satellites.
  </subtopic>
  <subtopic number="2" type="nav">
   Find satellite maps and geographic images.
  </subtopic>
  <subtopic number="3" type="nav">
   Find providers of satellite television.
  </subtopic>
  <subtopic number="4" type="inf">
   Find information about satellite telephones.
 <subtopic number="5" type="nav">
   Find providers of satellite internet service.
 </subtopic>
  <subtopic number="6" type="nav">
   Find providers of satellite radio systems.
  </subtopic>
</topic>
```

Below is the complete list of topics and sub-topic descriptions selected for the study:

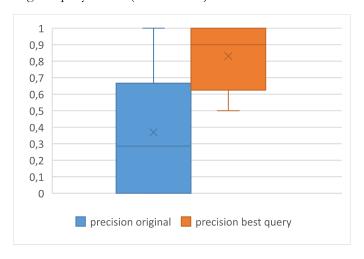
Task set	Query	Description	Precision in original query	Query with highest precision	Total number of relevant hits
1	grilling	Find recipies for grilling.*	0.67	Q8: 1.0	27

1	rock art	Find information on cave paintings all around the world.*	0.75	Q0: 0.75	20
1	iron	Find information about iron as an essential nutrient.	0.56	Q0: 0.56	8
1	Worm	Find information about worms in nature.	0.56	Q4: 0.9	40
1	Kiwi	Find information on kiwi fruit.	0.33	Q4,Q5: 1.0	32
2	Worm	Find information about computer worms, viruses, and spyware.	0.00	Q1: 1.0	17
2	Pvc	Find information about PVC pipes and fittings.	0.29	Q4: 1.0	16
2	Kiwi	Find information on kiwi birds.	0.11	Q2: 1.0	31
2	Keyboard review	Find reviews of computer keyboards.	0.88	Q1, Q5: 1.0	57
2	Joints	Find an explanation of the different types of joints used in woodworking.	0.00	Q9: 0.8	8
3	grilling	Find information on different type of barbecue grills.*	0.00	Q4: 0.5	7
3	Pvc	How are premature ventricular contractions treated?	0.14	Q6: 1.0	16
3	Joints	Find information about joints in the human body.	1.00	Q0, Q6: 1.0	32
3	Satellite	Find providers of satellite television hardware.*	0.00	Q6: 0.6	10
3	Dog heat	What is the effect of excessive heat on dogs?	0.00	Q8, Q9: 0.8	24
4	rock art	Where can I learn about rock painting or buy a rock-painting kit?	0.25	Q1,Q5: 7	21
4	Iron	Find information about the element iron (Fe).	0.44	Q5: 0.7	33
4	Keyboard review	Find reviews of electronic keyboards and digital pianos.	0.00	Q7: 0.5	6
4	Satellite	Find background information about man-made satellites.	0.67	Q1, Q2: 0.8	27
4	Dog heat	Find information on dogs' reproductive cycle. What does it mean when a dog is "in heat"?	1.00	Q0: 1.0	21

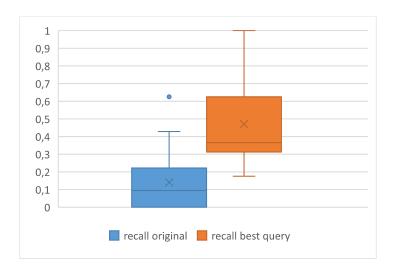
^{*} Slightly modified from original TREC description.

Mind, that the number of document surrogates for the original query is lower than 10 in our implementation, because the original query results are parsed directly from the Google results list, which is usually shorter than the list of 10 document surrogates delivered by the API (average number of document surrogates for the original query is 8.7 in our experiment).

The average precision of the *best* queries (see second to last column in table above) is 0.83 (median: 0.9). The average precision of the original query is 0.37 (median: 0.29):



The average recall of the best queries is 0.47 (median: 0.37). The average recall of the original query is 0.14 (median: 0.09):

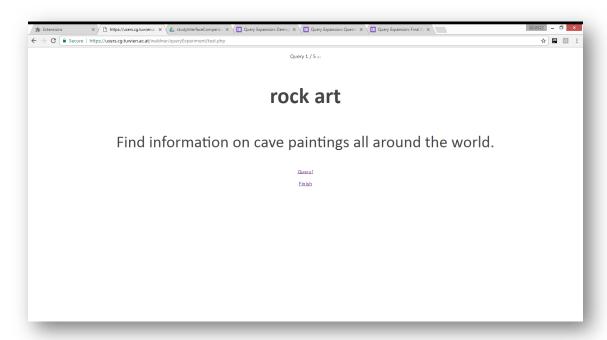


8.2 APPARATUS AND PROCEDURE

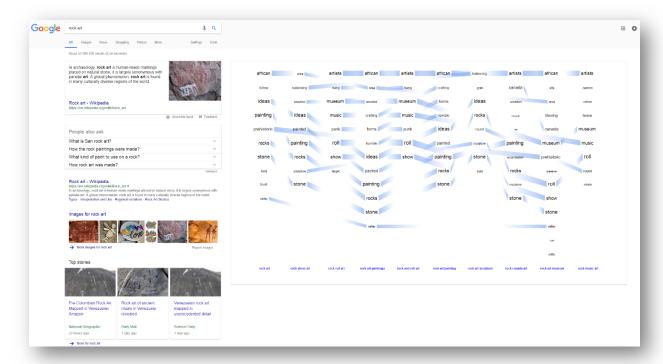
The complete procedure for the study was as follows:

- consent form
- demographic questionnaire
- task description
- for each interface:
 - o warm-up task (queries: spider [program] and jaguar [cat])
 - o task set with five sub-topics shown in random order
- questionnaire

Each task was preceded by a presentation of the query, together with its description:



After clicking the Query-button, the Google result page with the respective visualization (here Parallel Tag Clouds) was shown:



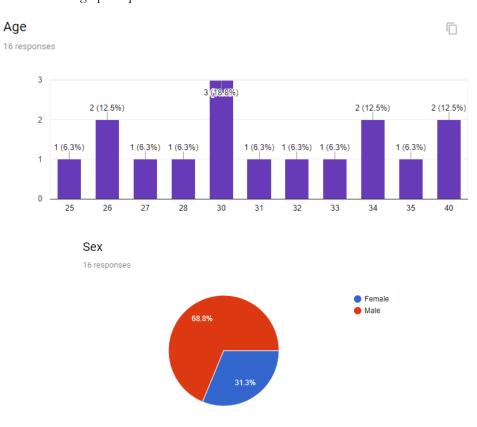
Task assignments to interfaces and presentation order of interfaces was balanced using a Graeco-Latin Square:

T = text, Q=Euler Diagram, P = Parallel Tag Clouds, L = Parallel Lists

User	Con.1	Task1	Con.2	Task2	Con.3	Task3	Con.4	Task4
1	Т	1	Q	3	P	4	L	2
2	Q	2	Т	4	L	3	P	1
3	P	3	L	1	Т	2	Q	4
4	L	4	P	2	Q	1	Т	3
5	Т	1	Q	3	P	4	L	2
6	Q	2	Т	4	L	3	P	1
7	P	3	L	1	Т	2	Q	4
8	L	4	P	2	Q	1	Т	3
9	Т	1	Q	3	P	4	L	2
10	Q	2	Т	4	L	3	P	1
11	P	3	L	1	Т	2	Q	4
12	L	4	P	2	Q	1	Т	3
13	Т	1	Q	3	P	4	L	2
14	Q	2	Т	4	L	3	P	1
15	P	3	L	1	Т	2	Q	4
16	L	4	P	2	Q	1	Т	3

8.4 PARTICIPANTS

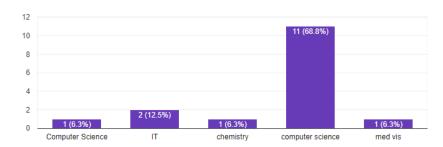
Responses from the demographic questionnaire:



Eye Sight 16 responses normal corrected to normal not corrected

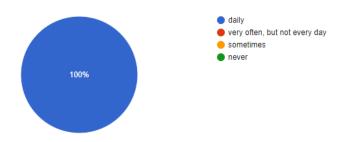
Domain (e.g., computer science, biology)

16 responses



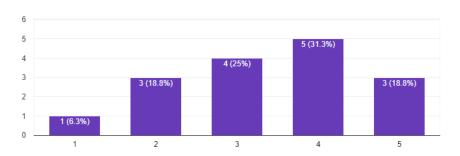
How often do you have to use online search engines (e.g., Google or Bing) for your work?

16 responses



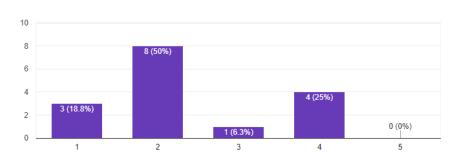
How often do you auto-complete a query suggested by the search engine? (see example below)

16 responses



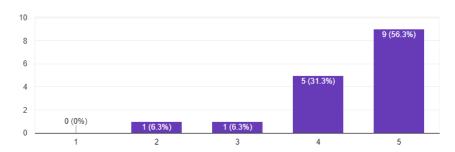
How often do you select an expanded query suggested by the search engine? (see example below)

16 responses



How familiar are you with visualization techniques (e.g., bar charts or pie charts)?

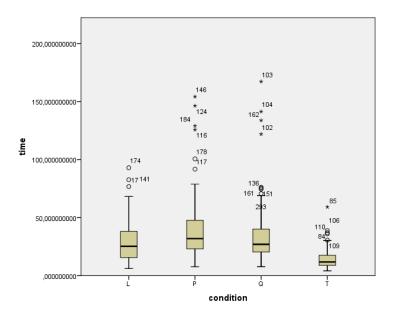
16 responses



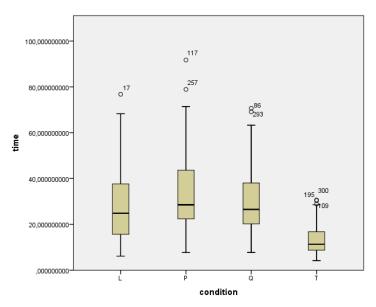
8.5 Analysis and Results

16 users conducted 5 tasks with 4 different interfaces, each, resulting in 320 samples for task completion time (TCT) and number of hits.

We first analyzed TCT and the number of hits for outliers. There were no outliers for number of hits, but we removed 18 samples (from 6 different users), because they were outliers in terms of TCT. Below, there is a box plot before outlier removal:



The same box plot after removing outliers:



For the remaining 302, we computed precision (number of relevant hits in selected query divided by number of document surrogates in the selected query) and recall (number of relevant hits in the selected query divided by the overall number of relevant hits in all queries). We aggregated the 302 per user and condition, resulting in 64 average TCT, precision, and recall values.

For precision, recall, and TCT, we conducted a repeated measures ANOVA with condition as within-subjects factor. In case of significance, we performed pairwise Bonferroni-corrected post-hoc comparisons between the 4 conditions. For the user ratings, we performed a Friedman test with Bonferroni-corrected Wilcoxon Signed-Rank post-hoc comparisons.

Task Completion Time

There is a significant difference of TCT between the conditions:

Tests of Within-Subjects Effects

Measure: TCT

Source		Type III Sum of Squares	df	Mean Square	F	Sig.	Partial Eta Squared
condition	Sphericity Assumed	4098,352	3	1366,117	34,923	,000	,700
	Greenhouse-Geisser	4098,352	2,323	1764,284	34,923	,000	,700
	Huynh-Feldt	4098,352	2,774	1477,368	34,923	,000	,700
	Lower-bound	4098,352	1,000	4098,352	34,923	,000	,700
Error(conditio	Sphericity Assumed	1760,327	45	39,118			
n)	Greenhouse-Geisser	1760,327	34,844	50,520			
	Huynh-Feldt	1760,327	41,611	42,304			
	Lower-bound	1760,327	15,000	117,355			

Condition 4 (text) had a significantly lower TCT than all other conditions:

Pairwise Comparisons

Measure: TCT

Measure. 101						
		Mean			95% Confidence Int	erval for Difference ^b
	(J)conditi	Difference (I-				
(I)condition	on	J)	Std. Error	Sig. ^b	Lower Bound	Upper Bound
1	2	-5,618	2,604	,286	-13,526	2,290
	3	-,926	1,781	1,000	-6,334	4,483
	4	15,633 [*]	2,639	,000	7,621	23,644
2	1	5,618	2,604	,286	-2,290	13,526
	3	4,692	2,280	,344	-2,230	11,614
	4	21,251*	2,128	,000	14,791	27,711
3	1	,926	1,781	1,000	-4,483	6,334
	2	-4,692	2,280	,344	-11,614	2,230
	4	16,558 [*]	1,642	,000	11,573	21,544
4	1	-15,633 [*]	2,639	,000	-23,644	-7,621
	2	-21,251 [*]	2,128	,000	-27,711	-14,791
	3	-16,558*	1,642	,000	-21,544	-11,573

Based on estimated marginal means

 $^{^{\}star}.$ The mean difference is significant at ,05 level.

b. Adjustment for multiple comparisons: Bonferroni.

Precision

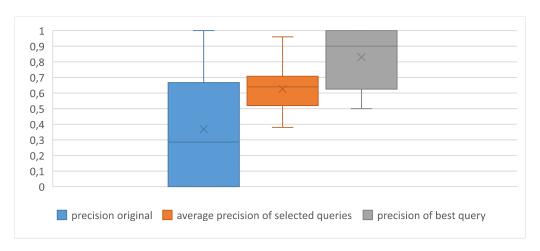
There is no significant difference in precision between the conditions:

Tests of Within-Subjects Effects

Measure: precision

Source		Type III Sum of Squares	df	Mean Square	F	Sig.	Partial Eta- Squared
condition	Sphericity Assumed	,109	3	,036	1,992	,129	,117
	Greenhouse- Geisser	,109	2,297	,048	1,992	,146	,117
	Huynh-Feldt	,109	2,735	,040	1,992	,135	,117
	Lower Bound	,109	1,000	,109	1,992	,179	,117
Error(condition)	Sphericity Assumed	,824	45	,018			
	Greenhouse- Geisser	,824	34,451	,024		ı	
	Huynh-Feldt	,824	41,031	,020			
	Lower Bound	,824	15,000	,055			

In addition, we also compared the precision between the *original* query (i.e., the precision without performing any query expansion), the *best* query (i.e., the optimal solution of all sub-topics), and the aggregated precision per task set and visualization of the users' *selected* queries from all conditions in the visualization.



We compared the precision of selected queries to the original query and the best query using a Kruskal-Wallis H test. The difference is significant:

Test Statistics^{a,b}

	precision
Chi-square	25,404
df	2
Asymp. Sig.	,000

a. Kruskal-Wallis-Test

b. Grouping Variable: query

We therefore compared the precision of the selected query to the original and the best query using Mann-Whitney U tests.

There is a significant difference between the selected query and the original query:

Test Statistics^a

	precision
Mann-Whitney-U	344,000
Wilcoxon-W	554,000
Z	-3,111
Asymp. Sig. (2-sided)	,002

a. Grouping Variable: query

There is also a significant difference between the selected query and the best query:

Test Statistics ^a	
------------------------------	--

precision

Mann-Whitney-U	263,000
Wilcoxon-W	2343,000
Z	-3,964
Asymp. Sig. (2-sided)	,000

a. Grouping Variable: query

This means that users could improve the precision of the retrieved documents by expanding the query. However, the selection of query expansions was not optimal.

Recall

There is also no difference between the conditions for recall:

Tests of Within-Subjects Effects

Measure: recall

Measure: recall							
Source		Type III Sum of Squares	df	Mean Square	F	Sig.	Partial Eta-Squared
condition	Sphericity Assumed	,025	3	,008	,919	,439	,058
	Greenhouse- Geisser	,025	1,941	,013	,919	,439	,058
	Huynh-Feldt	,025	2,225	,011	,919	,439	,058
	Lower Bound	,025	1,000	,025	,919	,439	,058
Error(condition)	Sphericity Assumed	,407	45	,009			
	Greenhouse- Geisser	,407	29,113	,014			
	Huynh-Feldt	,407	33,371	,012			
	Lower Bound	,407	15,000	,027			

User Ratings

User ratings were not normally distributed. We therefore compared the ratings using a Friedman test. The Friedman test was significant:

Statistics for Testa

N	16
Chi-Square	15,396
df	3
Asymp. Sig.	,002

a. Friedman-Test

We therefore conducted pairwise Wilcoxon Signed-Rank post-hoc comparisons with a Bonferroni-adjusted α of 0.0083. The results show that PTC was rated significantly lower than text and lists.

Statistics for Test^a

	euler - text	PTC - text	lists - text	PTC - euler	lists - euler	lists - PTC
Z	-,537 ^b	-2,829 ^b	,000°	-2,300 ^b	-,476 ^d	-3,119 ^d
Asymp. Sig. (2-sided)	,591	,005	1,000	,021	,634	,002

- a. Wilcoxon-Test
- b. Based on positive ranks.
- c. The sum of negative ranks equals the sum of positive ranks.
- d. Based on negative ranks.

User Feedback

Here, we list all user comments given for the final questionnaire:

Text list (text)

- easy to scan quickly, look for specifying keywords related to original query
- Fast overview, but not very detailed
- clean and linear alignment. quick overview based on the words. the sense is made by own experience and association of words
- simple but with less content
- i m quite used to the current way of queries in google, but i disliked that i couldnt actually see what each quey meant (i was still not sure what i should select)
- + Easy to understand not much additional information
- simple for fast seeing, bad for choices
- fast to read
- fast to find query, very used to it
- it's fast
- very easy to understand, not cluttered, visualization not appealing

Compact Euler Diagram (euler)

- hard to read
- Much too complex

- spatial alignment of blocks allows me to mentally associate the blocks with a position in space. I liked the clustering/bundelling of keywords together into blocks, this makes it easy to mentally divide between the meaning of words inside the blocks.
- straight forward but need to study how to use it
- i really liked that the most important terms in the queries were very easy to see at a glance, and with the links between the different queries, i could choose very easily what was the best result for me
- + good representation of query expansion slow / optically not that great
- good: interaction within the keywords, con:need a bit of time to understand how the representation works
- blocks provide a good overview, takes more time to understand
- a little bit crowded
- also fast finding
- best overview of connected terms
- visualization is nice, but confusing

Parallel Tag Clouds (PTC)

- hard to read
- hard to make connections
- initially, I was not aware that there are pillars associated with the expanded queries. the association is not quite easily visible, however highlighting helps. i did not know what the size of words is encoding. also the gradient confused me.
- all keywords are listed but hard to find its relationship
- i liked the fact that i could immediately see the most important keywords and based on the connections between the different queries i could choose the 'final' result
- - not very intuitive, requires a lot of space on screen
- nice idea too chaotic
- lot of redundant information
- a little bit confusing
- too many words to read, size of text distracts if the searched term is not very large.
- didn't get a good overview
- good overview, easy to select and understand, connections are interesting

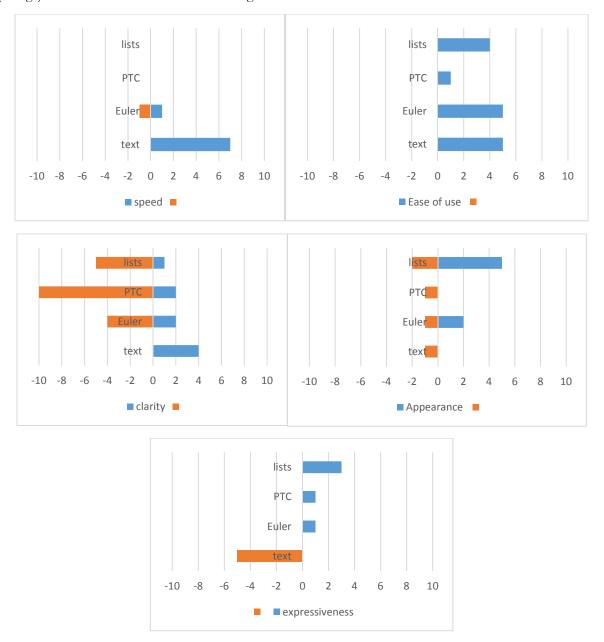
Lists View (lists)

- nice to see what keywords the different suggested queries had in common
- appealing view, easier to make connections
- i liked the left view, since it allows me to see an aggregated query and the optional expansion with the right view where more keywords can be seen. the connection are a bit hard to see without highlighting, so it needs interaction to expand. also, i am not sure how scalable this is, when there is a lot of different suggestions (however, this is true for most other visualizations)
- more text about the keywords but difficult to read when more items are associated
- i think that this is a good way of having an overview of the most important queries and what they mean, but i had to go through all of them to decide which one was the best match
- + very good representation of query expansion, easy to understand and handle takes a lot of space on screen
- easy to see and follow
- I like it

- sometimes hard to identify which words on the right side belong together
- right column is rather confusing
- nice visualization, easy to understand and select, interesting info about overlaps

Categorized Utterances

Below, we list the categories revealed during open coding, and the number of positive (blue) and negative (orange) utterances associated with these categories for each condition:

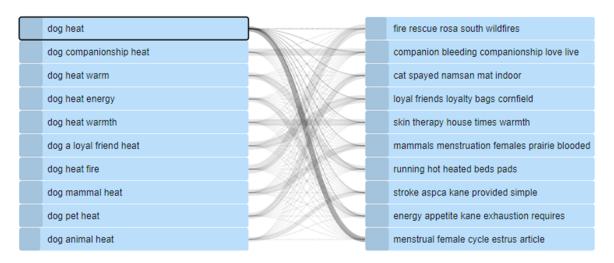


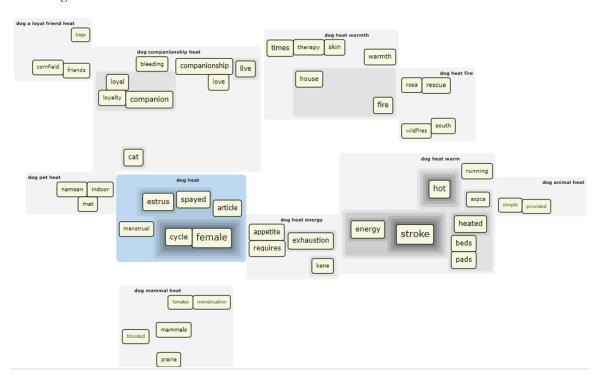
8 STIMULI

Here, all stimuli for each topic and interface condition from the user study are listed:

dog heat

Parallel List





List of alternate Queries for dog heat

dog companionship heat

dog heat warm

dog heat energy

dog heat warmth

dog a loyal friend heat

dog heat fire

dog mammal heat

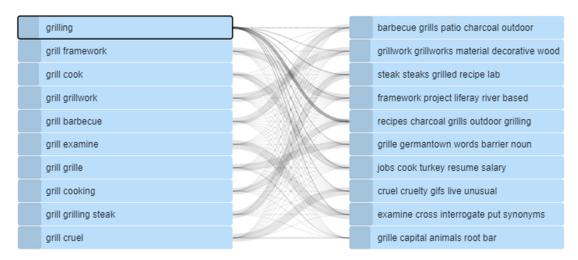
dog pet heat

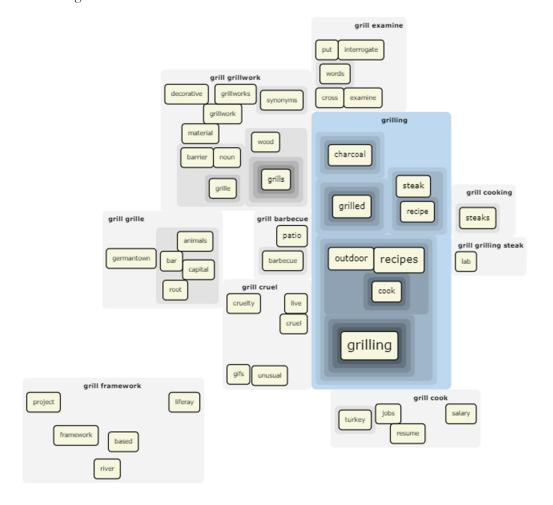
dog animal heat



Grilling

Parallel List





List of alternate Queries for grilling

grill framework

grill cook

grill grillwork

grill barbecue

grill examine

grill grille

grill cooking

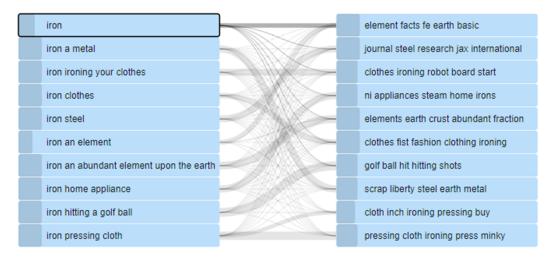
grill grilling steak

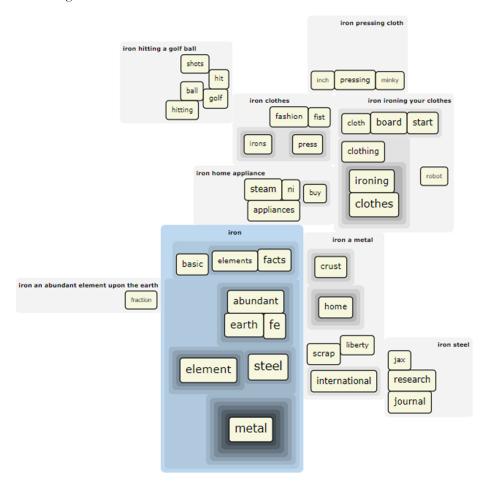
grill cruel



Iron

Parallel List





List of alternate Queries for iron

iron a metal

iron ironing your clothes

iron clothes

iron steel

iron an element

iron an abundant element upon the earth

iron home appliance

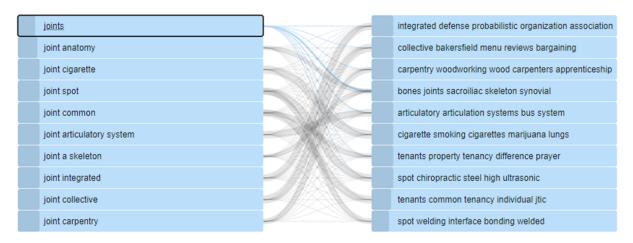
iron hitting a golf ball

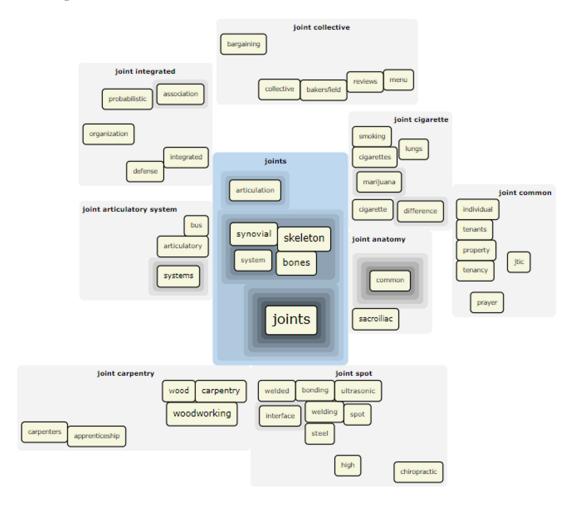
iron pressing cloth

abundant	abundant	board	clothes	element	abundant	abundant	appliances	ball	board
basic	crust	cloth	clothing	international	basic	crust	buy	golf	buy
earth	earth	clothes	fashion	jax	crust	earth	clothes	hit	cloth
element	element	clothing	fist	journal	earth	element	home	hitting	clothes
elements	fe	ironing	home	metal	element	elements	ironing	irons	home
facts	home	metal	ironing	research	elements	facts	irons	shots	inch
fe	international	robot	irons	steel	facts	fe	metal		ironing
metal	liberty	start	metal		fe	fraction	ni		minky
steel	metal		press		metal	metal	press		press
	scrap					steel	steam		pressing
	steel								start
iron	iron a metal	iron ironing your clothes	iron clothes	iron steel	iron an elem ent n a	n abundant element upon	the exact thome appliance	iron hitting a golf ball	iron pressing clot

Joints

Parallel List





List of alternate Queries for joints

joint anatomy

joint cigarette

joint spot

joint common

joint articulatory system

joint a skeleton

joint integrated

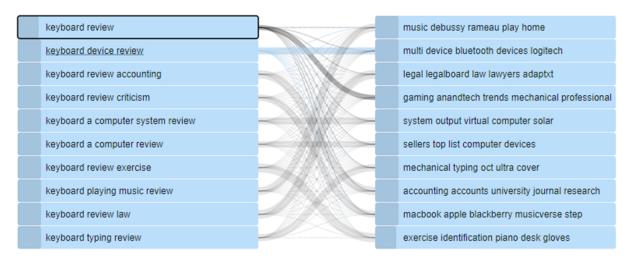
joint collective

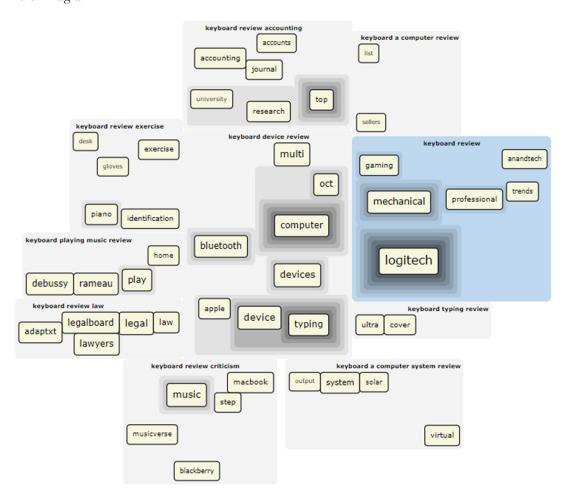
joint carpentry



keyboard review

Parallel List





List of alternate Queries for keyboard review

keyboard device review

keyboard review accounting

keyboard review criticism

keyboard a computer system review

keyboard a computer review

keyboard review exercise

keyboard playing music review

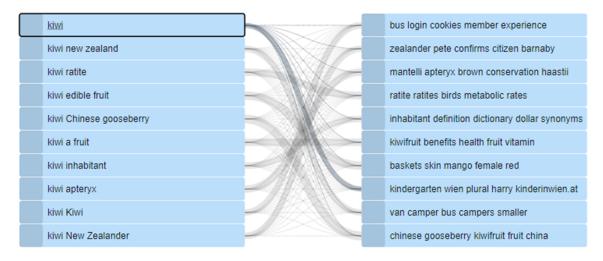
keyboard review law

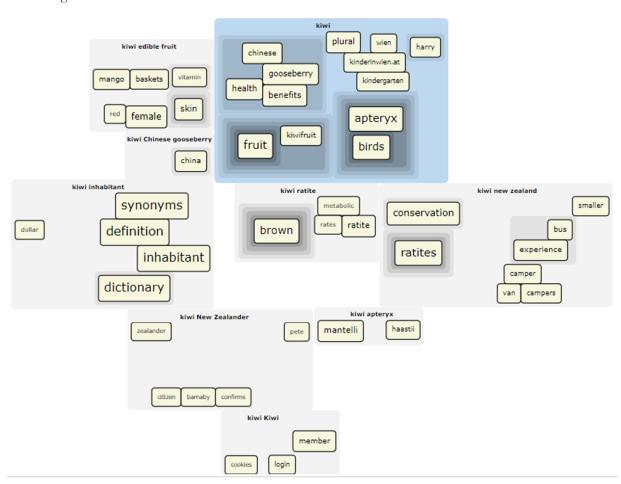
keyboard typing review



Kiwi

Parallel List





List of alternate Queries for kiwi

kiwi new zealand

kiwi ratite

kiwi edible fruit

kiwi Chinese gooseberry

kiwi a fruit

kiwi inhabitant

kiwi apteryx

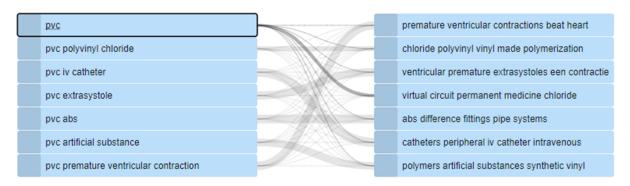
kiwi Kiwi

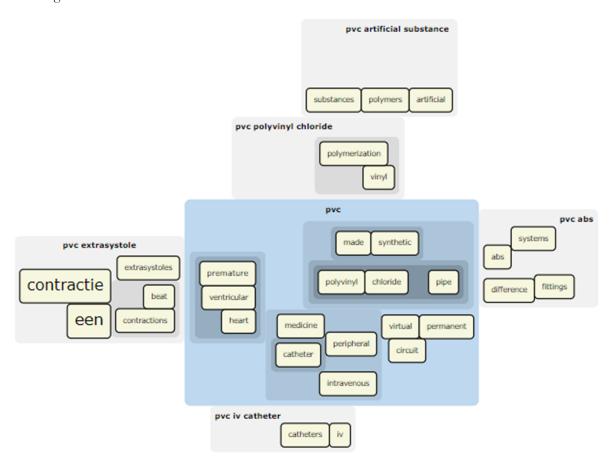
kiwi New Zealander



pvc

Parallel List





List of alternate Queries for pvc

pvc polyvinyl chloride

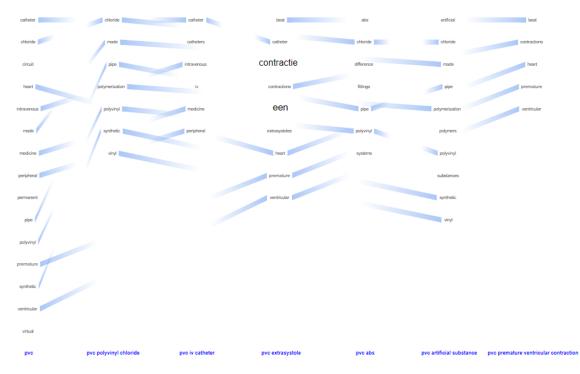
pvc iv catheter

pvc extrasystole

pvc abs

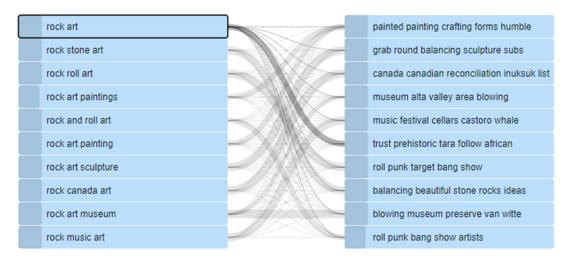
pvc artificial substance

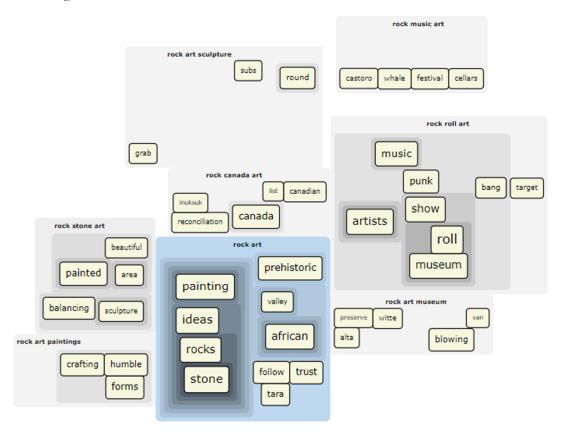
pvc premature ventricular contraction



rock art

Parallel List





List of alternate Queries for rock art

rock stone art

rock roll art

rock art paintings

rock and roll art

rock art painting

rock art sculpture

rock canada art

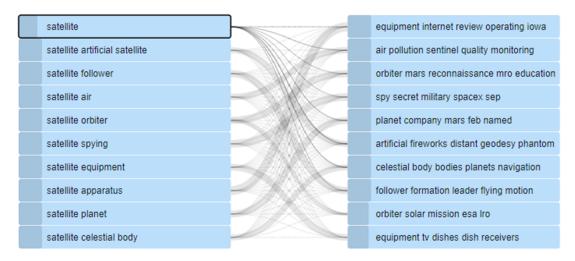
rock art museum

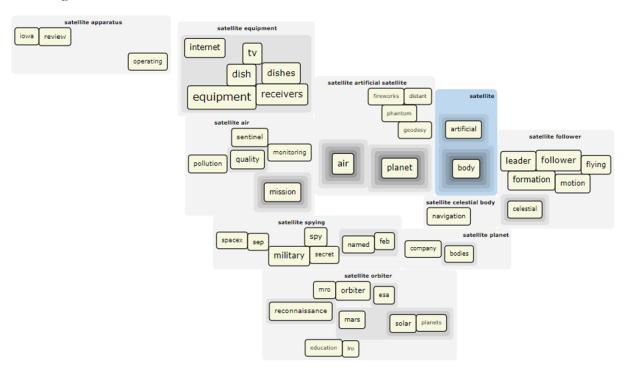
rock music art

african	area	artists	african	artists	african	balancing	artists	african	artists
follow	balancing	bang	area	bang	crafting	grab	canada	alta	castoro
ideas	beautiful	museum	beautiful	museum	forms	ideas	canadian	area	cellars
painting	ideas	music	crafting	music	humble	rocks	inuksuk	blowing	festival
prehistoric	painted	punk	forms	punk	ideas	round	list	canada	museum
rocks	painting	roll	humble	roll	painted	sculpture	painting	museum	music
stone	rocks	show	ideas	show	painting	stone	reconciliation	prehistoric	roll
tara	sculpture	target	painted		rocks	subs	rocks	preserve	round
trust	stone		painting		stone		sculpture	roll	whale
valley			rocks				stone	show	
			stone					stone	
			valley					valley	
								van	
								witte	
rock art	rock stone art	rock roll art	rock art paintings	rock and roll art	rock art painting	rock art sculpture	rock canada art	rock art museum	rock music art

Satellite

Parallel List





List of alternate Queries for satellite

satellite artificial satellite

satellite follower

satellite air

satellite orbiter

satellite spying

satellite equipment

satellite apparatus

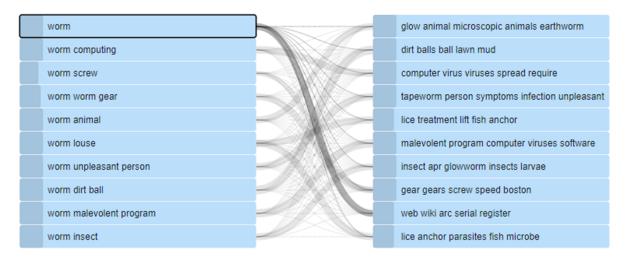
satellite planet

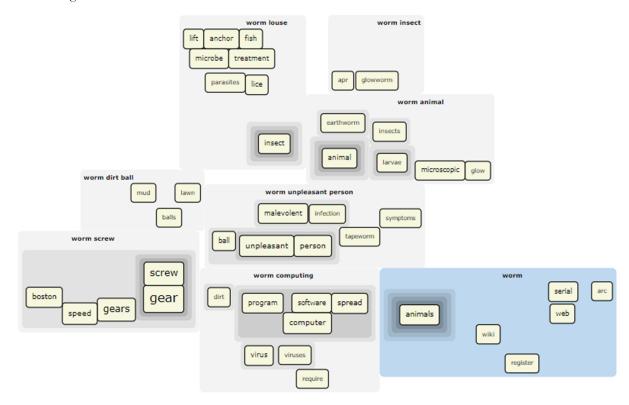
satellite celestial body



Worm

Parallel List





List of alternate Queries for worm

worm computing

worm screw

worm worm gear

worm animal

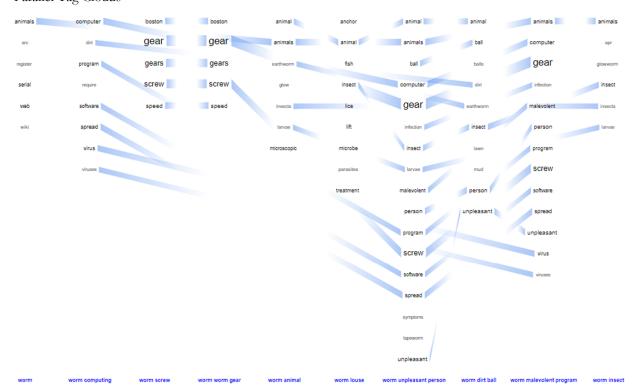
worm louse

worm unpleasant person

worm dirt ball

worm malevolent program

worm insect



7 PERFORMANCE TESTS

We performed a benchmark test of the three visualizations with five different queries. Currently, the natural language processing step is the biggest bottleneck in our pipeline. Also, the layout of ComED consumes considerable processing time. We therefore pre-computed all visualizations for the user study, so that the rendering step was the only limiting step.

The table below shows the average computation times (in milliseconds) for n=9 expansions of five queries (kiwi, worm, iron, grilling, and rock art) using k=10 topics and m=5 key terms per topic.

Pipeline step	Time (ms)
Expansion term retrieval from ConceptNet	500
Retrieval of document surrogates from Google Custom Search	1120
Part-of-Speech tagging and stop word removal	3280
Topic modeling (20 - 1000 NMF iterations, convergence value 0.001)	80-3530
Computation of visualization data structures	20
ComED layout computation (background script)	3040
Rendering of PTC and List View	20
Rendering of ComED	1700