SERPACT

EUROPEAN 2019 SEARCHAWARDS









SERP SEO & Semantic Content Optimization Hand-in-Hand

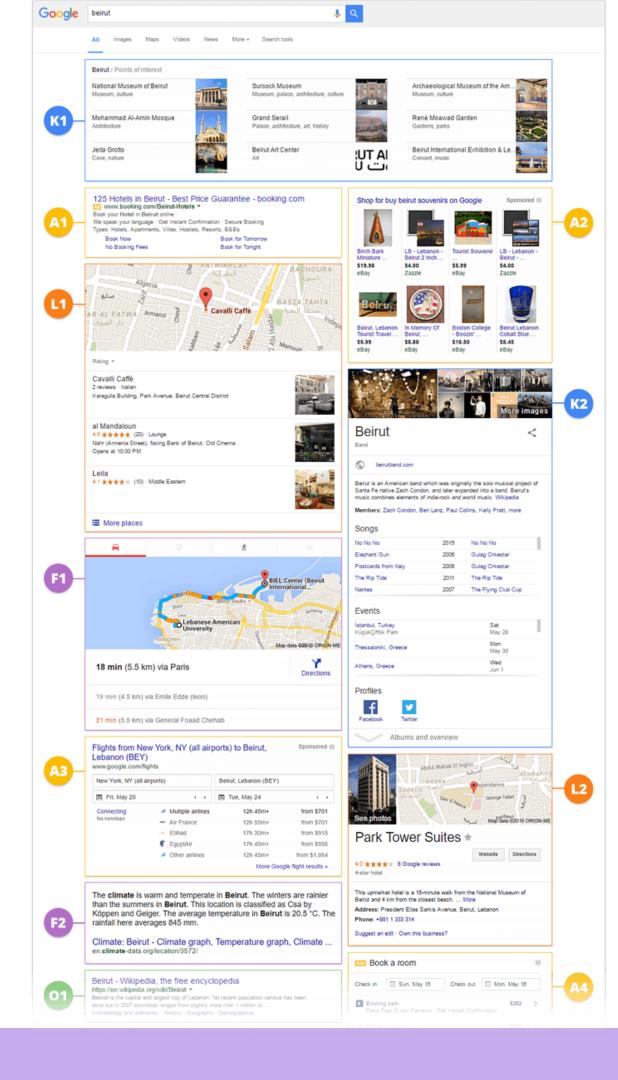
Prepared by Dido Grigorov, Head of SEO



Today's Discussion

Outline of Topics

- What is the new SERP SEO?
- SERP SEO Elements
- Why they are important?
- How to get there? Semantic SEO
- What is Semantic Search
- Semantic SEO begins with Semantic Keyword Research
- Steps of doing semantic keywords
- What is Bag of Words?
- What are ngrams?
- What is Part of Speech?
- What are Named Entities?
- Core keywords, Stemming keywords, Supporting keywords
- Word Associations
- Topic Tagging
- Text Classification
- Text Similarity
- Language Modeling and Semantic Keyword Research
- TF-IDF
- Calculating Co-Occurence
- Topic Modeling
- Content optimization on page tips and tricks
- Content optimization text tips and tricks
- Content optimization and Google Search Quality Rating Guidelines



What is the new SERP SEO?

So how can SEO professionals stay effective when the SERPs are becoming more competitive?

- Search engine result pages (SERPs) are more competitive than ever.
- We're starting to see many different types of media elements ranking.
- The decline of organic traffic for top performing keywords is rooted in the presence of new elements on Page 1.
- The organic Position 1 is now often bumped low behind ads, maps, media elements, and featured snippets.
- We, therefore, call these elements "position zero" because they outrank even the top organic results.
- One of the most competitive versions of SERPs is zero-click searches.
- A zero-click SERP is one where the answer is displayed directly on the search results page satisfying the search intent of the user without having to click further.

SERP SEO Elements

What are they?

- Knowledge Graph results Carousel, Brand
- Google Ads Top / Bottom
- Local Search Results Google My Business
- Featured Snippets Zero Result
- Organic Results





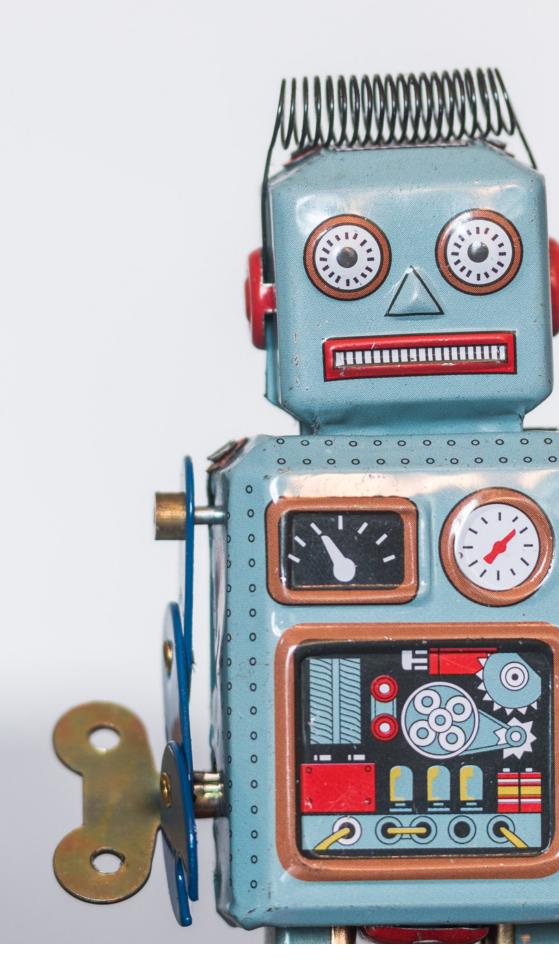
When doing your keyword research, defining your target pages and selecting target search terms, there's one important mindset shift to make to future-proof your SEO strategy:

- Search terms & Intent.
- Monthly search volumes.
- Keyword difficulty
- The presence of featured snippets
- Other media elements

A good question...

How To Get There?

ANSWER: SEMANTIC SEO



What is Semantic Search?

Semantic search is a data searching technique in a which a search query aims to not only find keywords, but to determine the intent and contextual meaning of the the words a person is using for search.

Semantic search provides more meaningful search results by evaluating and understanding the search phrase and finding the most relevant results in a website, database or any other data repository.

"dalmatian" and "dog" are semantically related.

"dalmatian" and "spotted" are more closely related than "dog" and "spotted." "dalmatian" is more frequently capitalized than other nouns. "spotted" can mean "seen" or "dotted."

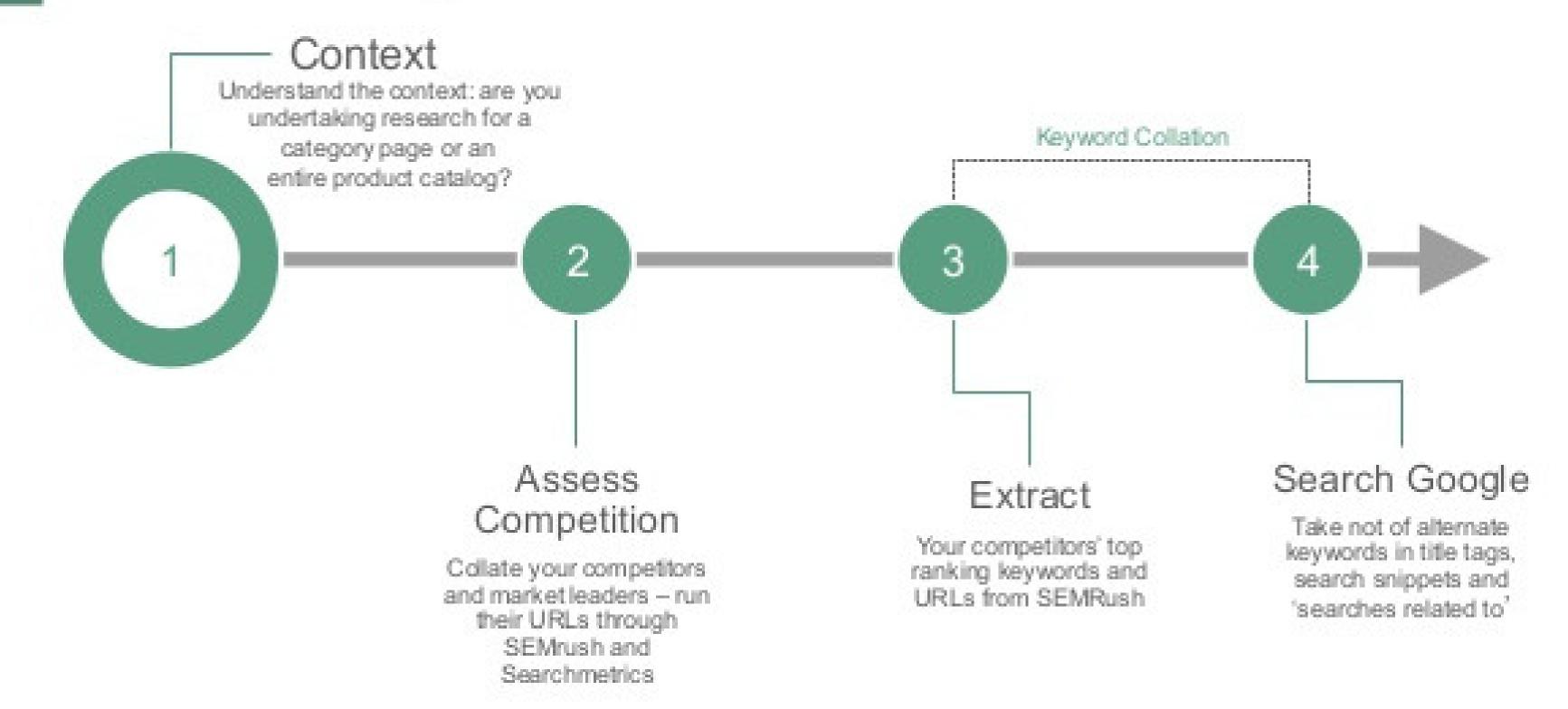
Semantic SEO begins with Semantic Keyword Research Steps of doing semantic keywords

Different types of keywords according to the:

- According to consumer expectations information, navigation, transaction, commercial
- According to their length short-tail, long-tail
- According to the brand branded or connected to the brand
- Separated by different classifiers color, size, volume, shape
- According to the trend seasons, appearing for a certain period of time (short or long term), evergreen keywords
- Product Defined
- Profile of the ideal buyer defining
- Geo-targeting keywords
- Wrong spelling words
- Semantic words related phrases, co-occurrence phrases, generic, stemming, supporting, synonyms and close Variants, entities
- Distributed to the customer journey stages
- Core keywords, Stemming keywords, Supporting keywords



Semantic Keyword Research Timeline



Semantic Keyword Research Timeline



the 'related keyword' list.

Collate competitors and market leaders - run their URLs through SEMrush

and Searchmetrics

keywords as complementary keywords to your 'core'

keyword

Semantic Keywords

Provide your copywriter with these semantically related keywords - that should be mentioned at least once in your copy in natural language targeted to shoppers.

Also use them as anchor text in your internal link strategy

Natural Languge Processing - a syntactic analysis is used to assess how the natural language aligns with the grammatical rules

- Lemmatization: It entails reducing the various inflected forms of a word into a single form for easy analysis.
- Tokenization separating sentences in words and make predictions about the meaning of the sentence
- Morphological segmentation: It involves dividing words into individual units called morphemes.
- Word segmentation: It involves dividing a large piece of continuous text into distinct units.
- Part-of-speech tagging: It involves identifying the part of speech for every word.
- Parsing: It involves undertaking grammatical analysis for the provided sentence.
- Sentence breaking: It involves placing sentence boundaries on a large piece of text.
- Stemming: It involves cutting the inflected words to their root form.

The bag-of-words model is a simplifying representation used in natural language processing and information retrieval (IR). In this model, a text (such as a sentence or a document) is represented as the bag (multiset) of its words, disregarding grammar and even word order but keeping multiplicity. The bag-of-words model has also been used for computer vision.

The bag-of-words model is commonly used in methods of document classification where the (frequency of) occurrence of each word is used as a feature for training a classifier.

An early reference to "bag of words" in a linguistic context can be found in Zellig Harris's 1954 article on Distributional Structure.[3]

What is Bag of Words?

What is Language Modelling?

representation vector language pairs LEU points

machine translation decoder multiple language model

shared factor pruning rate word forms

relative perplexity reductions

Word vectors

Word vectors

Languages Czech simple word vectors

additive representations

word similarity tasks normalised CSLM

morphological complexity

morphological complexity

respective neural-network

relative perplexity reductions

relative perplexity reductions

word vectors

real-valued vectors

test tokens rich languages

word similarity tasks normalised CSLM

morphological complexity

- A statistical language model is a probability distribution over sequences of words. Given such a sequence, say of length m, it assigns a probability to the whole sequence.
- The language model provides context to distinguish between words and phrases that sound similar. For example, in American English, the phrases "recognize speech" and "wreck a nice beach" sound similar, but mean different things.

What is language modelling - p.2?

- Estimating the relative likelihood of different phrases is useful in many natural language processing applications, especially those that generate text as an output. Language modeling is used in speech recognition, machine translation, part-of-speech tagging, parsing, Optical Character Recognition, handwriting recognition, information retrieval and other applications.
- In speech recognition, sounds are matched with word sequences. Ambiguities are easier to resolve when evidence from the language model is integrated with a pronunciation model and an acoustic model.
- Language models are used in information retrieval in the query likelihood model. There a separate language model is associated with each document in a collection. Documents are ranked based on the probability of the query Q in the document's language model. Commonly, the unigram language model is used for this purpose.

A good question

What is n-gram model?

UNIGRAM

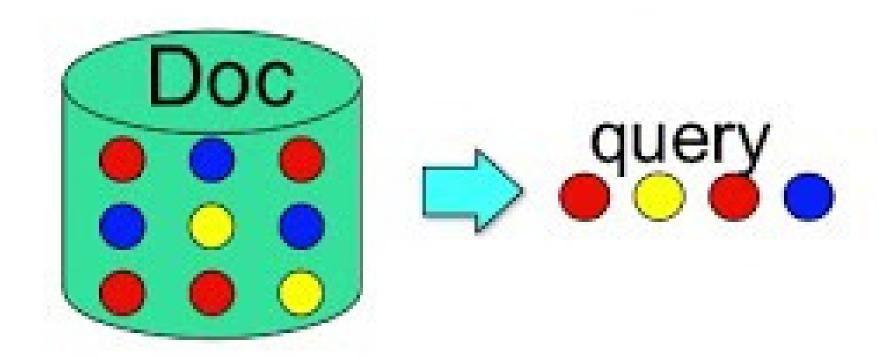
A unigram model can be treated as the combination of several one-state finite automata. It splits the probabilities of different terms in a context, e.g. from to.

In this model, the probability of each word only depends on that word's own probability in the document, so we only have one-state finite automata as units. The automaton itself has a probability distribution over the entire vocabulary of the model, summing to 1.



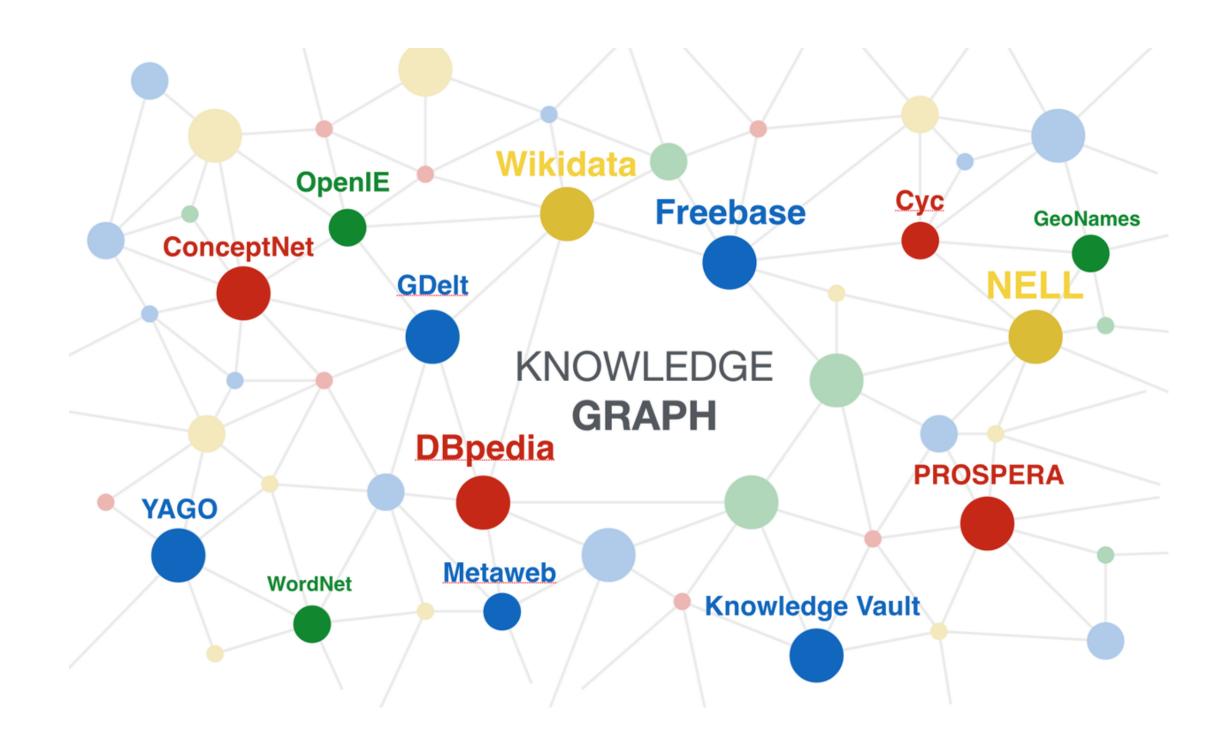
Unigram Language Models

- words are "sampled" independently of each other
 - metaphor: randomly pulling out words from an urn (w. replacement)
 - joint probability decomposes into a product of marginals
 - estimation of probabilities: simple counting



$$P(\bullet \circ \bullet \circ) = P(\bullet) P(\circ) P(\bullet) P(\bullet)$$

Knowledge Graph



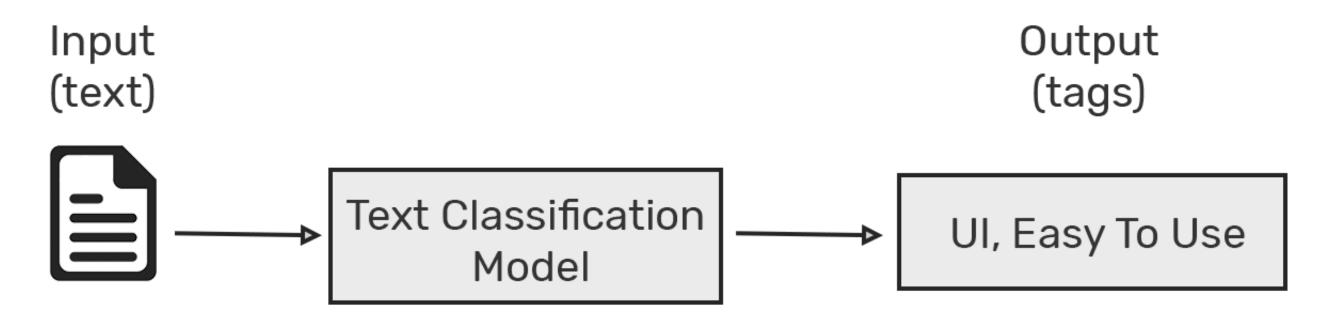
Text Classification

Text classification (a.k.a. text categorization or text tagging) is the task of assigning a set of predefined categories to free-text. Text classifiers can be used to organize, structure, and categorize pretty much anything. For example, new articles can be organized by topics, support tickets can be organized by urgency, chat conversations can be organized by language, brand mentions can be organized by sentiment, and so on.

As an example, take a look at the following text below:

"The user interface is quite straightforward and easy to use."

A classifier can take this text as an input, analyze its content, and then and automatically assign relevant tags, such as UI and Easy To Use that represent this text:



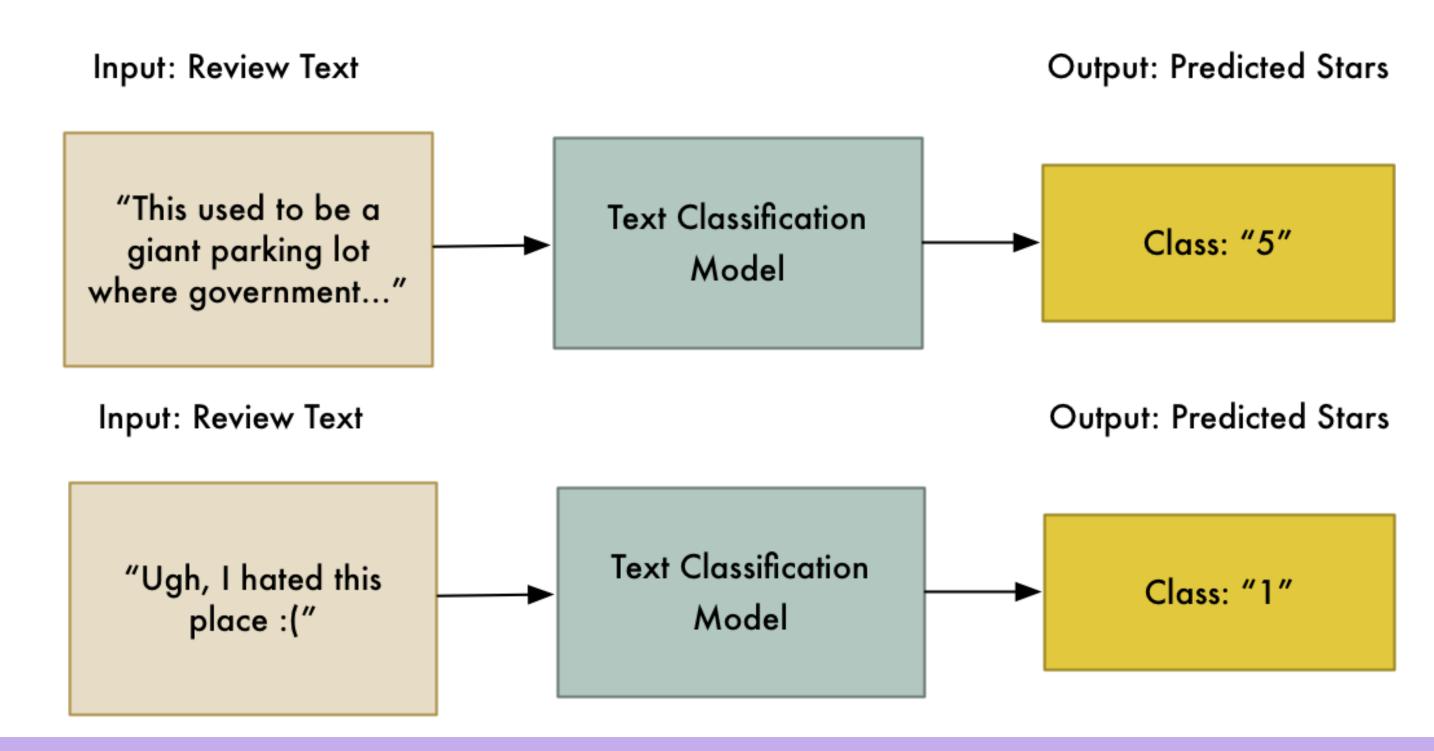
Text Classification

With text classification, the algorithm doesn't care whether the user wrote standard English, an emoji, or a reference to Goku. The algorithm is looking for statistical relationships between input phrases and outputs. If writing odcorrelates more heavily with 1-star and 2-star reviews, the algorithm will pick that up even though it has no idea what a "look of disapproval" emotion is.

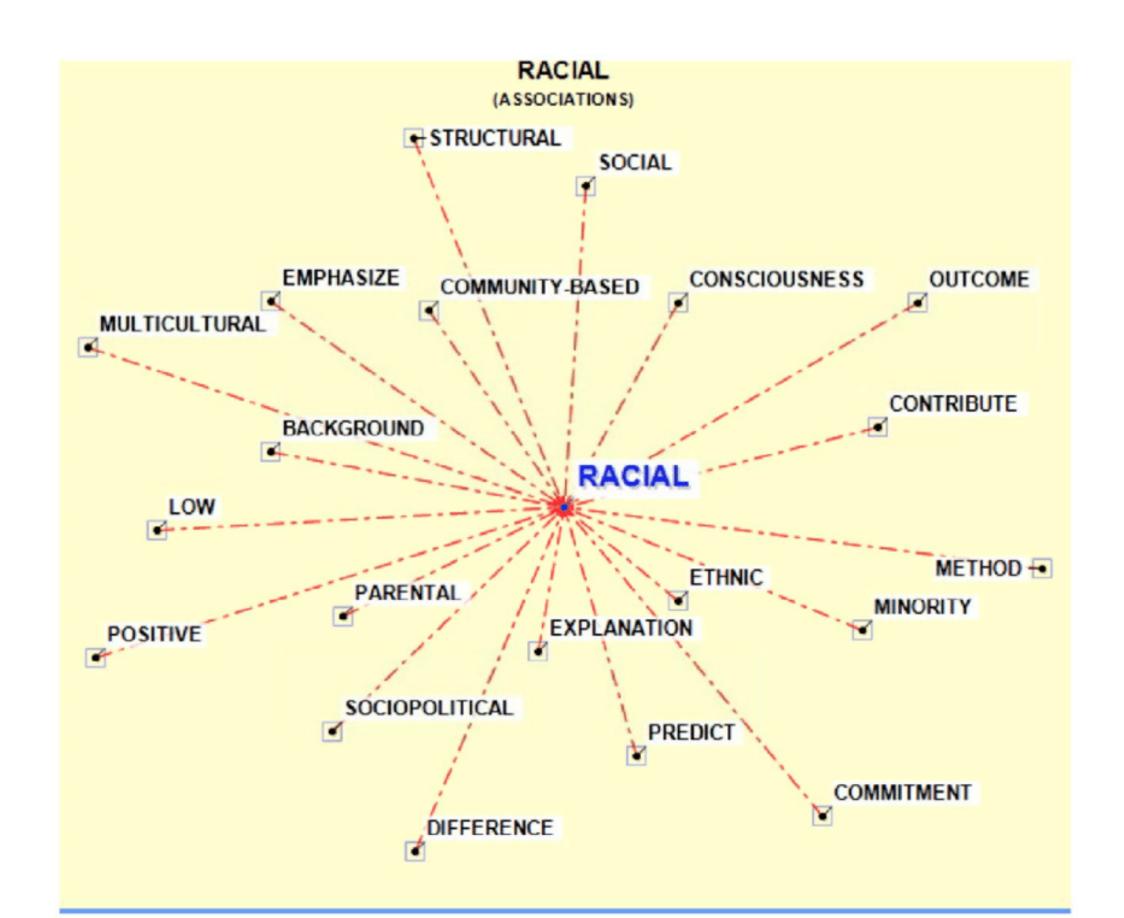
Word		(One hun	dred nur	nbers th	at enco	de the "	meaning	g" of ea	ch word							_
i	-0.010	0.118	-0.026	-0.097	0.097	0.078	-0.148	0.064	0.086	-0.056	0.029	0.084	-0.124	0.141	-0.139	0.072	
didn	-0.024	0.029	-0.015	-0.025	-0.017	0.034	-0.082	0.062	-0.020	0.033	0.008	0.001	0.021	-0.022	-0.060	-0.038	
1	-0.041	-0.178	0.081	0.112	0.101	0.044	-0.114	0.186	0.150	0.049	-0.019	0.125	-0.108	0.156	-0.254	0.025	
t	0.017	0.049	-0.051	-0.055	0.013	0.052	-0.066	0.027	0.009	-0.025	0.018	0.078	-0.062	0.032	-0.046	0.092	
love	0.051	0.315	-0.035	-0.191	0.243	0.181	0.026	-0.286	0.201	0.320	-0.242	-0.137	0.189	-0.049	0.242	-0.156	
this	0.068	-0.051	-0.022	0.054	0.142	0.112	0.016	-0.046	0.158	0.054	-0 .039	0.155	-0.112	0.177	-0.092	0.134	
place	-0.016	0.017	-0.032	-0.052	0.022	0.065	-0.034	-0.043	-0.010	0.051	-0 .002	-0.002	0.034	-0.002	0.060	0.014	
:	-0.060	0.003	-0.034	-0.042	-0.059	0.021	-0.118	0.106	-0.059	0.044	0.043	-0.013	0.041	-0.037	-0.085	-0.051	
(-0.061	-0.062	0.028	0.000	0.036	-0.019	-0.065	0.100	0.062	-0.056	0.028	0.066	-0.100	0.100	-0.205	0.025	

Text Classification

Instead of analyzing sentence structure and grammar, we'll just look for statistical patterns in word use.



Word Associations

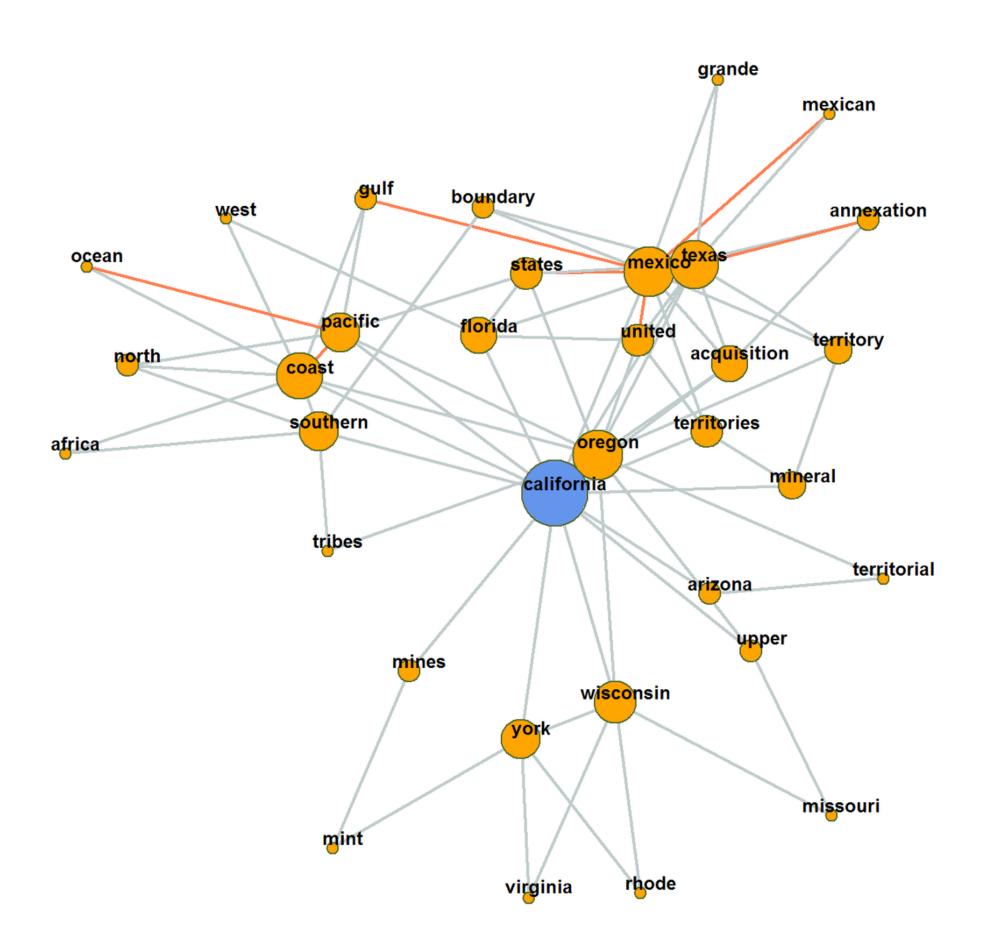


Co-occurence

- 1 Sentence detection
- 2 Counting co-occurrences
- 3 Statistical significance

```
## home introduction inventions manufactures
## home 698 5 1 11
## introduction 5 54 1 1
## inventions 1 1 18 1
## manufactures 11 1 1 160
```

california Graph

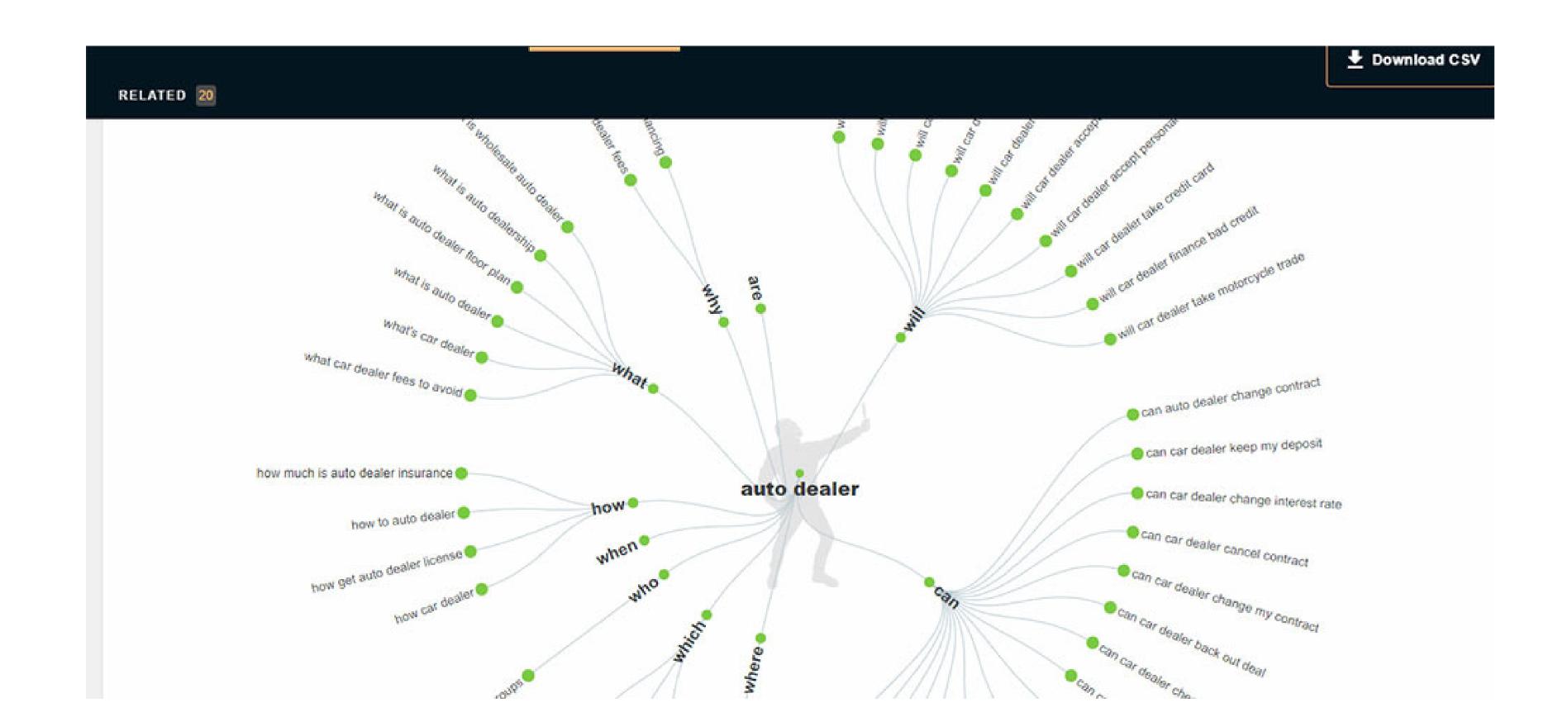


Google Search Quality Raters Guidelines

- E-A-T
- YMYL



Content Optimization and Knowledge Graph and SERP elements



Keyword Magic Tool: List 34 ~ ***



+ New keyword



Questions

Advanced filters 🗸

By num of keywords By volume

All keywords	2,529,164	<u>^</u>
> men	132,136 ⊙	
> size	123,332 •	
> women	122 , 171 •	
> nike	115,385 ⊙	
> run	113,871 ⊙	
> best	72,950 •	
> black	71,653 ⊙	
> adidas	60,034 ⊙	
> store	59,634 ⊙	
> dress	49,713 ⊙	
> white	49,677 ⊙	
N 2011	4E 702 🖎	

Total volume of all keywords. All keywo

Total volume 93,174,260 Average difficulty 8.92%

₫ Export	+ Add to Keyword Analyzer
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Keyword 🔷		Volume 🗘	Trend	KD 🜲	CPC 🔷	Com. 🔷	SERP Feat. 🜲	Results in SERP 🔷
shoes	Ξ,	1,220,000	~~	87.31	1.55	0.59	4	6.3B
nike shoes	Ξ,	1,000,000	~~~	84.75	0.72	1	5	98
shoe carnival	Ξ,	823,000	~~	88.83	0.08	0.26	7	100
adidas shoes	₩.	673,000	~~~	84.76	0.9	1	6	721 M
jordan shoes	Ξ,	450,000		77.65	0.74	1	4	665M
payless shoes	Ξ,	368,000	~~~	88.04	0.14	0.45	3	7.4M
dsw shoes	₩.	301,000		78.47	0.17	0.85	6	11.3M
rack room shoes	Ξ,	301,000	~~~	68.45	0.16	0.26	3	88.4M
basketball shoes	Ξ,	246,000		85.58	1.04	1	4	651M

Keyword Magic Tool: List 34 ~ "

→ Go to Keyword Analyzer

₫ Export

All lists 34

+ Add to Keyword Analyzer



shoes US V Apply Broad Match Phrase Match Exact Match Related 🚥

All Questions Advanced filters 🗸

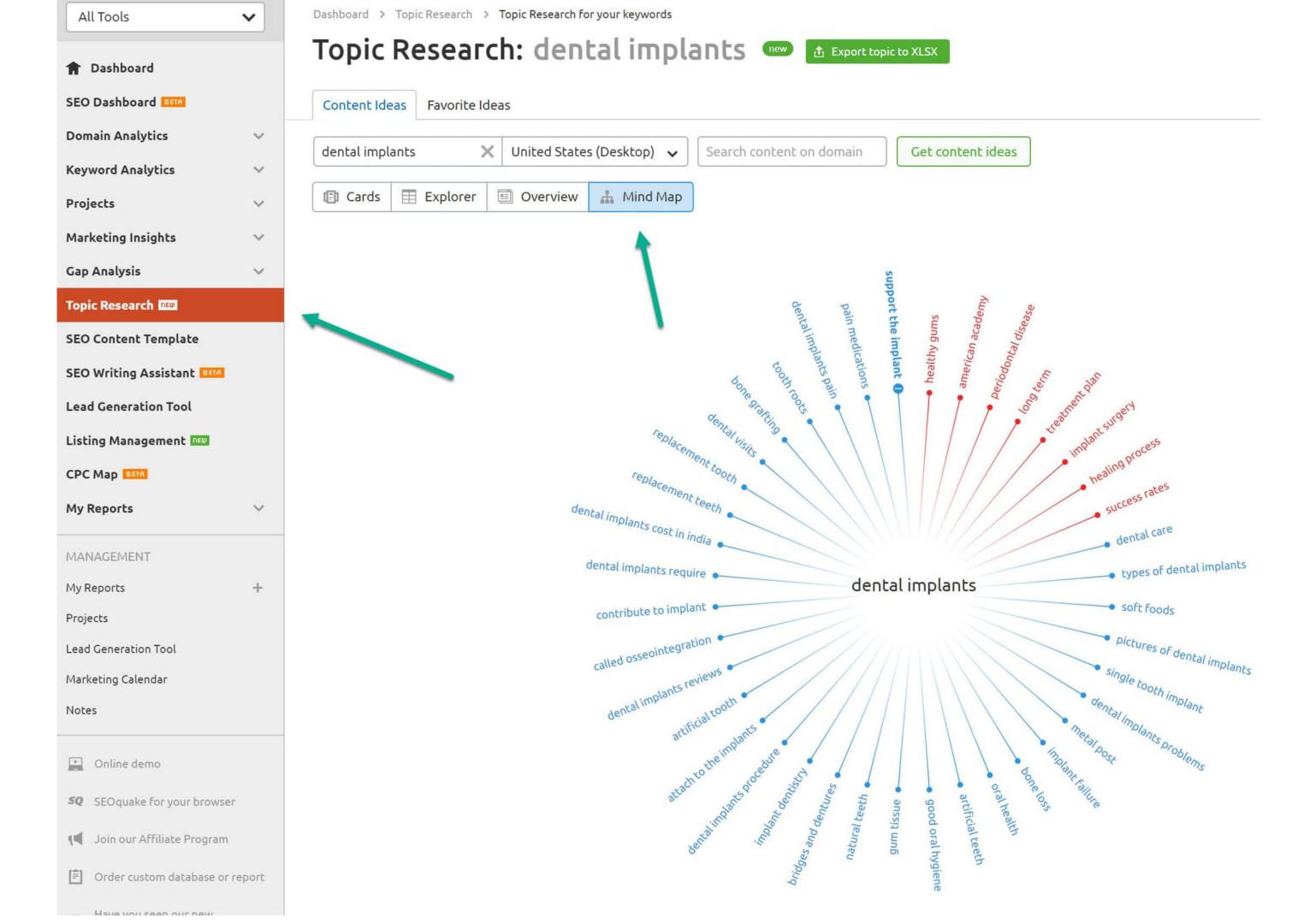
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> adidas	60,034 ⊙	
> store	59,634 ⊙	
> dress	49,713 ⊙	
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	45.700	

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7	basketball shoes	₽,	246,000		85.58	1.04	1	4	651M



Content optimization summary tips

- Use actionable speech
- Use h1, h2, h3 subheadings and create hierarchy of the content
- Use natural language phrases
- Research what your audience use in the form of phrases on the web
- Synonyms and Close Variants
- Semantic Distance and Term Relationships
- Phrase-Based Indexing
- Be clear with your topic
- Answer your current, its pre & post questions
- Use entities
- Use short paragraphs
- Provide clear answers
- Avoid too difficult words and sentences
- Be clear and descriptive with your titles
- Optimize your images properly alt tags, image titles, surrounding texts, captions
- Add videos