Balancing the Dimensions of User Intent



Trey Grainger Chief Algorithms Officer Lucidworks

HAYSTACK * EU * October 28, 2019 * * *

About Me

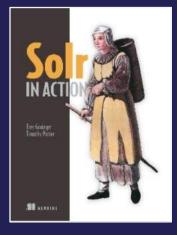


Trey Grainger Chief Algorithms Officer Lucidworks

- **Previously**: SVP of Engineering @ Lucidworks; Director of Engineering @ CareerBuilder
- Georgia Tech MBA, Management of Technology
- Furman University BA, Computer Science, Business, & Philosophy
- **Stanford University** Information Retrieval & Web Search

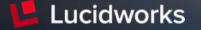
Other fun projects:

- Co-author of *Solr in Action*, plus numerous research publications
- Advisor to <u>Presearch</u>, the decentralized search engine
- Lucene / Solr contributor



Agenda

- About Lucidworks
- What is AI-powered Search?
- The Dimensions of User Intent
- Content Understanding:
 - Keyword Search
- User Understanding:
 - Collaborative Recommendations
- Content Understanding + User Understanding:
 - Personalized Search
- Domain Understanding:
 - Knowledge Graphs
- Domain Understanding + User Understanding:
 - Domain-aware Matching
- Content Understanding + Domain Understanding:
 - Semantic Search
 - Balancing Approaches:
 - Keyword vs. Vector vs. Knowledge Graph Search
 - Vector Search
 - Knowledge Graph Search
- Combining it all together



Who are we?

300+

EMPLOYEES

400+

CUSTOMERS ACROSS THE FORTUNE 1000



San Francisco, CA (HQ) Raleigh-Durham, NC Cambridge, UK Bangalore, India Hong Kong COMPANY BEHIND

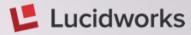
ACTIVATE The Search & Al Conference





FUSION

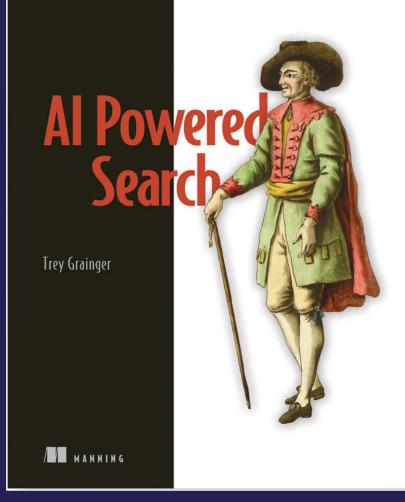
FUSION App Studio



Proudly built with open-source tech at its core: Apache Solr & Apache Spark Personalizes search with applied machine learning Proven on the world's biggest information systems



What is Al-Powered Search ?



(Haystack discount code: ctwhay19)

... is my new **book!**

http://aiPoweredSearch.com



Difference between machine learning and AI:

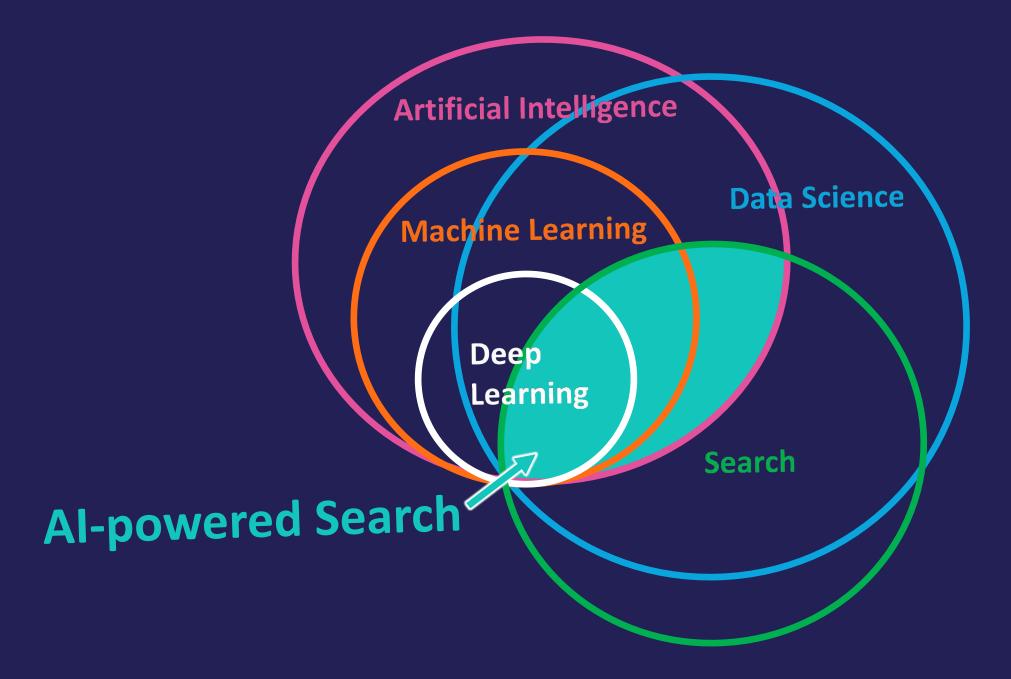
If it is written in Python, it's probably machine learning

 \bigcirc

If it is written in PowerPoint, it's probably AI

♡ 23.6K 9:25 PM - Nov 22, 2018

 \bigcirc 8,727 people are talking about this



ArtiAl-powered Search

Signals Boosting Models ine Learnin

De

- Learning to Rank
- Semantic Search
- Collaborative Filtering
- Personalized Search
- Content Clustering
- NLP / Entity Resolution
- Semantic Knowledge Graphs
- Document Classification
- etc.

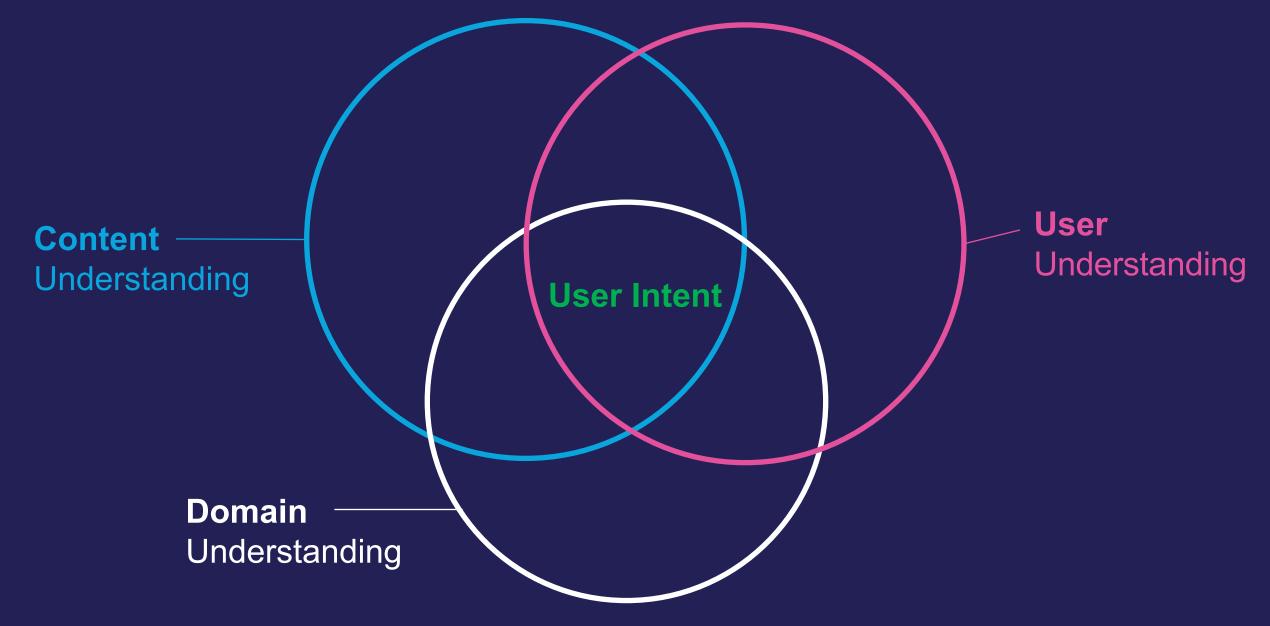
Data Science

- Question / Answer Systems
- Virtual Assistants
- Chatbots
- Rules-based Relevancy
- etc.
- Neural Search
- Word Embeddings
- Vector Search
- Image / Voice Search
- etc.

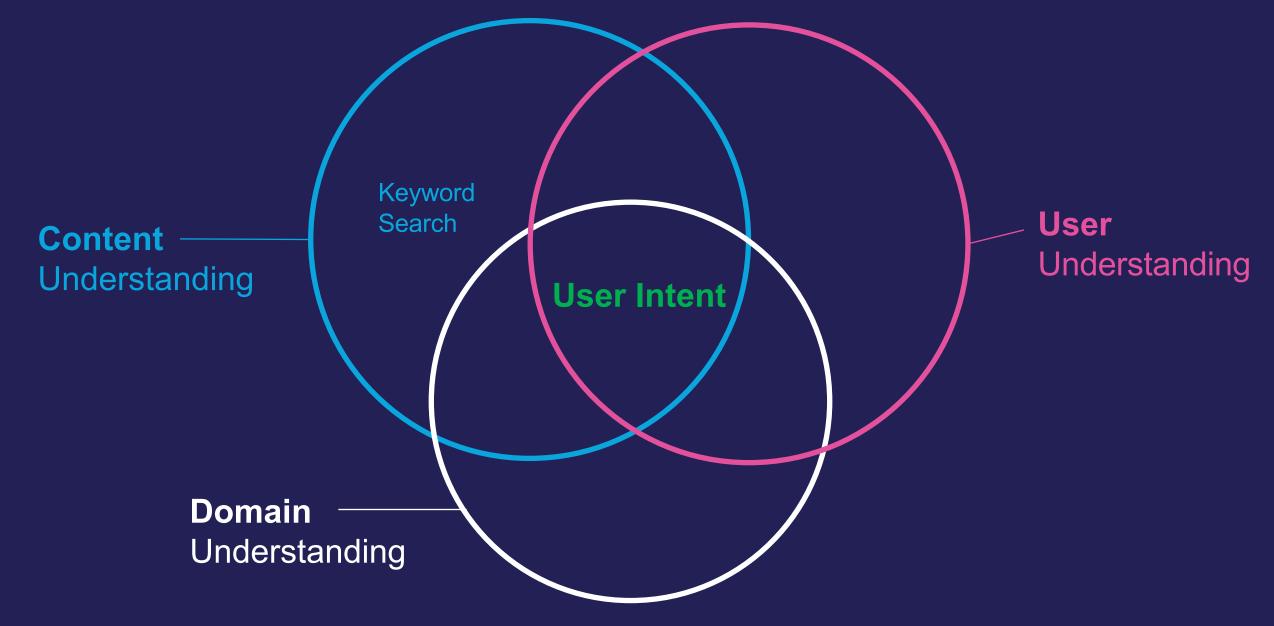
We have a big toolbox - great!

But how do we properly apply those tools?

Dimensions of User Intent



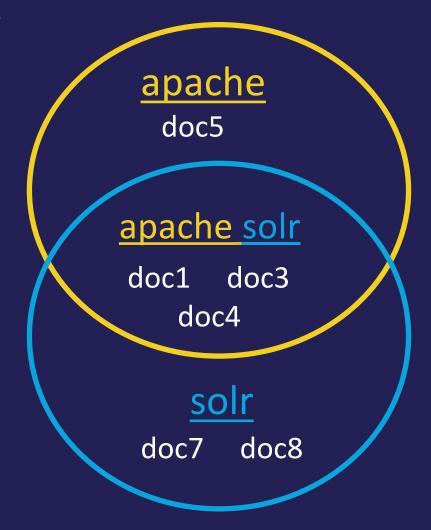
Dimensions of User Intent



Matching queries to documents

/solr/collection/select/?q=apache solr

Term	Documents
apache	doc1, doc3, doc4, doc5
lucene	doc2, doc4, doc6
solr	doc1, doc3, doc4, doc7, doc8

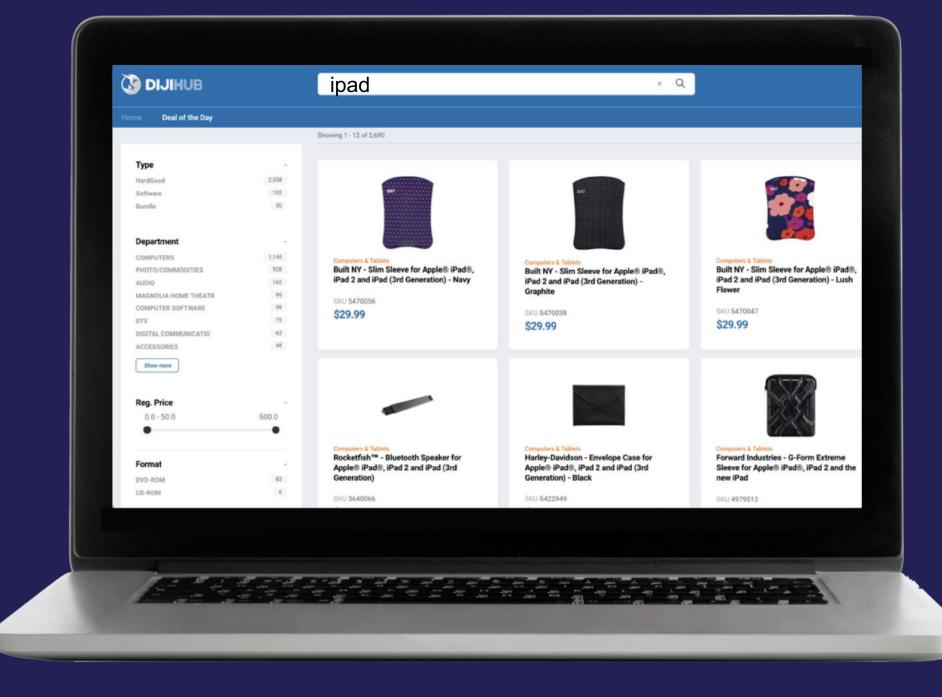


BM25 (Relevance Scoring between Query and Documents)

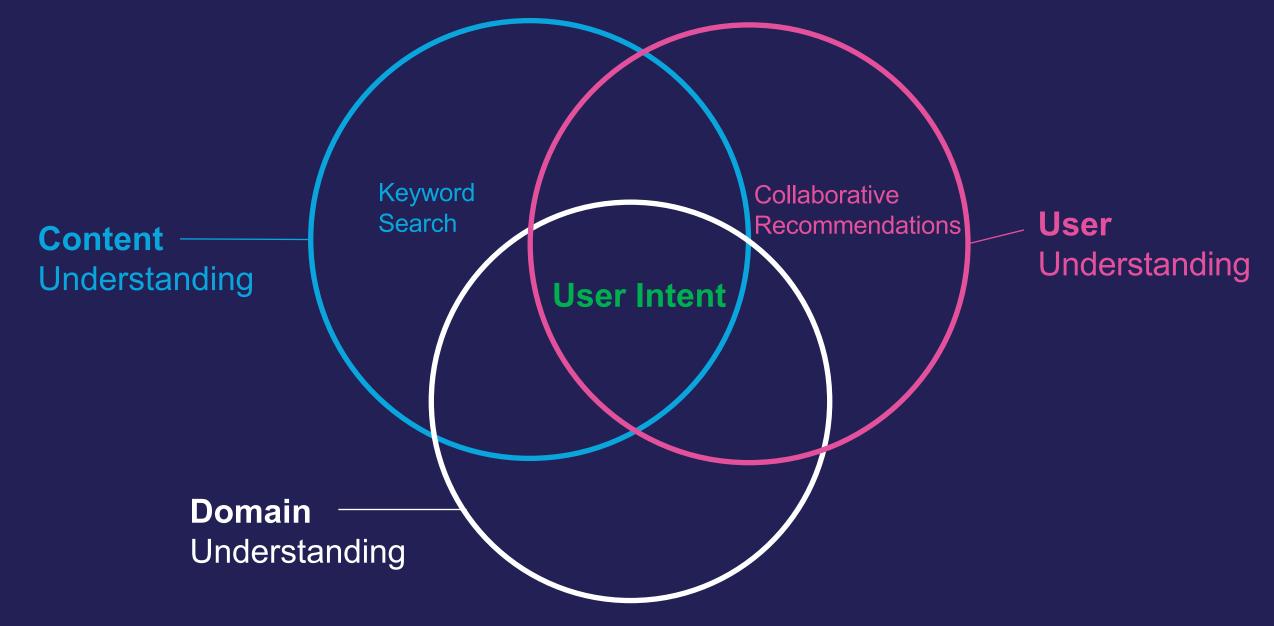
Score(q, d) = $\sum_{\substack{t \text{ in } q}} \frac{idf(t) \cdot (tf(t \text{ in } d) \cdot (k + 1)) / (tf(t \text{ in } d) + k \cdot (1 - b + b \cdot |d| / avgdl)}{t \text{ in } q}$

Where:

b = Free parameter. Usually ~0.75. Increases impact of document normalization.



Dimensions of User Intent



<

Computers & Tablets



9.7" widescreen display; 802.11a/b/g/n Wi-Fi; Bluetooth; iBooks support; measures just 0.34" thin and weighs only 1.35 lbs.



Trey, recommended for you!



TV & Home Theater LG - 47" Class / LED / 1080p / 120Hz / 3D / HDTV 3D Blu-ray Player Bundle

SKU 2970264

\$1099.99



LG - 47" Class - LCD - 1080p - 120Hz - HDTV

SKU 2138282 \$999.99

99.99



TV & Home Theater LG - 55" Class / LED / 1080p / 120Hz / 3D / HDTV 3D Blu-ray Player Bundle

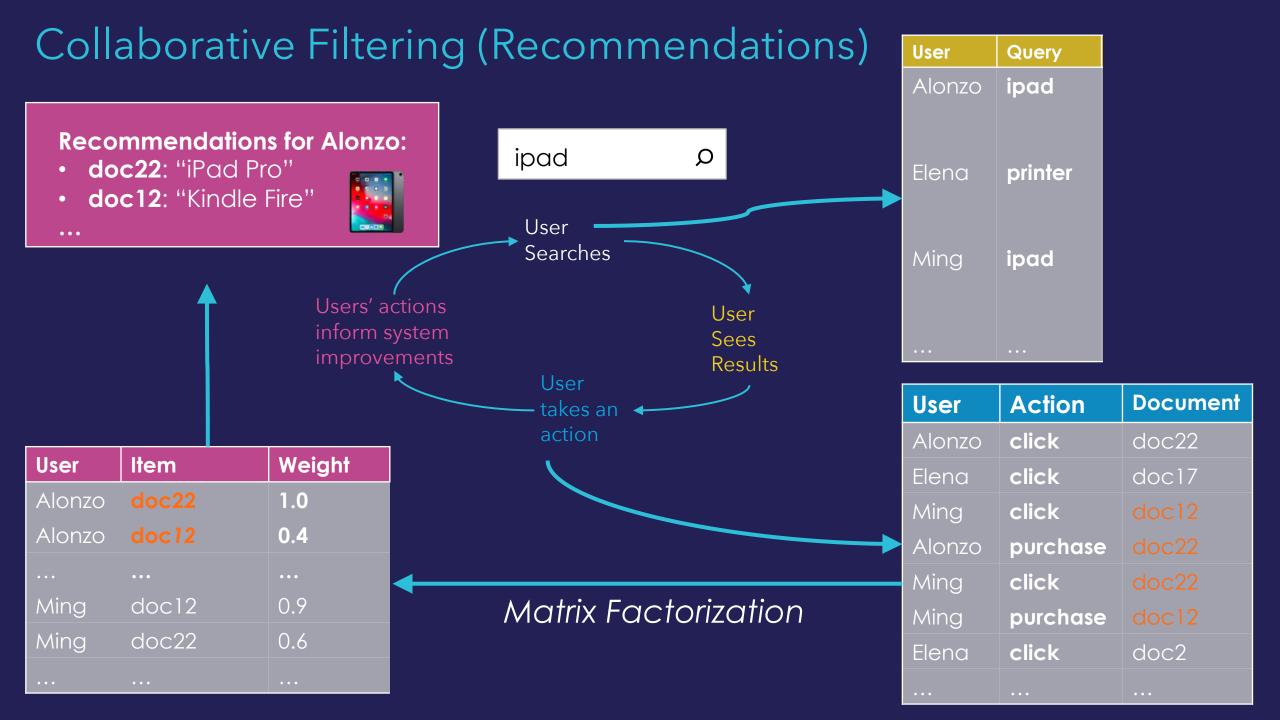
SKU 2970291

\$1599.99



TV & Home Theater LG - 55" Class / 1080p / 120Hz / LCD HDTV

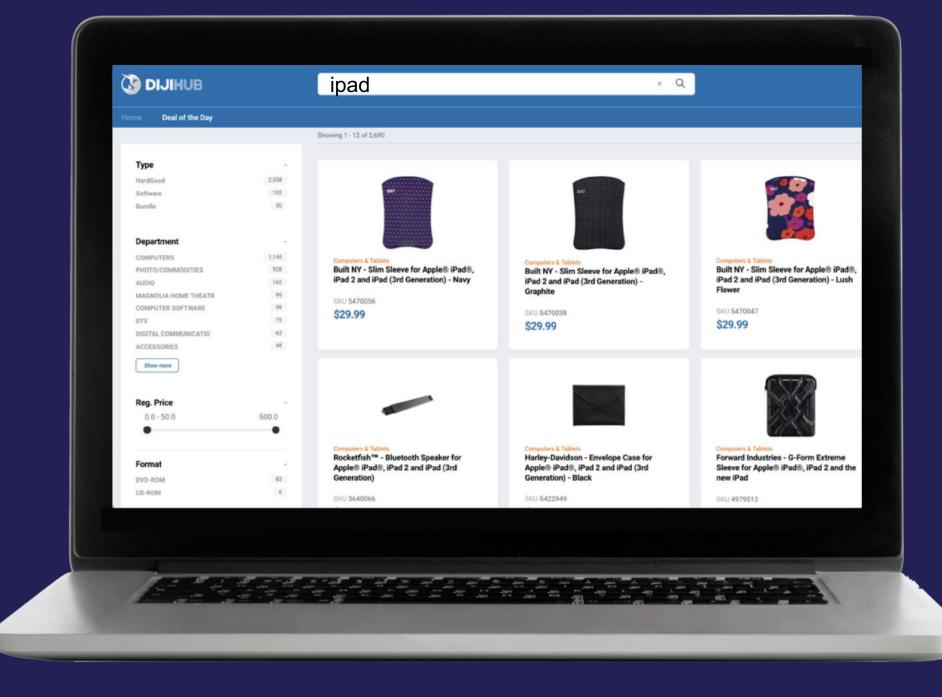
SKU 2138291 \$1499.99 >

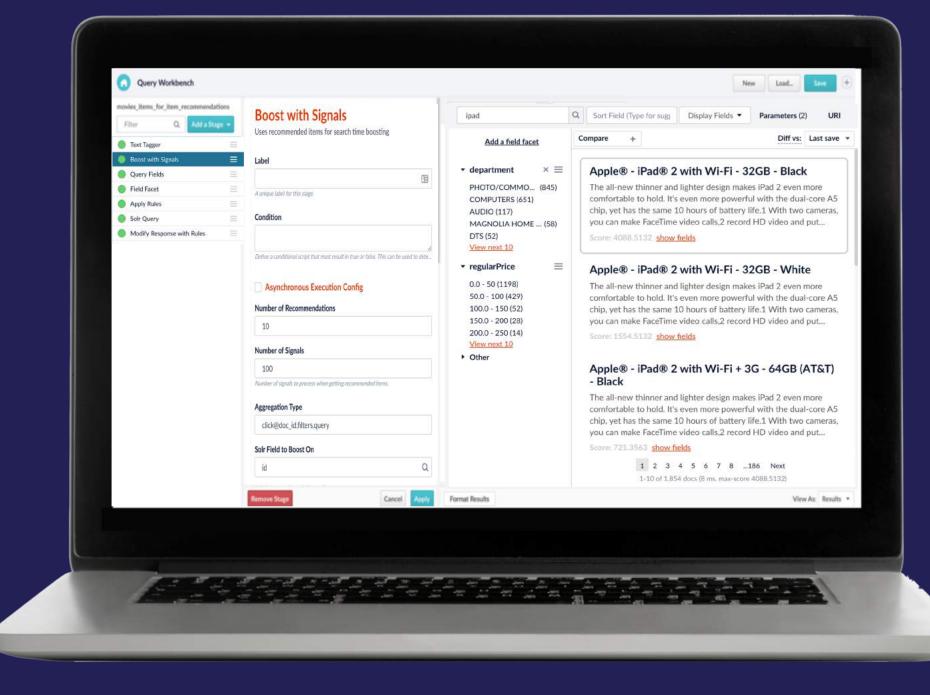


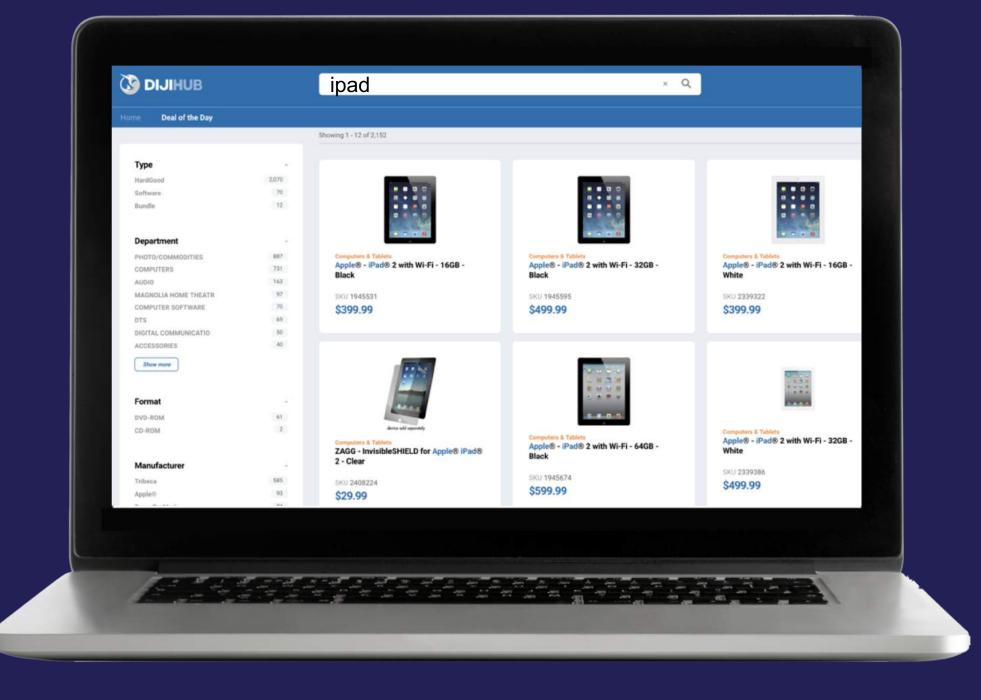
Recommendations (User-Item, Item-Item, Query-Item)

Recommendations for Alonzo: • doc22: "iPad Pro" • doc12: "Kindle Fire" 		 doc22: "iPad Pro" doc12: "Kindle Fire" 			Recommendations for "ipad": • doc22: "iPad Pro" • doc12: "Kindle Fire"			
User	ltem	Weight	Item	Item	Weight	Query	Item	Weight
Alonzo	doc22	1.0	doc22	doc22	1.0	ipad	doc22	0.98
Alonzo		0.4	doc22		0.85	ipad		0.6
	•••	•••		•••			• • •	•••
Ming	doc12	0.9	doc12	doc12	1.0	kindle	doc12	0.96
Ming	doc22	0.6	doc12	doc22	0.83	apple	doc22	0.90

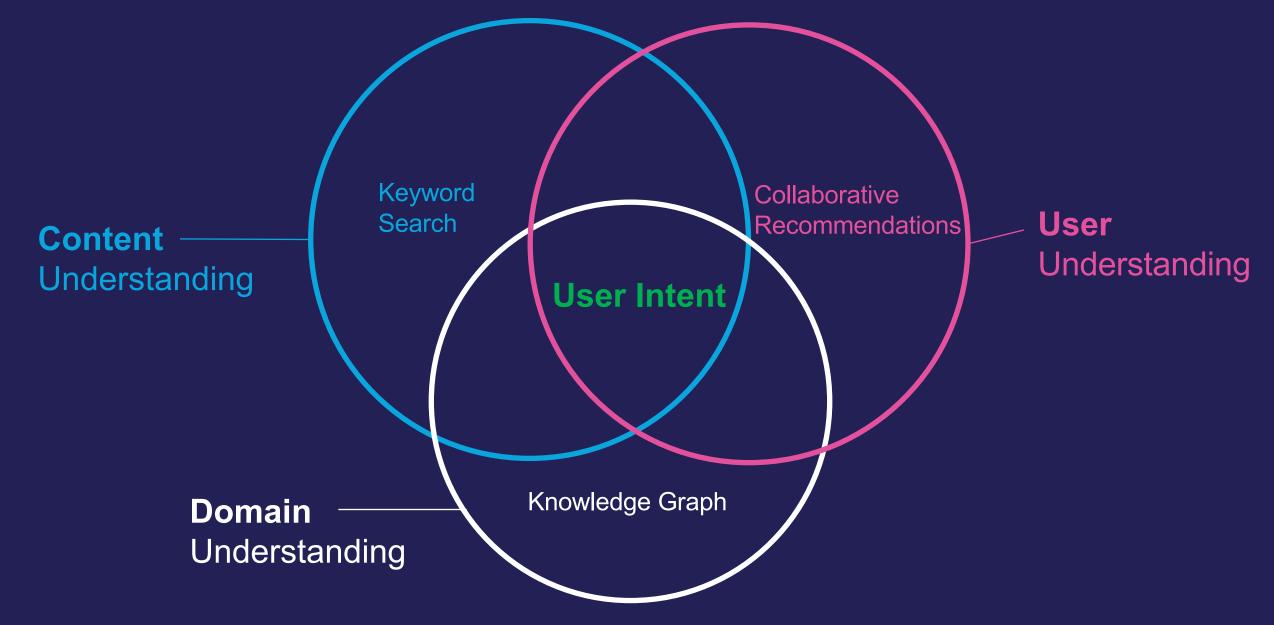
Matrix Factorization







Dimensions of User Intent



What is a Knowledge Graph? (vs. Ontology vs. Taxonomy vs. Synonyms, etc.)

Overly Simplistic Definitions

Alternative Labels: Substitute words with identical meanings [CTO => Chief Technology Officer; specialise => specialize]

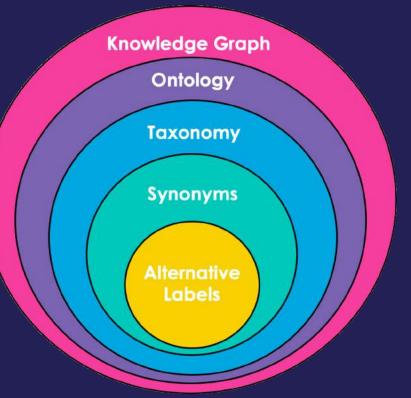
Synonyms List: Provides substitute words that can be used to represent the same or very similar things [human => homo sapien, mankind; food => sustenance, meal]

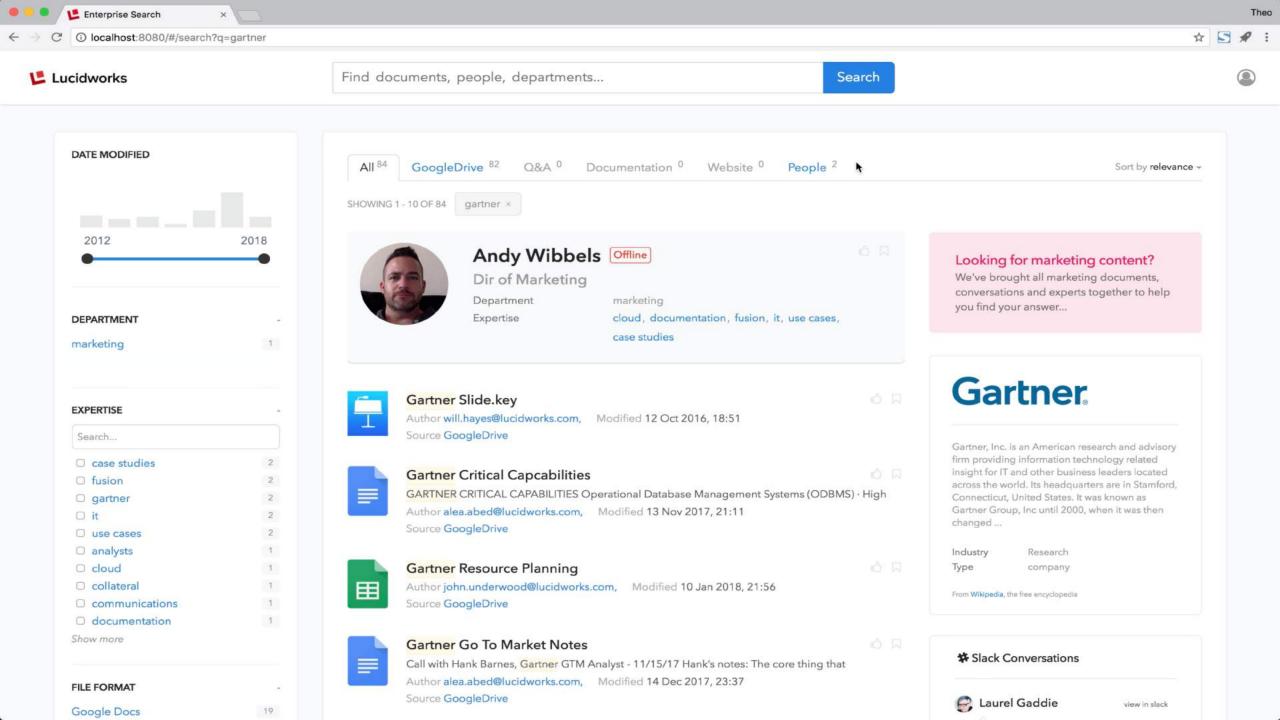
Taxonomy: Classifies things into Categories [john is Human; Human is Mammal; Mammal is Animal]

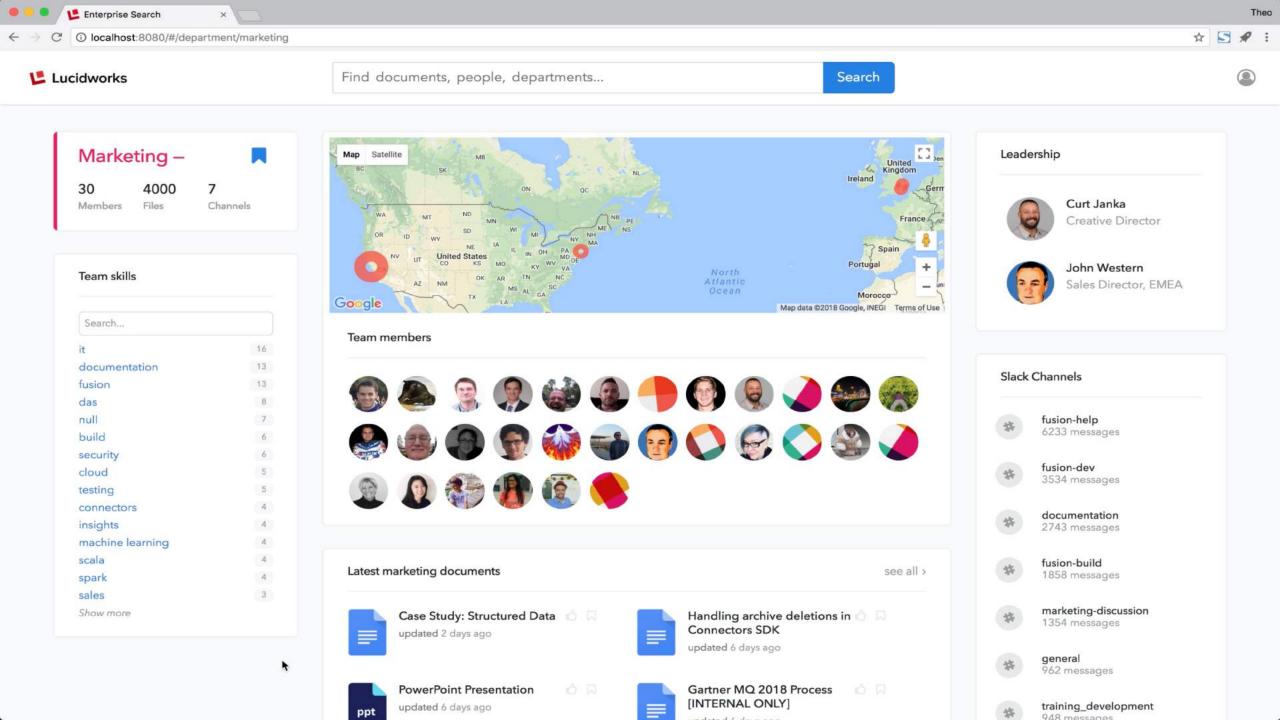
Ontology: Defines relationships between types of things [animal eats food; human is animal]

Knowledge Graph: Instantiation of an Ontology (contains the things that are related) [john is human; john eats food]

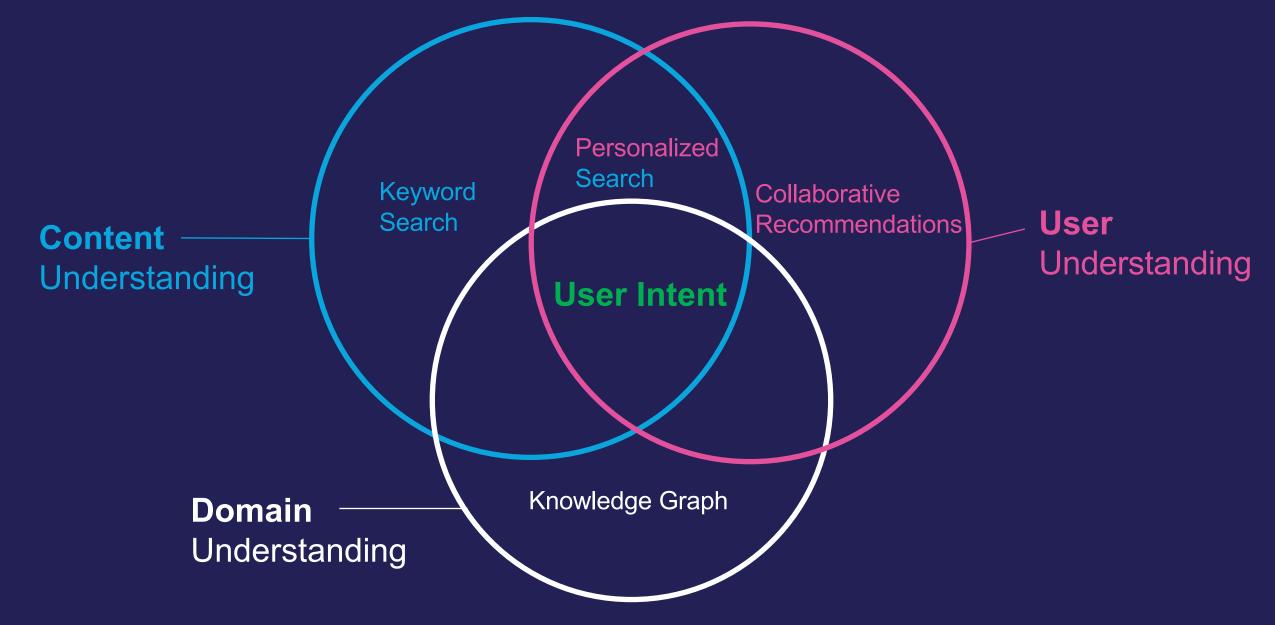
A Knowledge Graph subsumes the other types.







Dimensions of User Intent



Keyword Search (Completely User-specified)

Traditional Recommendations (Completely driven by user behavior)

Keyword Search (Completely User-specified)

User-guided Recommendations (Mostly driven by user profile, partially user-specified)

Personalized Queries (Mostly user-specified, partially driven by user profile) Traditional Recommendations (Completely driven by user behavior)

Personalized Search

Keyword Search (Completely User-specified)

User-guided Recommendations (Mostly driven by user profile, partially user-specified)

Personalized Queries (Mostly user-specified, partially driven by user profile)

Traditional Recommendations (Completely driven by user behavior)



User Details for 008f7b53f33aaeb279f13d9c222edca9610c0cb6

Launch Personalized Search Page

Personalization

User Clic	k History	Items for User	Personalization Aggregat	ion	
Image	Timestam	ıp	Query	Product	Doc Id
2	2018-10-2	23T08:53:06.400Z	samsung refrigerator	Samsung - 25.6 Cu. Ft. Side-by-Side Refrigerator with Thru-the-Door Ice and Water - Stainless-Steel	9747183
2	2018-10-2	23T08:50:28.420Z	samsung refrigerator	Samsung - 25.5 Cu. Ft. Side-by-Side Refrigerator with Thru-the-Door Ice and Water - Stainless-Steel	9791769
	2018-10-2	23T08:45:12.490Z	samsung range	Samsung - 30" Self-Cleaning Freestanding Electric Convection Range - Stainless-Steel/Black	9212504
	2018-10-2	23T08:44:09.910Z	samsung range	Samsung - 30" Self-Cleaning Freestanding Electric Convection Range - Stainless-Steel	9212498
	2018-10-0	06T16:08:07.830Z	samsung dishwasher	Samsung - 24" Tall Tub Built-In Dishwasher - Stainless-Steel	9980455
Ē	2018-10-0	04T15:57:40.500Z	lg range electric	LG - 30" Self-Cleaning Freestanding Electric Range - Stainless-Steel	2429379
	2018-10-0	04T15:56:10.880Z	lg range electric	LG - 30" Self-Cleaning Freestanding Double Oven Electric Range - Stainless-Steel	2428282



Trends Se

n Search with Que

uery Rewrite Search wi

re Personalization Aggregation

User Details for 008f7b53f33aaeb279f13d9c222edca9610c0cb6

Launch Personalized Search Page

User Click History Items for User

Personalization Aggregation

User Id	Department	Brand	Color	Count
008f7b53f33aaeb279f13d9c222edca9610c0cb6	Appliance		Stainless-Steel	22
008f7b53f33aaeb279f13d9c222edca9610c0cb6	Appliance	Samsung		9
008f7b53f33aaeb279f13d9c222edca9610c0cb6	Appliance	KitchenAid		6
008f7b53f33aaeb279f13d9c222edca9610c0cb6	Appliance	LG		5
008f7b53f33aaeb279f13d9c222edca9610c0cb6	Appliance		Monochromatic Stainless-Steel	3
008f7b53f33aaeb279f13d9c222edca9610c0cb6	Appliance	Bosch		2
008f7b53f33aaeb279f13d9c222edca9610c0cb6	Appliance	Frigidaire		2
008f7b53f33aaeb279f13d9c222edca9610c0cb6	Appliance	Whirlpool		1
008f7b53f33aaeb279f13d9c222edca9610c0cb6	Magnolia Home Theatr	Russound		1
008f7b53f33aaeb279f13d9c222edca9610c0cb6	Appliance	GE		1
008f7b53f33aaeb279f13d9c222edca9610c0cb6	Appliance	Electrolux		1
008f7b53f33aaeb279f13d9c222edca9610c0cb6	Appliance		Stainless-Steel/Black	1
008f7b53f33aaeb279f13d9c222edca9610c0cb6	Appliance		Black/Stainless-Steel	1
	All was dealed			

All results loaded

DIJIHUB

rends Search Se

Brand	

GE	293
Whirlpool	149
Sharp	101
Frigidaire	88
KitchenAid	78
LG	63
Panasonic	55
Maytag	38
Samsung	33
Electrolux	22
Goldstar	16
Bodum	13
Show more	

Department

		- 1		

Appliance	1,030
Computers	9
Interactive Software	3
Mobile Audio	3
Accessories	2
Photo/Commodities	2

microwave	×	Search Clear	
Search for "microwave"	Q		
PRODUCTS			
GE - 0.7 Cu. Ft. Compact Microwave - Black	GE		
Hello Kitty - 0.7 Cu. Ft. Compact Microwave	Hello Kitty		
GE - 1.1 Cu. Ft. Mid-Size Microwave - White	GE		
SUGGESTED SPELLINGS		microwave ovens ge o	ver the range microwave
panasonic microwave	Phrase => Phrase		
frigidaire microwave	Phrase => Phrase		
ge microwave	Phrase => Phrase		
SEARCH HISTORY			Sort by relevance -
microwave			
SUGGESTED WORDS/PHRASES			
microwave		Ft. Compact Microwave	0
microwave stainless		its in this microwave that mable recipes for customize	ed
microwave black		sign for optimal placement.	
microwave stainless steel		ties Brand Hello Kitty	
microwave white microwave package			
microwave package			
ID 9419729	Score 1034.6353	2	
\$59			



microwave

Search with Query Rewrite

Search with Signals Aggr

Compare Personalization Aggregation

Brand

GE	293
Whirlpool	149
Sharp	101
Frigidaire	88
KitchenAid	78
LG	63
Panasonic	55
Maytag	38
Samsung	33
Electrolux	22
Goldstar	16
Bodum	13

Department

Search...

Appliance	1,030
Computers	9
Interactive Software	3
Mobile Audio	3
Accessories	2
Photo/Commodities	2

Regular Search Results:

GE - 0.7 Cu. Ft.	Compact N	<mark>licrowave</mark> - Black	0
Cook or reheat yo	our meals in i	minutes with this com	pact
700-watt microwa	ave oven that	features instant-on Ll	ED
controls which he	at food with	the simple touch of a	
button. It also pro	vides peace	of mind with a contro	bl
lockout feature th	at locks the	microwave when it's n	
			IOT IN
use.			IOT IN

ID 9419729

\$59 Score 1373.8041

Hello Kitty - 0.7 Cu. Ft. Compact Microwave 0

Cook your favorite treats in this microwave that features 6 preprogrammable recipes for customized use and a compact design for optimal placement. Dept Photo/Commodities Brand Hello Kitty

ID 2817088

\$89

Score 1034.6353

GE - 1.1 Cu. Ft. Mid-Size Microwave - White 6 Prepare hassle-free meals and snacks with this



Personalized Search Results:

×

User: 008f7b53f33aaeb279f13d9c222edca9610c0cb6

Search

Clear

Samsung - Mid-Size Microwave - Stainless-steel (

This mid-size microwave is large enough for almost any meal, but compact enough to fit into almost any space. Take an interactive product tour of this Samsung Microwave! (479KB Flash demo) Dept Appliance Brand Samsung Color Stainless-Steel

ID 4564643 \$58 Score 1689.5975

Samsung - 1.7 Cu. Ft. Over-the-Range Microwave - Stainless-Steel





The capabilities of a toaster oven, convection oven, range hood and microwave combine in this microwave that features a 1.7 cu. ft. capacity for simultaneously cooking multiple dishes and sensor cook options to take the guesswork out of quick meal prep.









Nice - personalization is awesome!

Let's roll it out everywhere!



Trends

Search

Search with Query Rewrite

180

97

86

85

59

57

52

45

44

37

32

31

690

441

338

101

76

55

45

31

28

27

3



Search.

Tribeca

Incase

Apple

Belkin

IFrogz

Home

Sony

Mophie

Show more

Search...

Audio

DTS

Video

Color

-

Mobile Audio

Computers

Accessories

Department

Griffin Technology

Pangea Brands

Rocketfish Mobile

Digital Communication

Photo/Commodities

Computer Software Magnolia Home Theatr

Interactive Software

OtterBox

Suggested Searches

iphone unlock laguna beach unlocked iphone i phone iphone4 unlocked iphone verizon bodyguardz pantech breeze iphone refurbished refurbished iphone

iphone ×

SHOWING 1 - 10 OF 1,863 * Debug

1 2 3 4 5 next

Samsung - Audio Dock for Apple® iPod® and iPhone® 0

Listen to tunes on your Apple iPod or iPhone or compatible Samsung Galaxy mobile phone with this Samsung DAE570ZA audio dock that delivers 10W total system power for clear sound. Bluetooth technology makes it easy to wirelessly connect compatible devices. Dept Audio Brand Samsung

ID 5563122

\$179

Score 5.7202854

Samsung - Audio Dock for Apple® iPod®, iPhone® and 0 iPad®

Rock out to your favorite tunes with this Samsung DAE750ZA dock that features woven glass fiber speakers and a down-firing subwoofer for powerful audio. The dual docking system allows you to listen to tunes on your Apple iPod, iPhone, iPad and more. Dept Audio Brand Samsung

ID 5563159

\$699

Score 5.3621306

Samsung - Wireless Cradle for Apple® iPod® and iPhone® () - Black





Samsung - 840W 7.2-Ch. A/V Home Theater Receiver

Clear

Q ×

This powerful 7.2-channel home theater receiver features 12 DSP soundfields for a lush soundscape. An Apple® iPod® and iPhone® dock lets you listen to your favorite music. Dept Audio Brand Samsung

ID 1151846 \$499

Score 5.5088468

Samsung - Audio Dock for Apple® iPod®, iPhone® and 0 iPad®



Enjoy your favorite songs with this Samsung DAE670ZA dock that features a dual docking system for listening to tunes on your Apple iPod, iPhone, iPad or compatible Samsung Galaxy device. The woven glass fiber speakers deliver stunning audio. Dept Audio Brand Samsung

ID 5563104

\$359

Score 5.3621306

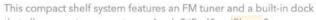


Samsung - 250W Compact Shelf System



0



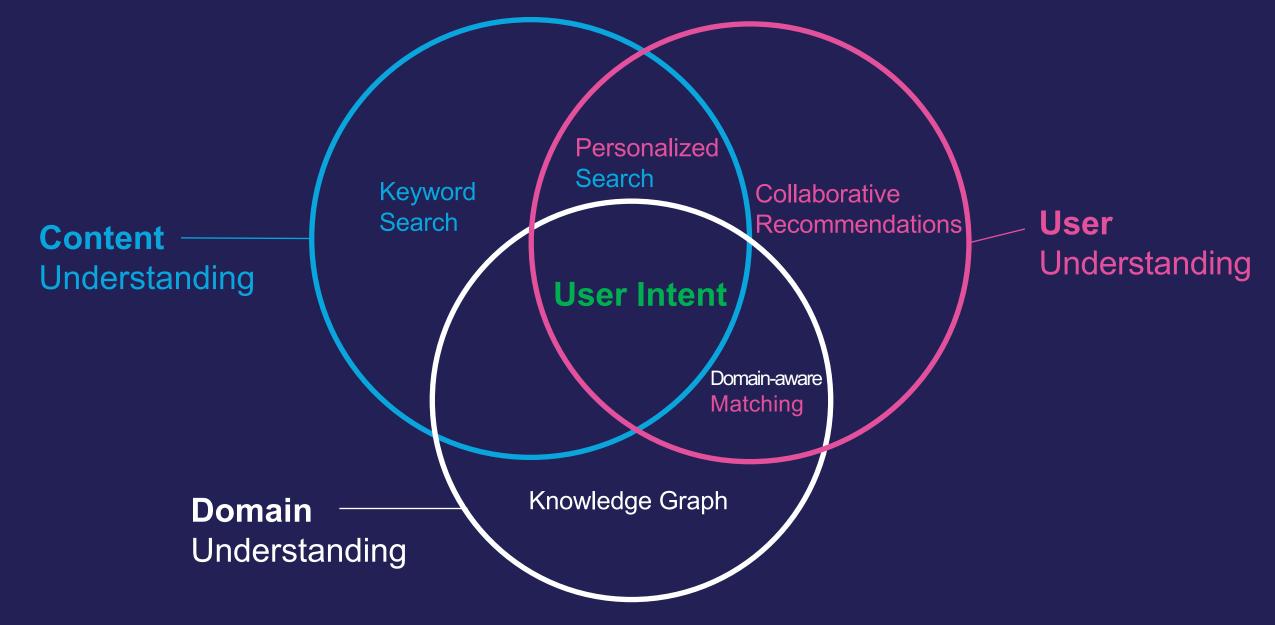






Sort by relevance -

Dimensions of User Intent



Collaborative Recommendations (Completely driven by user behavior)

Knowledge Graph

(Understanding conceptual and logical relationships between domain-specific entities)

Personas / User Profiles (User attributes and preferences in knowledge graph)

Collaborative Recommendations

(Completely driven by user behavior)

Knowledge Graph

(Understanding conceptual and logical relationships between domain-specific entities)

Multimodal Recommendations

(Recommendations combining collaborative filtering plus user-based profile attribute matching/ranking)

Domain-aware Matching

Personas / User Profiles (User attributes and preferences in knowledge graph)

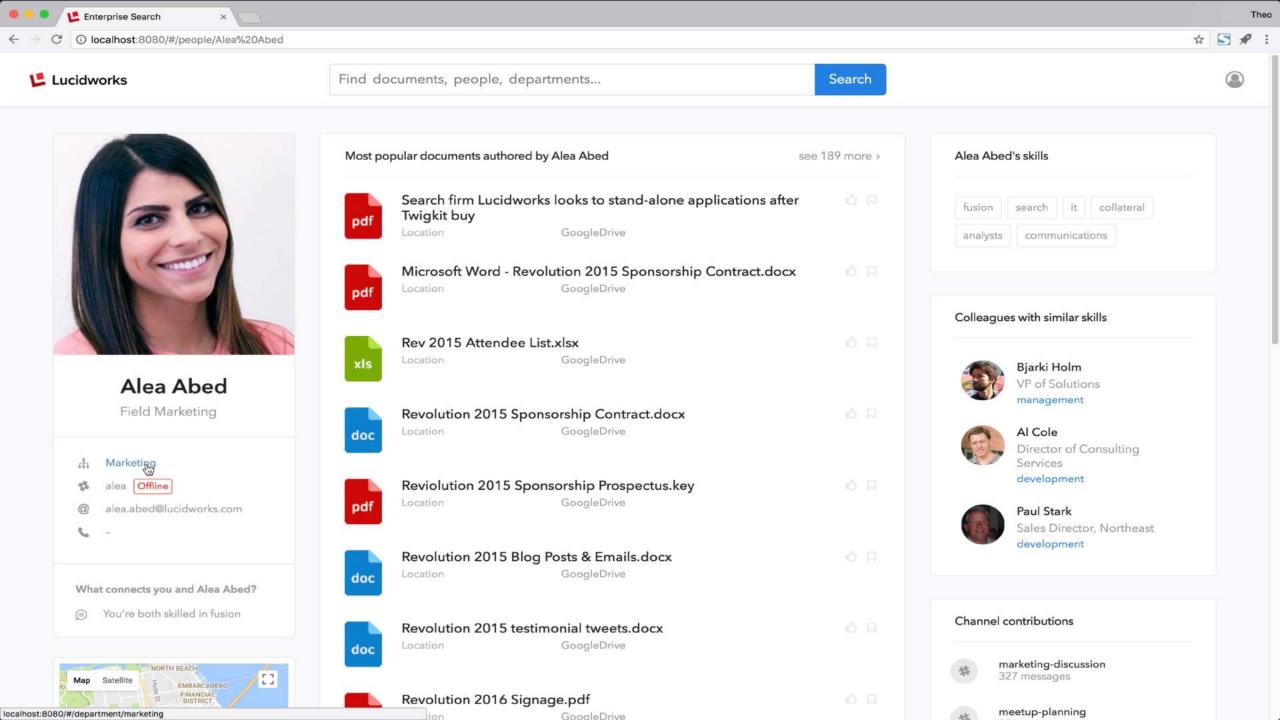
Collaborative Recommendations (Completely driven by user behavior)

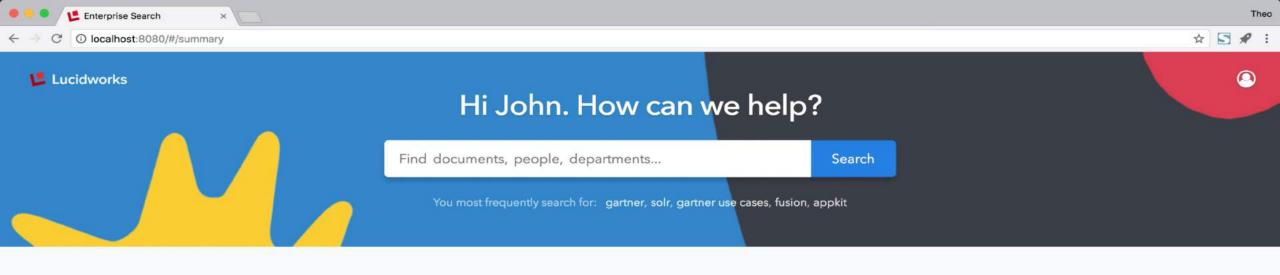
Knowledge Graph

(Understanding conceptual and logical relationships between domain-specific entities)

Multimodal Recommendations

(Recommendations combining collaborative filtering plus user-based profile attribute matching/ranking)



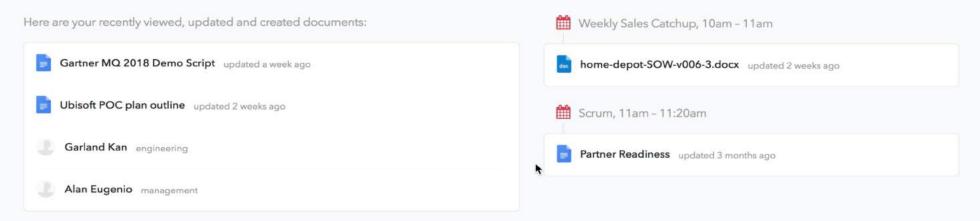


Jump right back in -

03:49

January 31, 2018





Latest Engineering GitHub notifications -

vlaw

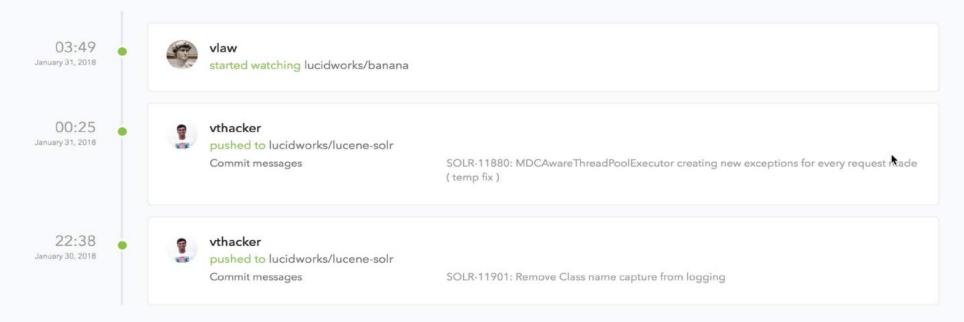
started watching lucidworks/banana

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🗧 😑 🖉 😢 Enterprise Search	×		Theo
← → C () localhost:8080/	/#/summary		☆ 🔄 🖋 🗄
	Garland Kan engineering	Partner Readiness updated 3 months ago	
	Alan Eugenio management		

Latest Engineering GitHub notifications -

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Explore departments -

more >

Marketing	
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20 people 2000 files 7 channels

Engineering

122 people 122 files 24 channels

HR

22 noopla 22 files 15 channels

• • •	• /1	Enterprise Search	×		т	hec
€ ⇒	C	localhost:8080/#	/summary	~ ☆	5 🖋	÷
			Detect Sentences Index Stage Steve Tsuchiyama support			

Latest Marketing department tweets -

𝕊 lucidworks >

Blog Post: #Search is critical for #Omnichannel #Retail customers. Here's why. #Retailtech #MachineLearning #AI https://t.co/eoXWFj4pIM

Are you an #NLP expert? Lucidworks is searching for an NLP Research Engineer to join our #CTO's team. #Hiring... https://t.co/FupVludArY

Check out @gsingers's newest article, How to Develop Tools for Great Customer Experience - DZone Agile https://t.co/1blZgkl8V4

Explore departments -

22:25

21:40

21:47

January 26, 2018

January 29, 2018

January 29, 2018

more >

HR Marketing Engineering 30 people. 3999 files. 7 channels. 132 people. 132 files. 24 channels. 22 people. 32 files. 15 channels.

eo

Multimodal Recommendations

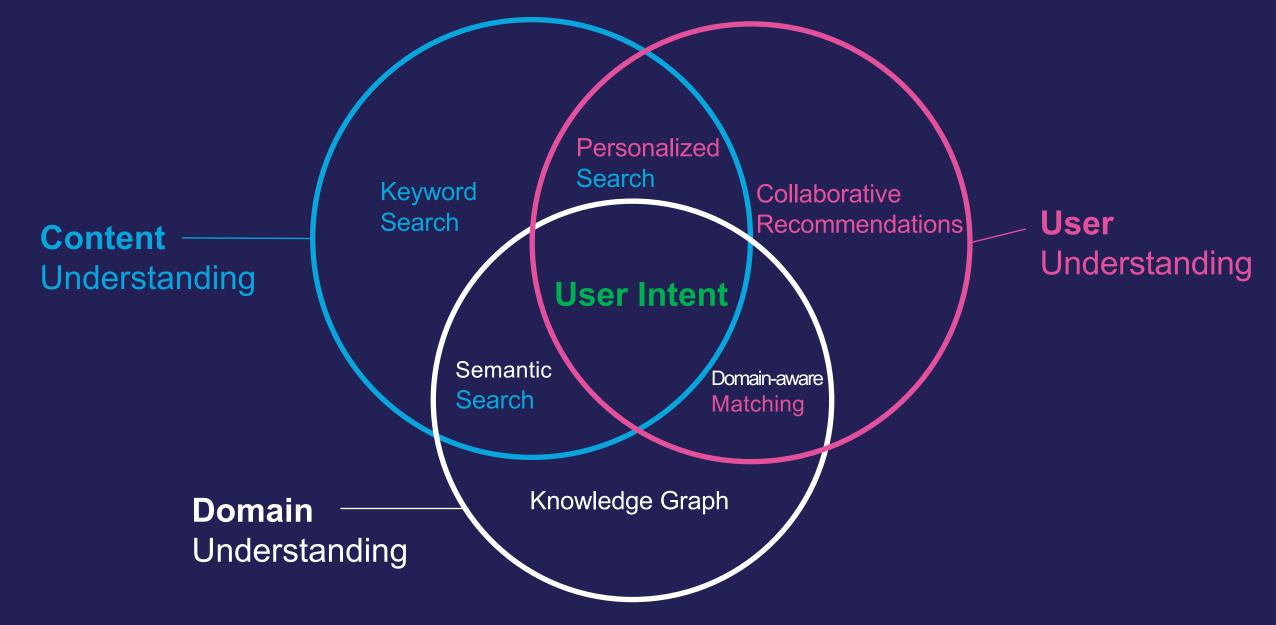
Jane is a nurse educator in Boston seeking between \$40K and \$60K

She has interacted with the same content as the following users: u99,u1,u50,u2311,u253,u70,u99

```
http://localhost:8983/solr/jobs/select/?
fl=jobtitle,city,state,salary&
q=(
    jobtitle:"nurse educator"^25 OR jobtitle:(nurse educator)^10
    )
    AND (
        (city:"Boston" AND state:"MA")^15
        OR state:"MA")
    AND (_val_:"map(salary, 40000, 60000,10, 0)"
    OR {!edismax mm=2}similar_users:(u99 u1 u50 u2311 u253 u70 u99))
```

*Example derived from chapter 16 of Solr in Action

Dimensions of User Intent



Keyword Search (Finding and Ranking Keyword)

Knowledge Graph

(Understanding conceptual and logical relationships between domain-specific entities)

Keyword Search (Finding and Ranking Keyword)

Terminology Understanding (Understanding domain-specific terms and conceptual meaning)

Language Understanding (Understanding syntax and query structure)

Knowledge Graph

(Understanding conceptual and logical relationships between domain-specific entities)

Semantic Search

Keyword Search (Finding and Ranking Keyword)

Terminology Understanding (Understanding domain-specific terms and conceptual meaning)

Language Understanding (Understanding syntax and query structure)

Knowledge Graph

(Understanding conceptual and logical relationships between domain-specific entities)

Google calls on Bert to look at the little words in effort to improve its answers

RICHARD WATERS - SAN FRANCISCO

Google is making one of its biggest changes to its ranking algorithm as it gives artificial intelligence a deepening role in the world's most popular search engine.

The change, which Google described as its most significant revision in at least five years, uses a new form of language analysis to understand users' queries better. It is set to affect its responses to one in 10 searches, according to Pandu Nayak, vice-president of search.

Google began to update its software in recent days to make the change for English language searches and will eventually apply it to other languages as well.

Until now, Google's algorithm has tried to single out the most important words in any search query, ignoring smaller or common words that seem less significant. This enables it to zero in on the main subject, but often results in it misunderstanding a precise request.

The new technique, known as Bert, relies on a language model built up from analysis of vast amounts of text online. Rather than read a string of search words sequentially it analyses them all at the same time — including smaller words that would have been ignored before.

One example Google gave of the types of question it could now handle — "How old was Taylor Swift when Kanye jumped on the stage?" — points to more complex queries that have been beyond its reach before.

since Google was founded 21 years ago periodic updates to its ranking algorithm have sometimes resulted in big changes in the amount of traffic it directs to outside websites, angering companies whose businesses depend on the search engine. But Google predicted the latest revision would not have a large impact on online traffic

The update marks the first application of a piece of research on natural language processing from last year that has drawn considerable attention in AI circles. Understanding language is one of the most difficult problems in AI given the fluidity of language depending on the context and the person using it.

The search company said the Bert technique would return more useful responses to many queries. But it also said that in some instances the new algorithm had produced worse results than before. Jeff Dean, Google's head of AI, said: "I don't think we're anywhere close to solving language, but this is a good step."

Powerless

California fires

The revolut of 1989 Misha Glenny on the downfall

of the Soviet Union OFFE ARTS

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Can my wife drive on my insurance?

Does My Husband Have To Be On My Car Insurance?

Can Husband And Wife Have Separate Auto Insurance?
Does My Spouse Have To Be On My Auto Insurance?
Can My Girlfriend Be On My Car Insurance?
Does My Spouse Have My Auto Insurance?
Can I Keep My Ex Wife On My Car Insurance?
Is My Spouse Covered On My Auto Insurance?
Can I Add My Girlfriend To My Auto Insurance Policy

Can I Get Car Insurance Without My Husband?

Does My Husband Have To Be On My Auto Insurance If He Doesn't Drive The Car?

Score: 0.85

The answer is yes, unless your husband has coverage on a separate auto insurance policy. In Wisconsin at least every person in the household must be listed on an auto insurance policy or they must be listed on anouther auto insurance policy. Typically having more than one person listed on a policy will lessen the cost of coverage as opposed to having separate policies. Contact your local agent to find out the premium difference between a combined policy and individual policies.

Score: 0.81

The answer to this question could be yes or no, depending on a variety of factors. If your husband has an individual policy, not related to his employer, than you can certainly remain on his policy. If your husband's coverage is employer sponsored, then the rules set by the employer's plan would be the guiding factor in this answer.

Score: 0.81

In California all licensed drivers residing in the household must be included on the car policy. That information is more readily available and your insurance company and agent probably know about your wife and her driving record. She can, however, be excluded. In that case the insurance company is not obligated to honor any claim when your wife is operating the car. Of course, that is a very scary proposition.

Score: 0.8

Yes, it is possible; but is typically frowned upon. Typically you are able to secure better rates by having both husband and wife on the same policy. If your husband secures auto insurance on his own then it would be possible for you to do the same. Not all carriers will allow this though. Contact a local agent to discuss your coverage needs and pricing in your area.

Score: 0.79

+

Based on my education and experience in the industry up to this point, yes it is possible for a husband and wife to have separate auto insurance. This practice is frowned upon though by the carriers and will typically make the cost of the auto insurance more costly for both the husband and the wife.

Score: 0.77

Search

Thought Vectors

Word Embeddings: [5, 1, 3, 4, 2, 1, 5, 3] [4, 1, 3, 0, 1, 1, 4, 2]

Sentence Embeddings: [2, 3, 2, 4, 2, 1, 5, 3] [5, 3, 2, 3, 4, 0, 3, 4]

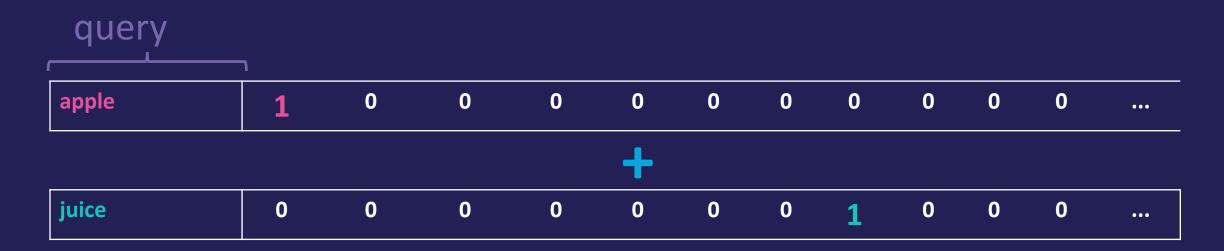
Paragraph Embeddings: [5, 1, 4, 1, 0, 2, 4, 0] [1, 1, 4, 2, 1, 0, 0, 0]

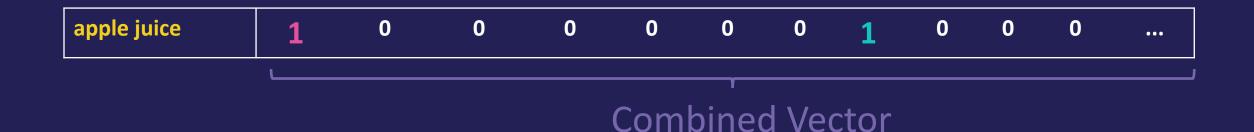
Document Embedding: ____ [4, 1, 4, 2, 1, 2, 4, 3]

Single Term Searches (as a Vector)

query	exact term lookup in inverted index											
· · ·	apple	caffeine	cheese	coffee	drink	donut	food	juice	pizza	tea	water	term N
cappuccino	0	0	0	0	0	0	0	0	0	0	0	••••
apple	1	0	0	0	0	0	0	0	0	0	0	
juice	0	0	0	0	0	0	0	1	0	0	0	
cheese	0	0	1	0	0	0	0	0	0	0	0	
pizza	0	0	0	0	0	0	0	0	1	0	0	
donut	0	0	0	0	0	1	0	0	0	0	0	
green	0	0	0	0	0	0	0	0	0	0	0	
tea	0	0	0	0	0	0	0	0	0	1	0	
bread	0	0	1	0	0	0	0	0	0	0	0	
sticks	0	0	0	0	0	0	0	0	0	0	0	••••

Multi-term Query Vectors





Multi-term Searches

query	exact term lookup in inverted index											
· · ·	apple	caffeine	cheese	coffee	drink	donut	food	juice	pizza	tea	water	term N
latte	0	0	0	0	0	0	0	0	0	0	0	
cappuccino	0	0	0	0	0	0	0	0	0	0	0	
apple juice	1	0	0	0	0	0	0	1	0	0	0	
cheese pizza	0	0	1	0	0	0	0	0	1	0	0	
donut	0	0	0	0	0	1	0	0	0	0	0	
soda	0	0	0	0	0	0	0	0	0	0	0	
green tea	0	0	0	0	0	0	0	0	0	1	0	
water	0	0	0	0	0	0	0	0	0	0	1	
cheese bread sticks	0	0	1	0	0	0	0	0	0	0	0	
cinnamon sticks	0	0	0	0	0	0	0	0	0	0	0	

avaat tarm laakun in invartad indav

Dimensionality Reduction

	food	drink	dairy	bread	caffeine	sweet	calories	healthy
apple juice	0	5	0	0	0	4	4	3
cappuccino	0	5	3	0	4	1	2	3
cheese bread sticks	5	0	4	5	0	1	4	2
cheese pizza	5	0	4	4	0	1	5	2
cinnamon bread sticks	5	0	1	5	0	3	4	2
donut	5	0	1	5	0	4	5	1
green tea	0	5	0	0	2	1	1	5
latte	0	5	4	0	4	1	3	3
soda	0	5	0	0	3	5	5	0
water	0	5	0	0	0	0	0	5

Vector Similarity Scoring

Vector:

Phr	ase:
apple	juice:

pple juice:	[0, 5, 0, 0, 0, 4, 4, 3
appuccino:	[0,5,3,0,4,1,2,3
heese bread sticks:	[5, 0, 4, 5, 0, 1, 4, 2
heese pizza:	[5, 0, 4, 4, 0, 1, 5, 2
innamon bread sticks:	[5, 0, 4, 5, 0, 1, 4, 2
onut:	[5, 0, 1, 5, 0, 4, 5, 1
reen tea:	[0,5,0,0,2,1,1,5
atte:	[0,5,4,0,4,1,3,3
oda:	[0, 5, 0, 0, 3, 5, 5, 0
vater:	[0, 5, 0, 0, 0, 0, 0, 5

Vector Similarity (a, b): $cos(\theta) = \frac{a \cdot b}{|a| \times |b|}$

Ranked Results: Cheese Pizza		Ranked Results: Green Tea	
0.99	cheese bread sticks	0.94	water
0.91	cinnamon bread sticks	0.85	cappuccino
0.89	donut	0.80	latte
0.47	latte	0.78	apple juice
0.46	apple juice	0.60	soda
•••			
0.19	water	0.19	donut

Performance Considerations

Problem: Vector Scoring is Slow

• Unlike keyword search, which looks up pre-indexed answers to queries, Vector Search must instead calculate similarities between the query vector and every document's vectors to determine best matches, which is slow at scale.

Solution: Quantized Vectors

- "Quantization" is the process for mapping vectors features to discrete values.
- Creating "tokens" which map to a similar vector space, enables matching on those tokens to perform an ANN (Approximate Nearest Neighbor) search
- This enables converting vector scoring into a search problem (term lookup and scoring), which is fast again, at the expense of some recall and scoring accuracy

Recommended Approach: Quantized Vector Search + Vector Similarity Reranking

• Combine the best of both worlds by running an initial ANN search on a quantized vector representation, and then re-rank the top-N results using full Vector similarity scoring.

Solr Implementation Options

Option 1: Streaming Expressions

Streaming Expressions

Send Documents to Solr:

```
curl -X POST -H "Content-Type: application/json" \
http://localhost:8983/solr/food/update?commit=true \
 --data-binary '
 {"id": "1", "name s": "donut", "vector fs": [5.0,0.0,1.0,5.0,0.0,4.0,5.0,1.0]},
 {"id": "2", "name s":"apple juice",
             "vector fs": [1.0,5.0,0.0,0.0,0.0,4.0,4.0,3.0] },
 {"id": "3", "name s":"cappuccino",
             "vector fs": [0.0,5.0,3.0,0.0,4.0,1.0,2.0,3.0]},
 {"id": "4", "name s":"cheese pizza",
             "vector fs": [5.0,0.0,4.0,4.0,0.0,1.0,5.0,2.0] },
 {"id": "5", "name s":"green tea",
             "vector fs": [0.0,5.0,0.0,0.0,2.0,1.0,1.0,5.0]},
 {"id": "6", "name s":"latte", "vector fs":[0.0,5.0,4.0,0.0,4.0,1.0,3.0,3.0]},
 {"id": "7", "name s":"soda", "vector fs":[0.0,5.0,0.0,0.0,3.0,5.0,5.0,0.0]},
 {"id": "8", "name s":"cheese bread sticks",
             "vector fs": [5.0,0.0,4.0,5.0,0.0,1.0,4.0,2.0] },
 {"id": "9", "name s":"water", "vector fs": [0.0,5.0,0.0,0.0,0.0,0.0,0.0,5.0]},
 {"id": "10", "name s":"cinnamon bread sticks",
              "vector fs": [5.0,0.0,1.0,5.0,0.0,3.0,4.0,2.0] }
```

Solr	Streaming Expression (expr) top(select(search(food, q="*:*", fl="id,vector_fs,name_s", sort="id asc"),
Dashboard	cosineSimilarity(vector_fs, array(5.2,0.0,1.0,5.0,0.0,4.0,5.0,1.0)) as cos, id, name_s), n=5,
ڬ Logging	sort="cos desc"
Collections	Execute in with explanation
<u>[</u>] Java Properties	http://localhost:8983/solr/food/stream?expr=top(select(search(food%2C%20q%3D%22*%3A*%22%2C%20fl%3D%22id%2Cvector_fs%2Cname_s%22%2C%2
📄 Thread Dump	<pre>{ "result-set": {</pre>
Suggestions	"docs": [
food •	{ "name_s": "donut", "cos": 0.99984612,
1 Overview	"id": "1"
T Analysis	}, {
🛃 Dataimport	"name_s": "cinnamon bread sticks",
Documents	"cos": 0.98628004, "id": "10"
📙 Files	},
🔎 Query	{ "name_s": "cheese pizza",
ିଅଟ୍ଟି Stream	"cos": 0.89078353,
🕒 Schema	"id": "4"
	}, {

Option 2: Streaming Expressions Query Parser

Streaming Expressions Query Parser

Request:

```
http://localhost:8983/solr/food/select?q=*:*&fl=id,name_s&
fq={!streaming_expression}top(
    select(
        search(food, q="*:*", fl="id,vector_fs", sort="id asc"),
        cosineSimilarity(vector_fs, array(5.1,0.0,1.0,5.0,0.0,4.0,5.0,1.0)) as cos, id),
        n=5, sort="cos desc"
```

Response:

```
{ "responseHeader":{
    ... },
    "response":{"numFound":5,"start":0,"docs":[
        { "name_s":"donut", "id":"1"},
        { "name_s":"apple juice", "id":"2"},
        { "name_s":"cheese pizza", "id":"4"},
        { "name_s":"cheese bread sticks", "id":"8"},
        { "name_s":"cinnamon bread sticks", "id":"10"}]
}}
```

Option 3: Solr Vector Scoring Plugin

Solr Vector Scoring Plugin

Send Documents to Solr:

curl -X POST -H "Content-Type: application/json"
http://localhost:8983/solr/{your-collection-name}/update?commit=true -data-binary `

{"name":"example 0", "vector":"0|1.55 1|3.53 2|2.3 3|0.7 4|3.44 5|2.33"},
{"name":"example 1", "vector":"0|3.54 1|0.4 2|4.16 3|4.88 4|4.28 5|4.25"},
{"name":"example 2", "vector":"0|1.11 1|0.6 2|1.47 3|1.99 4|2.91 5|1.01"},
{"name":"example 3", "vector":"0|0.06 1|4.73 2|0.29 3|1.27 4|0.69 5|3.9"},
{"name":"example 4", "vector":"0|4.01 1|3.69 2|2 3|4.36 4|1.09 5|0.1"},
{"name":"example 5", "vector":"0|0.64 1|3.95 2|1.03 3|1.65 4|0.99 5|0.09"}

Solr Vector Scoring Plugin

Request:

http://localhost:8983/solr/{your-collection-name}/query?fl=name,score,vector&q={!vp f=vector vector="0.1,4.75,0.3,1.2,0.7,4.0"

Response:

```
{ "responseHeader":{ "status":0, "QTime":1}},
```

```
"response":{ "numFound":6,"start":0,"maxScore":0.99984086,
```

```
"docs":[
```

```
{ "name":["example 3"], "vector":["0|0.06 1|4.73 2|0.29 3|1.27 4|0.69 5|3.9 "],
```

```
"score":0.99984086},
```

```
{ "name":["example 0"], "vector":["0|1.55 1|3.53 2|2.3 3|0.7 4|3.44 5|2.33 "], "score":0.7693964},
{ "name":["example 5"], "vector":["0|0.64 1|3.95 2|1.03 3|1.65 4|0.99 5|0.09 "], "score":0.76322395},
{ "name":["example 4"], "vector":["0|4.01 1|3.69 2|2 3|4.36 4|1.09 5|0.1 "], "score":0.5328145},
{ "name":["example 1"], "vector":["0|3.54 1|0.4 2|4.16 3|4.88 4|4.28 5|4.25 "], "score":0.48513117},
{ "name":["example 2"], "vector":["0|1.11 1|0.6 2|1.47 3|1.99 4|2.91 5|1.01 "], "score":0.44909418}]
}}
```

Option 4: Solr Vector Scoring + LSH Plugin

Solr Vector Scoring + LSH Plugin

Send Documents to Solr:

curl -X POST -H "Content-Type: application/json" http://localhost:8983/solr/{your-collectionname}/update?update.chain=LSH&commit=true --data-binary '

{"id":"1", "vector":"1.55,3.53,2.3,0.7,3.44,2.33"}, {"id":"2", "vector":"3.54,0.4,4.16,4.88,4.28,4.25"}

Request:

http://localhost:8983/solr/{your-collection-name}/query?fl=name,score,vector&q={!vp f=vector vector="1.55,3.53,2.3,0.7,3.44,2.33" lsh="true" reRankDocs="5"}&fl=name,score,vector_vector_lsh_hash_

Solr Vector Scoring + LSH Plugin

Request:

http://localhost:8983/solr/{your-collection-name}/query?fl=name,score,vector&q={!vp f=vector vector="1.55,3.53,2.3,0.7,3.44,2.33" lsh="true" reRankDocs="5"}&fl=name,score,vector, vector , lsh hash

Response:

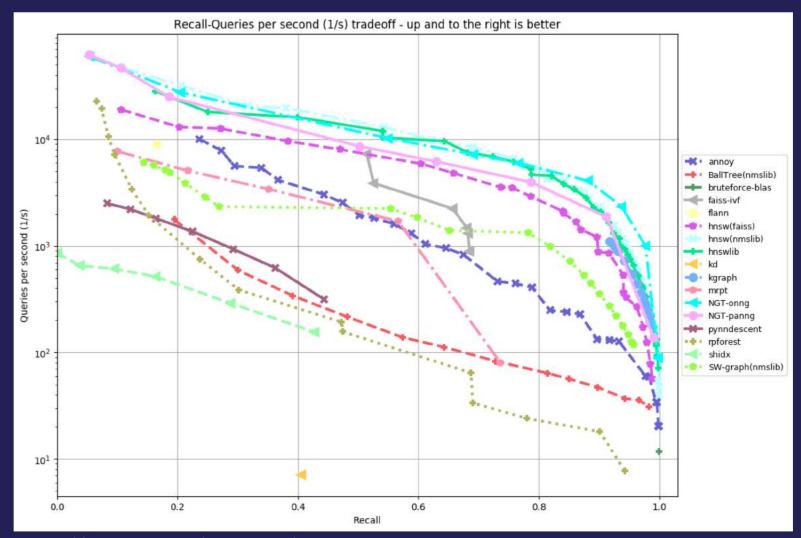
"responseHeader":{ "status":0, "QTime":8, "response":{"numFound":1,"start":0,"maxScore":36.65736, "docs":[

- { "id": "1", "vector": "1.55,3.53,2.3,0.7,3.44,2.33",
- "_vector_":"/z/GZmZAYeuFQBMzMz8zMzNAXCj2QBUeuA==",
- "_lsh_hash_":["0_8", "1_35", "2_7", "3_10", "4_2", "5_35", "6_16", "7_30", "8_27", "9_12", "10_7", "11_32", "12_48", "13_36", "14_10", "15_7", "16_42", "17_5", "18_3", "19_2", "20_1", "21_0", "22_24", "23_18", "24_42", "25_31", "26_35", "27_8", "28_1", "29_24", "30_47", "31_14", "32_22", "33_39", "34_0", "35_34", "36_34", "37_39", "38_27", "39_27", "40_45", "41_10", "42_21", "43_34", "44_41", "45_9", "46_31", "47_0", "48_4", "49_43"], "score":36.65736}

Option 5 (Work in Progress): First-class Vector Fields in Lucene/Solr

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= /	APACHE SOFTWARE FOUNDATION	Dashboards 🗸	Projects 🗸 Issues 🗸		≡	APACHE SOFTWARE FOUNDATION	Dashboards 🗸	Projects 🗸 Issues	~	Search
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N 3	Solr / SOLR-1: Vector Se		(Umbrella Issu	le)	N		/ LUCENE-9004	vector search	1	
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	Type: Priority: Affects Version/s: Component/s: Labels: Description	New Feature	Status: Resolution: Fix Version/s:	OPEN Unresolved None		Type: Priority: Affects Version/s: Component/s: Labels: Lucene Fields:	 New Feature Major None None None None None New 	Status: Resolution: Fix Version/s:	OPEN Unresolved None	Assignee: Unassign Reporter: Michael S Votes: 3 Vote for t
	 We have recently corsolr, and have even we chose to use the suppicked can be easily in a binary field. Perhaps an addition of same properties to coshould be considered ✓ Issue Links is duplicated by SOLR-13500 Verticates to 	vorked on a so all of erbit algorithe, but chagned), are store of an LSH URP in co alculate LSH(or may	DC: Ve used an URP the code is designed access to ring ther njunction with a query be ktree, or some othe	a va th algorithm	n	te ns, queries and d er ine. SOLR- 2890 vector-based scoring The idea here is to ex found an approach b neighbor relation at r nearest neighbor cal- implementing HNSW approximate nearest search). At a high level the wa has a partial encodin long-distance links. I small world property	cument us b count of several on g v g functions. To is is a xplore approximate r based on navigating a multiple scales can p culations) at a reaso ((hierarchical naviga vector search (ofter ay this algorithm wor og of the nearest neig if this graph is built i), then you can effic	approaches withis bounoff issue from the nearest-neighbor sea a graph that partially provide accuracy > 99 mable cost. This issue able small-world) graph n referred to as KNN rks is this. First assur- ghbor relation, with s n the right way (has to iently traverse it to fin	ure for a podern's prch including providing hat. Arch. Researchers have encodes the nearest 5% (as compared to exact e will explore ons for the purpose of or k-nearest-neighbor and you have a graph that ome short and some the hierarchical navigable and nearest neighbors	 Watchers: 11 Start wat Dates Created: 4 days ago Updated: Yesterday
»	links to				»	this idea was pioneer	red in [1]. The great			

ANN Benchmarks (Approximate Nearest Neighbor)



https://github.com/erikbern/ann-benchmarks

Vector Encoders

Vector Encoders

- Take queries, documents, sentences, paragraphs, etc. and transform them into vectors.
- Usually leverage deep learning, which can discover rich language usage rules and map them to combinations of features in the vector
- Popular Libraries:
 - Bert
 - Elmo
 - Universal Sentence Encoder
 - Word2Vec
 - Sentence2Vec
 - Glove
 - fastText
 - many more ...

Google calls on Bert to look at the little words in effort to improve its answers

RICHARD WATERS - SAN FRANCISCO

Google is making one of its biggest changes to its ranking algorithm as it gives artificial intelligence a deepening role in the world's most popular search engine.

The change, which Google described as its most significant revision in at least five years, uses a new form of language analysis to understand users' queries better. It is set to affect its responses to one in 10 searches, according to Pandu Nayak, vice-president of search.

Google began to update its software in recent days to make the change for English language searches and will eventually apply it to other languages as well.

Until now, Google's algorithm has tried to single out the most important words in any search query, ignoring smaller or common words that seem less significant. This enables it to zero in on the main subject, but often results in it misunderstanding a precise request.

The new technique, known as Bert, relies on a language model built up from analysis of vast amounts of text online. Rather than read a string of search words sequentially it analyses them all at the same time — including smaller words that would have been ignored before.

One example Google gave of the types of question it could now handle — "How old was Taylor Swift when Kanye jumped on the stage?" — points to more complex queries that have been beyond its reach before.

Since Google was founded 21 years ago periodic updates to its ranking algorithm have sometimes resulted in big changes in the amount of traffic it directs to outside websites, angering companies whose businesses depend on the search engine. But Google predicted the latest revision would not have a large impact on online traffic.

The update marks the first application of a piece of research on natural language processing from last year that has drawn considerable attention in Al circles. Understanding language is one of the most difficult problems in Al given the fluidity of language depending on the context and the person using it.

The search company said the Bert technique would return more useful responses to many queries. But it also said that in some instances the new algorithm had produced worse results than before. Jeff Dean, Google's head of AI, said: "I don't think we're anywhere close to solving language, but this is a good step."



The revolut of 1989 Misha Glenny on the downfall

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WeWork is 'bl



Powerless

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Keyword Search vs. Vector Search

Query Type	Likely Outcome
Obscure keyword combinations <i>Q. (software OR hardware) AND enginee*</i>	 Keyword search succeeds Vector Search fails
Natural Language Queries Q. Can my wife drive on my insurance?	 Keyword search might get lucky, but probably fails Vector Search succeeds

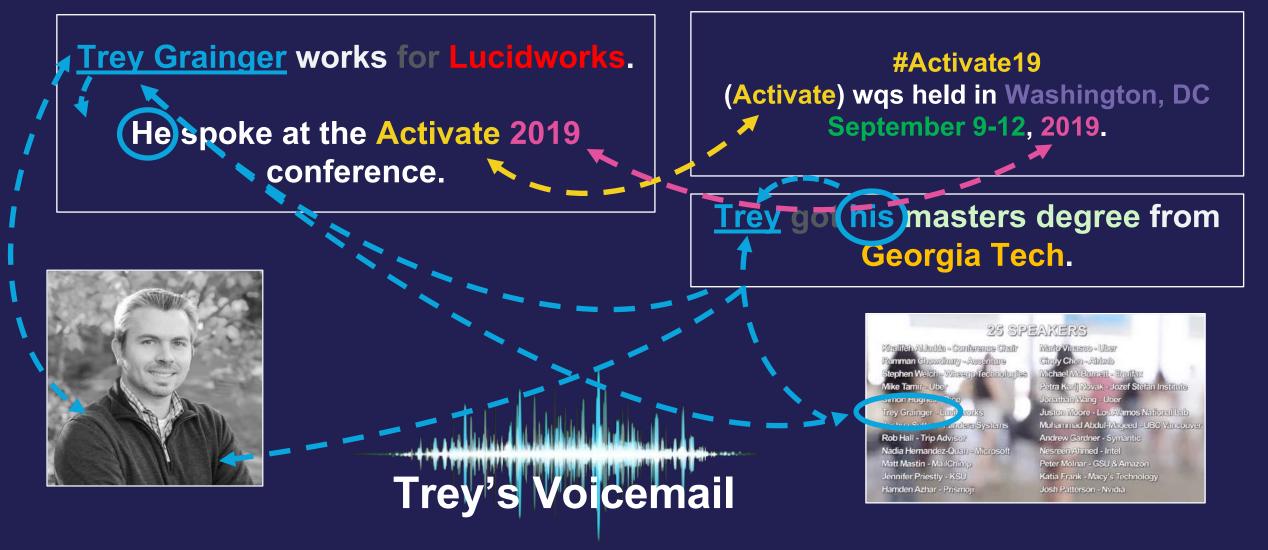
Fuzzy Language Queries Q. famous french tower

Structured Relationship Queries *Q. popular bbq near Activate*

- Keyword search mismatch yields poor results
- Vector Search succeeds
- Keyword search fails
- Vector search fails
- Need a Knowledge Graph!

Giant Graph of Relationships...





Semantic Knowledge Graph

Documents

id: 1
job_title: Software Engineer
desc: software engineer at a
great company
skills: .Net, C#, java

id: 2

job_title: Registered Nurse desc: a registered nurse at hospital doing hard work skills: oncology, phlebotemy

id: 3

job_title: Java Developer
desc: a software engineer or a
java engineer doing work
skills: java, scala, hibernate

Docs-Terms Forward Index

field	doc	term
	1	а
	T	at
		company
		engineer
		great
		software
	2	а
	Ζ	at
		doing
		hard
desc		hospital
		nurse
		registered
		work
	3	а
	5	doing
		engineer
		java
		or
		software
		work
job_title	1	Software Engineer

Terms-Docs Inverted Index

field	term	postings list		
		doc	pos	
		1	4	
	а	2	1	
		3	1, 5	
		1	3	
	at	2	4	
	company	1	6	
	daina	2	6	
	doing	3	8	
	•	1	2	
	engineer	3	3, 7	
desc	great	1	5	
	hard	2	7	
	hospital	2	5	
	java	3	6	
	nurse	2	3	
	or	3	4	
	registered	2	2	
	software	1	1	
	soitware	3	2	
	work	2	10	
	work	3	9	
job_title	java developer	3	1	



Source: Trey Grainger, Khalifeh AJJadda, Mohammed Korayem, Andries Smith. "The Semantic Knowledge Graph: A compact, auto-generated model for real-time traversal and ranking of any relationship within a domain". DSAA 2016.

Related term vector (for query concept expansion)

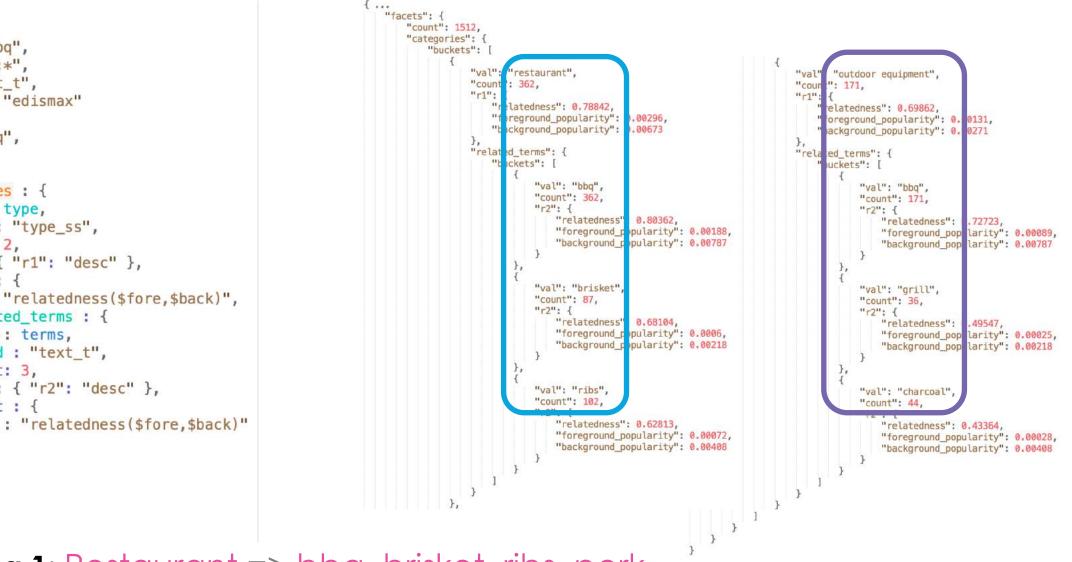
http://localhost:8983/solr/stack-exchange-health/skg

```
{
{
 "queries": [
    "advil"
 ],
  "compare": [
      "type": "Body_t",
      "limit": 20,
      "sort": "relatedness"
```

```
"data": [
        "type": "Body_t",
        "values": [
                "name": "advil",
                "relatedness": 0.62076,
                "popularity": 909,
                "foreground_popularity": 909,
                "background_popularity": 909
            },
                "name": "motrin",
                "relatedness": 0.51264,
                "popularity": 649,
                "foreground_popularity": 649,
                "background_popularity": 779
            },
                "name": "aleve",
                "relatedness": 0.4601,
                "popularity": 390,
                "foreground_popularity": 390,
                "background_popularity": 390
            },
                "name": "naproxen",
                "relatedness": 0.26633,
                "popularity": 520,
                "foreground_popularity": 520,
                "background_popularity": 2598
            },
                "name": "ibuprofen",
                "relatedness": 0.2508.
```

Disambiguation by Category Example

params: { fore: "bbg" back: "*:*" qf: "text_t", defType: "edismax" }, query: "bbq", limit: 0. facet: { categories : { type : type, field : "type_ss", limit: 2, sort: { "r1": "desc" }, facet : { r1 : "relatedness(\$fore,\$back)", related terms : { type : terms, field : "text t", limit: 3, sort: { "r2": "desc" }, facet : { r2 : "relatedness(\$fore,\$back)"



Meaning 1: Restaurant => bbq, brisket, ribs, pork, ... Meaning 2: Outdoor Equipment => bbq, grill, charcoal, propane, ...

← Query Rewriting Dashboard > Misspelling Detection

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Q Search

Misspelling Detection SIMULATOR >>

92 unpublished changes

PUBLISH

STATUS

Fusion replaces misspelled query terms with their spelling corrections for better search results. Schedule the Token and Phrase Spell Correction job to get AI-generated spelling corrections based on your signals collection.

PUB	LISF	IED	
		870)

Auto (30962)

No (92)

CONFIDENCE

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0

Misspelling	Suggested Correction	Confidence	Edited On $oldsymbol{\psi}$	Status	Published
laptopo	laptop	0.5	04/09/19 08:39	Auto ^	Yes
joybird	jaybird	1	04/09/19 08:39	Approved	Yes
e case	encase	0.5	04/09/19 08:39	Denied	Yes
shredde	shredder	0	04/09/19 08:39	Auto 🛩	Yes
flatsceen	flatscreen	1	04/09/19 08:39	Auto 🗸	Yes
i live	ilive	1	04/09/19 08:39	Auto 🗸	Yes
antholigy	anthology	1	04/09/19 08:39	Auto 🗸	Yes
dazzale	dazzle	1	04/09/19 08:39	Auto 🗸	Yes
rep blic	republic	1	04/09/19 08:39	Auto 🗸	Yes
reveng	revenge	0.5	04/09/19 08:39	Auto 🗸	Yes
acvessory	accessory	0.5	04/09/19 08:39	Auto 🗸	Yes
storagw	storage	0.5	04/09/19 08:39	Auto 🗸	Yes
correl	corel	1	04/09/19 08:39	Auto 🗸	Yes
beat bry	beat by	0.5	04/09/19 08:39	Auto 🗸	Yes
a droid	android	0.5	04/09/19 08:39	Auto 🗸	Yes
flyerhtc	flyer htc	1	04/09/19 08:39	Auto 🗸	Yes

Displaying 1 - 20 of 30962 results

Previous 1 2 3 Next >



Clear

Search with Query Rewrite Personalization Aggregation Compare Brand Info Query rewrite triggered for spell laptopo, replace with laptop Search... Modified query is samsung black laptop Samsung 2,280 Search results are based on dcommerce_main2_query_rewrite pipeline HP 763 Sony 745 samsung black laptopo × Tribeca 736 GE 727 502 Lenmar SHOWING 1 - 10 OF 34,186 * Debug Whirlpool 461 Toshiba 441 2 3 4 5 next Frigidaire 397 Dell 371 **KitchenAid** 300

Sort by relevance -

This laptop features a 640GB hard drive for abundant storage and a built-in high-speed wireless LAN for an easy Internet connection.Windows 7 Home Premium installed Learn more. Dept Computers Brand Samsung Color Black

ID 3603163

294

6,098

5,571

5,163

4,058

3,342

2,414

Metra

Show more

Search...

Computers

Appliance

Audio

Department

Photo/Commodities

Video/Compact Disc

Digital Communication

\$646 Score 22.64946



This Samsung NP305V5A-A09US laptop features a 1.3MP high-definition webcam for video chatting and wireless networking for a quick and easy connection to the Internet. The double-layer DVD±RW/CD-RW drive allows you to create customized CDs and DVDs.Windows 7 Home Premium installed Learn more.

Dept Computers Brand Samsung Color Black

ID 4700152 \$556 Score 22 105953



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Q Search

Synonym Detection SIMULATOR >>

9 278 unpublished changes

PUBLISH

ÎÌ,

STATUS Pending (278) Approved (1) Denied (1)

PUBLISHED Yes (2) No (278)

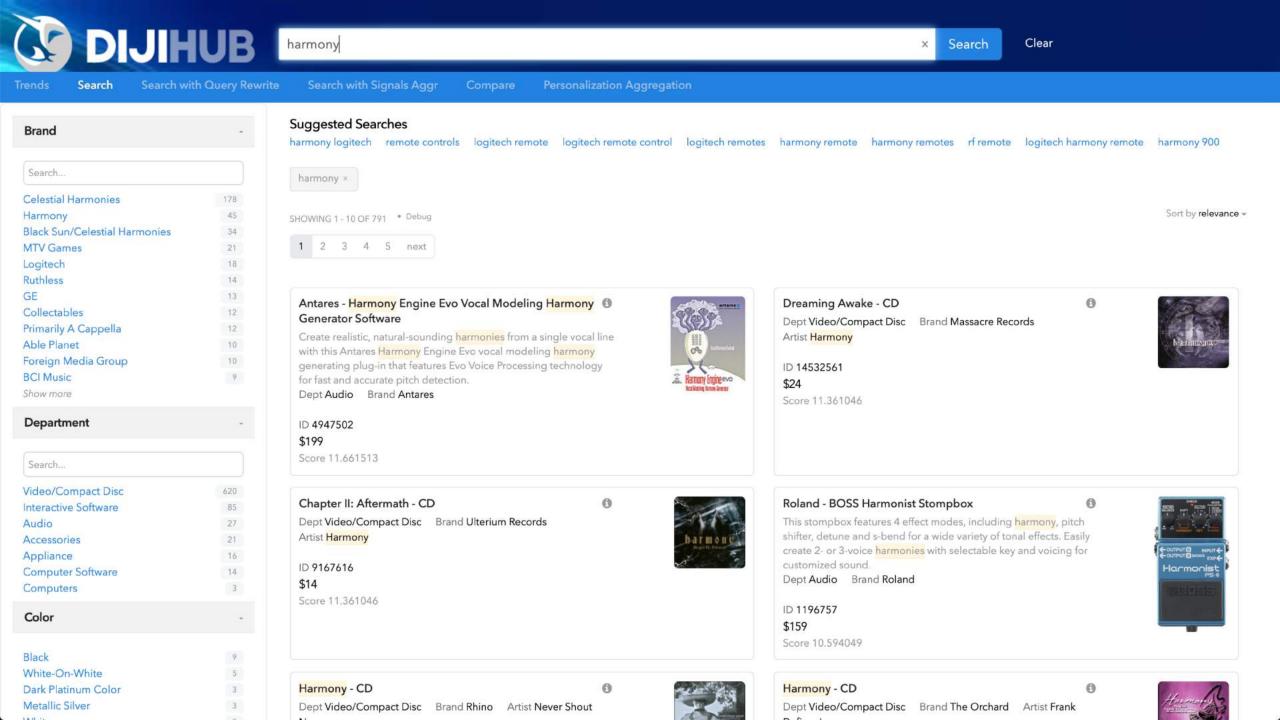
CONFIDENCE

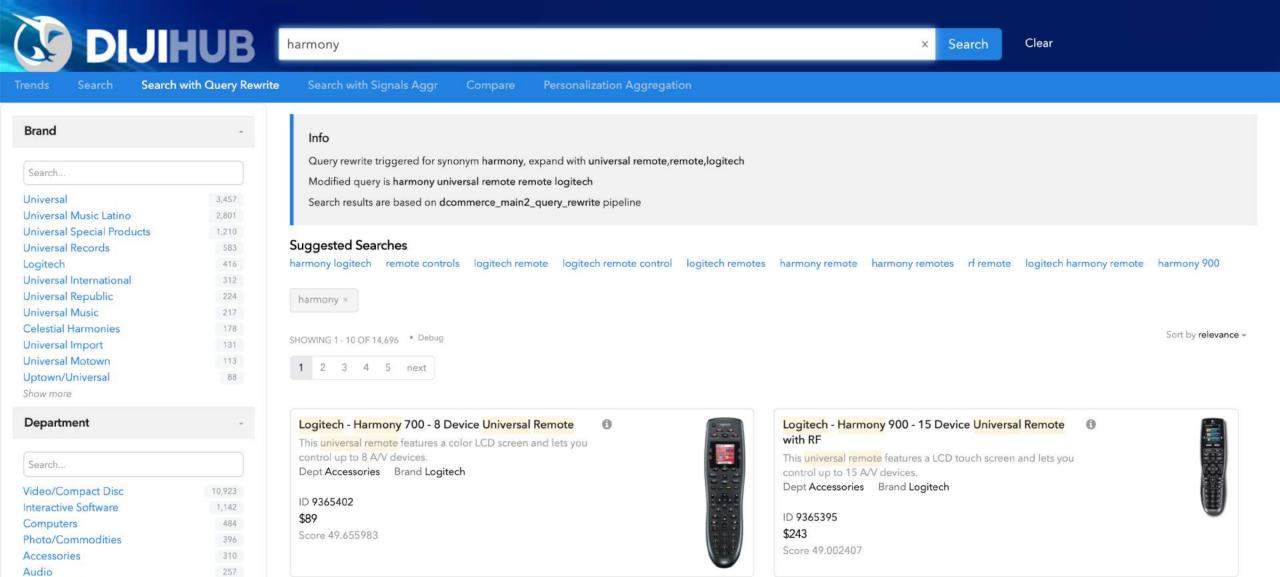
0

Search Term		Synonyms	Confidence	Edited On ${f V}$	Status	Published
narmony	→*-	universal remote, remote, logitech	0.02843	04/19/19 10:21	Approved V	Yes
nusic	→←	reciever, piano	0.16679	04/16/19 16:13	Denied ~	Yes
peaker	→*-	wireless speaker	0.02403		Pending ^	No
blueray	→←	complete, movies	0.02130		Approved	No
iteracy	→*-	logictech	0.1875		Denied	No
neadset	→*-	wireless headset	0.02892		Pending 💙	No
ndr	→*-	hdrcx	0.01718		Pending 😽	No
nvisibleshield	→←	zagg	0.01199		Pending 👻	No
loor	→←	vacuum	0.06024		Pending 👻	No
screen	→←	screen protector	0.17502		Pending 👻	No
audio cable	→←	cable	0.01482		Pending 😽	No
gps	→←	navigation	0.01358		Pending 😽	No
cam	→←	camcorder, video camera, camra	0.01665		Pending 👻	No
computer speaker	→←	promedia	0.05882		Pending 👻	No
clean	→+-	lens	0.03759		Pending 🗸	No

Displaying 1 - 20 of 280 results

< Previous 1 2 3 Next >





Logitech Harmony 1100 Universal Remote & Wireless Extender Package

Take command of your expansive home entertainment center with the Logitech Harmony 1100 15-device universal remote. The package also includes a free Logitech wireless extender, which gives your remote a 100' range through cabinets and walls. Plus, save \$50 on the remote.



0

Logitech - Harmony 670 Advanced Universal Remote with LCD Screen

Stop fumbling with multiple remotes. This universal remote lets you take control of your home entertainment system with the help of a clear, interactive LCD screen. Learn about the advantages of Harmony Setup and Programming Resources. Click here to see more products from Logitech.

Dept Accessories Brand Logitech



Color

Video

Show more

Mobile Audio

Digital Communication

Magnolia Home Theatr

Computer Software

Appliance

248 211

148

118

48

-

Example Query:

bbq near haystack eu

```
"responseHeader":{
  "status":0,
  "QTime":0,
  "params":{
    "sort": "popularity desc",
    "matchText":"true",
    "json.nl":"map",
    "field": "name tag",
    "echoParams":"all",
    "fl":"id,type,canonical form,name,country:countrycode s,admin area:admin code 1 s,popularity,* p,command function"}},
"tagsCount":2,
"tags":[{
    "startOffset":4,
    "endOffset":8,
    "matchText": "near",
    "ids":["5",
      "1",
      "6"]},
    "startOffset":9,
    "endOffset":20,
    "matchText": "haystack eu",
                                                  bbg near haystack eu
    "ids":["27"]}],
"response":{"numFound":4,"start":0,"docs":[
      "type": "command",
      "popularity":10.0,
                                                         TAGGED: bbg {near} {haystack eu}
      "command_function":"cmd_text_distance(query
      "canonical form":"{text distance}",
      "id":"5"},
      "type": "command",
                                                         PARSED: { type:keyword, known: false, surface form: "bbg"}
      "popularity":90.0,
                                                         {"type":"command","popularity":90,"command function":"cmd location distance(query,
      "command function": "cmd location distance(
                                                         position)","canonical form":"{location distance}","id":"1"}
      "canonical form":"{location distance}",
                                                         {"type":"event","popularity":100,"canonical_form":"haystack eu conference","id":"27"}
      "id":"1"},
      "type": "event",
      "popularity":100.0,
                                                         TO SOLR: bbg^0.91032 ribs^0.65674 brisket^0.63386 doc_type:"restaurant" {!geofilt d=50
      "canonical_form": "haystack eu conference",
                                                         sfield="coordinates_pt" pt="52.554930,13.383930"}
      "id":"27"},
    {
      "type": "command",
      "popularity":10.0,
      "command function": "cmd text distance(query, position)",
      "canonical form":"{text distance}",
      "id":"6"}]
}}
```

{

```
"responseHeader":{
 "status":0,
 "OTime":0,
 "params":{
   "sort": "popularity desc",
   "matchText":"true",
   "json.nl":"map",
   "field": "name tag",
   "echoParams":"all",
   "fl":"id,type,canonical form,name,country:countrycode s,admin area:admin code 1 s,popularity,* p,command function"}},
"tagsCount":2,
"tags":[{
    "startOffset":4,
   "endOffset":8,
   "matchText": "near",
   "ids":["5",
     "1",
      "6"]},
    "startOffset":9,
   "endOffset":17,
   "matchText": "activate",
                                                  bbg near activate
   "ids":["26"]}],
"response":{"numFound":4,"start":0,"docs":[
      "type": "command",
      "popularity":10.0,
                                                         TAGGED: bbg {near} {activate}
      "command function": "cmd text distance(quer
      "canonical form":"{text distance}",
      "id":"5"},
      "type": "command",
                                                         PARSED: { type:keyword, known: false, surface_form: "bbg"}
      "popularity":90.0,
                                                         {"type":"command","popularity":90,"command_function":"cmd_location_distance(query,
      "command function": "cmd_location_distance(
                                                         position)","canonical form":"{location distance}","id":"1"}
      "canonical form":"{location distance}",
                                                         {"type":"event","popularity":100,"canonical_form":"activate conference","id":"26"}
      "id":"1"},
      "type": "event",
      "popularity":100.0,
                                                         TO SOLR: bbg^0.91032 ribs^0.65674 brisket^0.63386 doc_type:"restaurant" {!geofilt d=50
      "canonical form": "activate conference",
                                                         sfield="coordinates_pt" pt="38.916120,-77.045220"}
      "id":"26"},
      "type": "command",
      "popularity":10.0,
      "command function": "cmd text distance(query, position)",
      "canonical form":"{text distance}",
      "id":"6"}]
```

Q

}}

{

Demo Data

[{

bbqnear activate

Places (also includes geonames database)

"id": "1". "type": ["event"], "name": "Activate Conference", "description": "The Search and AI Conference", "has location": true. "venue": "Washington Hilton" 'popularity': 96 "id": "1", "type": ["event"], "name": "Haystack Conference", "description": "The Search Relevance Conference", "has location": true, "venue": "Violet Crowne", "popularity": 95 }, "id": "13", "type": ["hotel"], "name": "Washington Hilton", "description": "Hilton in Washington, DC", Ubac locationly true "coordinates": "38.916120,-77.045220", "address": "1919 Connecticut Ave. NW, Washington DC, 20009" popularity . 02 }, "id": "3", "type": ["restaurant"],

[Web crawl of restaurant and product reviews sites]

Entities (includes search commands)

34, is, is, keyword, 100

1,near,"{location distance}",command,90,"cmd location distance(query, position)" 2, IN, {location distance; , command, 100, cmg location distance(query, position) 3, around, "{location_distance}", command, 100, "cmd_location_distance(query, position)" 4, by, "{location_distance}", command, 90, "cmd_location_distance(query, position)" 5, by, "{text within one edit distance}", 10, "cmd text within one edit distance(query, position)" 6, near, "{text distance}", command, 10, "cmd text distance(query, position)" 7, popular, {popular}, command, 100, "cmd_popularity(query, position)" 8, top, {popular}, command, 100, "cmd popularity(query, position)" 9, best, {popular}, command, 100, "cmd_popularity(query, position)" 10, good, {popular}, command, 100, "cmd popularity(query, position)" 11, hilton, hilton, brand, 100 12, washington hilton, washington hilton, hotel, 100 13, washington dc hilton, washington hilton, brand, 100 14. washington d.c. hilton washington hilton, hotel, 100 15, activate activate conference, event 100 16, activate conf, activate conference, event, 100 17, activate conference, activate conference, event, 100 18, activat, activate conference, event, 100 19, activat conf, activate conference, event, 100 20, activat conference, activate conference, event, 100 21, lucidworks accivate, activate conference, event, 100 22, violet, violet, color, 100 23, violet crowne, violet crowne, brand, 100 24, violet crowne charlottesville, violet crowne charlottesville, movie theater, 100 25, violet crown, violet crowne, brand, 100 26, violet crown charlottesville, violet crowne charlottesville, movie theater, 100 27, haystack, haystack conference, event, 100 28, haystack conf, haystack conference, event, 100 29, havstack conference, havstack conference, event, 100 30, heystack, haystack conference, event, 100 31, heystack conf, haystack conference, event, 100 32, heystack conference, haystack conference, event, 100 33, around, "{location_distance}", command, 100, "cmd_location_distance(query, position)"

Q

Solr Knowledge Graph Traversal Query

```
params: {
 unknown_phrase_1: "bbg", location_keyword_1: "activate",
 content_field: "text", doc_type_field: "type", name_field: "name_s", name_search_field: "name_t",
 venue_field: "venue", address_field: "address", coordinates_field: "coordinates",
 all docs: "*:*", echoParams: none
},
query: "{!edismax v=$all_docs}", limit: 0,
facet: {
 $unknown phrase 1: {
     type: query, query: "{!type=edismax qf=$content field v=$unknown phrase 1}",
     facet: {
         doc_type : {
           type : terms, field : ${doc_type_field}, limit: 1, sort: { "r1": "desc" },
           facet : {
             r1 : "relatedness($unknown phrase 1,$all docs)",
             related terms : {
               type : terms, field : "${content_field}", limit: 4, sort: { "r2": "desc" },
               facet : {
                 r2 : "relatedness($unknown phrase 1,$all docs)"
 },
 location_1: {
      type: query,
     query: "{!type=graph from=${name_field} to=${venue_field} returnOnlyLeaf=true}{!type=edismax gf=${name_search_field} v=${location_keyword_1}}",
      facet: {
         "venue": {
              type:terms, field:${name_field}, limit:1,
              facet: {
               "address": {type:terms, field:${address_field}},
               "coordinates": {type:terms, field:${coordinates_field}}
```

```
{
  "responseHeader":{
   "status":0,
    "QTime":0,
    "params":{
      "sort": "popularity desc",
      "matchText":"true",
      "json.nl":"map",
      "field": "name tag",
      "echoParams": "all",
      "fl":"id,type,canonical form,name,country:countrycode s,admin area:admin code 1 s,popularity,* p,command function"}},
  "tagsCount":3,
  "tags":[{
      "startOffset":0,
      "endOffset":3,
      "matchText": "top",
     "ids":["8"]},
      "startOffset":8,
      "endOffset":10,
      "matchText":"in",
      "ids":["2"]},
      "startOffset":11,
                                                     top bbg in berlin
      "endOffset":17,
      "matchText":"berlin",
      "ids":["2950159",
        "4930431",
                                                            TAGGED: {top} bbg {in} {berlin}
        "4556518",
        "5083330",
        "5245497",
        "4348460",
        "4500771",
                                                            PARSED: {"type":"command","popularity":100,"command_function":"cmd_popularity(query,
        "3614789",
                                                            position)","canonical_form":"{popular}","id":"8"} { type:keyword, known: false, surface_form:
        "3587266"]}],
                                                            "bbg"}"type":"command","popularity":100,"command_function":"cmd_location_distance(query,
  "response":{"numFound":11,"start":0,"docs":[
                                                            position)","canonical_form":"{location_distance}","id":"2"}
                                                            {"location_p":"52.52437,13.41053","type":"city","popularity":0.15354161869920396,"name":
        "location p":"52.52437,13.41053",
                                                            ["Berlin"],"canonical_form":"Berlin","id":"2950159","country":"DE","admin_area":"16"}
        "type":"city",
        "popularity":0.15354161869920396,
        "name":["Berlin"],
        "canonical form": "Berlin",
        "id":"2950159",
                                                            TO SOLR: {!func}mul(min(popularity,1),100) bbq^0.91032 ribs^0.65674 brisket^0.63386
        "country": "DE",
                                                            doc type:"restaurant" {!geofilt d=50 sfield="location p" pt="52.52437,13.41053"}
        "admin area":"16"},
        "type": "command",
        "popularity":100.0,
        "command function": "cmd popularity(query, position)",
        "canonical form":"{popular}",
        "id":"8"},
        "type": "command",
        "popularity":100.0,
        "command function": "cmd location distance(query, position)",
        "canonical form":"{location distance}",
        "id":"2"},
        "location p":"42.3812,-71.63701",
        "type":"city",
        "popularity":1.0853455319837705E-4,
        "name":["Berlin"],
        "canonical form": "Berlin",
```

Why this Semantic Nuance Matters

Q

bbq near activate

TAGGED: bbq {near} {activate}

PARSED: { type:keyword, known: false, surface_form: "bbq"} {"type":"command","popularity":90,"command_function":"cmd_location_distance(query, position)","canonical_form":"{location_distance}","id":"1"} {"type":"event","popularity":100,"canonical_form":"activate conference","id":"26"}

TO SOLR: bbq^0.91032 ribs^0.65674 brisket^0.63386 doc_type:"restaurant" {!geofilt d=50 sfield="coordinates_pt" pt="38.916120,-77.045220"}

bbq grill

TAGGED: bbq {grill}

PARSED: { type:keyword, known: false, surface_form: "bbq"} {type: product_category, canonical_form: "grill"}

TO SOLR: ("bbq" OR "grill"^0.722 "propane"^0.646 OR "charcoal"^0.613) AND type:"outdoor appliance" product_category:"grill"

Q

Other Knowledge Graph Search examples:

popular barbeque near Haystack EU
(popular same as "good", "top", "best")

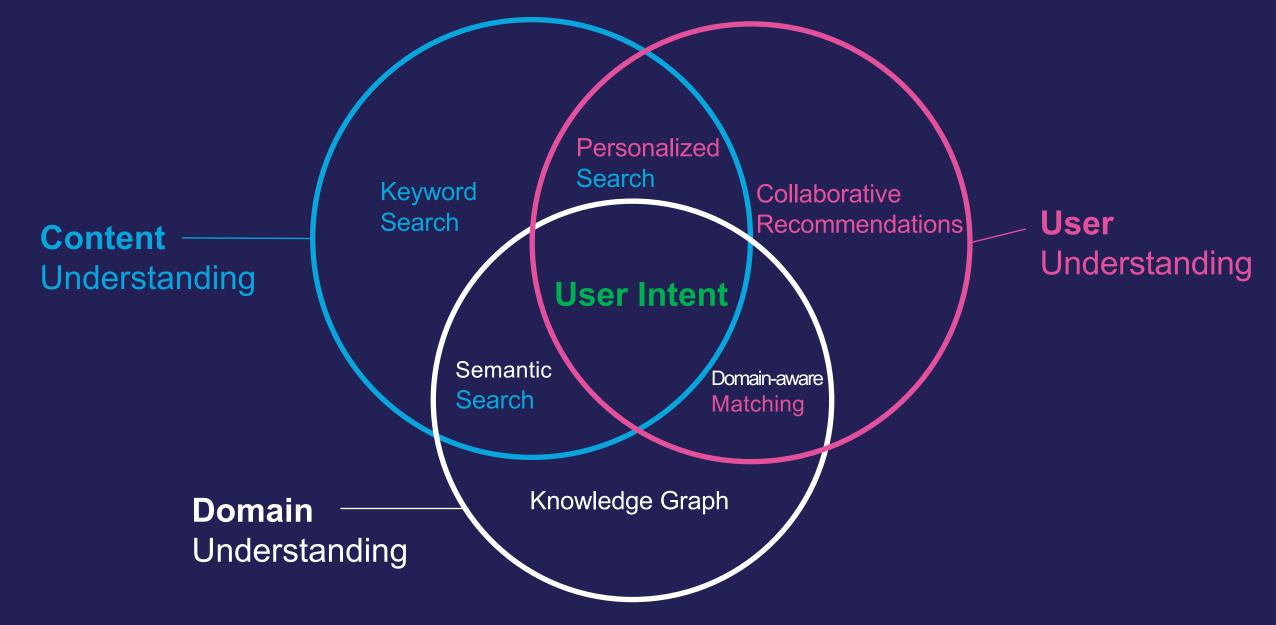
hotels near Haystack EU

hotels near popular BBQ in Berlin

BBQ near airports near Berlin

hotels near movie theaters in Berlin ...

Dimensions of User Intent



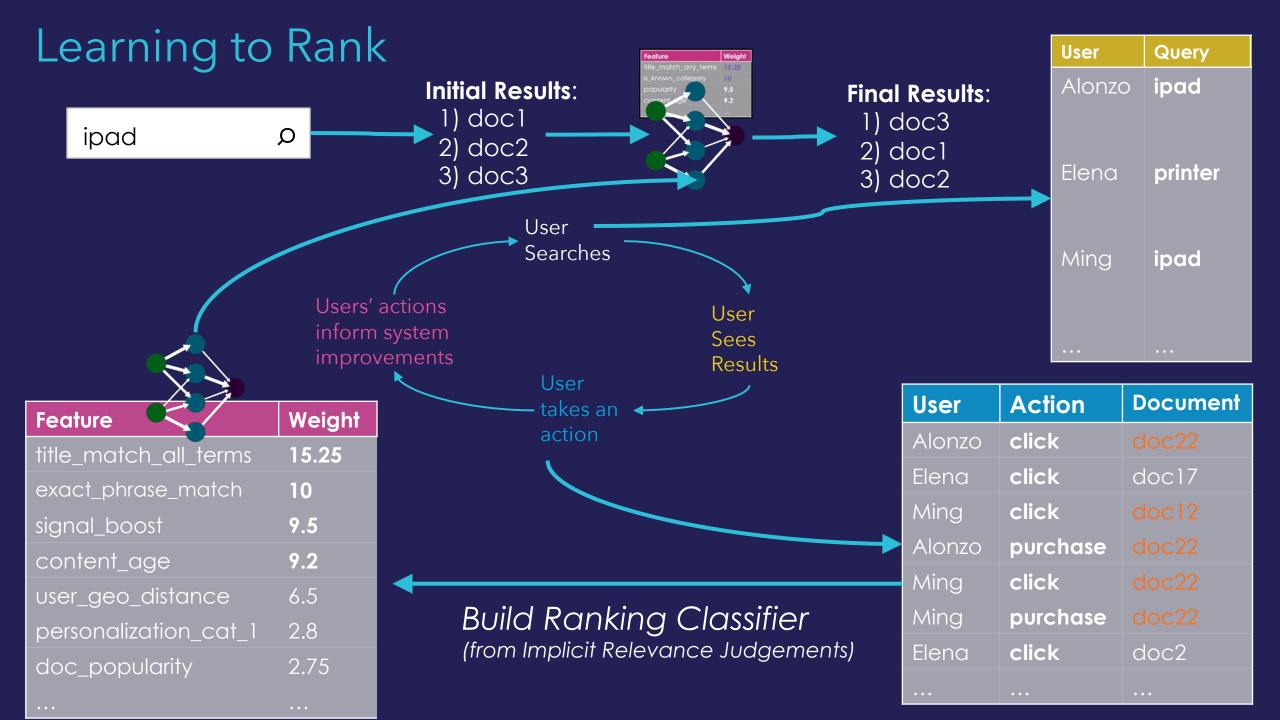
The right ranking algorithm is domain and context-dependent

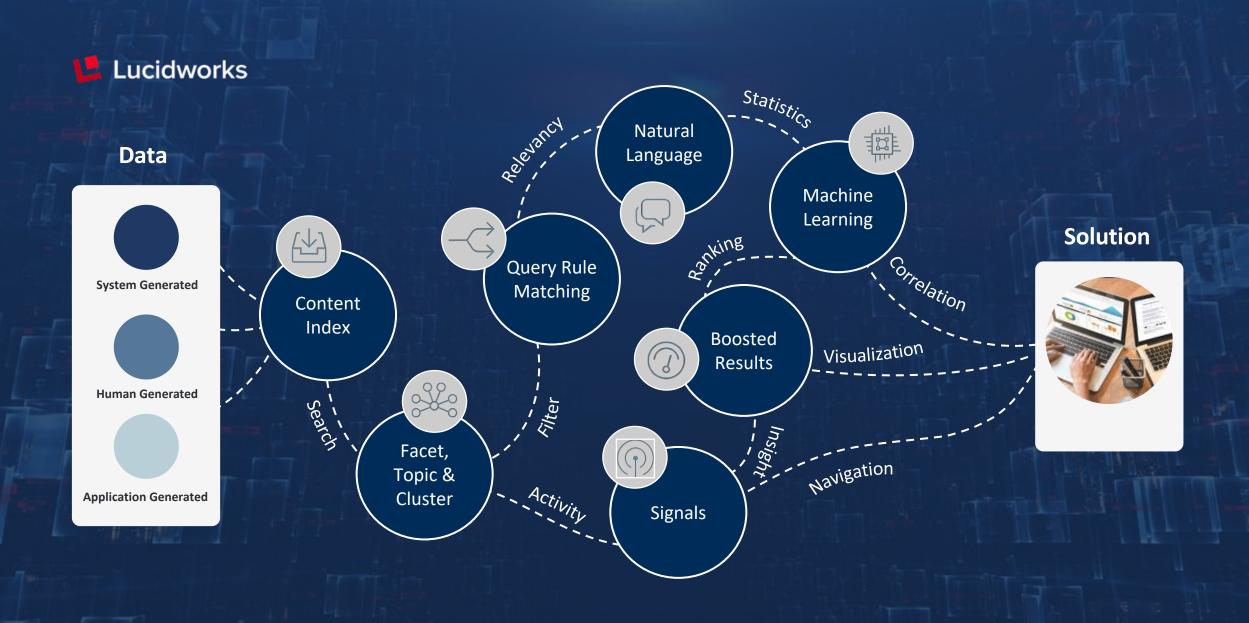
- News Search : popularity and freshness drive relevance
- Restaurant Search: geographical proximity and price range are critical
- Ecommerce: likelihood of a purchase is key
- Movie search: More popular titles are generally more relevant
- Job search: category of job, salary range, and geographical proximity matter

Example Combining Content + Domain + User Context

News website:

But how do we figure out the right balance of weights?







We operationalize AI for the largest businesses on the planet.

Questions?



Thank You!

Trey Grainger



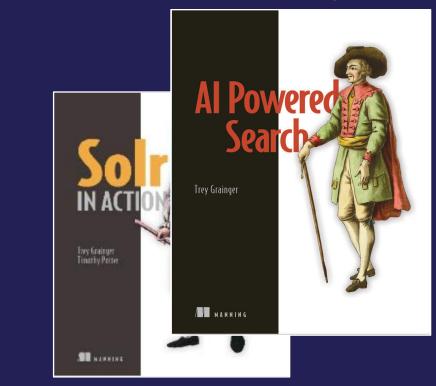
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