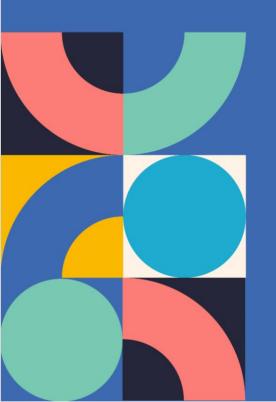
# brighton SEO. April 2022





How to Implement
Machine Learning in Your
Internal Linking Audit
LAZARINA STOY. | INTREPID DIGITAL





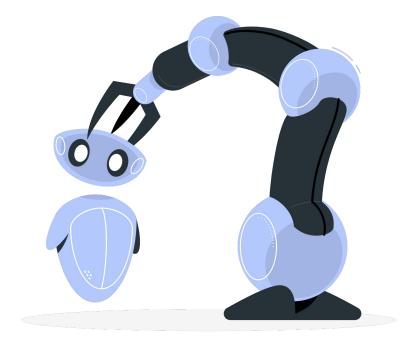


#### **(THANK YOU FOR BEING HERE).**

(IF YOU NEED A TRANSCRIPT OF THE VIDEO RECORDING OF MY TALK, CLICK HERE)



## BECAUSE OF THE NATURE OF MY TALK...





#### DISCLAIMER

## I AM...

- SEO & Data Science Manager
- @ Intrepid Digital
- Content-creator
- Growth & efficiency-obsessed





## I AM NOT...

X A developer A machine learning expert

An 'expert'

A 'guru'













# THE GOAL



# THE CHALLENGE



## THE OPPORTUNITY



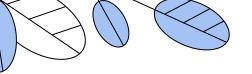




#### NO

# END-TO-END, CUSTOM-BUILT, GROUND-BREAKING, NEW TECHNOLOGIES

TODAY...









#### **ONLY A**

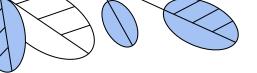
#### **TESTED CLAIM**

**AND** 

### SIMPLE PROMISE:







#### IN JUST 20 MINUTES...







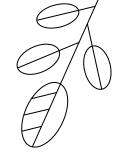
**Q2 ANALYZE** 



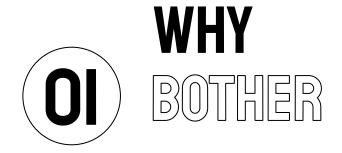


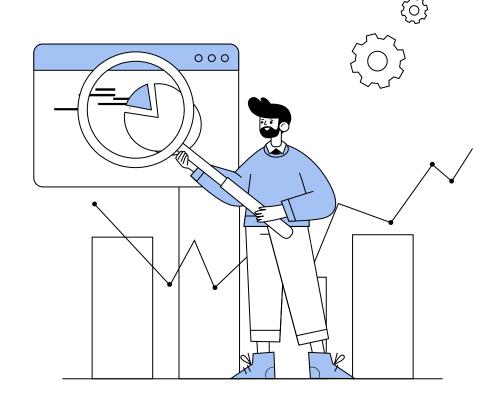












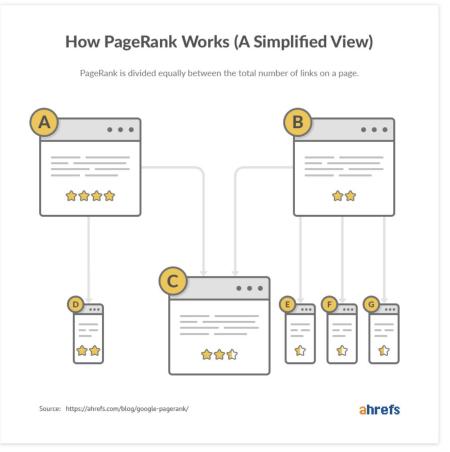




#### **PAGERANK**

Google's first algorithm relied (mostly) on links,

quality and quantity.

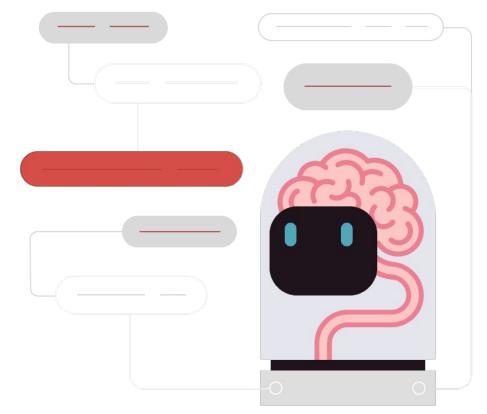




#### **RANKBRAIN**

Focus on intent-driven structure and topic comprehension.

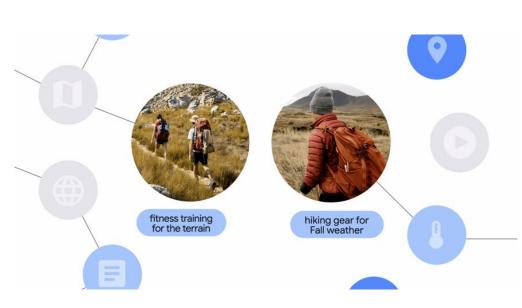
Things over strings.





#### MUM

Surfacing subtopics, semantic relationships, multi-modal comprehension

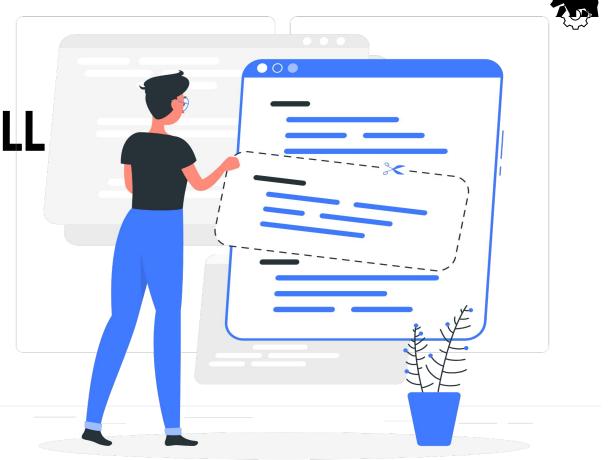


#### LINKS ARE STILL

#### **CRUCIAL FOR**

#### **DISCOVERY**

(but...)







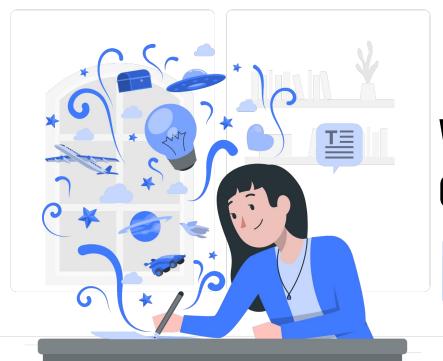
# CONTENTIS KING.









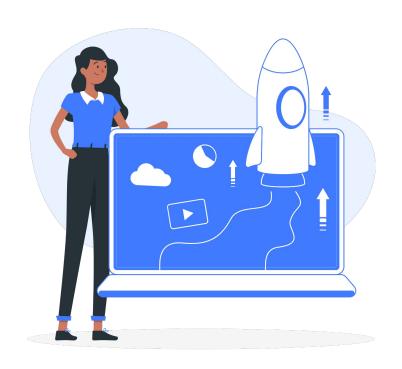


# WITHOUT GOOD CONTENT, CONTEXT BECOMES

MEANINGLESS.







# IF THE QUALITY IS THERE, CONTEXT BECOMES THE

# **VESSEL**

TO PUSH THE CONTENT FORWARD.





#### RELEVANT IN 2022?

48 MENTIONS OF INTERNAL LINKS.

FOR BACKLINKS.



"It is hard to come across a resource these days that has input from so many brilliant and knowledgeable people in the SEO community. This is a must-read for anyone who is looking to take their SEO knowledge to the next level."

Joy Hawkins, Sterling Sky Inc.

SEOIN2022.COM











Aleyda Solis = ② @aleyda · Oct 12, 2020

What's the most frequent \*high impact\* but often overlooked Technical SEO issue that you see a lot across your projects happening again and again and try to fix asap?

In my case, it's likely poor **internal linking** of critical (priority to rank) pages! What's yours?

0

55

37

0

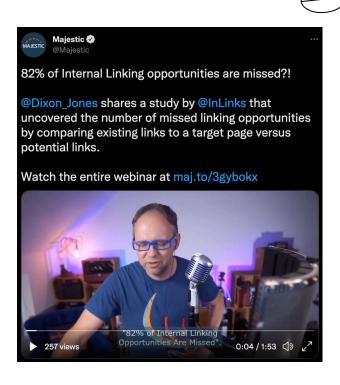
210

,1





82 OF INLINK
OPPORTUNITIES
% ARE MISSED



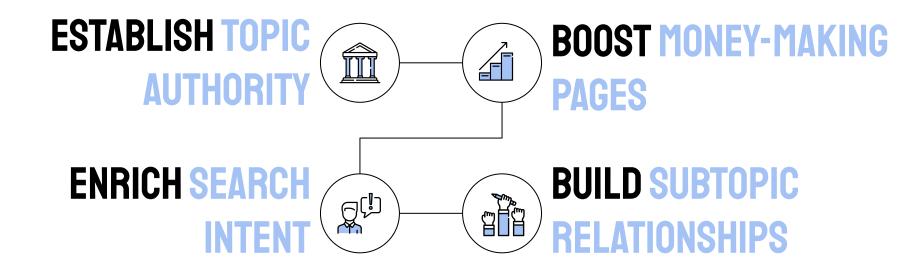




**€** 



#### AN INTERNAL LINKING STRATEGY SHOULD AIM TO:





#### **AND WHO DOESN'T NEED**





BETTER STRUCTURE

# IMPROVED INDEXING

NAVIGATION & UX













# 02

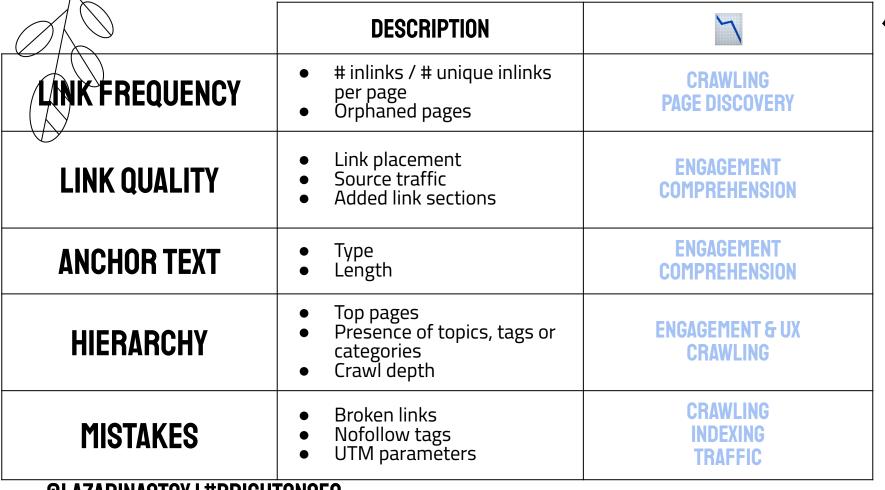
#### **HOW TO ANALYZE**

INTERNAL LINKING STRUCTURE





## WHAT SHOULD **AN INTERNAL** LINK AUDIT **ASSESS?**





# MOST OF THIS CAN BE IDENTIFIED WITH A QUICK CRAWL & CLEVER DATA STUDIO TEMPLATE

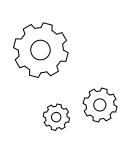


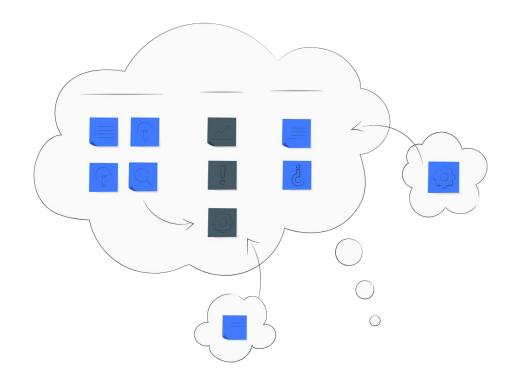






# BUT TO REALLY DIG DEEP YOU CAN INCORPORATE MACHINE LEARNING TO HELP YOU

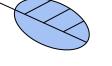


















#### Classify Anchor Text N-Grams for Interlinking Insights with Python

Home / Python / Automation / Classify Anchor Text N-Grams for Interlinking Insights with Python

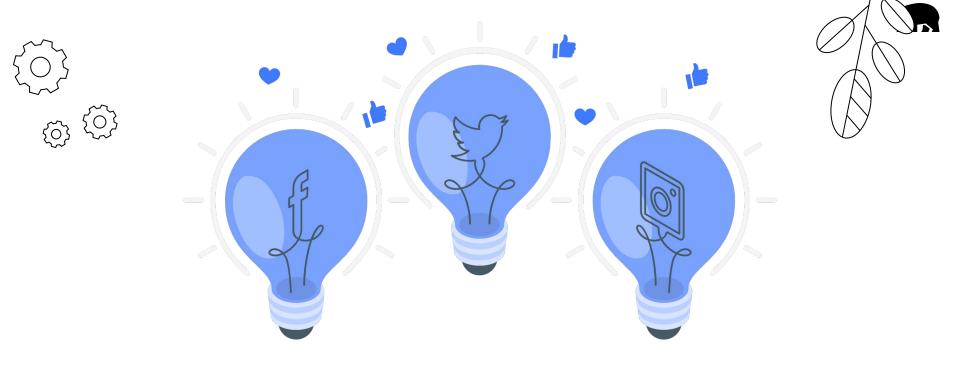


**Greg Bernhardt** 



# How to use GPT3 to classify Keywords

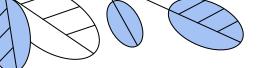




## IDENTIFY ENTITIES













# Analyze entities in a string Inspect text for known entities (proper nouns such as public figures and landmarks), and return information about those entities. Go JAVA NODE.JS PHP PYTHON View sample View in documentation Analyzing Entities Coogle Natural Language API

Sentiment Analysis & Entity Extraction with AWS Comprehend

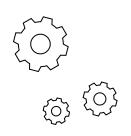
Ram Vegiraju









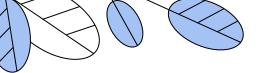






















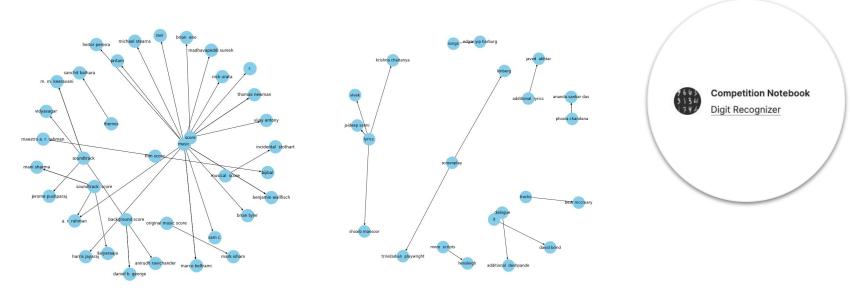








#### Knowledge Graph & NLP Tutorial-(BERT, spaCy, NLTK)







## MORE









RIGHTONSEO







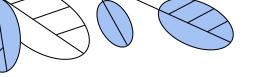
# BETTER

# RECOMMENDATIONS











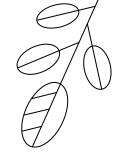




#### **SEE YOUR CONTENT AND LINKS**

## THE WAY SEARCH ENGINES SEE IT

TO UNDERSTAND HOW TO IMPROVE IT.



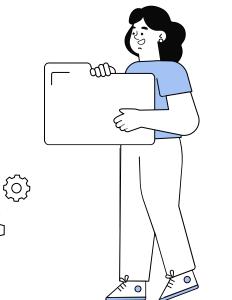


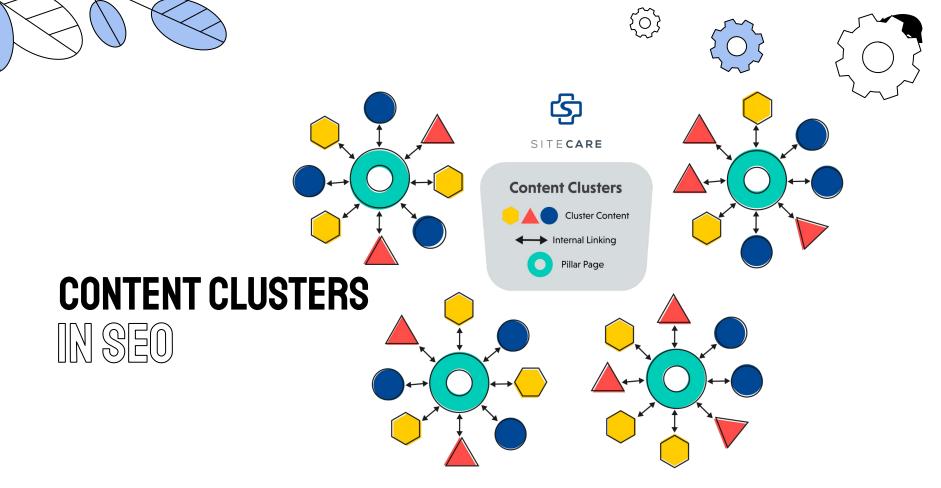


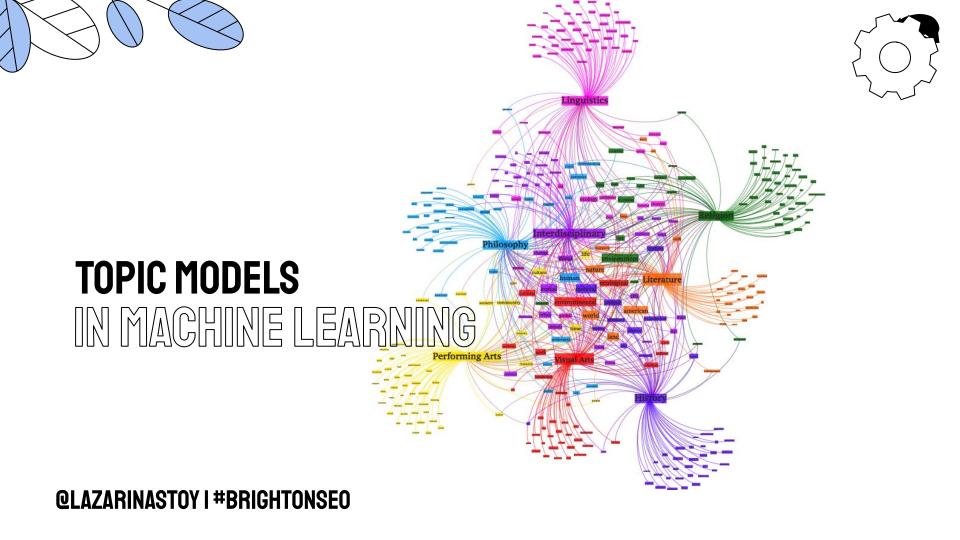




# CLUSTER CONTENT WITH ML









#### THE FUNDAMENTALS ARE THE SAME











### HOW DID TOPIC MODELING COME ABOUT?



OF THE FIRST TOPIC MODELING ALGORITHM





ξ<sup>ο</sup>}

#### **PROBLEM**

Collective knowledge increases

→information

discovery gets more

difficult



## SOLUTION

help organize, search, and understand vast amounts of information





£

## **PROBLEM**

Current systems of query & links are good, but can be limiting



### SOLUTION

We need systems that organise documents based on themes (topics) and subtopics.





### **PROBLEM**

No ability to zoom in and out of topics of interest,

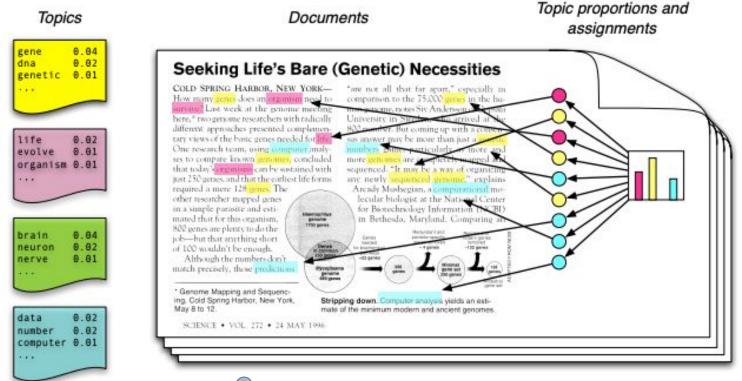
highlighting patterns.



## SOLUTION

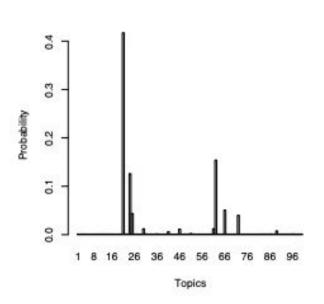
We need topic models that operate under the assumption that a document has multiple topics.





Introduction to Probabilistic Topic Models (Blei, 2012)





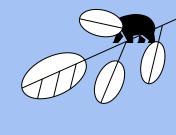
"Genetics"	"Evolution"	"Disease"	"Computers"
human	evolution	disease	computer
genome	evolutionary	host	models
dna	species	bacteria	information
genetic	organisms	diseases	data
genes	life	resistance	computers
sequence	origin	bacterial	system
gene	biology	new	network
molecular	groups	strains	systems
sequencing	phylogenetic	control	model
map	living	infectious	parallel
information	diversity	malaria	methods
genetics	group	parasite	networks
mapping	new	parasites	software
project	two	united	new
sequences	common	tuberculosis	simulations



Introduction to Probabilistic Topic Models (Blei, 2012)







# TOPIC MODELING IS PATTERN RECOGNITION

IN LARGE, TEXT-BASED CORPUSES OF DATA.









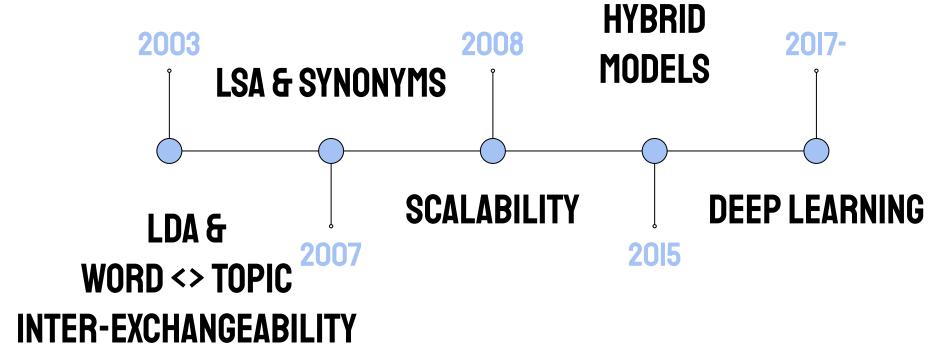
















## **7**55

# BUT THE PROBLEM WE FACE REMAINS THE SAME







# SO, WHY NOT TRY AN ML-BASED SOLUTION?





#### TWO METHODS TO KICKSTART THIS JOURNEY





Topic Modeling in Python: Latent Dirichlet Allocation (LDA)

Shashank Kapadia



**Latent Semantic Analysis using Python** 



















## WATCH THE DETAILS LATER

I've recorded a step-by-step tutorial on doing **topic modelling** using a no-code, publicly-available, web-based app using LDA.

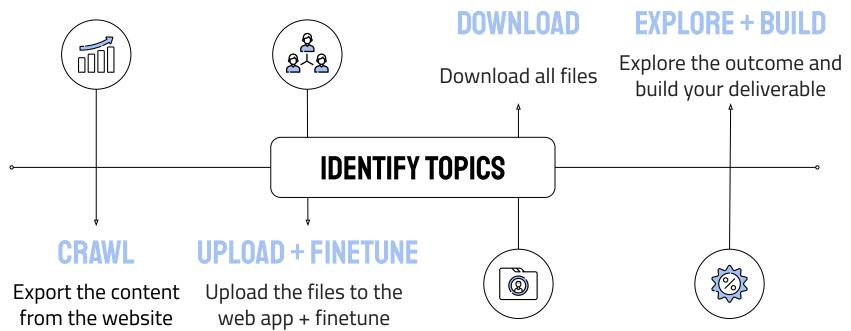








#### **BUT IN A NUTSHELL...**











#### **TOPIC MODELS**

Α	В	С	D	E	F	G	Н	ı	J	K
	health mental staff workplace home support group working ehl students	corporate volunteering giving purpose social responsibility grawehr stéphanie	time content media support volunteer	alaya platform data user services users policy information conditions general	csr business social companies strategy responsibility corporate initiatives	volunteering program employees corporate giving programs matching benefits	it's back feel start make mission that's	employees impact engage purpose activities make platform community	engagement employees work team engaged good teams find make virtual	nonprofits season donors carmen amell nonprofit fundraising make strategy story
					J					
AN HAVE BUT SATISFACE THAN SAME!	0.00%	17.65%	-44.77%		-16.00%	-104.73%	-4.67%	9.29%	-5.99%	
posts occidently polysome and prodots positi digitate impagnion.	17.65%	0.00%	-21.13%	-88.64%	-8.61%	-62.86%	-36.37%	-23.42%	27.18%	6.80
spolf obstance the potent radio again before responds done box	-44.77%	-21.13%	0.00%	9.88%	-26.01%	-58.78%	12.52%	16.87%	-56.55%	70.50
po platform ligo par service part palls, screeker condition general.		-88.64%	9.88%	***************************************	-30.66%	-	-157.96%	-53.92%		
Surrect sold conjunite bridge percebble organise retailmer paids (seems)	-16.00%	-8.61%	-26.01%	-30.66%	0.00%	60.18%	54.64%	26.11%	-0.59%	-218.54
Criticality program anglisses inspirally regularization, habiting benefits gove to	-104.73%	-62.86%	-58.78%		60.18%	0.00%	35.22%	30.17%	-0.59%	-77.019
agin company time to have been don't make some fluid.	-4.67%	-36.37%	12.52%	-157.96%	54.64%	35.22%	0.00%	28.20%	27.18%	-46.269
plicate figure riguy access allella. In pathon compute, orbit and	9.29%	-23.42%	16.87%	-53.92%	26.11%	30.17%	28.20%	0.00%	-1.53%	-40.319
pagement employees sort from empaged on layers that make othank.	-5.99%	27.18%	-56.55%		-0.59%	-0.59%	27.18%	-1.53%	0.00%	-29.359
sports assert from prior and suppli- change rate distances		6.80%	70.50%		-218.54%	-77.01%	-46.26%	-40.31%	-29.35%	0.00

Topic to Topic Similarity 🕶

Topic Modelling per Page

Topic to Topic Similarity ▼



**PAGE INFO** 

#### **TOPIC MODELS**

ontent xport = Address	normalised										
		0.00%	0.00%	50.00%	0.00%	0.00%	50.00%	0.00%	0.00%	0.00%	0.00
		0.00%	0.00%	1.49%	0.37%	0.00%	20.07%	1.86%	33.09%	0.00%	0.0
		1.33%	0.00%	2.00%	0.36%	5.74%	18.55%	6.20%	10.51%	7.64%	0.0
		0.00%	0.00%	0.00%	0.00%	0.60%	14.83%	1.40%	38.48%	1.20%	0.0
		0.00%	48.53%	0.74%	0.00%	0.00%	0.00%	0.00%	0.00%	8.09%	4.0
		0.00%	0.00%	9.38%	0.00%	0.00%	0.00%	0.00%	3.13%	0.00%	53.1
		0.00%	3.51%	15.59%	2.10%	0.86%	2.73%	6.63%	5.22%	0.00%	8.9
		2.99%	0.48%	4.31%	1.08%	4.55%	1.20%	7.19%	6.23%	15.81%	0.9
		6.50%	3.58%	3.17%	0.41%	6.81%	7.16%	7.98%	4.35%	15.05%	0.1
		9.84%	7.81%	9.12%	1.74%	3.91%	0.00%	5.64%	9.41%	2.03%	0.0
		0.09%	1.29%	0.76%	0.09%	25.68%	5.38%	8.90%	5.47%	4.09%	0.0
		4.12%	1.17%	9.48%	1.58%	1.37%	2.68%	5.36%	4.53%	0.76%	15.5
		1.40%	0.97%	7.97%	2.37%	3.47%	4.50%	5.36%	4.81%	1.58%	7.6
		0.00%	2.53%	27.09%	2.61%	2.61%	0.00%	3.15%	6.45%	2.23%	8.2
		0.00%	49.14%	0.00%	0.00%	0.00%	0.00%	0.69%	0.00%	9.62%	5.8
		0.00%	5.18%	8.83%	6.91%	7.49%	1.92%	4.41%	15.36%	0.58%	0.5
		0.00%	0.00%	43.24%	0.00%	0.00%	0.00%	2.70%	0.00%	0.00%	18.9
		0.00%	19.63%	8.89%	0.00%	0.00%	0.00%	0.00%	0.00%	0.00%	32.5
		0.00%	62.77%	0.00%	0.00%	0.00%	0.00%	0.00%	0.00%	0.00%	0.0
		7.07%	51.52%	0.00%	0.00%	0.00%	0.00%	0.34%	0.00%	5.39%	1.6
		4.34%	1.42%	5.54%	0.78%	2.35%	0.28%	10.73%	3.98%	0.57%	13.1
		0.00%	54.29%	2.86%	2.86%	2.86%	0.00%	0.00%	0.00%	0.00%	0.0
		1.93%	4.16%	12.01%	1.97%	6.22%	13.46%	4.72%	4.25%	2.19%	0.0
		2.59%	4.25%	3.22%	0.00%	5.97%	3.14%	7.63%	12.74%	12.50%	0.3
		2.03%	1.92%	1.05%	0.17%	23.55%	8.31%	14.42%	2.56%	3.43%	0.0

**QLAZARINASTOY I #BRIGHTONSEO** 

Topic Modelling per Page ▼

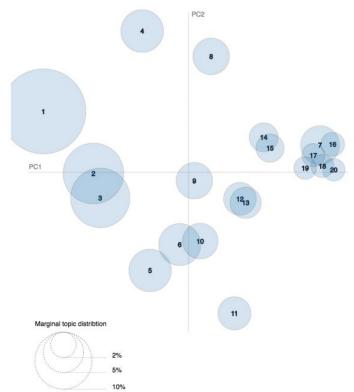


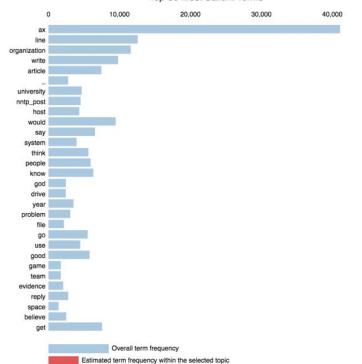




Top-30 Most Salient Terms1







1. saliency(term w) = frequency(w) \* [sum\_t p(t | w) \* log(p(t | w)/p(t))] for topics t; see Chuang et. al (2012) 2. relevance(term w | topic t) =  $\lambda * p(w | t) + (1 - \lambda) * p(w | t)/p(w)$ ; see Sievert & Shirley (2014)











# SAVED A TON OF TIME

Took you less than 30 minutes. Hooray!















# **BASELINE OVERVIEW**

You've got a clear overview of the main topic clusters and their relationship







## TESTED SOMETHING NEW

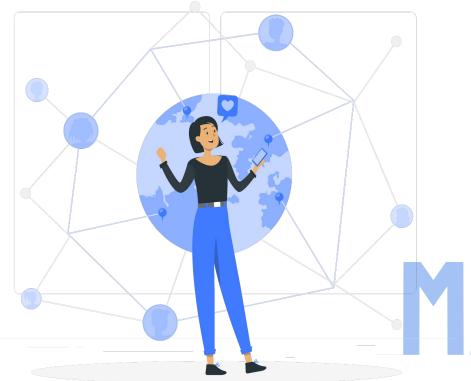
Scrutinize the output.

Tell your client how you have improved on it.









### **ENTER**

# **FUZZY**

# MATCHING





# A QUICK AND DIRTY WAY FOR CALCULATING

### SIMILARITY BETWEEN TWO STRINGS







# WATCH THE DETAILS LATER

I've recorded a step-by-step tutorial on using fuzzy matching for things like:

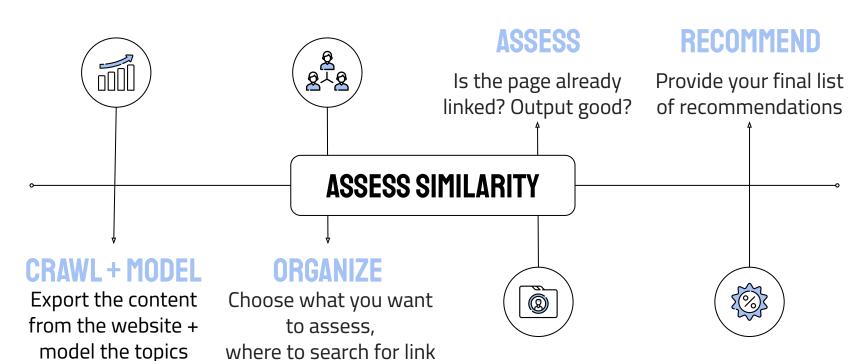
- Identifying link opportunities
- String Similarity Analysis
- Other cool things





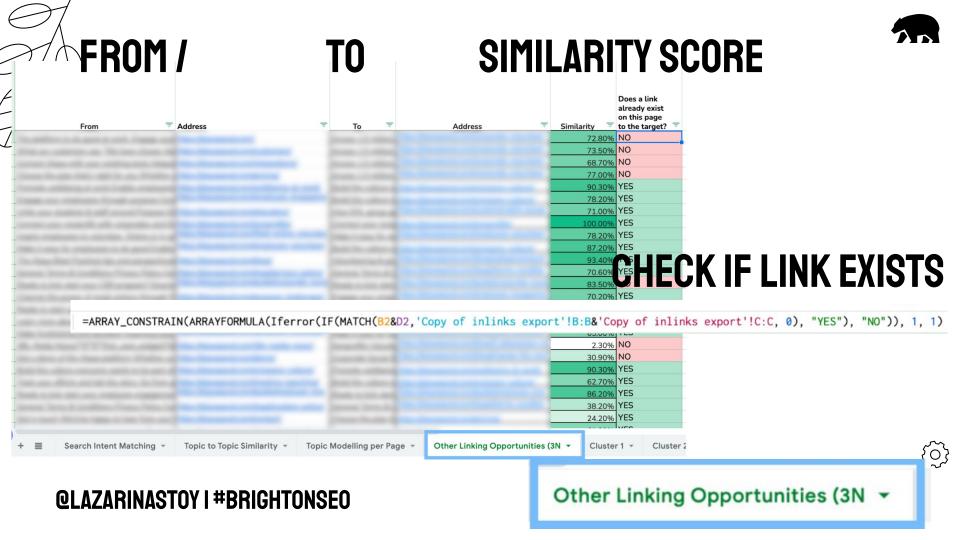


#### BUT IN A NUTSHELL...



opportunities



















MY ACCOUNT

PROFILE

**CHANGE PASSWORD** 









#### **About Us**

**Phlox** blog

#### **Popular Posts**

#### **Most View**

#### video











Review Of Healthy Breakfast Meals For... 2 weeks ago | 30,451 views | 204



Review Of Healthy Breakfast Meals For... 2 weeks ago | 30,451 views | 20.4



Review Of Healthy Breakfast Meals For... 2 weeks ago | 30,451 views | 204



Review Of Healthy Breakfast...

2 weeks ago | 30,451 views | 20;45



Review Of Healthy Breakfast...

2 weeks ago | 30,451 views | 20:45



Review Of Healthy Breakfast...

1 month ago | 30,451 views | 20:45

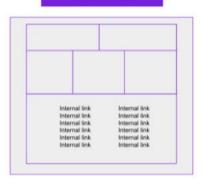
Top Ads	15
Gadgets	14
Telephone	13
Electronic	24
Game Play	37
Best Of	12
Animations	19
• Series	26
• Cars	17

**QLAZARINASTOY I #BRIGHTONSEO** 

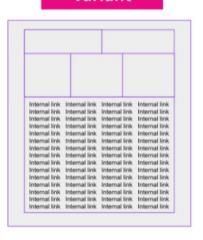
Wordpress Theme Built By Phlox.

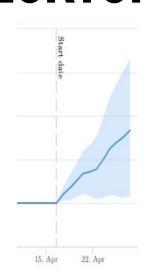
# +I0% +5% DESKTOP MOBILE

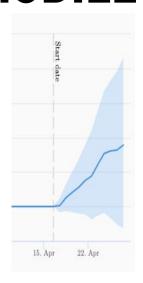
#### Control



#### Variant



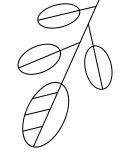






**Searchpilot Case Study** 









# HOW TO FIND WHERE TO LINK





Clear All Filters

#### **QLAZARINASTOY I #BRIGHTONSEO**

Contains

Contains

Enter Regular Expression

Enter Regular Expression



~ ➡ ₱ 100% ▼	£ % .0 .00 123 v	11 • B I S A	→ ⊞ EE - <u>= -</u> T -	·   + · ▷ · GD	Ψ Σ Ψ
→ fx					
A	В	С	D	E	F
	= INLINK OPPORTUNITIES COUNT= =	0 =	0 =	0 =	0 =
	ENTER TARGET KEYWORDS HERE =				
	ENTER URL TO LINK TO =>				
TO THIS PAGE	TEXT ON PAGE				
		Keyword not contained	Keyword not contained	Keyword not contained	Keyword not contained
		Keyword not contained	Keyword not contained	Keyword not contained	Keyword not contained
		Keyword not contained	Keyword not contained	no w roy ot a ve	er vo
		Keyword not contained	Keyword not contained	Ke ware not cental e	er vo li ot contemed
		Keyword not contained	Keyword not contained	yw rd ot cental e	ey of ot contained
		Keyword not contained	Keyword not contained	Keyword not contained	Keyword not contained
		Keyword not contained	Keyword not contained	Keyword not contained	Keyword not contained
		Keyword not contained	Keyword not conta	keyword rut con aine	Ke word no lid a led
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		Keyword not contained	Keyword not contained	Keyword not contained	Keyword not contained
		Keyword not contained	Keyword not contained	Keyword not contained	Keyword not contained
		Keyword not contained	Keyword not contained	Keyword not contained	Keyword not contained
		Keyword not contained	Keyword not contained	Keyword not contained	Keyword not contained
		Keyword not contained	Keyword not contained	Keyword not contained	Keyword not contained
		Keyword not contained	Keyword not contained	Keyword not contained	Keyword not contained
		Keyword not contained	Keyword not contained	Keyword not contained	Keyword not contained





The following query searches across all columns of the Recipes table for the value toast and returns the rows that contain this value.





THE SAME









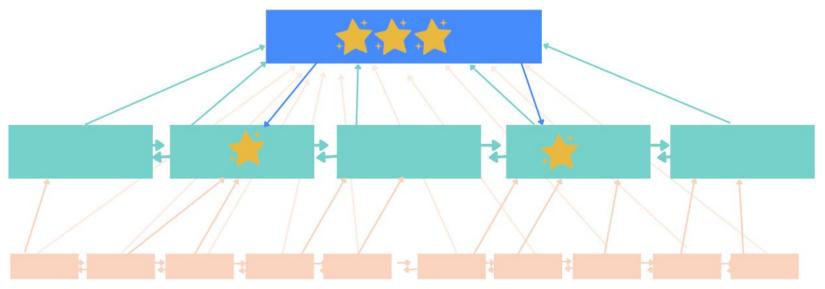
#### SEARCH FOR INTENT-BASED KEYWORD VARIATIONS

**SEARCH FOR DIFFERENT ANCHOR TYPES** 

INCORPORATE ENTITY-BASED LINKING

THINK ABOUT MONEY-MAKING PAGES





#### **Informational Intent Page**

- Olinks to other informational content, based on similarity;
- Olinks to pillar page (money page, highest converting)
- Slinks to one other high intent page

#### **High Intent page**

- ⊗links to pillar page (money page, highest converting)
- $\mathscr{O}$ links to one other high intent pages

#### Pillar Page ( conversion rate)

- Olinks to 1-2 high intent pages in the cluster (with the second and third highest conversion rates)
- receives links from multiple informational pages











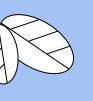
# BUILD MEASURE 06

LEARN





## WHAT SHOULD YOU INCLUDE IN YOUR



# DELIVERABLE?







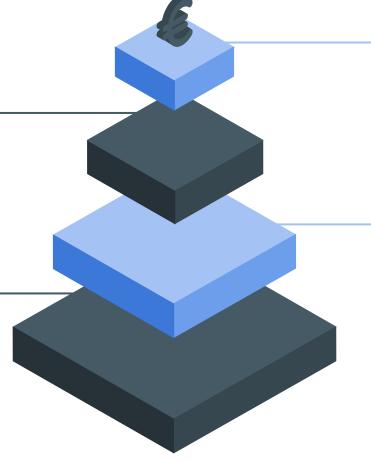


## LINK OPPORTUNITIES

Highlight additional link opportunities, anchor strategies.

#### **ANALYSIS**

Present your analysis in a format that allows for benchmarking



#### PRIORITIZE, BUDGET

Talk about link implementation, timeline, tracking.

#### **TOPIC MODELS**

After finalising the categories and structure, create custom dimensions for the topics in GA/GDS/GSC







# WHAT SHOULD YOU MEASURE?



# YOUR REPORTING SHOULD INCLUDE

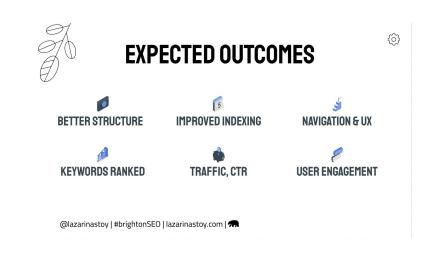


	DESCRIPTION	٦
LINK FREQUENCY	# inlinks / # unique inlinks     per page     Orphaned pages	CRAWLING PAGE DISCOVERY
LINK QUALITY	Link placement     Source traffic     Added link sections	ENGAGEMENT COMPREHENSION
ANCHOR TEXT	Type Length	ENGAGEMENT Comprehension
HIERARCHY	Top pages Presence of topics, tags or categories Crawl depth	ENGAGEMENT & UX Crawling
MISTAKES	Broken links     Nofollow tags     UTM parameters	CRAWLING INDEXING TRAFFIC

QLAZARINASTOY I #BRIGHTONSEO

### BENCHMARKING

### **GROWTH METRICS**





### FOR GROWTH REPORTING, TRACK



- Quicker Indexing
- # of pages indexed
- # of pages discovered
  - Organic traffic
- Inter-cluster user flow

TRAFFIC



- CTR
- Average time per session
  - Clicks
  - Interactions

**ENGAGEMENT** 



- # of ranking queries
- Length of queries
- Performance (visibility; engagement, rankings) split per topic cluster

**#QUERIES** 











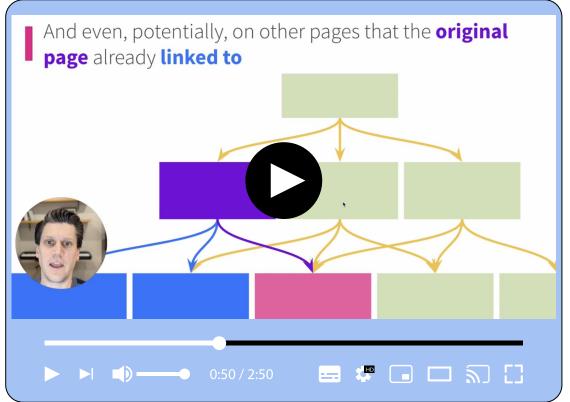








**WILL CRITCHLOW** EXPLAINS...







### INTERNAL LINKING EFFECT MEASUREMENT IS



# HARD



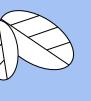








### **OUR INTUITION ABOUT LINK STRUCTURES IS**

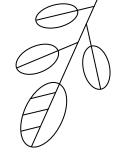


POOR

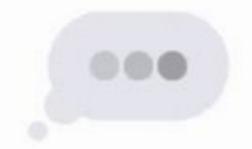


ZAP STOY I #BRIGHTONSEO









## IN OTHER WORDS...



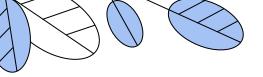


# IS PART OF THE PROCESS.



{O}

















TESTING

EXPERIMENTS



LEARNING











THIS DECK,

REFERENCED WORK

OF OTHER COOL PEOPLE, AND A

**TALK WRITE-UP** 

**WILL BE AVAILABLE** 







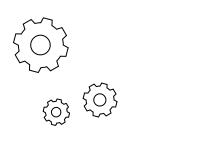


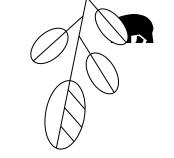




https://lazarinastoy.com/how-to-incorporate-machine-learning-in-internal-linking-audits/









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THANK YOU.



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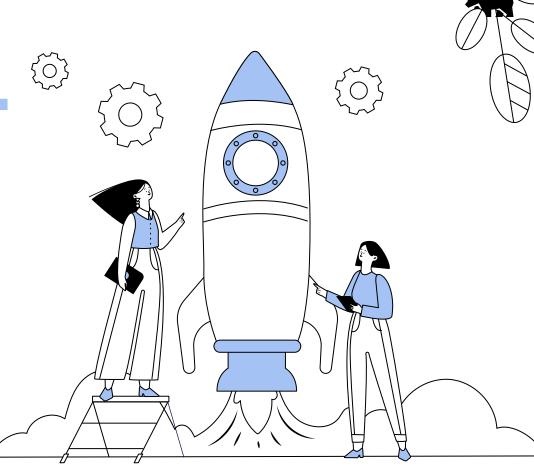


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