



How AI/ML is changing information retrieval

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Accessing Information

As soon as we started writing stuff down, we needed to find it

Ancient Sumer

Callimachus—*Pinakes* ca. 245 BCE

Card catalogs

Organization

Static

Not interactive

But you needed to talk to a Librarian



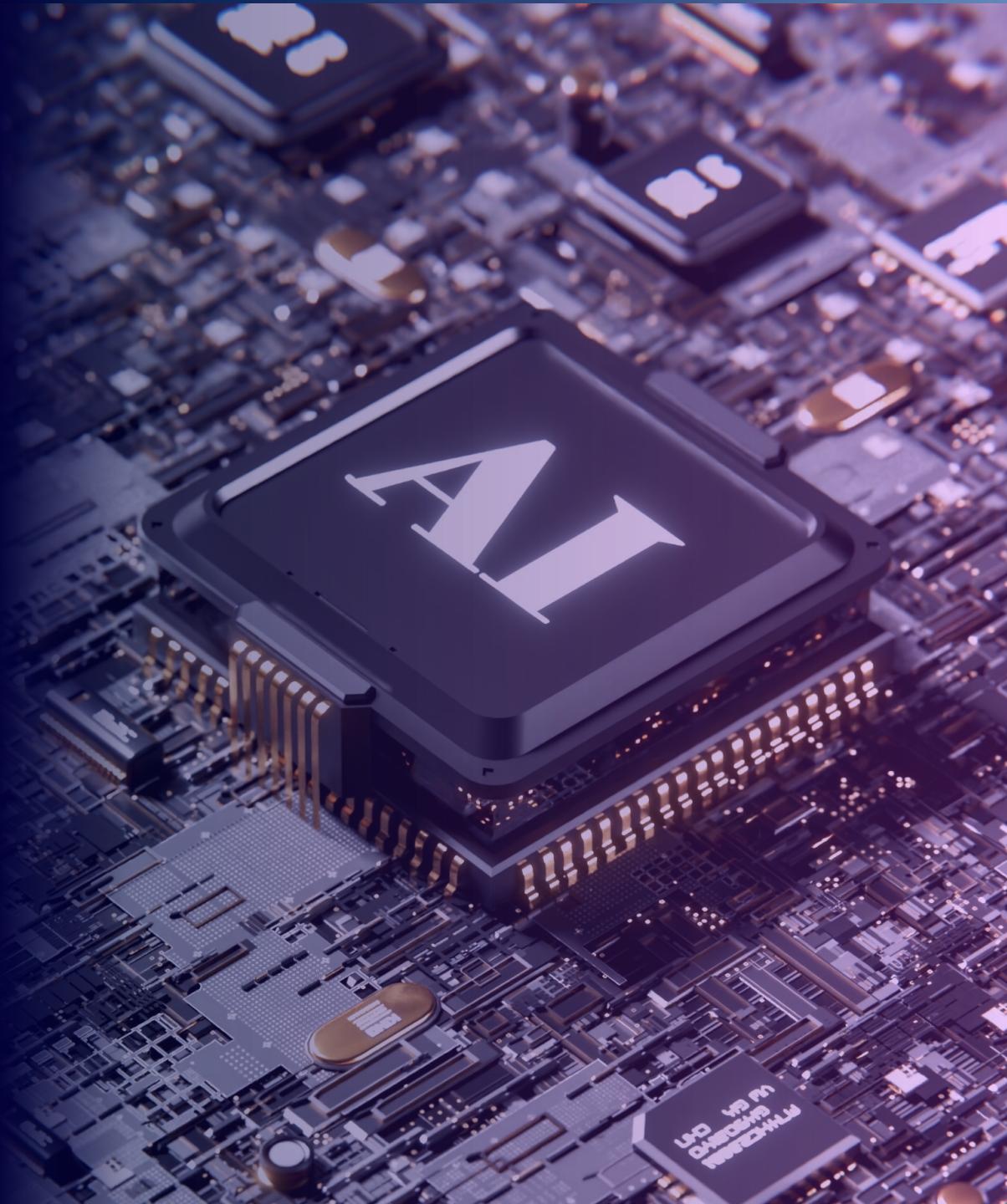
The boom in AI/ML

Large language models (LLM) are powering a new wave of generative AI

Embedding generating LLMs capture meaning from language

Text generating LLMs enable human-like chats in AI assistants and chatbots

Natural language is becoming the currency and yardstick for search



Where we're headed



10 blue links will still be useful for some cases like parts search, where you know the exact name of the thing you want



Finding and interacting with information will become more long-term, and more based on conversation between human and computer

Lexical Search

The basic idea

Words are units of meaning that code the concepts behind them, from concrete to abstract

Sun ← → Love

Words created for a purpose express that purpose through the assembly of concrete and abstract meaning

Search queries express user purpose through words (and some UI elements)

Match words in order to match meaning

The more matches the better, and the more information-specific to the purpose, the better

Search engines match words and score

Information goal



Text

Facets

Geo

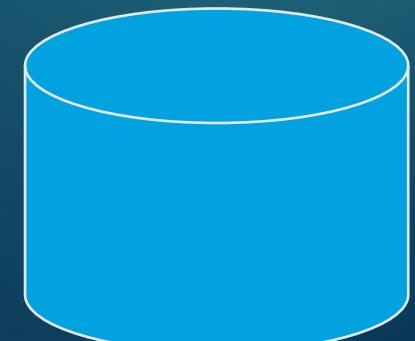
Representation

Similarity



OpenSearch

Encoding



Stuff

Ranked stuff that matches
the information goal

Encoding

Search documents represent search targets

Fields provide scoped units of meaning

Field names carry meaning as well, but only for the builders

```
{  
    "question_text": "Im 5'8 and pretty thin.  
    Should i go for a small or a medium?\\"",  
    "category_name": "t-shirts",  
    "question_type": "WH",  
    "answer_aggregated": "N/A",  
    "answers": [  
        {  
            "answer_text": "I'm 5'8 also, usually wear women's M.  
            I ordered a small and it's fitting  
            but good. I would say a small for you.\\"",  
            "gender": "other",  
            "user_lon": -63.723771,  
            "name": "Michelle Davenport",  
            "user_lat": 6.134147,  
            "age": 9,  
            "product_rating": 3  
        }  
    "chunk": "This Parks & Recreation T-shirt features  
    a I Met Li'l Sebastian design on an  
    adult-sized cotton tee. QINOL Parks &  
    Recreation - Lil' Sebastian T-Shirt Heather  
    (S) Grey 100% Cotton Fully Machine  
    Washable Fast And Free Shipping  
    Printed in the U.S Brand New",  
    "brand_name": "",  
    "item_name": "QINOL Parks & Recreation -  
    Lil' Sebastian T-Shirt  
    Heather (S) Grey",  
    "asin": "B005TGLE64",  
    "product_description": "This Parks & Recreation  
    T-shirt features a I Met  
    Li'l Sebastian design on  
    an adult-sized cotton tee.",  
    "question_id": "Tx125VNPD96ZLI6",  
    "bullets": "100% Cotton Fully Machine Washable  
    Fast And Free Shipping Printed in  
    the U.S Brand New"  
}
```

Text analysis: prep for matching

Source

Miss Kobayashi's Dragon Maid Tohru Cosplay Dress Outfit Package:dress+tie+belt+gloves
Fabric:Cotton 90% & Polyester blend.
Washable! Practical cotton cosplay!
Christmas, Halloween, Birthday parties, Barbecue Party AND Daily Kitchen.

Analyzed

miss kobayashi dragon maid tohru cosplai dress outfit package:dress tie belt glove fabric:cotton 90 polyest blend washabl practic cotton cosplai christma halloween birthdai parti barbecue parti daili kitchen

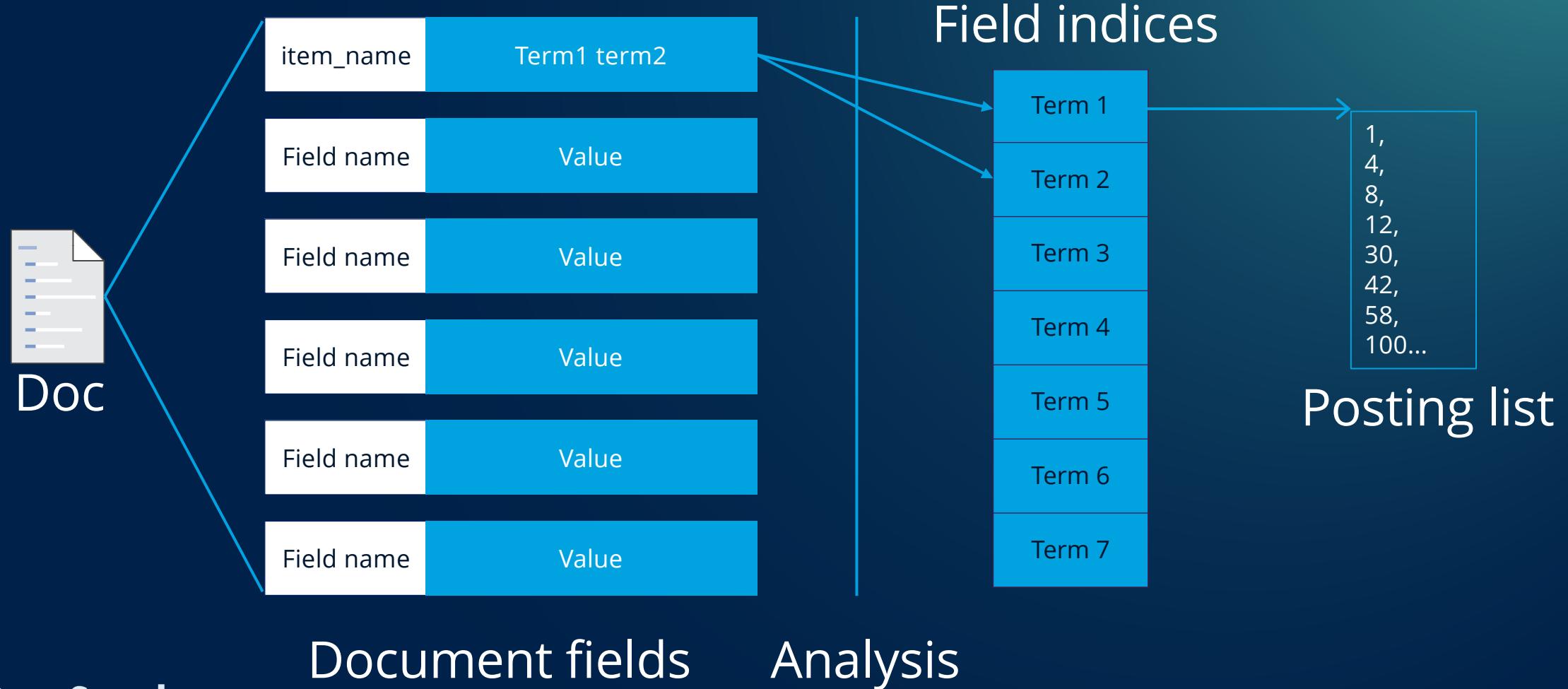
Stemming: brings words to a common form

Stop words: remove common terms that don't provide matching or discriminatory value

Synonyms: Add to increase matchability

All of these transformations work at the token level to improve the *meaning* of a token-token matching

Search indices map terms to posting lists



Searching: query

```
GET amazon_pqa/_search
{
  "query": {
    "multi_match": {
      "query": "cotton washable
                clothes",
      "fields": ["bullets",
                  "product_description",
                  "item_name"],
      "operator": "and"
    }
  }
}
```

22: miss kobayashi dragon maid tohru cosplai dress outfit
package:dress tie belt glove fabric:cotton 90 polyest blend
washabl practic cotton cosplai christma hallowen birthdai
parti barbecu parti daili kitchen washable cotton kitchen
apron cosplai woman m 155 160cm miss kobayashi
dragon maid tohru

38: veri soft silicon vinyl stuf pp cotton bodi babi 100
handmad craftwork hand print dimens 55cm 22 inch real
babi size deviat exist due differ measur wai weight about
1.2 kg type silicon vinyl doll doll can put water ey close ey
hair mohair implant hand high fidel washabl can comb
cloth just pictur shown

Searching: analysis

1. Analyze query: “cotton washable clothes” becomes “cotton washabl cloth”

22: miss kobayashi dragon maid tohru cosplai dress outfit package:dress tie belt glove fabric:cotton 90 polyest blend washabl practic cotton cosplai christma hallowen birthdai parti barbecu parti daili kitchen washable cotton kitchen apron cosplai woman m 155 160cm miss kobayashi dragon maid tohru

38: veri soft silicon vinyl stuf pp cotton bodi babi 100 handmad craftwork hand print dimens 55cm 22 inch real babi size deviat exist due differ measur wai weight about 1.2 kg type silicon vinyl doll doll can put water ey close ey hair mohair implant hand high fidel washabl can comb cloth just pictur shown

Searching: map

1. Analyze query: “cotton washable clothes” becomes “cotton washabl cloth”
2. Match terms and retrieve posting lists

canva	11	23	42	60	85
cloth	1	19	38		
cotton	3	12	18	22	38
great	14	22	38	47	
park	22	38	90		
shirt	22	42	43		
washabl	12	19	22	35	38

22: miss kobayashi dragon maid tohru cosplai dress outfit package:dress tie belt glove fabric:cotton 90 polyest blend washabl practic cotton cosplai christma hallowen birthdai parti barbecu parti daili kitchen washable cotton kitchen apron cosplai woman m 155 160cm miss kobayashi dragon maid tohru

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cotton	washabl	cloth
3	12	1
12	19	19
18	22	38
22	35	
	38	
38		
	86	

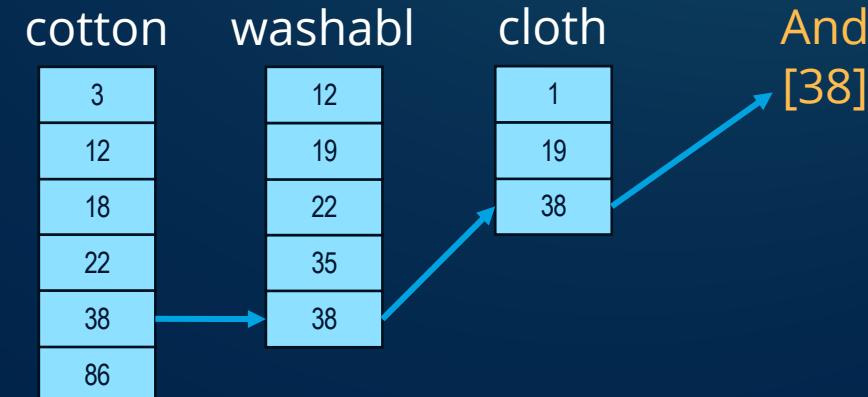
Searching: reduce

1. Analyze query: “cotton washable clothes” becomes “cotton washabl cloth”
2. Match terms and retrieve posting lists
3. Merge posting lists

canva	11	23	42	60	85
cloth	1	19	38		
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cotton	washabl	cloth
3	12	1
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38	38	
		86

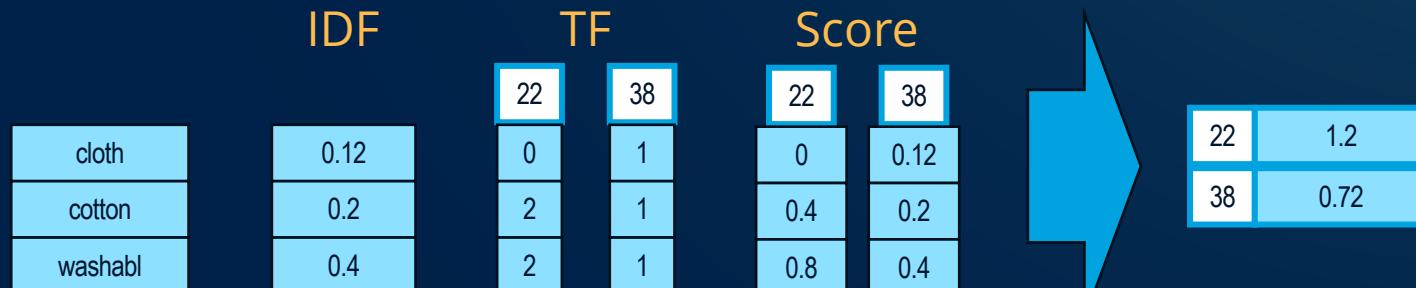
Or
[1, 3, 12,
18, 19, 22,
35, 38, 86]

Searching: score

1. Analyze query: “cotton washable clothes” becomes “cotton washabl cloth”
2. Match terms and retrieve posting lists
3. Merge posting lists
4. Score

22: miss kobayashi dragon maid tohru cosplai dress outfit package:dress tie belt glove fabric:cotton 90 polyest blend washabl practic cotton cosplai christma hallowen birthdai parti barbecue parti daili kitchen washabl cotton kitchen apron cosplai woman m 155 160cm miss kobayashi dragon maid tohru

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Okapi BM25

$$score(D, Q) = \sum IDF(q_i) f(q_i, D) \cdot \frac{f(q_i, D) (k_1 + 1)}{f(q_i, D) + k_1(1 - b + b \frac{|D|}{avgdl})}$$

The diagram illustrates the Okapi BM25 formula with four components represented by overlapping circles:

- IDF**: The first circle on the left.
- TF**: The second circle, partially overlapping the first.
- Term saturation**: The third circle, partially overlapping the second.
- Weighted document saturation**: The fourth circle, partially overlapping the third.

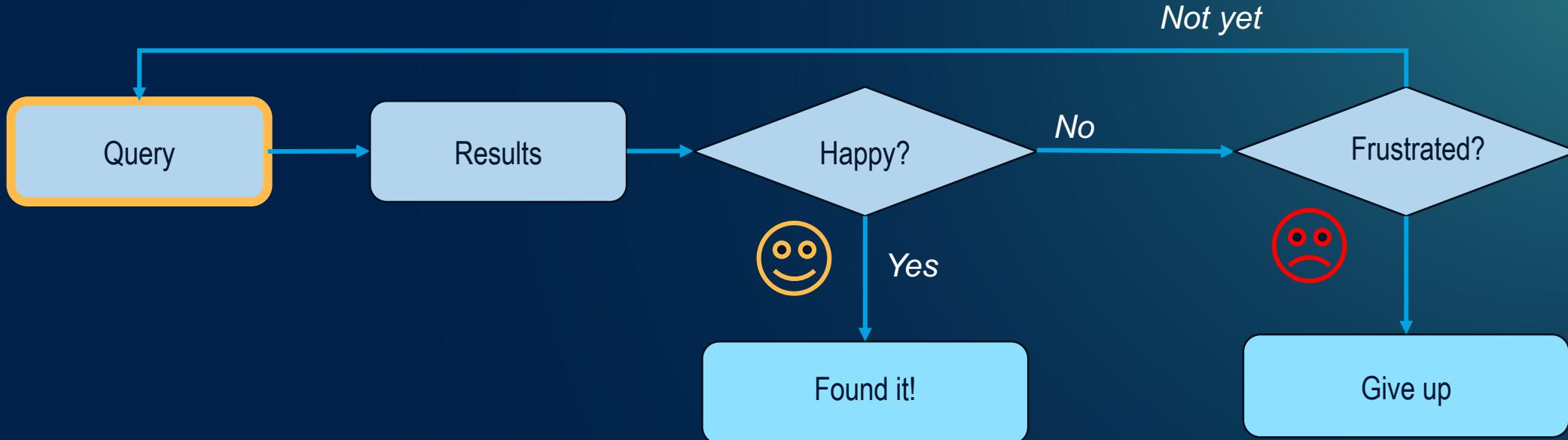
A blue arrow points from the text "Document saturation" to the term $\frac{|D|}{avgdl}$ inside the "Weighted document saturation" circle.

Based on probabilistic retrieval model

Considers term saturation & document length

Reward short documents, while penalizing matches in long documents

Did you get what you came for?



Lexical search leads to an iterative pattern of one-shot searches

To go beyond requires matching beyond single terms

Vectors

Why vectors?

Some ML Models (Large Language Models—LLMs) capture the meaning of words or blocks of text

They can emit vectors with values in many dimensions to represent this text

Related words and concepts cluster together



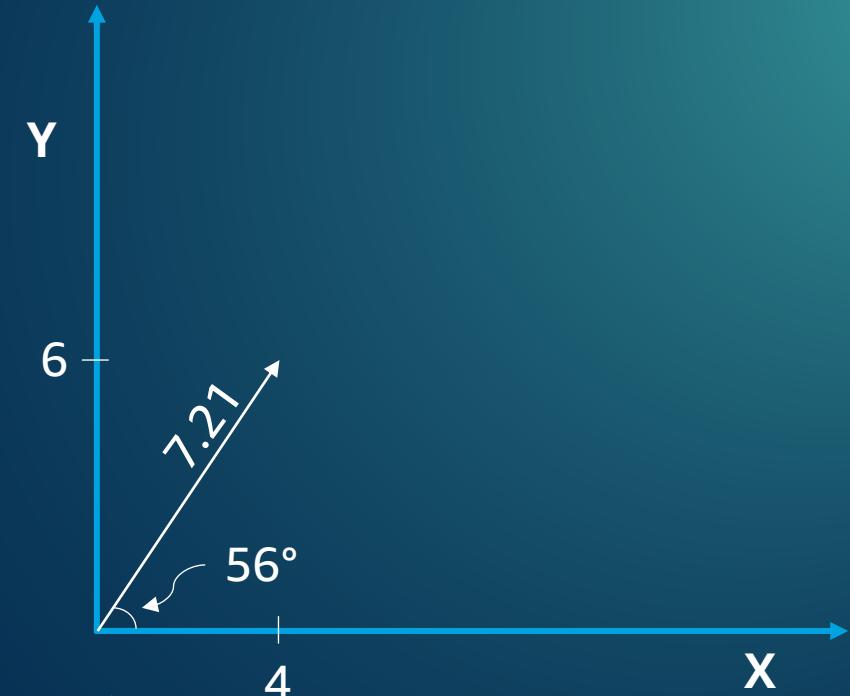
Vectors

A vector is a quantity with a magnitude and a direction

In two dimensions, you can represent it with **X** and **Y** coordinates: (4, 6), on perpendicular axes

But these two directions could mean anything!

E.g.: Y axis corresponds to “Orange”, X axis corresponds to “Apple”, we could represent “Apple” as a vector



One-hot coding

Apple	[1, 0, 0, ...]
Aardvark	[0, 1, 0, ...]
Banana	[0, 0, 1, ...]
Fruit	:
Red	:
Vegetable	
White	[..., 0, 0, 1]

Dimensions are the same as the number of words in the corpus

Still useful – some ML techniques use 1-hot

Generalization is limited: Apple + Banana = ?

Sparse coding

TOMS Charcoal Felt Men's Searcher Boot. The Searcher Boot was constructed with the world traveler or urban explorer in mind. Featuring a combination of felt and leather on the upper, a ventilated footbed for breathability and TPR outsole for comfortable support and durability. Tongue gussets keeps out unwanted elements.

Dimensions are << the number of words in the corpus

Some generalization from collapse of tokens

Relationship maintained between tokens and values

```
"##ed": 0.08905113488435745,  
"##r": 0.33547675609588623,  
"##er": 0.9349609017372131,  
"man": 0.45929819345474243,  
"found": 0.39697781205177307,  
"best": 0.37575557827949524,  
"men": 0.9220025539398193,  
"black": 0.19649049639701843,  
"top": 0.015254381112754345,  
"felt": 0.9697568416595459,  
"near": 0.07436800748109818,  
"##man": 0.028174642473459244,  
"find": 0.7214985489845276,  
"research": 0.29030007123947144,  
"support": 0.31443119049072266,  
"feel": 0.4629647135734558,
```

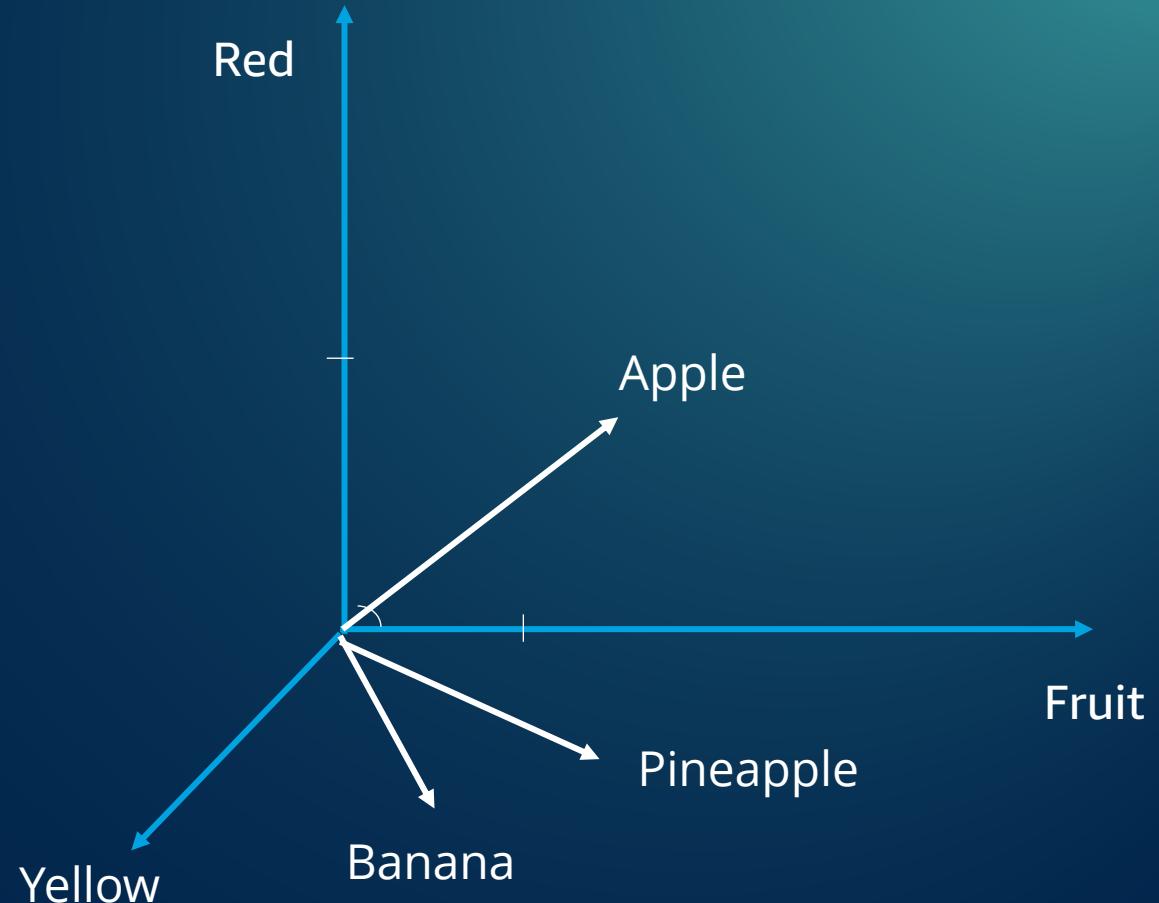
Dense coding

Select a few concepts to represent everything else

E.g. Y axis "Red", X axis "Fruit" – "Red" + "Fruit" = "Apple"

Each dimension adds additional capacity

How to pick? Neural nets + backpropagation. Via backprop, the neural network learns the important concepts from your corpus



Dense coding

Dimensions << # of words

Generalization is high... but hallucinations

Language is complex, non-linear!

The general equation for a line in n dimensions is

$$a_1x_1 + a_2x_2 + \cdots + a_n x_n = c$$

To represent language you need a complicated function to represent the space, billions of parameters

[0.46289062, -0.39257812, 0.25195312, 0.016845703, 0.20996094, 0.13867188, -0.20410156, 0.00041770935, 0.002029419, -0.15332031, 0.046142578, 0.4765625, 0.08642578, -0.07763672, -0.009399414, -0.123535156, 0.23730469, 0.15234375, 0.115234375, 0.079589844, 0.15527344, 0.078125, -0.24023438, 0.38867188, 0.078125, -0.3203125, -0.0058898926, 0.061035156, 0.33007812, -0.29101562, 0.46875, -0.078125, -0.11816406, -0.31640625, 0.20898438, -0.13183594, 0.07714844, 0.1640625, 0.34765625, -0.19824219, 0.35351562, -0.26171875, -0.12988281, -0.27148438, 0.0007209778, -0.16992188, -0.22460938, 0.14746094, 0.16894531, 0.28515625, -0.12109375, -0.07421875, 0.609375, -0.15039062, 0.55859375, 0.47265625, 0.20996094, 0.029541016, -0.15625, 0.064941406, 0.296875, 0.100097656, -0.31640625, 0.4375, 0.063964844, -0.011413574, 0.06201172, -0.14453125, 0.111328125, -0.038330078, -0.4296875, 0.09423828, -0.27734375, -0.5703125, 0.17285156, 0.0020141602, -0.048095703, 0.18066406, 0.30078125, -0.38085938, -0.107421875, 0.34765625, -0.045654297, 0.29101562, -0.0061035156, 0.31640625, -0.26953125, -0.2421875, 0.00013828278, -0.32226562, 0.19921875, 0.38476562, 0.62890625, -0.035888672, 0.025634766, 0.12792969, 0.33789062, -0.076660156, -0.91015625, -0.17382812, -0.15527344, -0.16503906, 0.06225586, 0.08300781, -0.010559082, 0.24121094, 0.46484375, 0.24121094, 0.16015625, 0.11621094, 0.328125, 0.026855469, -0.41210938, 0.24804688, -0.22753906, -0.34765625, -0.23730469, -0.20996094, -0.39453125, 0.36132812, 0.5625, 0.0065612793, 0.034179688, 0.018554688, -0.05834961, -0.24804688, -0.03173828, 0.19140625, 0.06640625, 0.18261719, 0.09667969, -0.265625, 0.006652832, 0.29101562, 0.203125, -0.19042969, -0.25, -0.09863281, 0.22363281, -0.0022125244, -0.40234375, -0.20996094, -0.1796875, -0.28320312, 0.011108398, -0.42773438, -0.10595703, 0.1015625, -0.24023438, -0.18359375, -0.087890625, -0.20898438, -0.3359375, 0.6875, 0.4453125, 0.025512695, -0.4921875, 0.50390625, -0.16210938, -0.3671875, 0.15234375, -0.16894531, 0.23339844, 0.03515625, -0.22070312, -0.15625, 0.19921875, 0.32421875, 0.46679688, 0.025268555, -0.032226562, 0.19726562, -0.16601562, -0.13085938, 0.08544922, 0.56640625, 0.40625, -0.13476562, -0.74609375, 0.33007812, 0.12988281, -0.5234375, 0.036132812, -0.049072266, 0.5390625, 0.16308594, 0.41601562, 0.009460449, -0.16503906, 0.14941406, -0.026733398, -0.118652344, 0.045410156, 0.036376953, 0.15625, -0.2578125, 0.28320312, -0.06738281, -0.24511719, -0.110839844, -0.037597656, -0.08300781, -0.39453125, -0.22753906, -0.30273438, -0.11376953, -0.21386719, -0.056396484, 0.067871094, 0.20117188, -0.32421875, -0.46289062, -0.14941406, -0.2578125, 0.26953125, 0.080566406, 0.1015625, 0.22851562, -0.33398438, 0.46484375, -0.06738281, -0.30078125, -0.06933594, -0.44921875, -0.24511719, -0.099121094, 0.28515625, -0.13671875, 0.23046875, -0.07910156, 0.52734375, -0.048583984, 0.19628906, -0.66015625, 0.08203125, -0.16699219, -0.27148438, ...]

Natural Language and Search

New paradigms for searching



Sparse

Sparse models more closely encode source tokens and retain better relevance for exact matching



Multi-modal

LLMs and other technologies can create embeddings for text, images, audio, etc. People are using these additional media to improve search relevance



Hybrid

Lexical search and semantic search both have their benefits and right applications. Hybrid blends scores from vector and lexical search to improve overall relevance



Conversational

As NL capabilities improve and chatbots are commonplace, people are turning to these bots for simple searching.

The basic idea

Information goal



Words

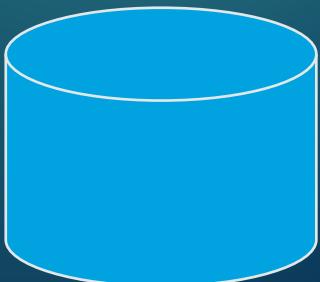
Structured
information

Images

Intent

Vectors

Encoding



Stuff

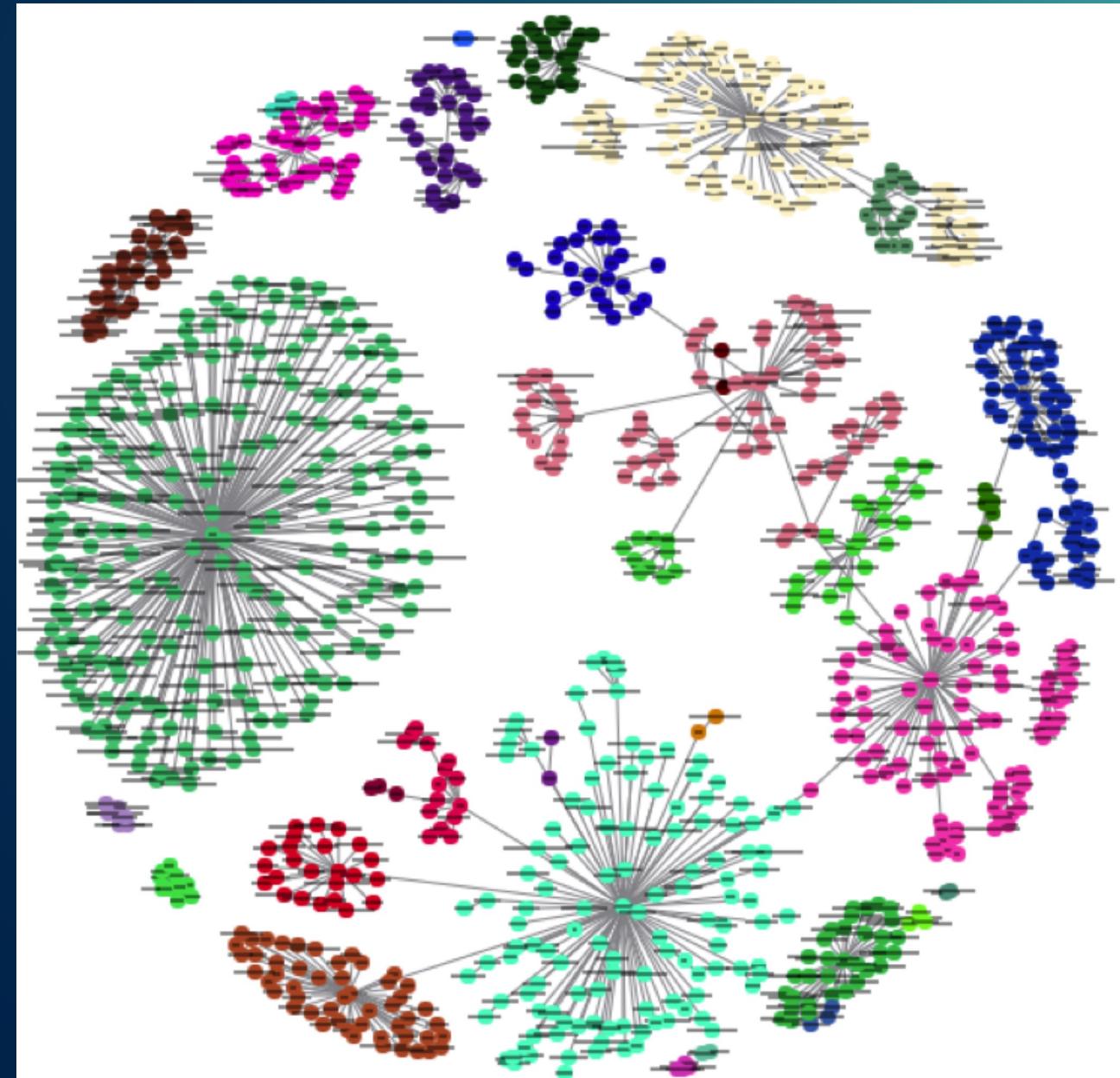
Ranked stuff that matches
the information goal

Vector similarity

Documents with similar “meaning” are encoded “near” one another

The score for a query-document pair is measured by distance

Different distance measures include L1, L2 (Euclidean), L_{Inf}, cosine, inner product, and Manhattan,



Neural Plugin

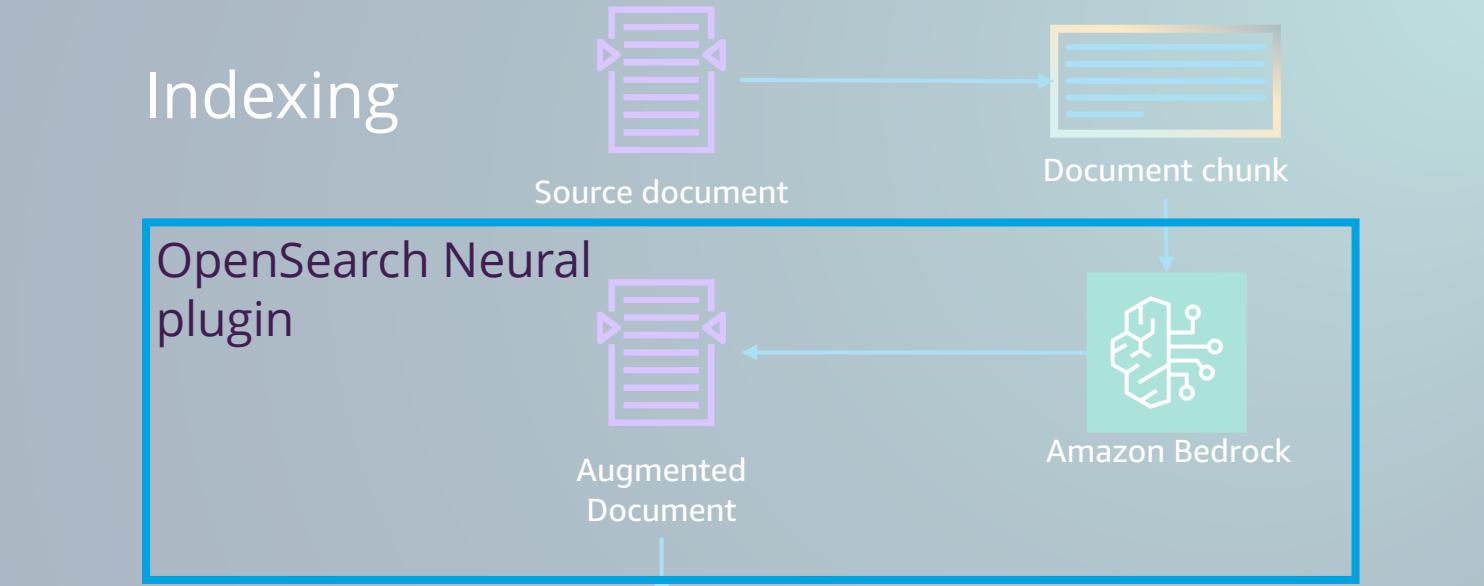
OpenSearch provides connectors to 3P model-hosting systems – e.g., Amazon Bedrock

Indexing: Select chunks from the source document, send to the 3P system for embeddings

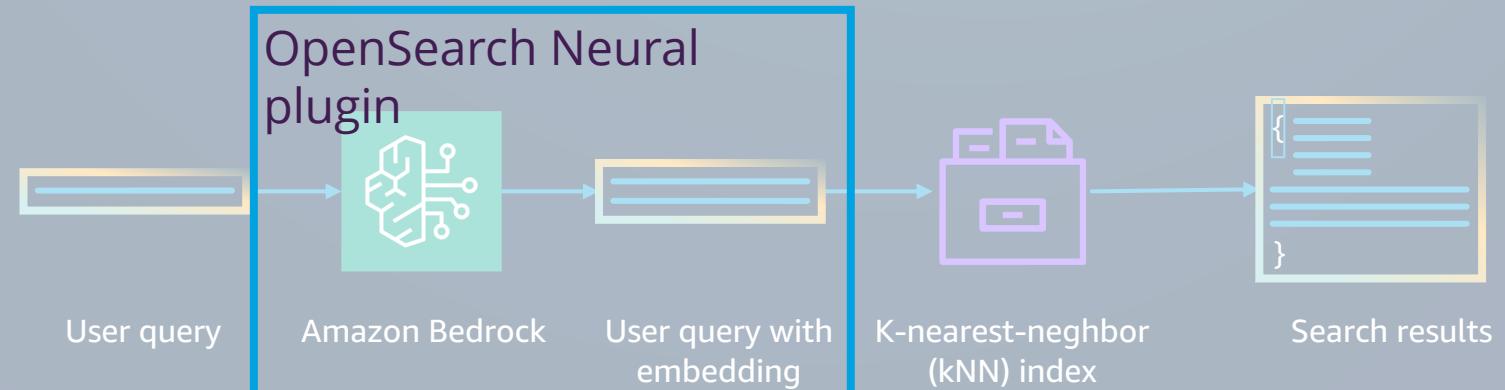
Search: Create embedding for the query then find nearest neighbors



Indexing



Search



Ingest, search pipelines for Neural

```
PUT /_ingest/pipeline/nlp-ingest-pipeline
{
  "description": "Text embedding pipeline",
  "processors": [ {
    "text_embedding": {
      "model_id": "xxxx",
      "field_map": {
        "chunk": "chunk_embedding"
      }
    }
  } ]
}
```

```
PUT /_search/pipeline/nlp-search-pipeline
{
  "description": "Hybrid search",
  "phase_results_processors": [ {
    "normalization_processor": {
      "normalization": {
        "technique": "min_max"
      },
      "combination": {
        "technique": "arithmetic_mean",
        "parameters": {
          "weights": [ 0.3, 0.7 ]
        }
      }
    }
  }]
}
```

Hybrid

bullets: Crafted with premium materials, these versatile black sneakers feature a sleek, minimalist look to complement any outfit while providing lasting comfort for urban exploration and adventures.
chunk_embedding: [0.078125, -0.24023438, ...]



```
PUT /_search/pipeline/nlp-search-pipeline
{
  "description": "Hybrid search",
  "phase_results_processors": [
    {
      "normalization_processor": {
        "normalization": {
          "technique": "min_max"
        },
        "combination": {
          "technique": "arithmetic_mean",
          "parameters": {
            "weights": [
              0.3,
              0.7
            ]
          }
        }
      }
    }
  ]
}
```

```
GET amazon_pqa/_search?search_pipeline=nlp-search-pipeline
{
  "query": {
    "hybrid": {
      "queries": [
        {
          "match": {
            "bullets": {
              "query": "stylin kicks"
            }
          }
        },
        {
          "neural": {
            "chunk_embedding": {
              "query_text": "stylin kicks",
              "model_id": "70fGJY8BoFwiDML8j-zu",
              "k": 5
            }
          }
        }
      ]
    }
  }
}
```

Sparse

Crafted with premium materials, these versatile black sneakers feature a sleek, minimalist look to complement any outfit while providing lasting comfort for urban exploration and adventures.

```
PUT /amazon_pqa_sparse
{
  "settings": {
    "default_pipeline": "sparse"
  },
  "mappings": {
    "properties": {
      "chunk_sparse": {
        "type": "rank_features"
      },
      "chunk": {
        "type": "text"
      }
    }
  }
}
```



Sparse

Crafted with premium materials, these versatile black sneakers feature a sleek, minimalist look to complement any outfit while providing lasting comfort for urban exploration and adventures.

```
PUT /_ingest/pipeline/sparse
{
  "description": "Sparse encoding",
  "processors": [
    {
      "sparse_encoding": {
        "model_id": "70fGJY8BoFwiDML8j-zu",
        "field_map": {
          "chunk": "chunk_sparse"
        }
      }
    }
  ]
}
```



OpenSearch



Amazon Sagemaker
sparse model

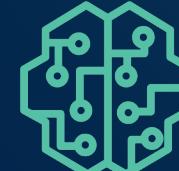


Sparse

Crafted with premium materials, these versatile black sneakers feature a sleek, minimalist look to complement any outfit while providing lasting comfort for urban exploration and adventures.



OpenSearch



Amazon Sagemaker
sparse model

```
PUT /amazon_pqa_sparse
{
  "settings": {
    "default_pipeline": "sparse"
  },
  "mappings": {
    "properties": {
      "chunk_sparse": {
        "type": "rank_features"
      },
      "chunk": {
        "type": "text"
      }
    }
  }
}
```

Query: Styling Kicks

```
GET amazon_pqa/_search
{
  "query": {
    "neural": {
      "chunk_embedding": {
        "query_text": "stylin kicks",
        "model_id": "70fGJY8BoFwiDML8j-zu",
        "k": 100
      }
    }
  }
}
```

Multimodal



Crafted with premium materials, these versatile black sneakers feature a sleek, minimalist look to complement any outfit while providing lasting comfort for urban exploration and adventures.



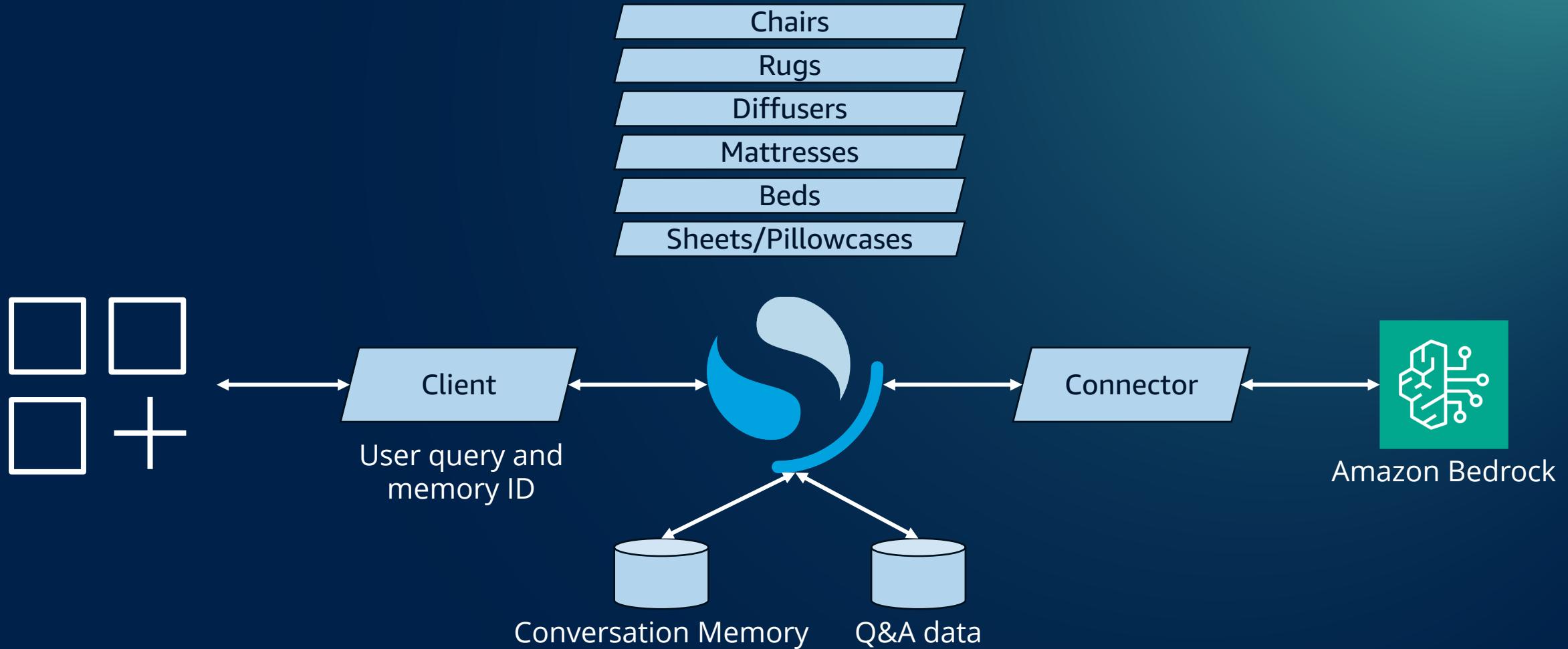
```
PUT /_ingest/pipeline/multimodal
{
  "description": "multimodal pipeline",
  "processors": [
    {
      "text_image_embedding": {
        "model_id": "afYQAosBQkdnhhBsK593",
        "embedding": "chunk_embedding",
        "field_map": {
          "text": "chunk",
          "image": "image_binary"
        }
      }
    }
  ]
}
```



Amazon Bedrock

```
GET /my-nlp-index/_search
{
  "size": 10,
  "query": {
    "neural": {
      "chunk_embedding": {
        "query_text": "stylin kicks",
        "query_image": "iVBORw0KGgoAAAANSU...",
        "model_id": "-fYQAosBQkdnhhBsK593",
        "k": 5
      }
    }
  }
}
```

Conversational search



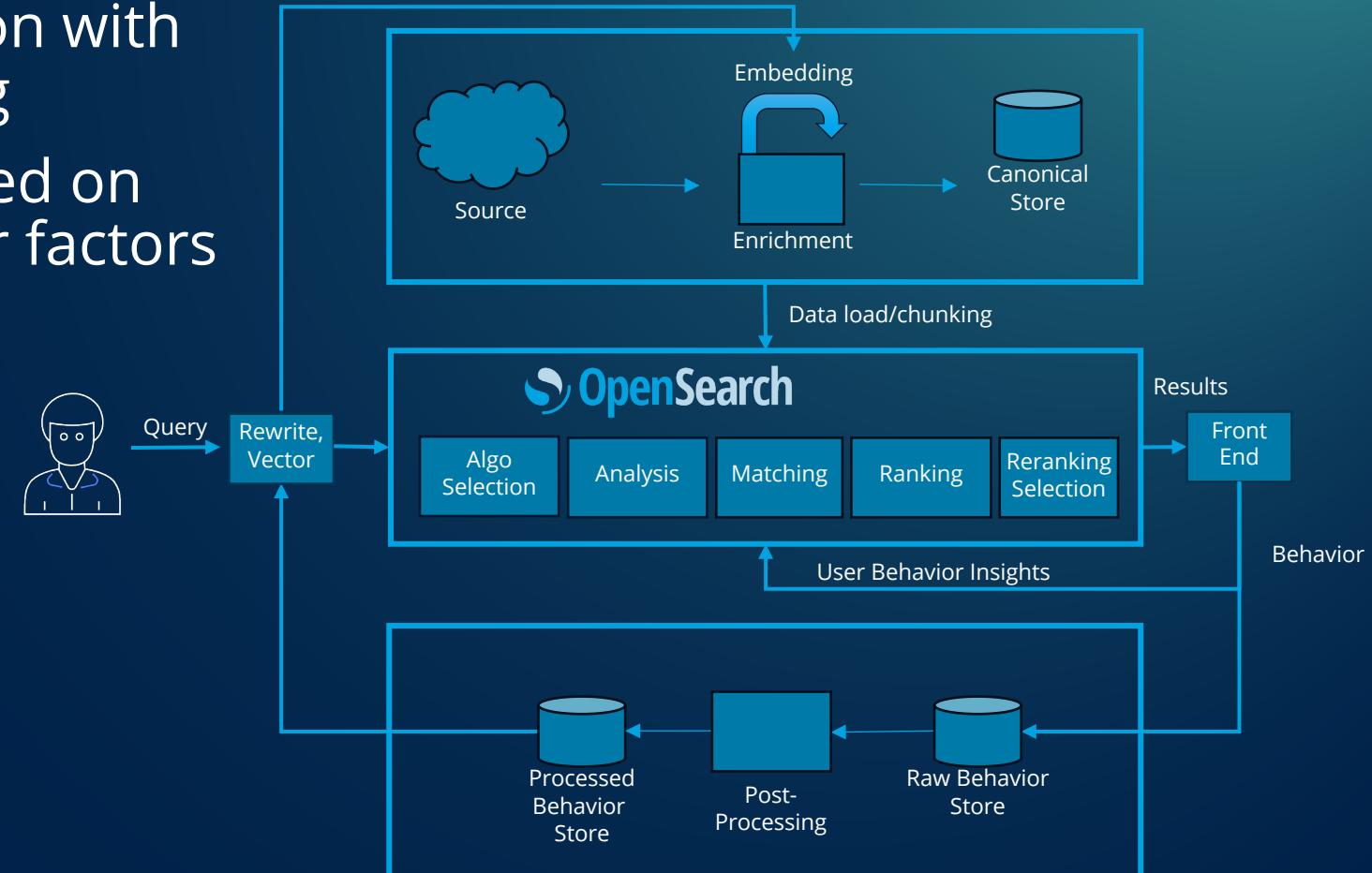
Where to from here?

PREPARING, ANALYZING, VECTORIZING, RANKING, AND RERANKING DOCUMENTS

Employing behavior information with User Behavior Insights tracking

Query algorithm selection based on concrete vs. abstract and other factors

Reranking selection based on past user behavior and other factors



O'REILLY®



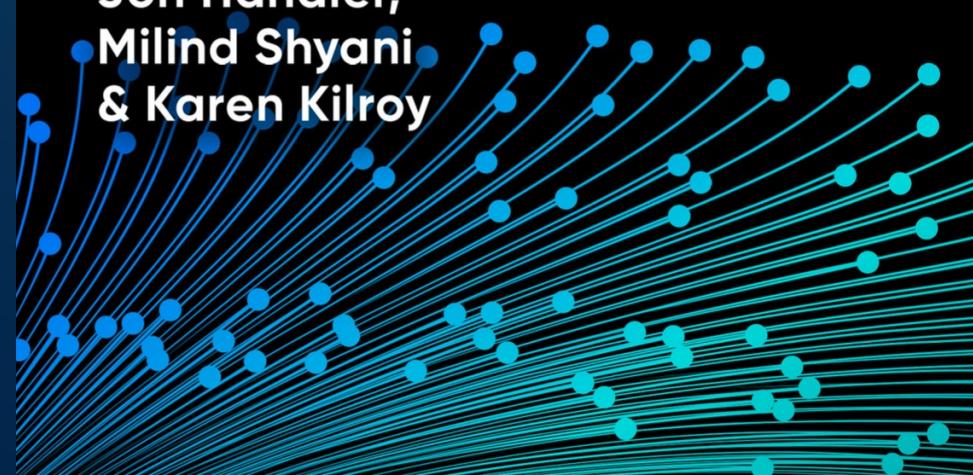
https://d1.awsstatic.com/aws-analytics-content/OReilly_book_Natural-Language-and-Search_web.pdf

 OpenSearch

Natural Language and Search

Large Language Models (LLMs)
for Semantic Search and
Generative AI

Jon Handler,
Milind Shyani
& Karen Kilroy

A large, abstract graphic on the right side of the page features a dense network of blue dots connected by thin blue lines, resembling a neural network or a complex semantic graph. The dots are concentrated in several distinct clusters along the bottom edge of the frame.

REPORT

Wrap

It was always about language – language to record information and language to find information

Improvements in language processing have enabled the capture and generation of language

How you interact with information in your enterprise, and in the world, mediated by technology is changing