Information Retrieval

Oct 1, 2024 @ Introduction to Human Language Technology

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Slides borrowed from SIGIR24 Tutorial "Neural Methods for Cross-Language Information Retrieval" What is Information Retrieval? (relevant) <u>Retrieve</u> information from a <u>storage</u> based on <u>user's</u> information <u>need</u>

Don't we have Google?

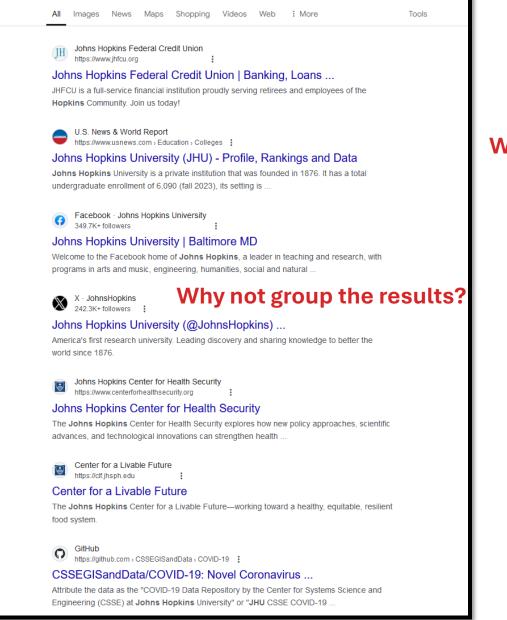
Yes, but Google is not all.

What if I'm looking for the person?

Why in a list?

Google

Johns Hopkins



X 🔱 💽 🔍

Why not read my mind?

Why asking me to read?

Google Search is just one implementation

Google trained us well!

- Even faster?
- Smarter?
- Cross language?

Hard Matching Problem

- Text to text
 - Search in notes
 - Cross language search
 - Cross domain search
- Text to other modalities
 - Image search
 - Video search

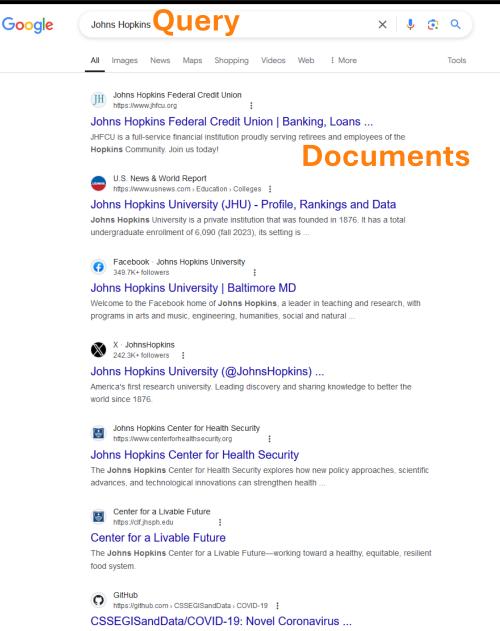
Different Search Process

- Iterative search
 - e.g., electronic discovery and systematic review
- Conversational search
 - Alexa search
- Recommendation systems
 - Implicit queries
- (Set Retrieval)

Core Problem

- Rank relevant document at top
- Do it fast

Ranked List



Attribute the data as the "COVID-19 Data Repository by the Center for Systems Science and Engineering (CSSE) at Johns Hopkins University" or "JHU CSSE COVID-19 ...

Design Space

Effectiveness

- Definition of relevancy
- How to model relevancy

Efficiency

- How fast
- Fast at what stage

Agenda

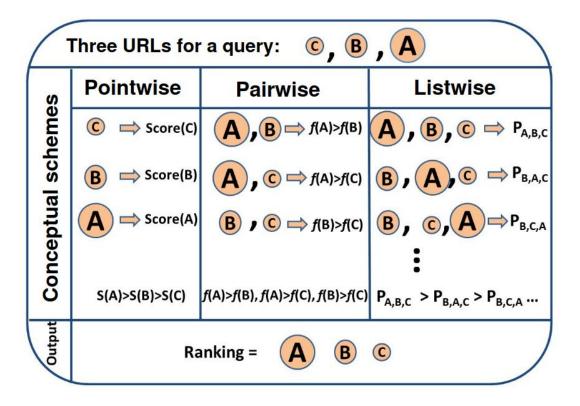
- What is information retrieval?
- Retrieval Modeling and Pipeline
 - Statistical and Neural
- Evaluation
- State of IR Research and active research problems

Retrieval Modeling and Pipeline

Modeling relevancy and do it fast

Three main modeling strategies

- Pointwise
- Pairwise
- Listwise
- And combinations of them



https://medium.com/vptech/learning-to-rank-at-veepeeed420fd828e5

Statistical Models

В

$$score(D, Q) = \sum_{i=1}^{n} \text{How important the term is} \times \text{How often the term appear in the D}$$
For each query term
$$score(D, Q) = \sum_{i=1}^{n} \text{Inverted document frequency} \times \text{Term frequency}$$
For each query term
$$TF-IDF$$

$$score(D, Q) = \sum_{i=1}^{n} log \frac{N}{n_t} \times log(f(q_i, D) + 1)$$

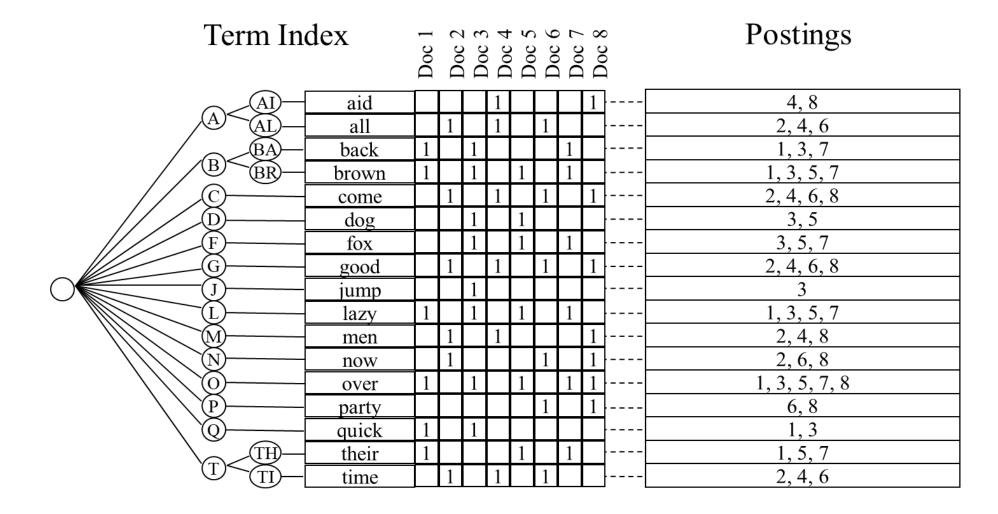
$$BM25$$

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

How to make it fast?

- "Fast" in responding to queries
- Better data structure
- Preprocess the data

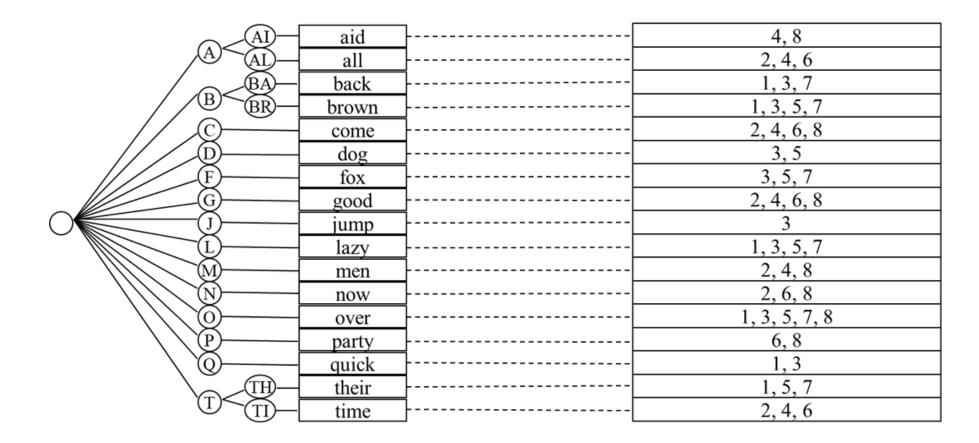
Inverted Index



Inverted Index

Term Index

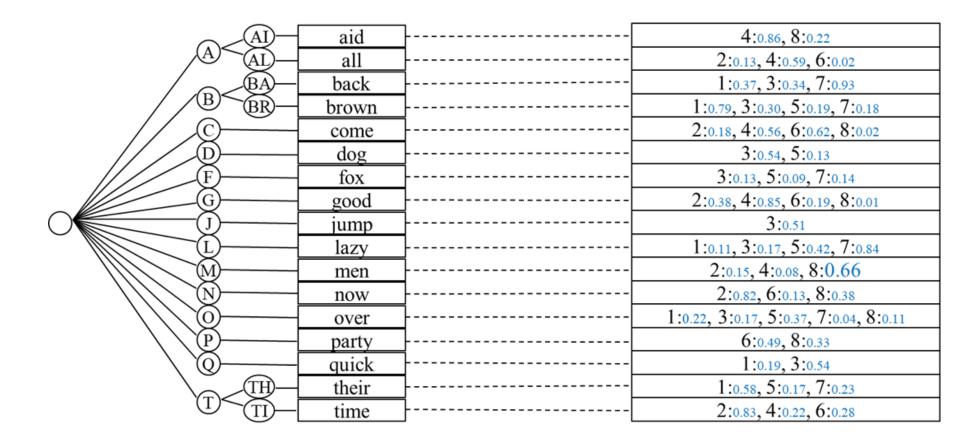




Inverted Index

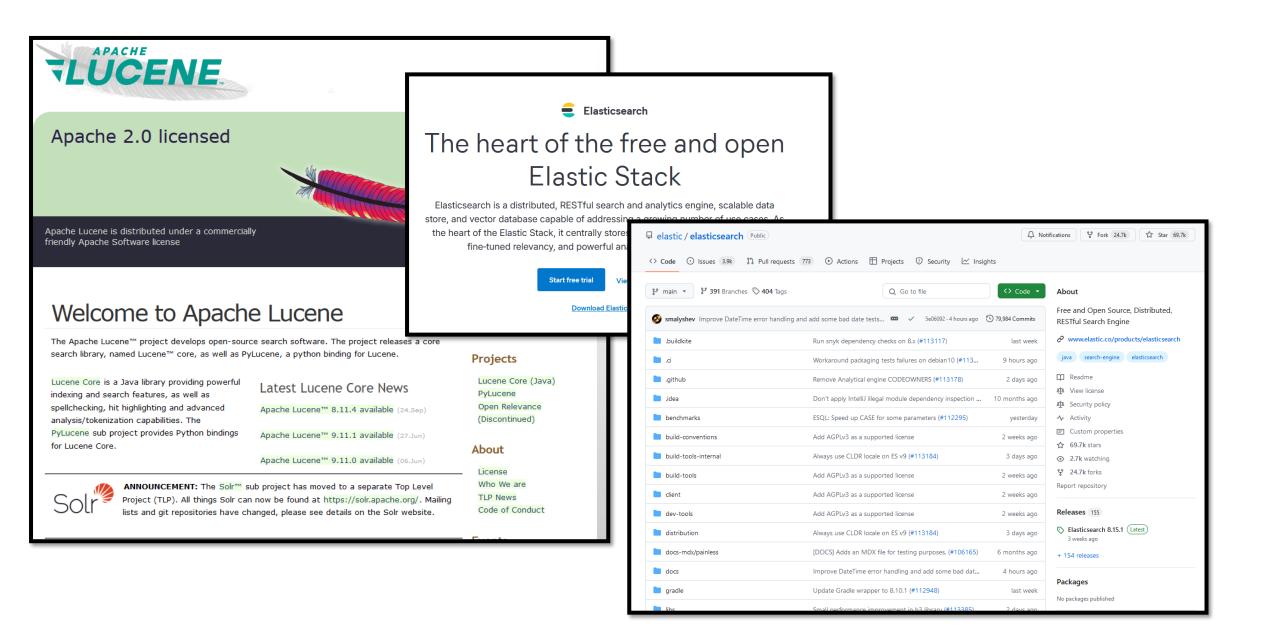
Term Index

Postings



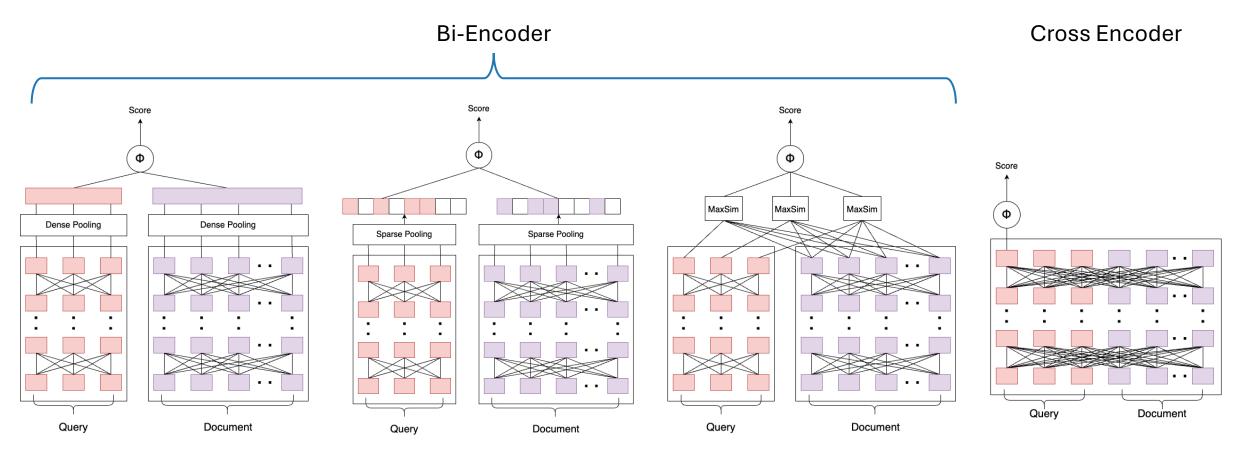
Two-Stage System

- Offline preprocessing and indexing
 - Define retrieval unit
 - Tokenization
 - Build the inverted index
- Online query serving
 - Traverse the inverted index and score it



Can we go beyond surface forms?

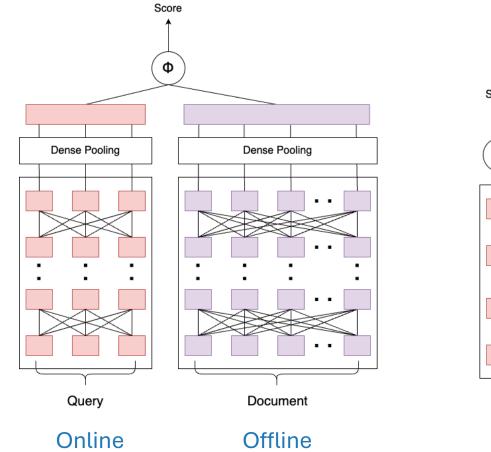
neural language models



One Dense Vector Per Sequence e.g., DPR

One <u>Sparse</u> Vector Per Sequence e.g., SPLADE

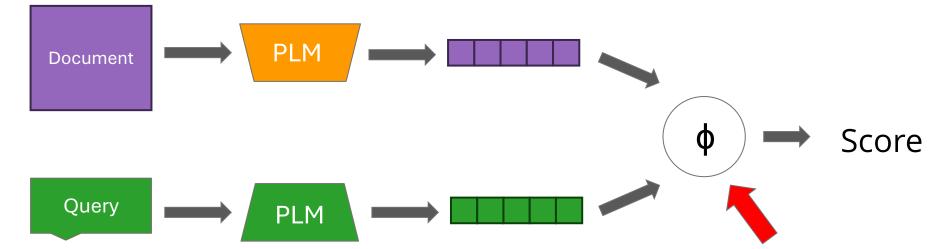
Multiple Dense Vectors Per Sequence e.g., ColBERT Joint Encoder e.g., monoBERT

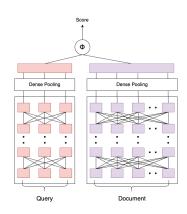


Both Online

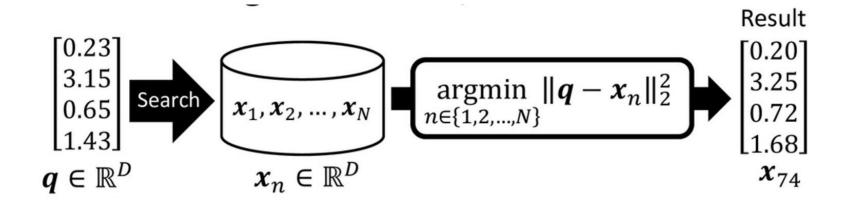
Separate query and document processing

One Vector per Query, One Vector per Document

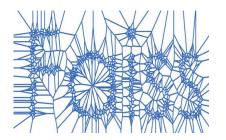




Nearest Vectors aka Neighbors



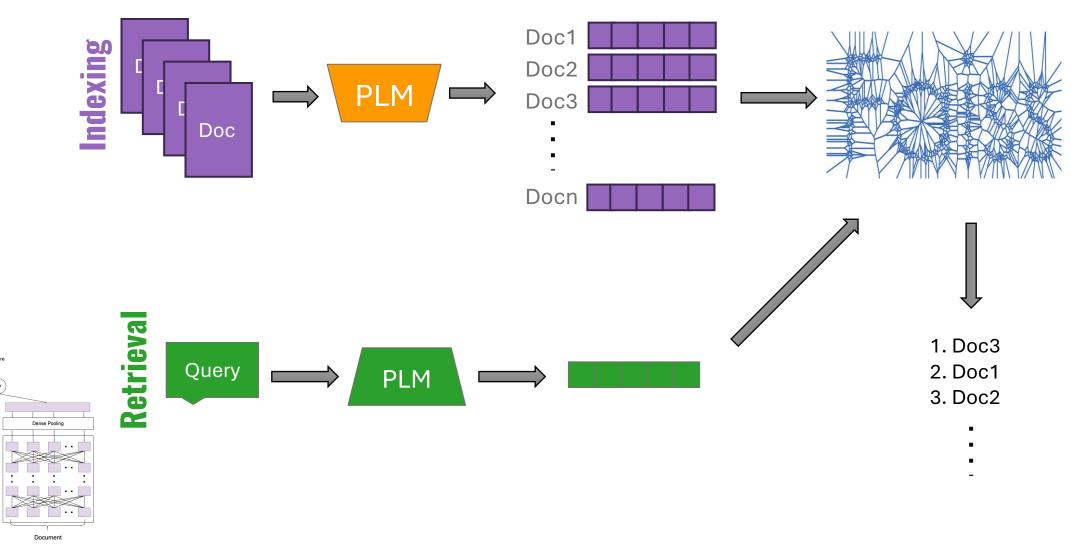
- Linear Search
 - Slow (scales linearly in size of document collection)
- Approximate Methods (e.g., Product Quantization) → ANN
 - Faster Search
- Runtime Efficiency vs Effectiveness

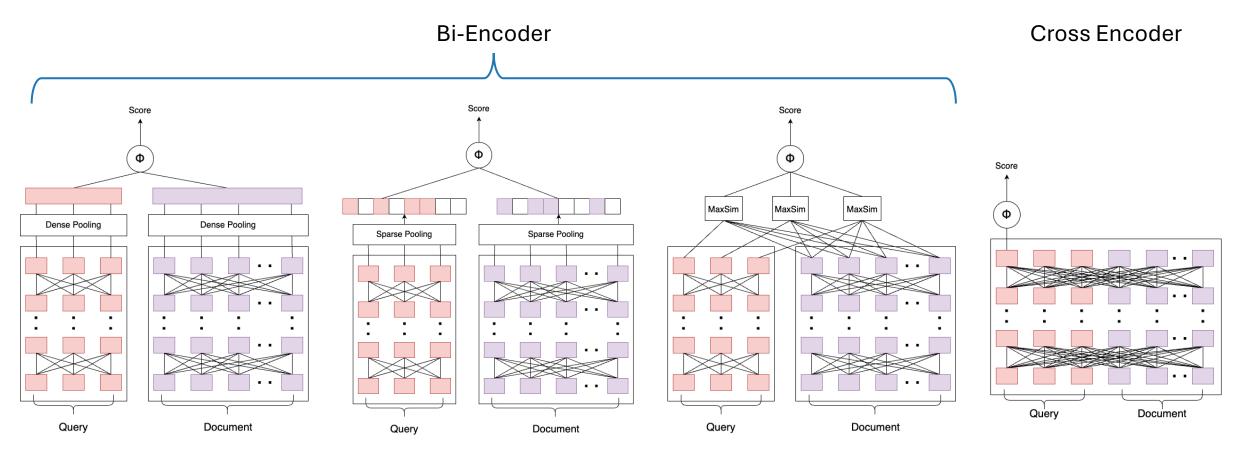


DPR Indexing and Retrieval

Dense Pooling

Query

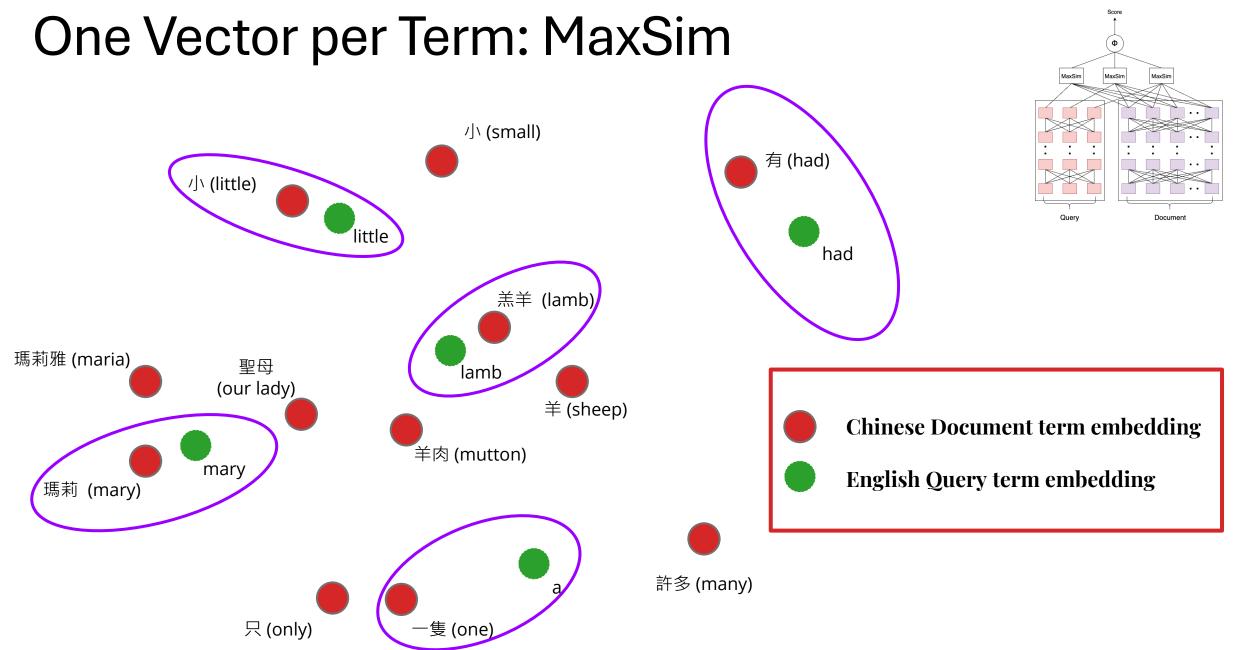


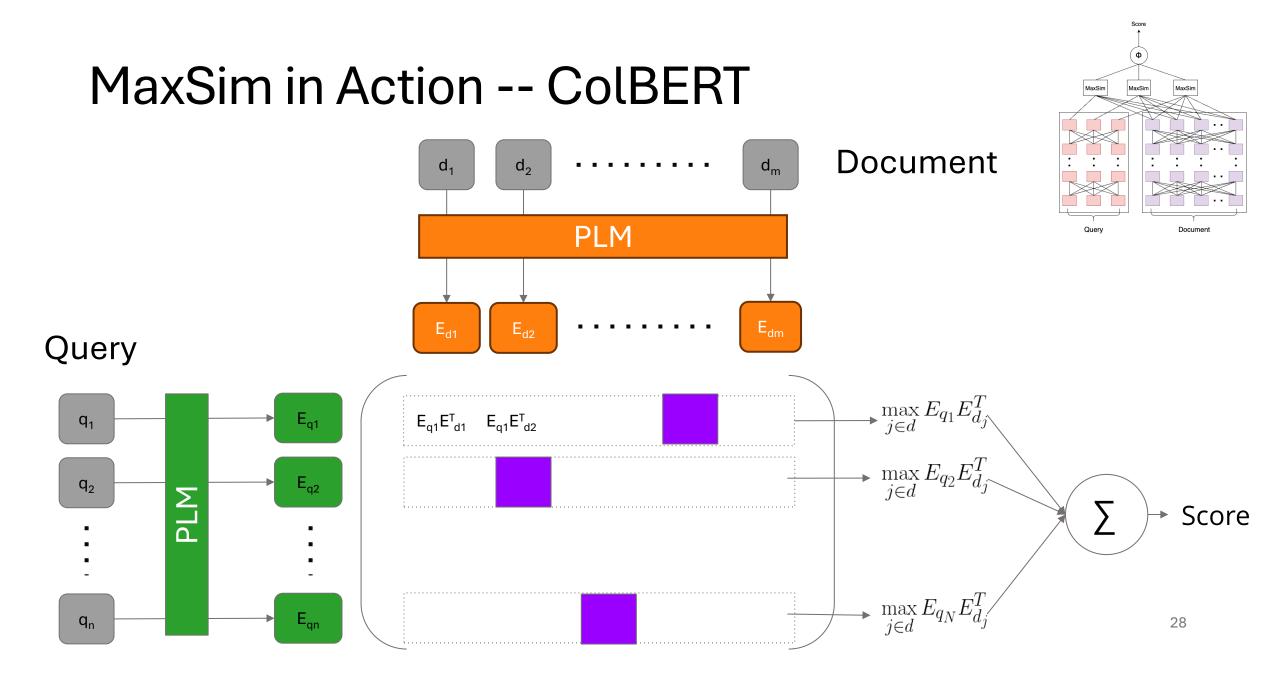


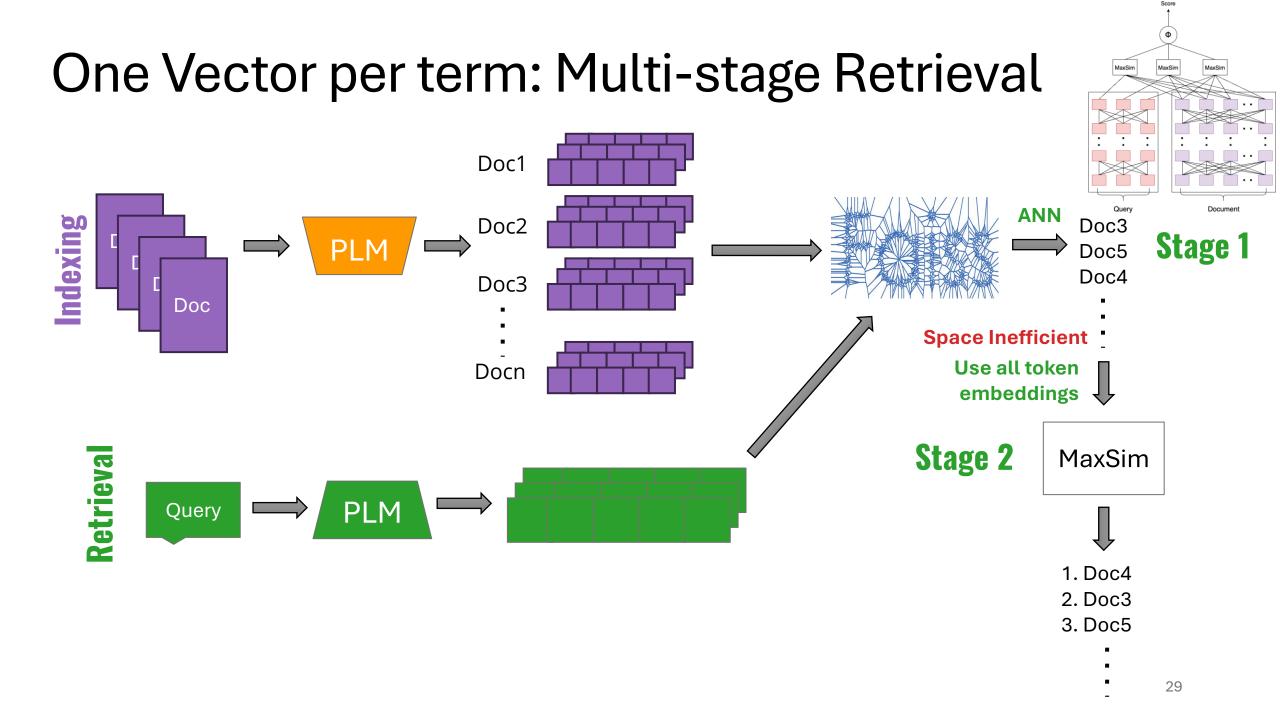
One Dense Vector Per Sequence e.g., DPR

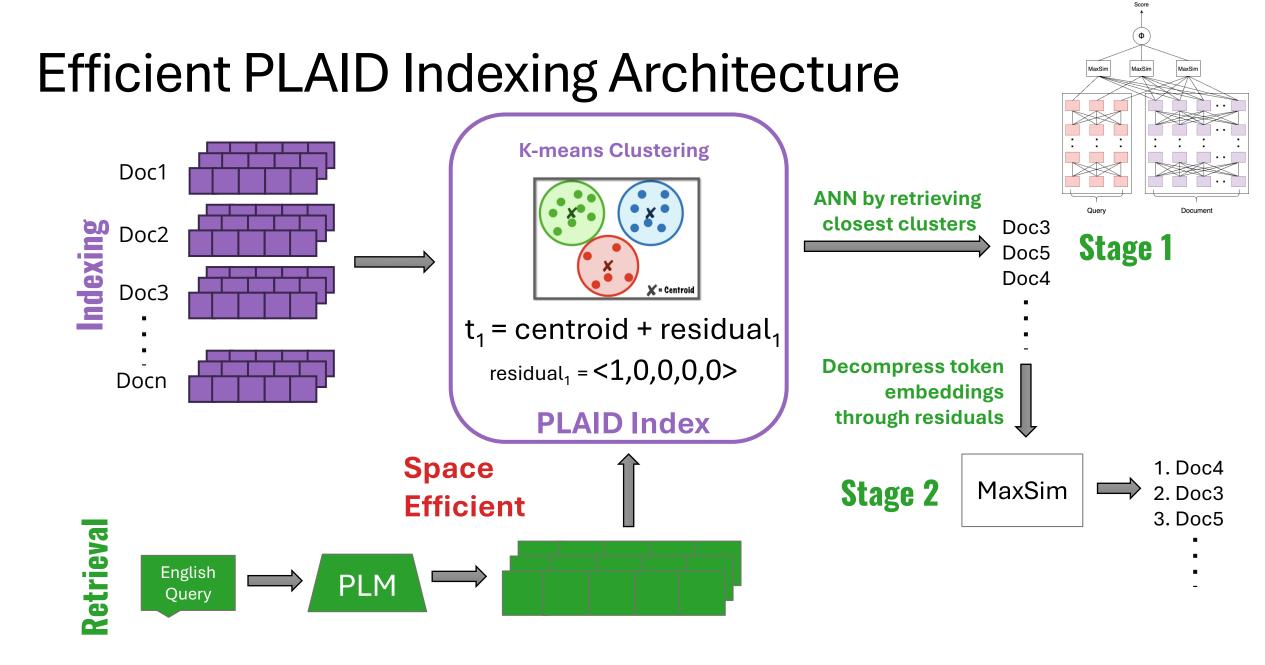
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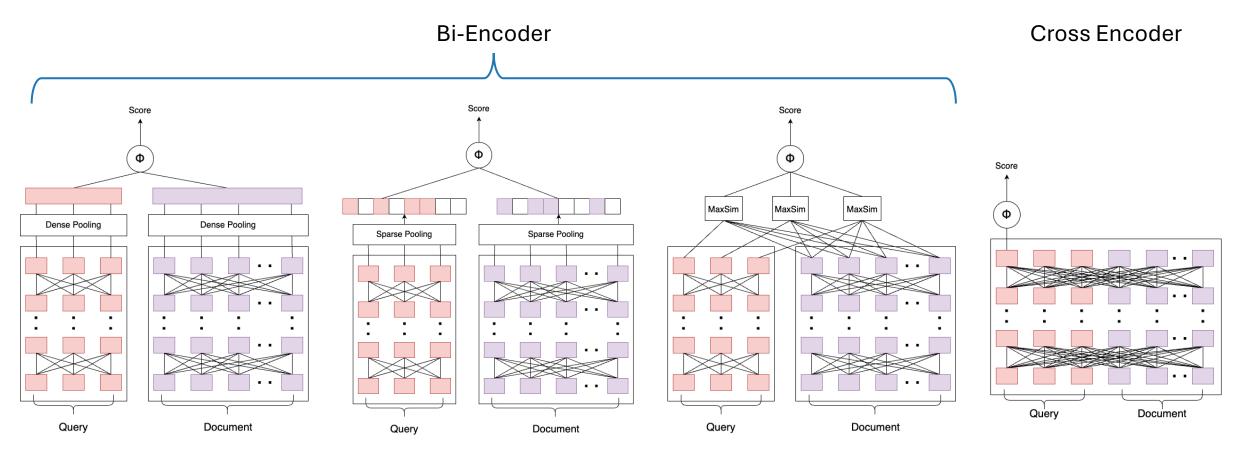








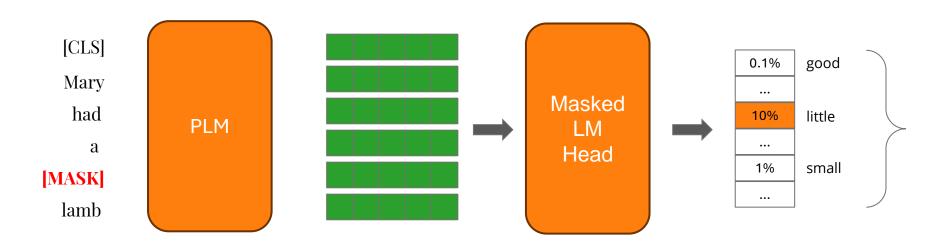
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docs	update theme	2 years ago 최	docs	Updated setup.py + intro nb	10 months ago	Collection of trained model and teacher scores for distillation for paper "Translate-Distill" Code: https://
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README.md	Update README.md (#8)	4 months ago Re	LoTTE.md	Add LoTTE download link	2 years ago	Translate-Distill: Learning Cross-Language Dense Retrieval by Translation and Distillation
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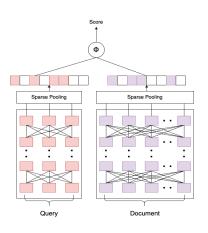
One Dense Vector Per Sequence e.g., DPR

One <u>Sparse</u> Vector Per Sequence e.g., SPLADE

Multiple Dense Vectors Per Sequence e.g., ColBERT Joint Encoder e.g., monoBERT



High-dimensional Vector: Masked LM



PLM Vocabulary

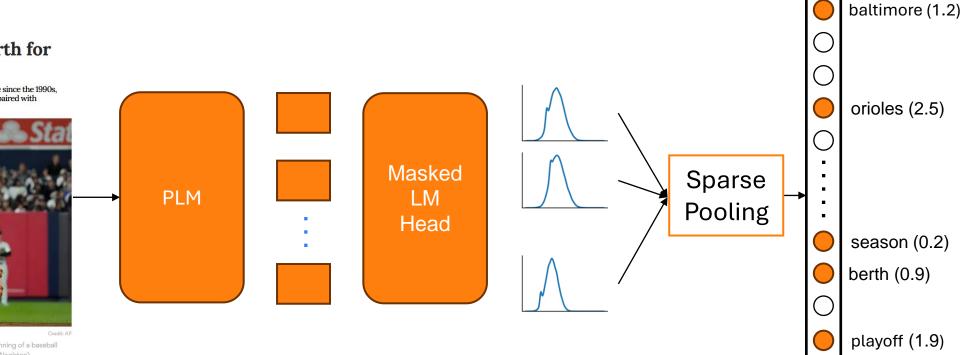
SPLADE

Baltimore Orioles clinch playoff berth for 2nd straight season

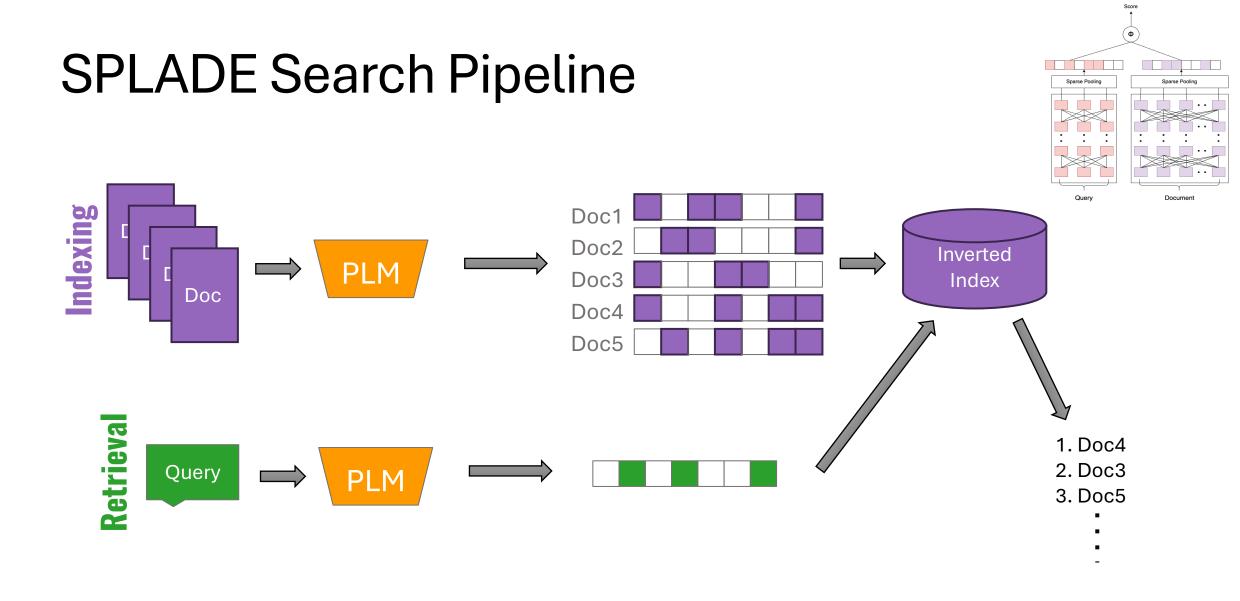
The Baltimore Orioles are headed to the playoffs in consecutive years for the first time since the 1990s, clinching no worse than a wild-card berth with a 5-3 win over the New York Yankees paired with Minnesota's loss to Miami



Baltimore Orioles' Anthony Santander runs the bases after hitting a home run during the sixth inning of a baseball game against the New York Yankees, Tuesday, Sept. 24, 2024, in New York. (AP Photo/Bryan Woolston)

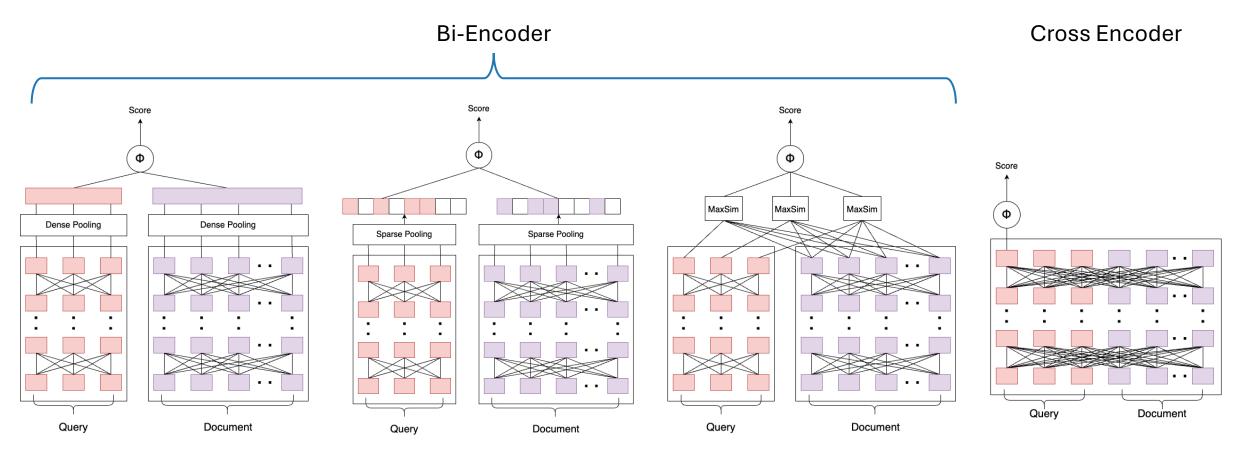


Predicted Vocabulary



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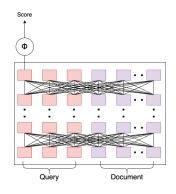
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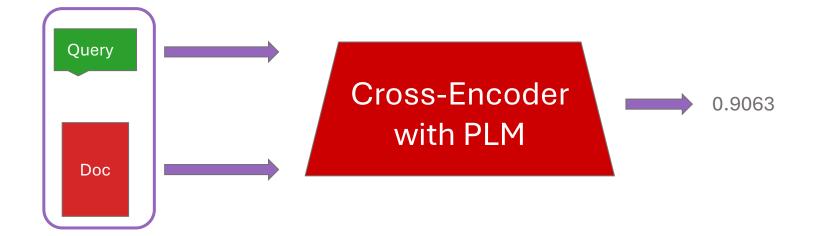
One Dense Vector Per Sequence e.g., DPR

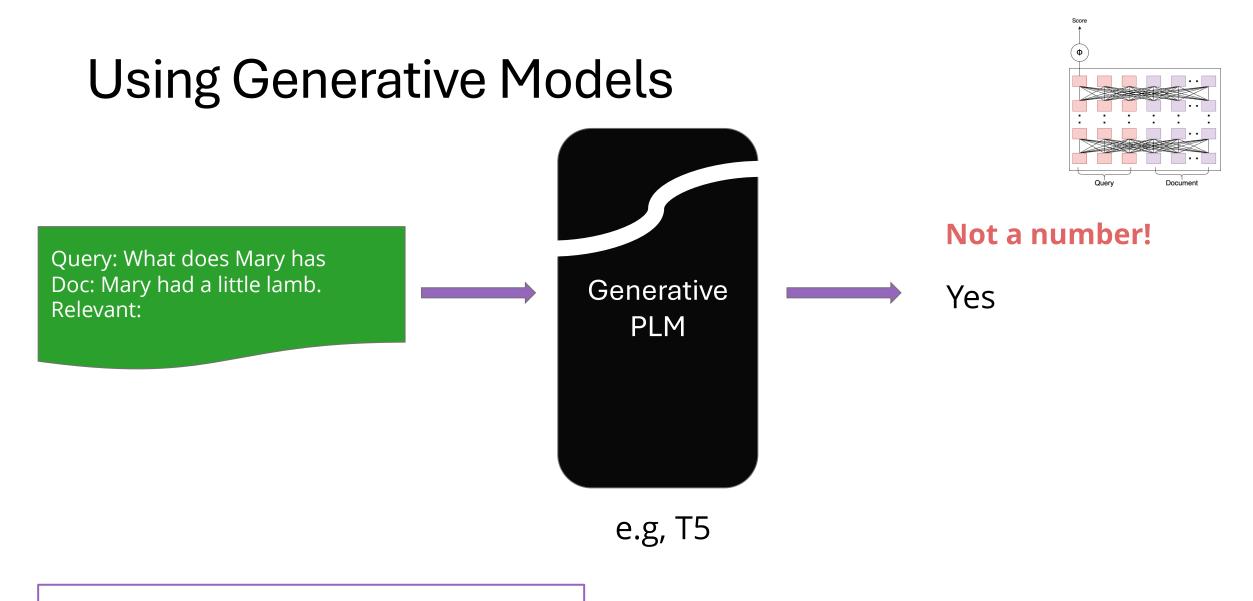
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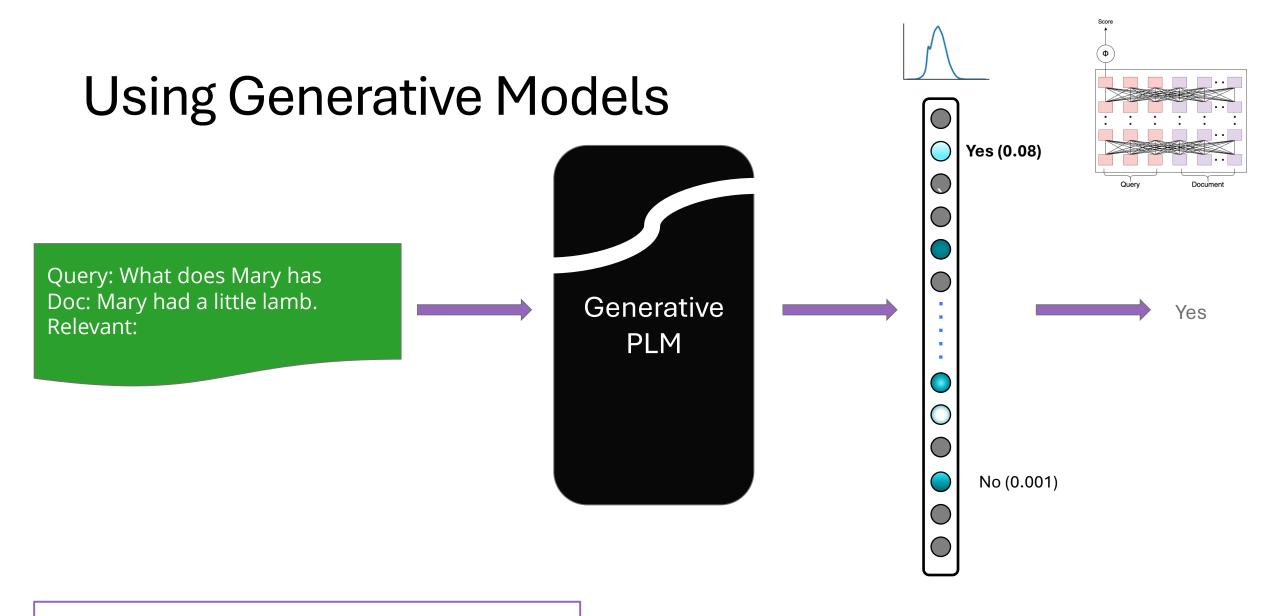
Multiple Dense Vectors Per Sequence e.g., ColBERT Joint Encoder e.g., monoBERT

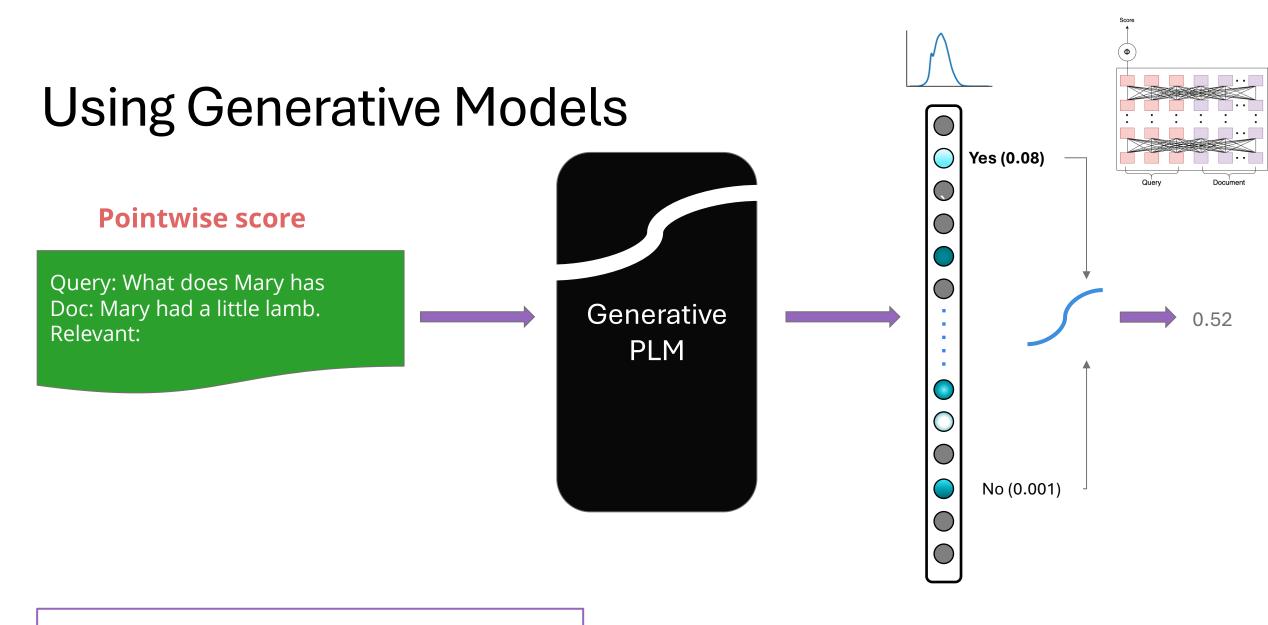


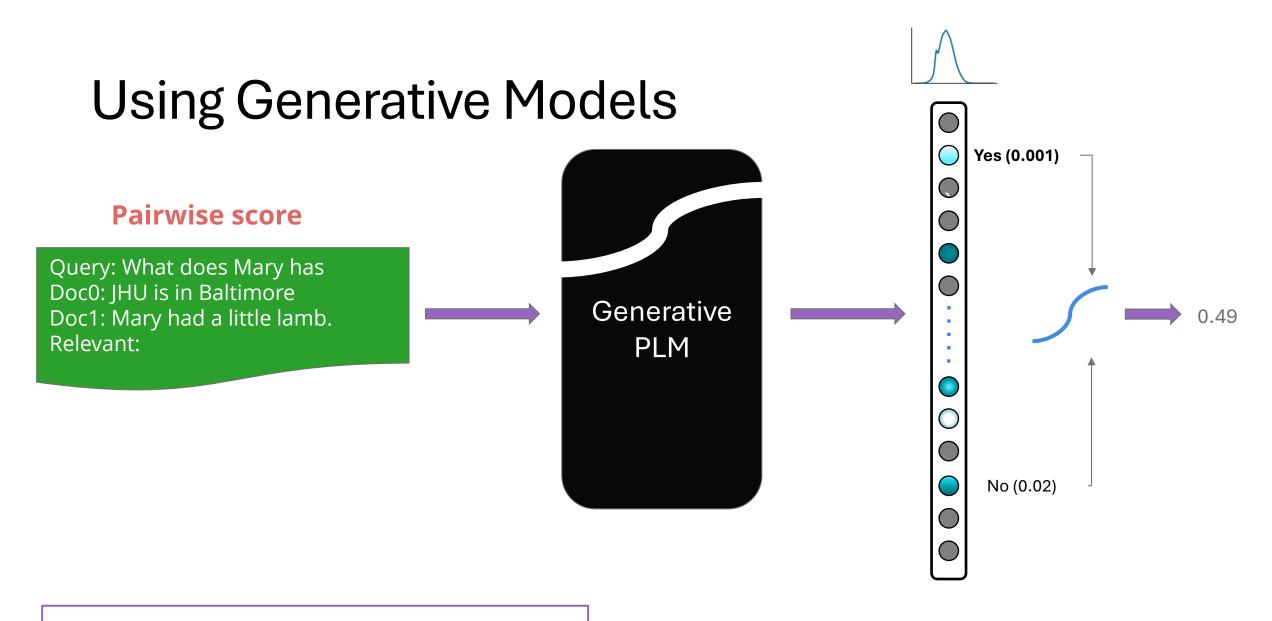
Cross-Encoder

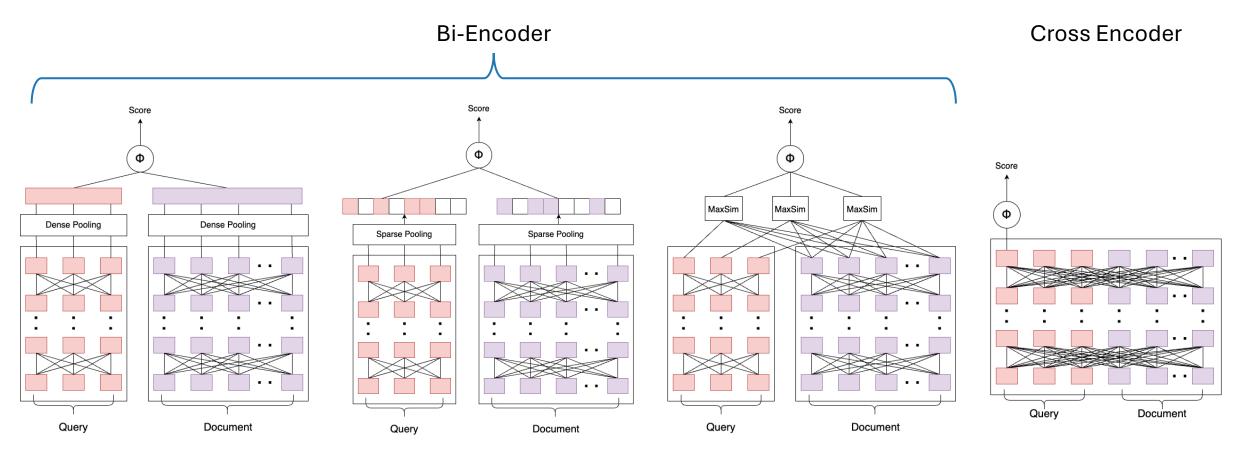








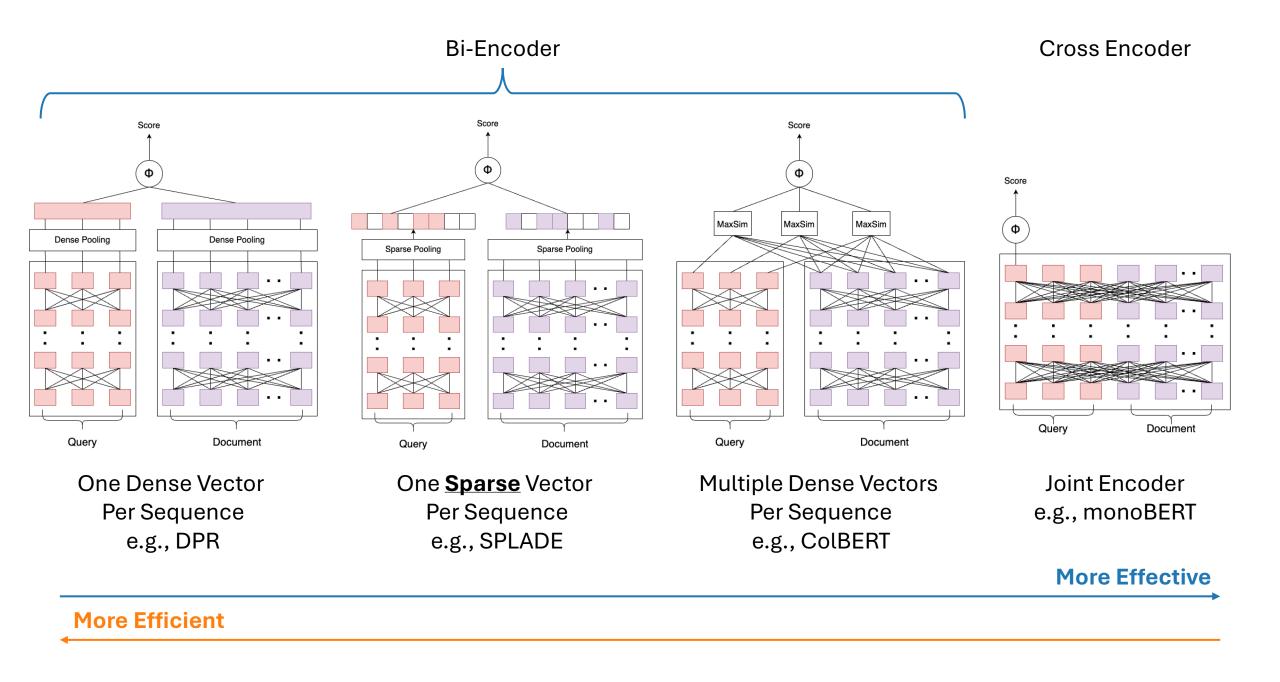




One Dense Vector Per Sequence e.g., DPR

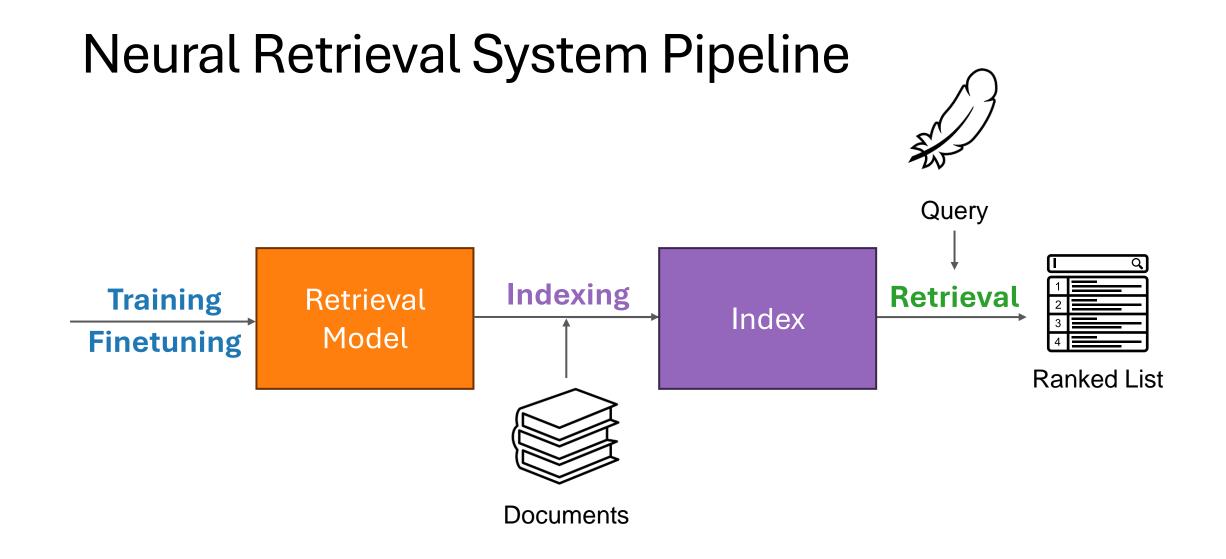
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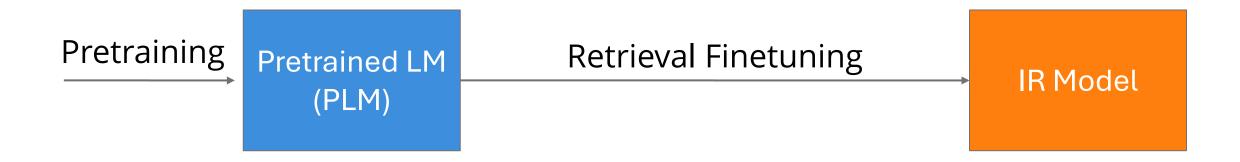


Retrieve-and-Rerank System Combinations





PLM to IR Model



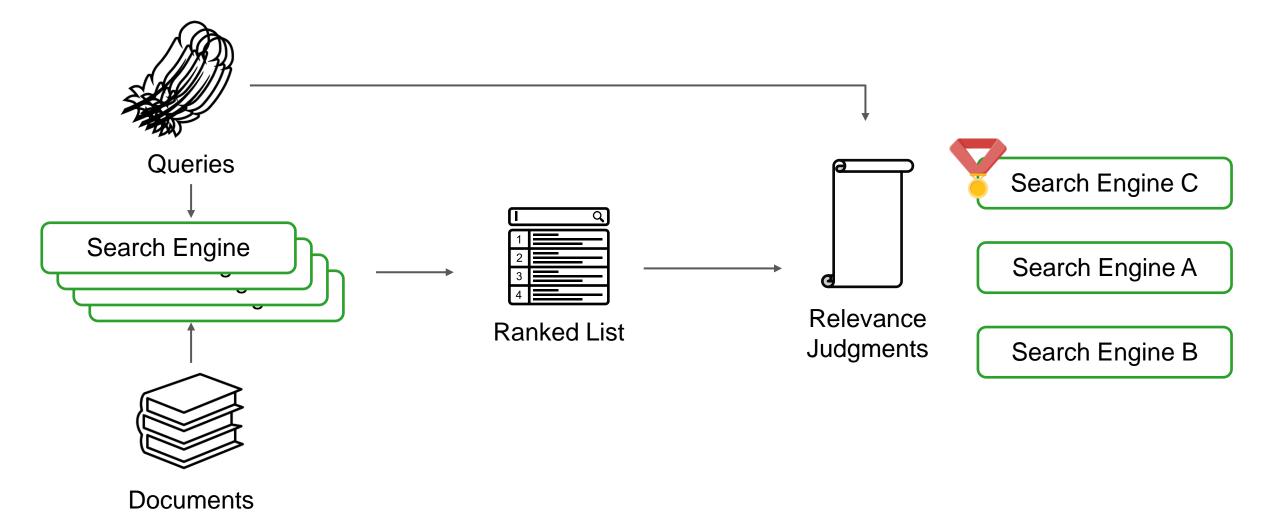
- Align the representation
- Model "relevancy"

Evaluation

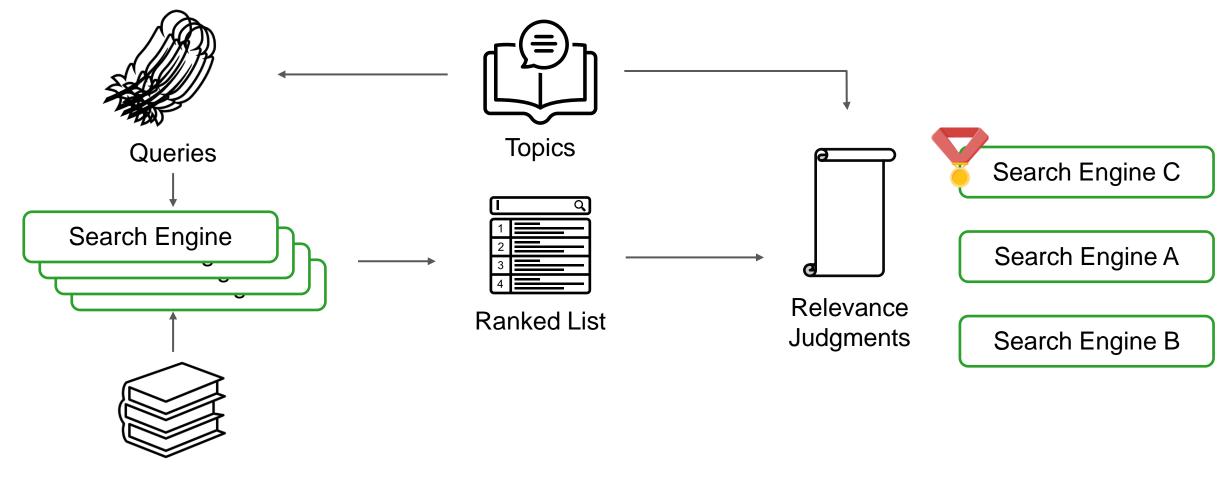
Which system is better?

What is Information Retrieval? (relevant) <u>Retrieve</u> information from a <u>storage</u> based on <u>user's</u> information <u>need</u> Which system retrieve more relevant information?

Cranfield Paradigm Evaluation

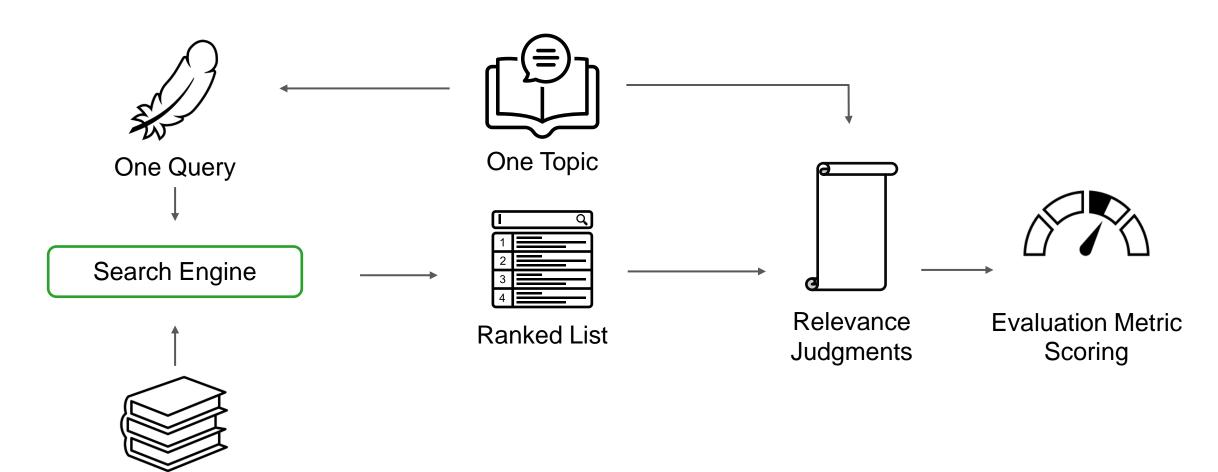


Cranfield Paradigm Evaluation



Documents

Cranfield Paradigm Evaluation





IR-Specific Issues

- Topics vs Queries
 - Clear intent vs an expression of such intent
- Relevant vs related
 - Fulfilling the information need or not
- Relevance Judgements vs Labels
 - Opinion vs "fact"
- Ranked retrieval metrics
 - Measuring the quality/effectiveness of a ranked list

55

IR Metrics

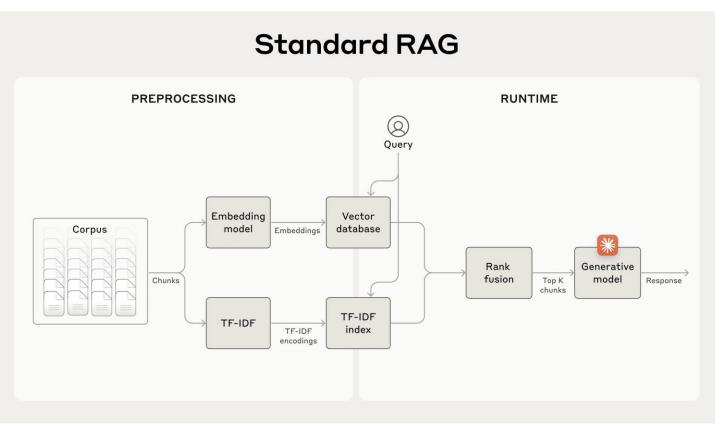
- Effective Metrics
 - Mean Average Precision
 - Normalized Discounted Cumulative Gain
 - Recall@k
- Efficiency Metrics
 - Indexing time
 - Index disk space
 - Query latency (average search time per query)



State of IR Research

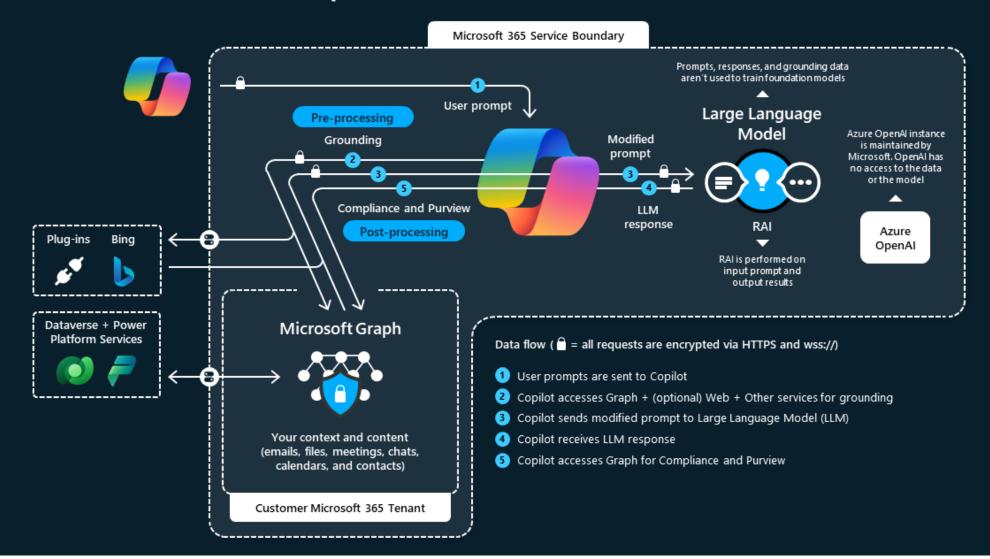
Retrieval-Augmented Generation

- Is everything a RAG problem?
- What is the right retrieval model/system for RAG?
- IR going away?

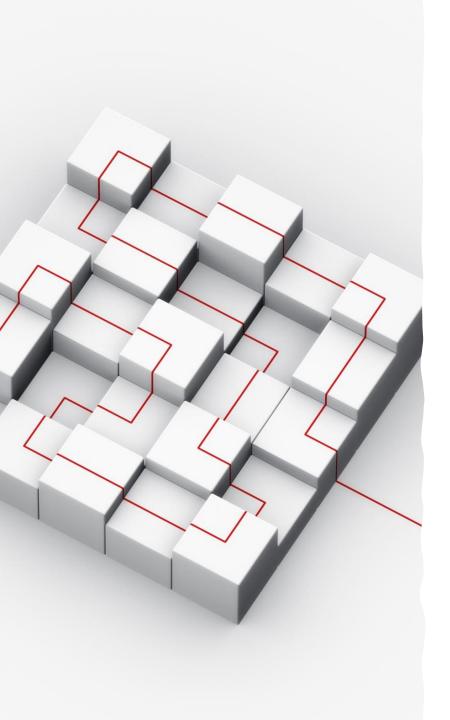


https://www.anthropic.com/news/contextual-retrieval

Microsoft Copilot for Microsoft 365 architecture

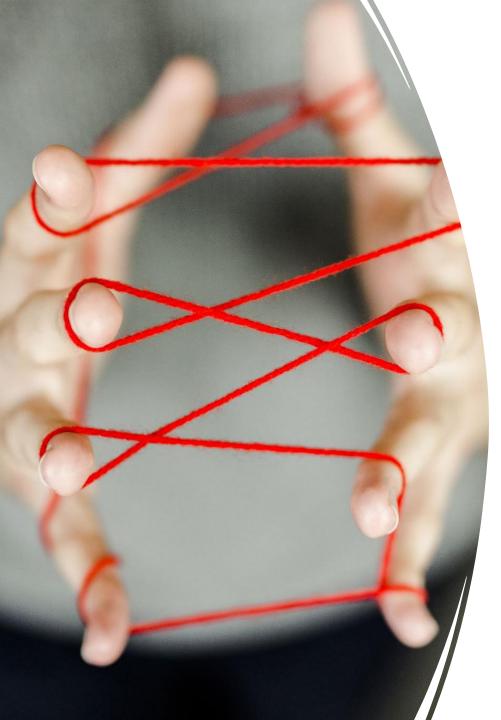


https://learn.microsoft.com/en-us/copilot/microsoft-365/microsoft-365-copilot-overview



Better Retrieval Models

- More effective
 - Better/larger neural models
 - Better architecture?
 - Under harder setup, e.g., scholar search, multilingual, cross-modal, etc
- More efficient
 - Faster at query time
 - Less resource footprint, e.g., memory, storage, compute, etc
- Other qualities
 - Fairness, diversity, etc



Other Retrieval Problems

- Conversational
 - Guessing intent, finding the "right" information to serve
- Iterative/interactive/human-in-theloop
 - Rounds of interactions
- Generative
 - Returning a piece of text



Evaluation

- What to measure
 - and when would it fail
- How to measure
 - Generative text? Citations?
- "Better" evaluation collection
 - Not necessarily larger