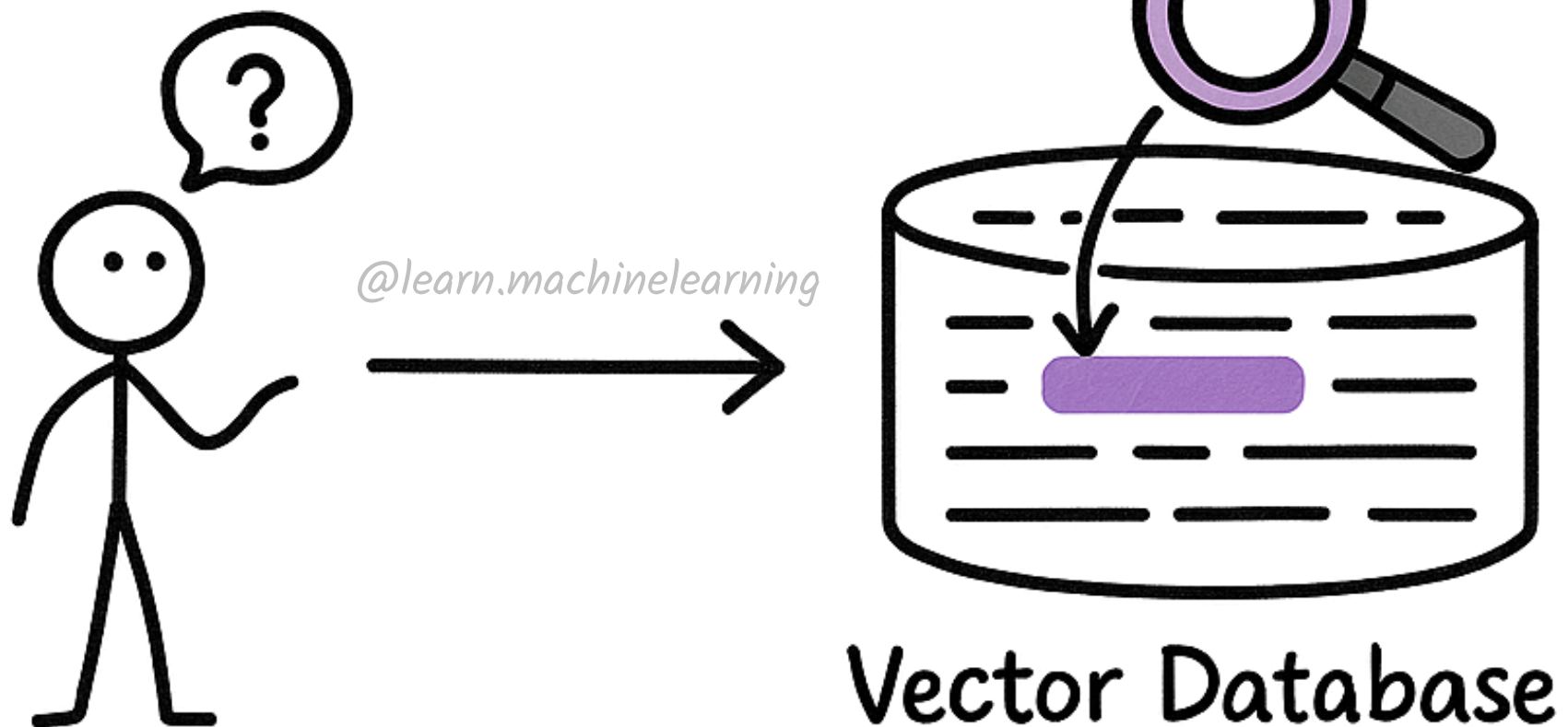


GENERATIVE AI SERIES: DAY 20

The Art of Retrieval: Finding the Right Information

Yesterday, we built the AI's knowledge library.
Today, we learn how it searches that library to
find exactly what you need, from millions of
pieces of info!



Your Question vs. & Millions of Chunks

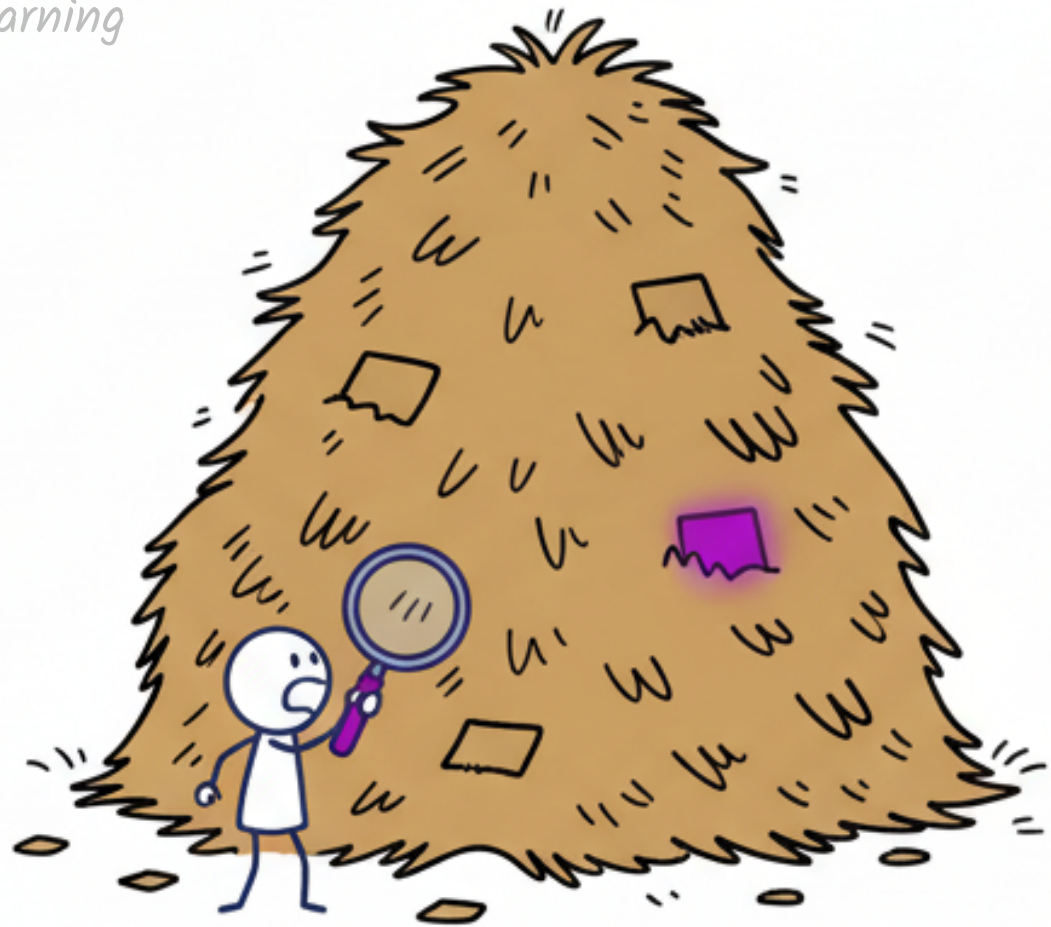
Imagine your RAG system has processed thousands of documents, resulting in *millions* of small, embedded text chunks.

User Question: "What are the Q3 sales projections?"

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The Problem: How does the system instantly find the few relevant chunks from its massive library?

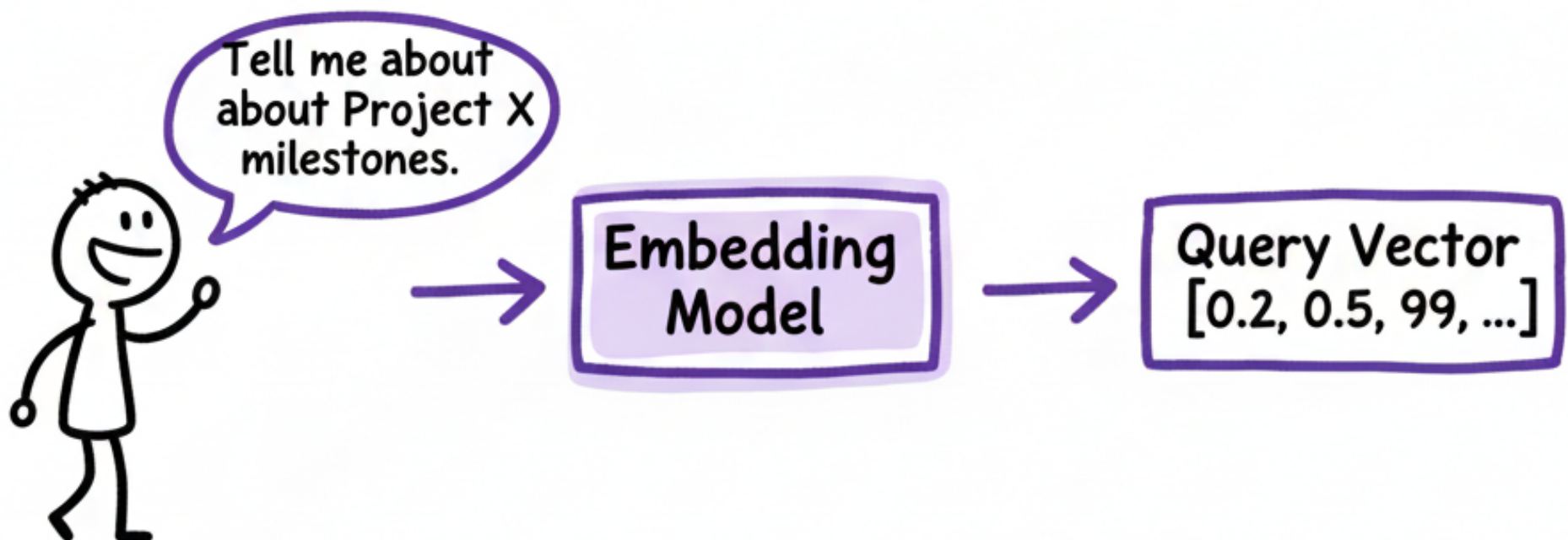
The Goal: Fast, accurate, and contextually rich



Day 20: The Art of Retrieval

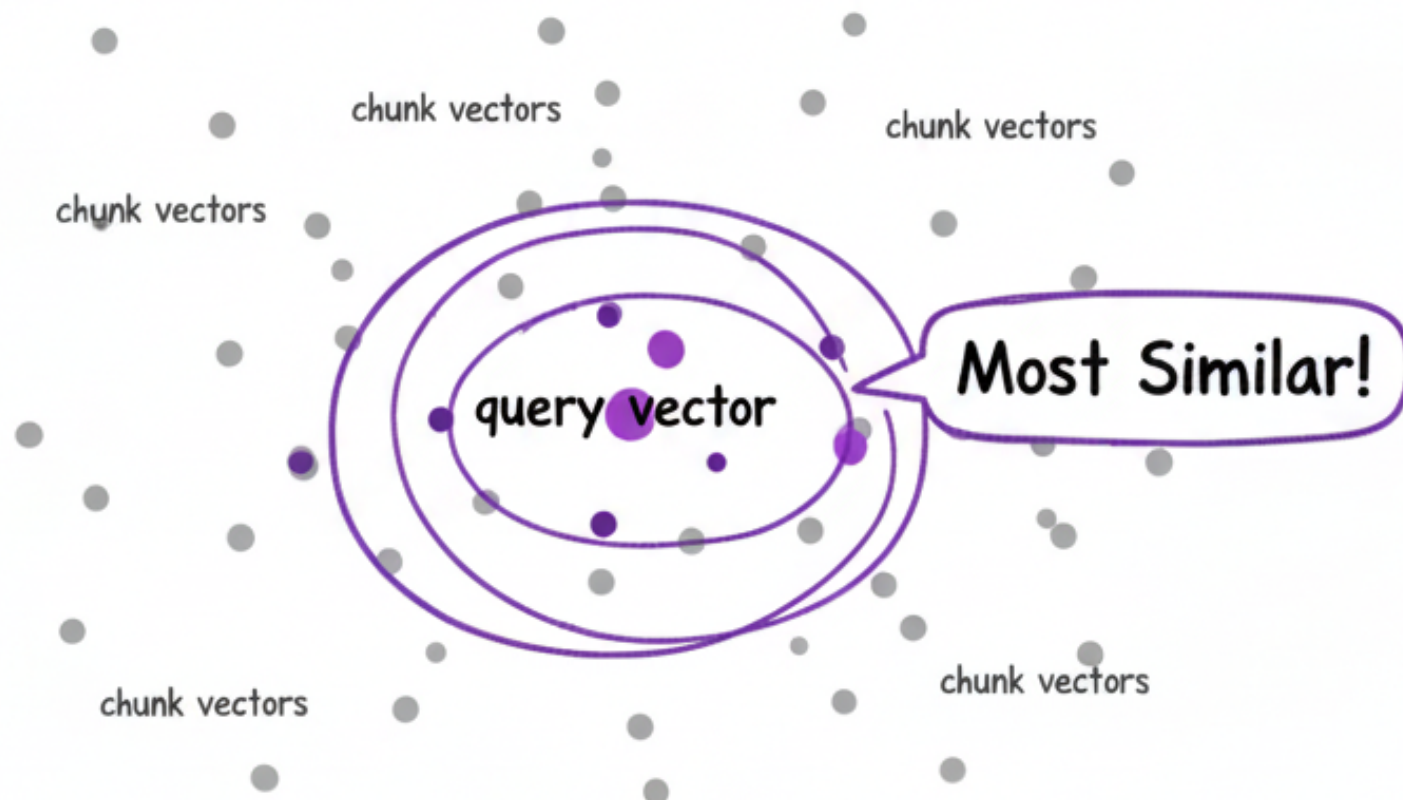
Step 3: The First Step: Turning Your Question into Numbers

- Just like we embedded our document chunks into vectors, we do the same for your question.
- **User Asks:** Tell me about Project X milestones.
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- **Embedding Model:** Converts this question into query vector (a list of numbers).
- **Why?** Because computers are great at comparing numbers to find similarities!



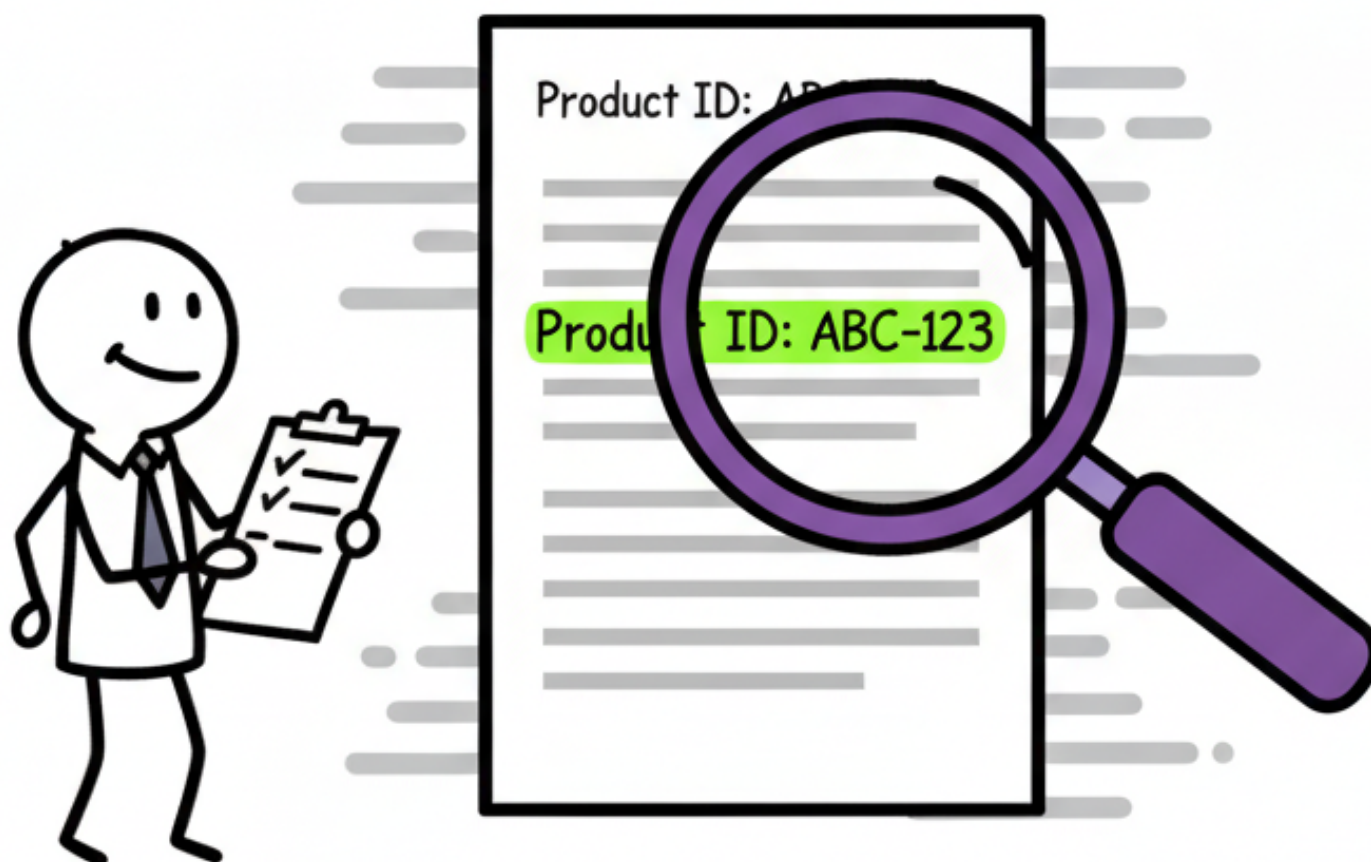
Vector Search: Finding Semantic Neighbors

- Now we compare the query vector to all the document chunk vectors stored in our database.
- Similarity Metrics: We use math (like Cosine Similarity) to measure how “close” two vectors are in multi-dimensional space.
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- Closer Vectors = More Relevant Meaning!
- Result: The system retrieves the top K (e.g, 5-10) most similar chunks. These are our potential answers!



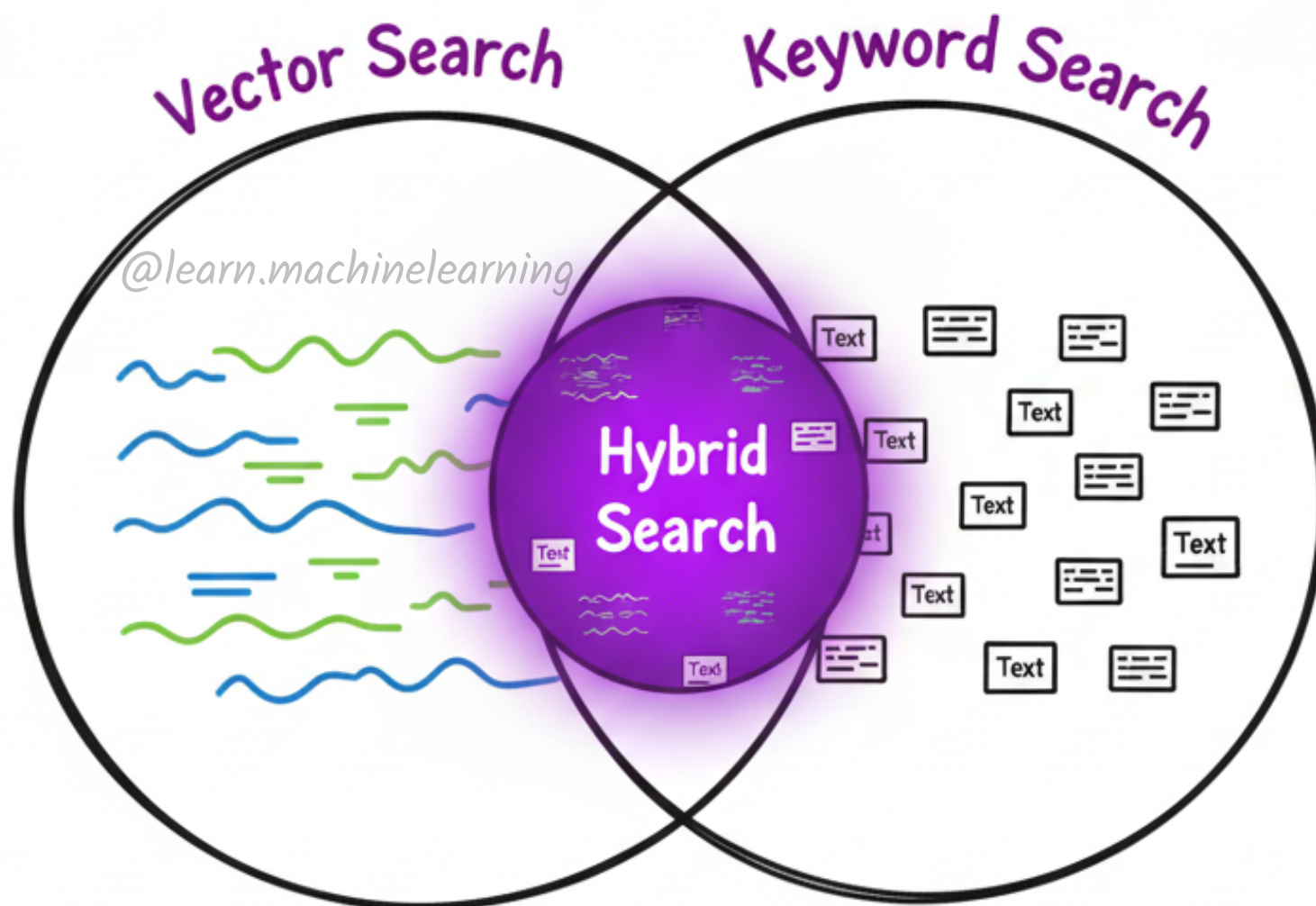
Beyond Vectors: Keyword Search (BM25)

- Sometimes, pure semantic similarity isn't enough. What if you're looking for exact names, product codes, or very specific terms?
- Keyword Search (e.g., BM25): This method focuses on finding exact matches and their frequency within documents.
- Pros: Great for precise term matching, less prone "hallucinating" if specific keywords are crucial. *@learn.machinelearning*
- Cons: Can miss synonyms or context if the exact words aren't present.



Hybrid Search: The Best of Both Worlds

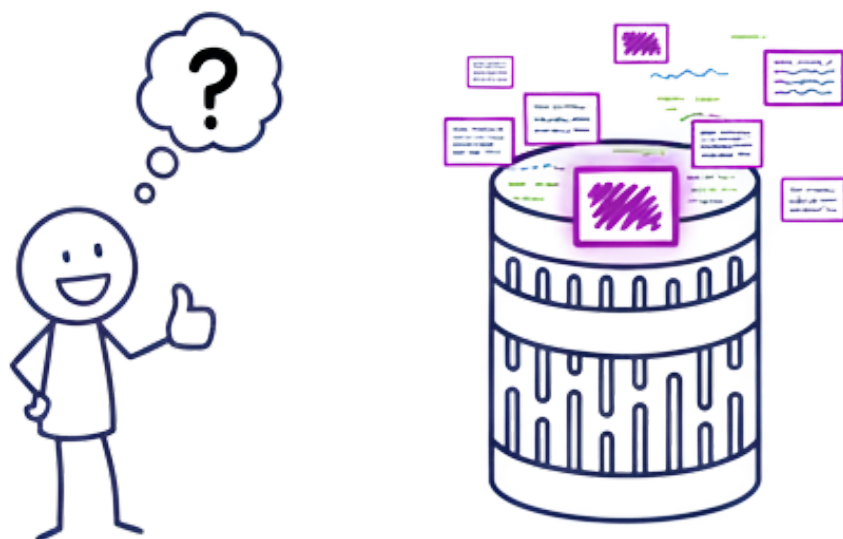
- Why choose when you can have both?
- Hybrid Search: Combines the strengths of Vector Search (semantic understanding) and Keyword Search (exact matching).
- Benefit: Catches both conceptually similar information and exact term matches, leading to higher quality answers.



Day 20 Takeaway:

Retrieval is a Smart Search Party!

- Retrieval is the critical “live” act of finding relevant information. By transforming questions into vectors, using
 - Vector Search for semantic understanding, and sometimes combining it with *Keyword Search* in a *Hybrid* approach, your RAG system becomes an expert librarian, always finding the right book.
-



Vector Database

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Next Up (Day 21):



Day 21: The Generation
Step: Synthesizing
Answers with Context

What kind of documents would you want your own AI to learn from? ↓

Follow @[learn.machinelearning](#) for the full RAG Deep Dive!

SAVE this post if you want your AI to find the right answers every time!

