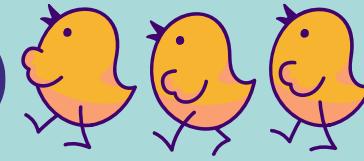
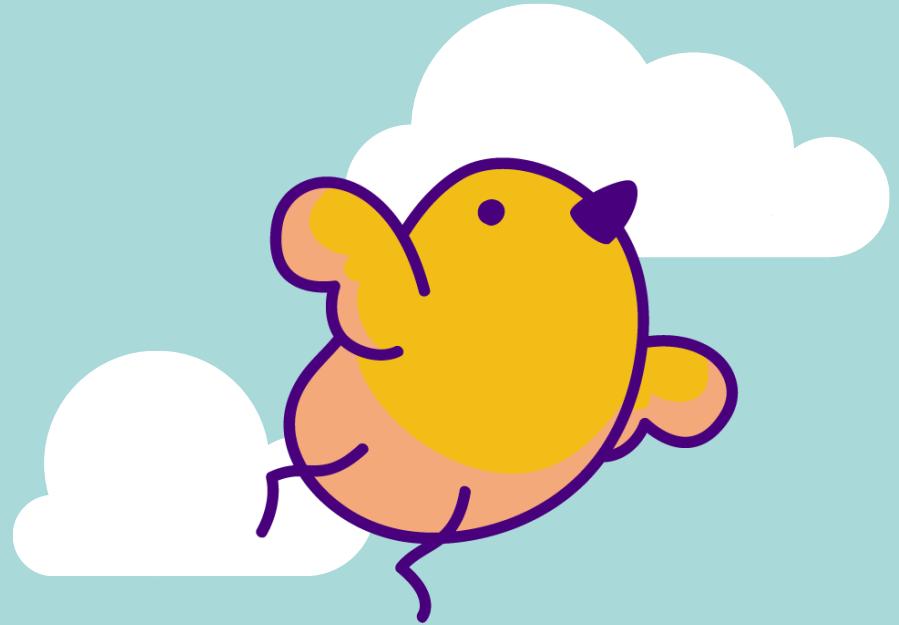


LangCon 2024



# 더 빠른 대규모 벡터 연산을 위해 FPGA를 이용해 가속을 해볼까?

김주현(메티스엑스) + 박진형(사이오닉AI)



FPGA? 벡터 검색 가속?  
(CXL은 또 무엇?)

LangCon 2024

## Who?



**CXL Computational Memory**를 개발  
하는 Fabless start-up

## 김주현

Co-founder, Chief Product Officer  
프로세서, 시스템SW  
하드웨어+소프트웨어 통합 설계  
SoC, SSD, Memory, Database, AI, ...

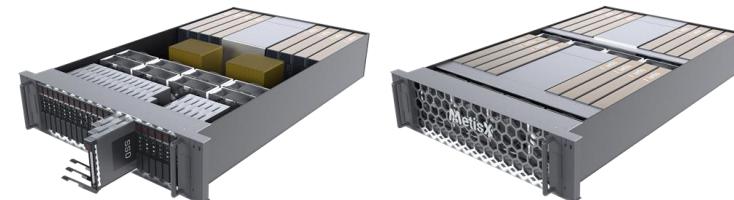
## CXL



GPU/SSD를 꽂던 PCIe 슬롯에 메모리를!  
**캐시 일관성 지원**  
Storage처럼 Memory도 Disaggregation

## Computational Memory

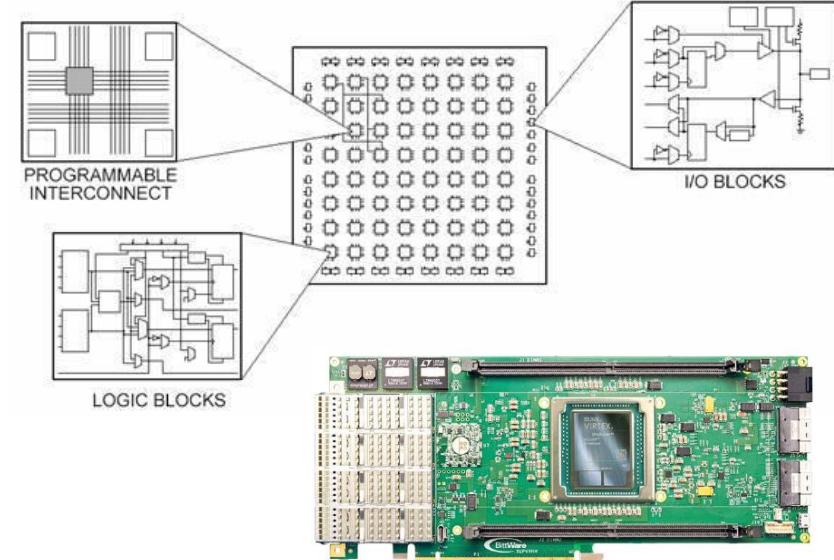
데이터 연산을 해주는 똑똑한 메모리  
**Near Data Processing**  
→ 벡터/DB 쿼리/ETL/...



## FPGA

## Field Programmable Gate Array

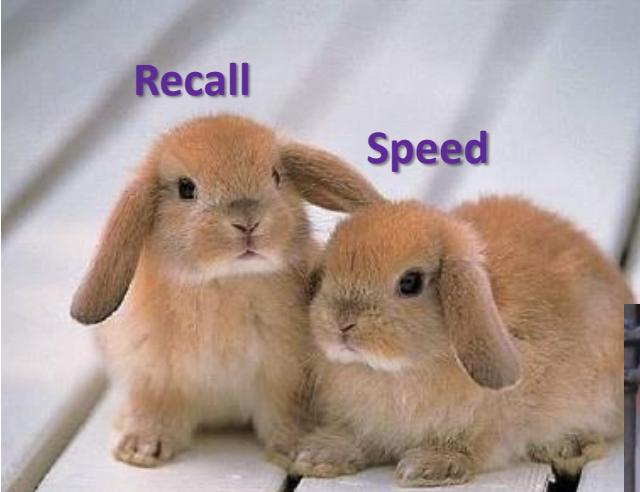
가속 하드웨어를 Programmable하게  
만들어 볼 수 있는 칩



# 왜 벡터 HW 가속이 필요할까?

Language Conference

- CPU 보다 빠르고 정확하게, GPU 보다 크고 싸게!

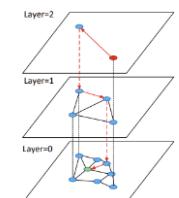
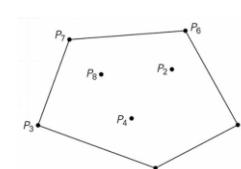


Recall  
Speed  
Bigger Data

Rank	Model	Model Size (GB)	Embedding Dimensions
1	SFR-Embedding-Mistral	14.22	4096
2	voyage-lite-02-instruct		1024
3	GritLM-7B	14.48	4096
4	e5-mistral-7b-instruct	14.22	4096
5	GritLM-8x7B	93.41	4096
6	UAE-Large-V1	1.34	1024
7	text-embedding-3-large		3072
8	voyage-lite-01-instruct		1024
9	Cohere-embed-english-v3.0		1024
10	multilingual-e5-large-instruct	1.12	1024

<https://huggingface.co/spaces/mteb/leaderboard>

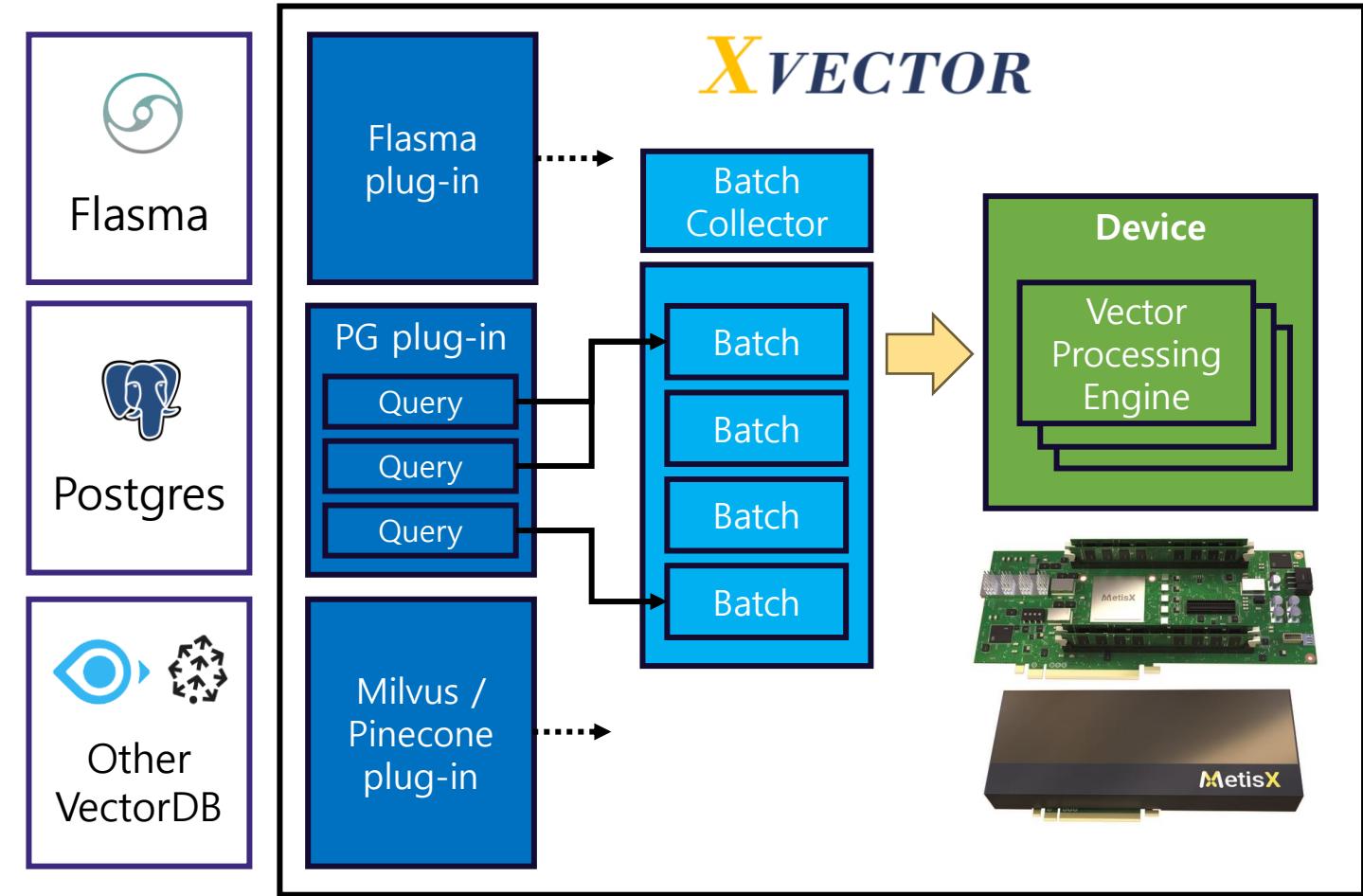
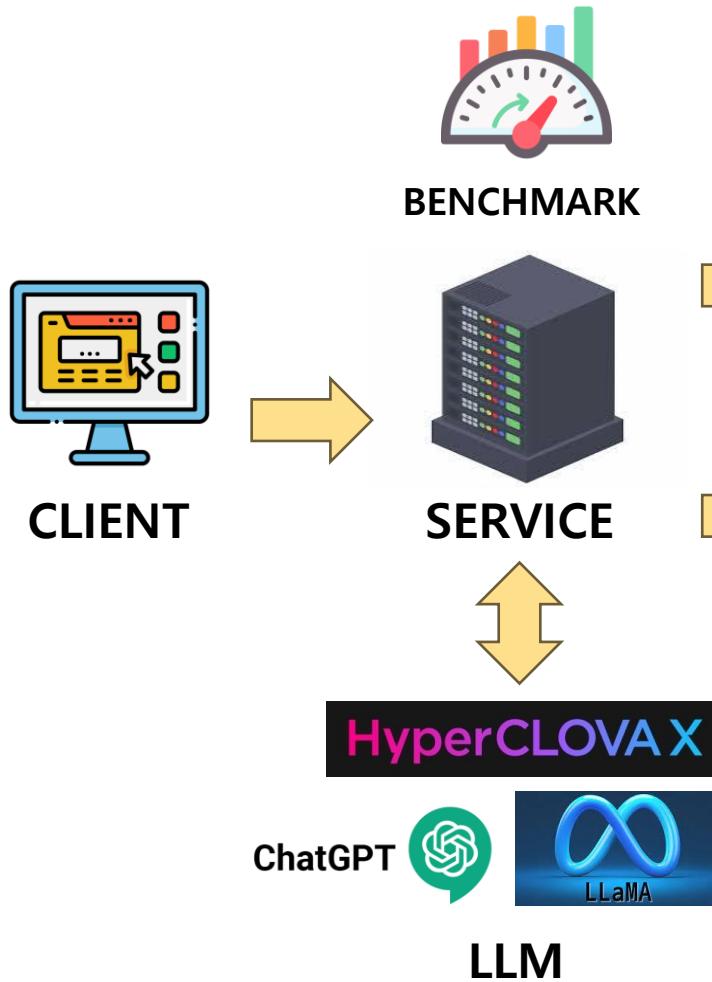
Higher dimension  
Larger dataset



	Recall	Speed	Size
Flat / Brute-force			
ANN (HNSW, ...)			
GPU Accelerated			
CXL Near Memory Vector Engine			

# SionicAI + MetisX 콜라보

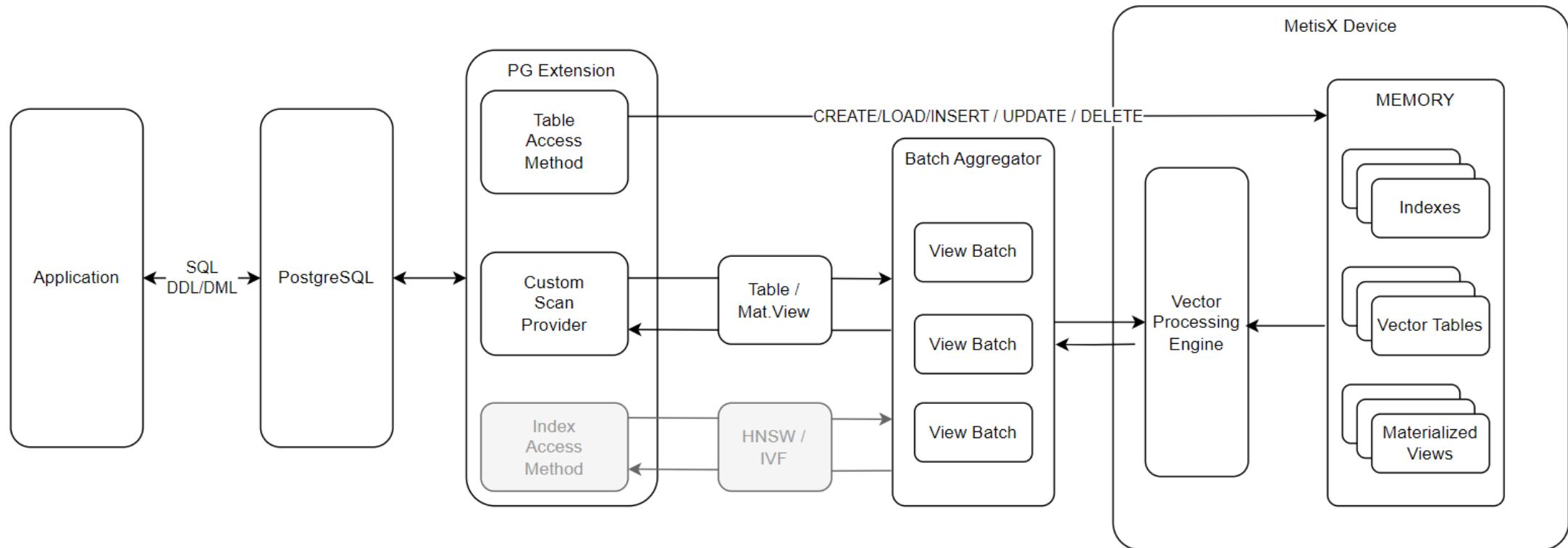
Language  
Conference



# pg-Xvector 구조

- PG Extension을 이용하여 Vector Table / Index Scan 가속
- Table, View 기반 Full Search / IVF Index 지원

기존 pg\_vector 에 pg\_Xvector plug-in 만 설치하면 적용 끝! (no schema/sql change!!)



<https://www.postgresql.org/docs/current/custom-scan.html>  
<https://www.postgresql.org/docs/current/tableam.html>  
<https://www.postgresql.org/docs/current/indexam.html>

# 벡터 쿼리 성능

Language  
Conference

- 400 QPS(FPGA) >> 260 QPS(3090GPU) (쿼리 당 벡터 100,000개 비교해서 Top-K를 뽑았을 때)
- 카드당 벡터 데이터베이스 크기 ~512GB x 슬롯 추가 확장
- ASIC(2025년 출시 예정)에서는 성능 10배, 용량 4배 이상 - Higher Recall & Speed, Bigger Data



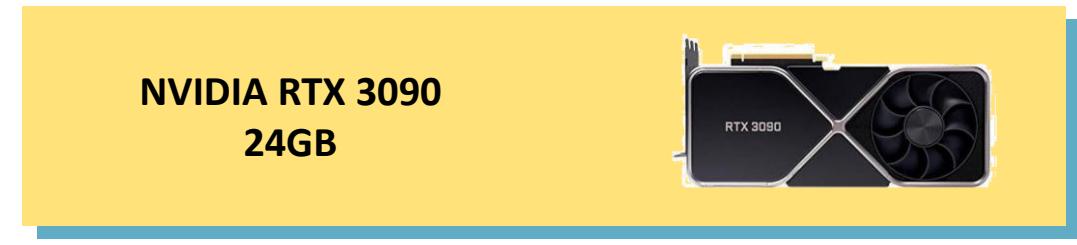
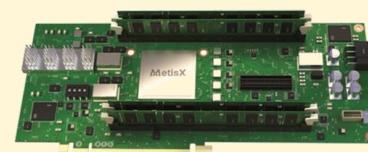
Vector search 성능 : 40M Vector/s @ 2K dim.(fp32)

Target vector 100,000개 환산 400 Query/s (12 batch)

Vector database size : max 512GB



4000 QPS, max 2TB



number of target vectors = 100,000 <https://github.com/sionic-ai/flasma>

table value = RPS(request per second)

table value 0 -> Cuda Out Of Memory

batch/dim	100	300	512	768	1024	2048	30522
1	393	389	383	365	355	192	31
2	822	705	555	432	394	241	42
4	1420	968	684	505	459	260	0
8	2187	1242	836	605	506	0	0
16	3006	1442	936	655	0	0	0
32	3750	1579	1001	0	0	0	0
64	4184	1645	0	0	0	0	0
128	4424	0	0	0	0	0	0

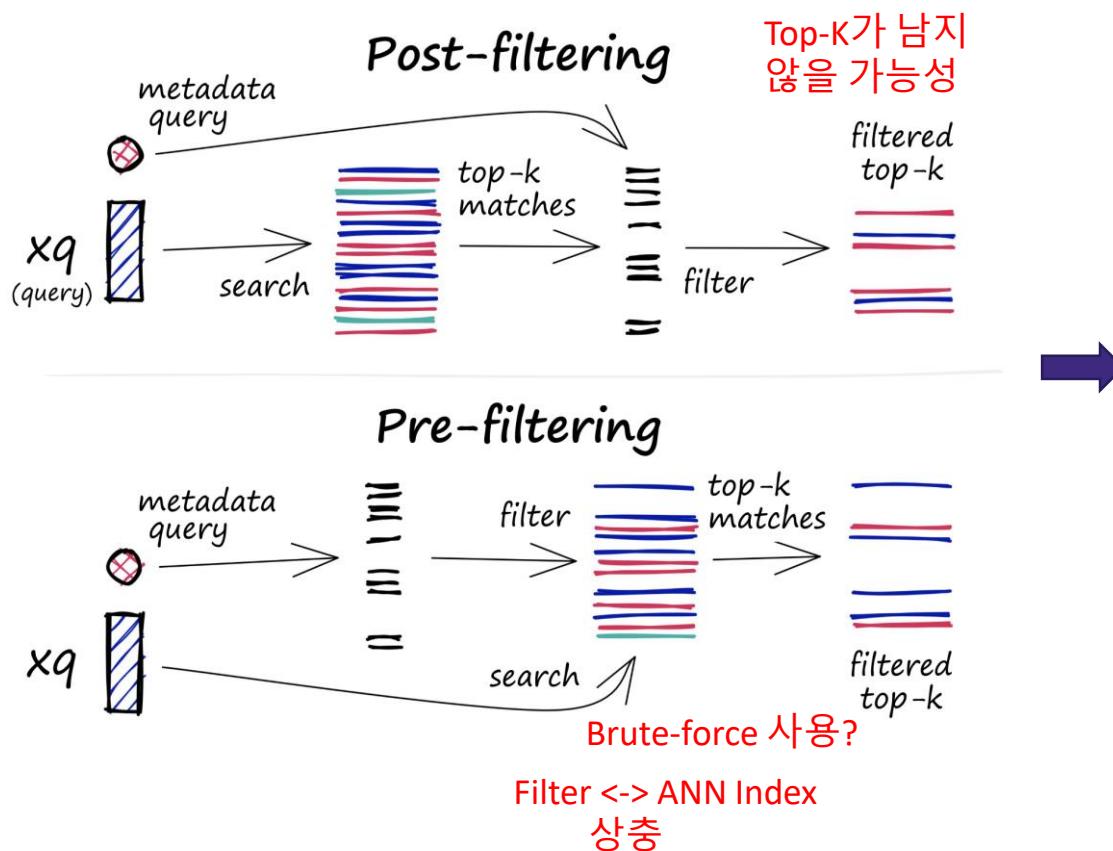
“메모리 용량 문제”

# RAG를 더 똑똑하게

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그러나, 벡터 거리만으로 찾기에는 현실 세계가 너무 복잡/방대합니다.

- WHERE clause in vector search,
- top-k where {"volume" > 14 and "department" == "finance"}



## "In-Query Filtering w/ metadata"

```
index.query(queries=[xq.tolist()], top_k=3,  
            filter={'lang': {'$eq': 'en'}})  
  
{  
    'context': 'Il Collegio di Ingegneria è stato istituito nel 1920,  
    'metadata': {'lang': 'it', 'title': ''}  
}  
{  
    'context': 'La KU School of Engineering è una scuola di ingegneri  
    'metadata': {'lang': 'it', 'title': ''}  
}  
{  
    'context': 'The KU School of Engineering is an ABET accredited, p  
    'metadata': {'lang': 'en', 'title': 'University_of_Kansas'}
```

<https://www.pinecone.io/learn/vector-search-filtering/>