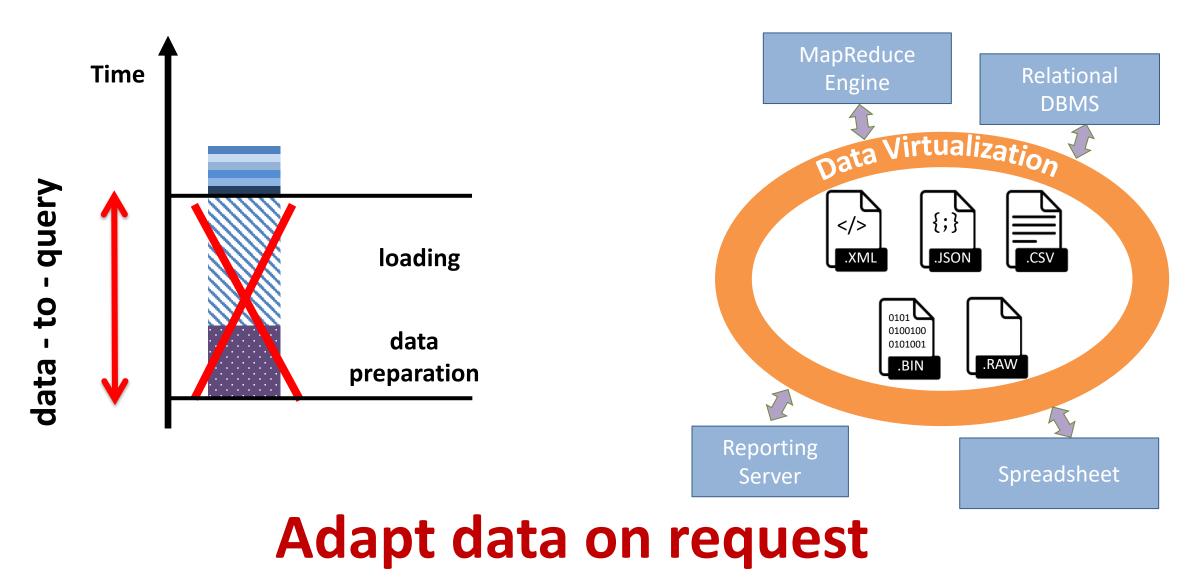
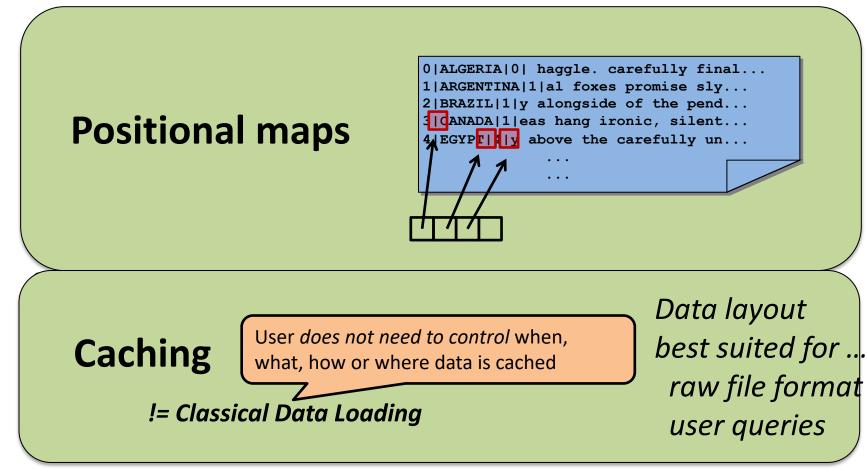


NoDB: minimize data-to-query time

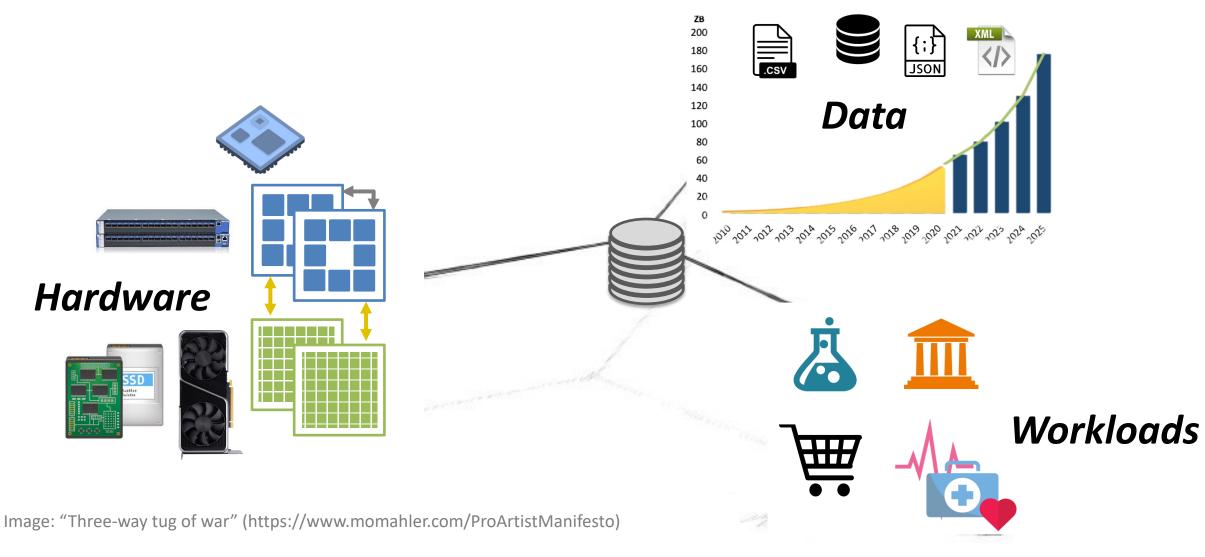
SIGMOD 2022 Test-of-Time award



NoDB idea: Decouple functionality from performance



An incessantly evolving landscape



Data management faces its most critical challenges

biological disease signatures

coupling

clinical measurements with *validated biomarkers*

Example: Alzheimer's disease

Example: Alz	zheimer's dis	ease (myloid			
Clinical - Phenotype	Proteomic Biomarkers	ease Genomic Biomarkers atrophy/amyloid			
Challenge:					
Real-time	integratio	on of heterogeneous data			

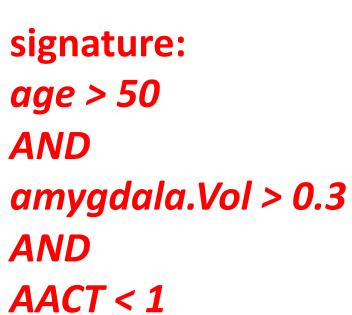
clinical+genetic+imaging data \rightarrow signature

Patients (CSV)

id **Protein:** Phenotype Age ••• AACT 1.4 45 1 Trauma ... Chronic 2 2 55 . . . Symptoms 3 0.2 56

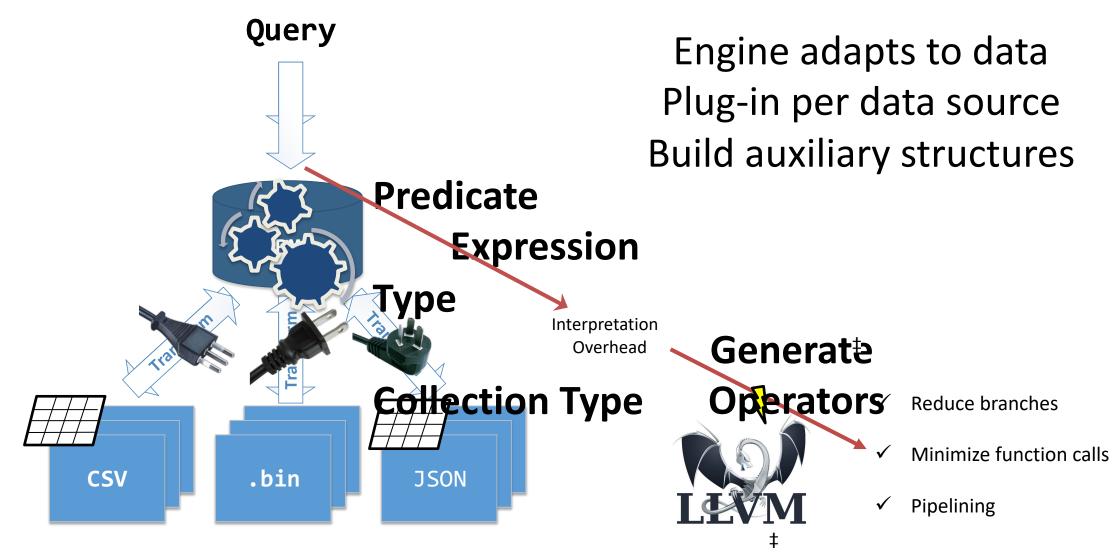
	0	1	•••	n
0	0.45	0.75	•••	0.1
1	0.33	0.3		0.38
•••			•••	
m	0.12	0	•••	0.47

Brain_GrayMatter (Binary)



BrainRegions (JSON)

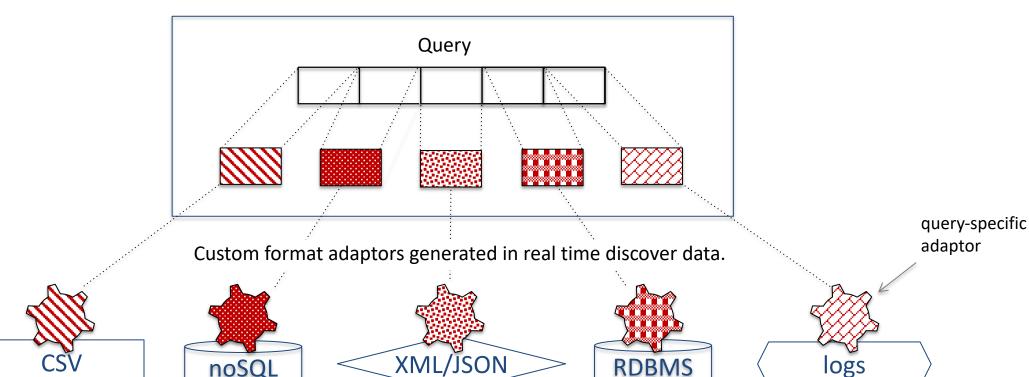
From LotsOfCode to NoCode



Codegen operators, continuously adapting engine

RAW: a *single* engine for all data





RAW Query is automatically split up for each data source.

Data is integrated transparently and on-demand.

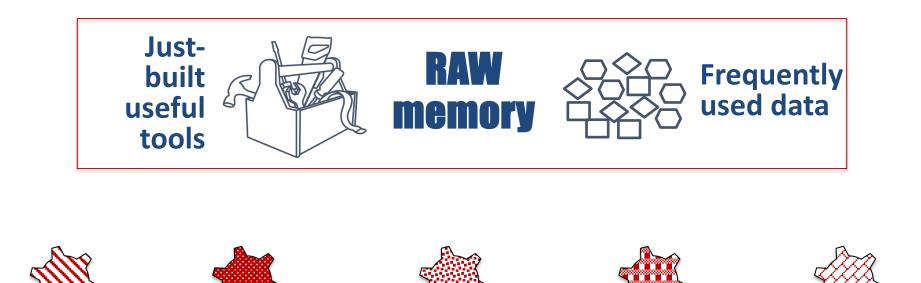
Users think of all of their data as a unified database, without preparation

Zero-cost (virtual) database



logs

RDBMS



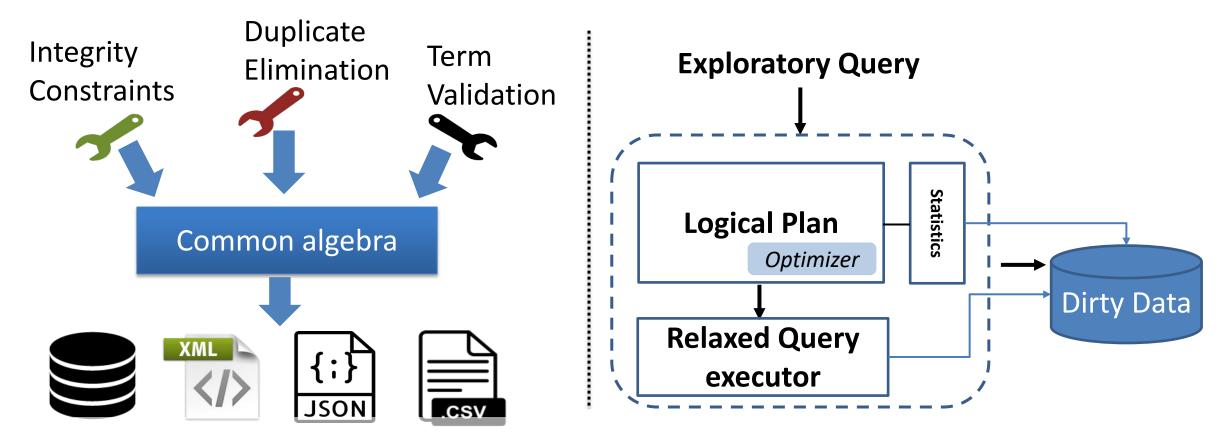
RAW is now a platform for trusted, live & secure data delivery

XML/JSON

noSQL

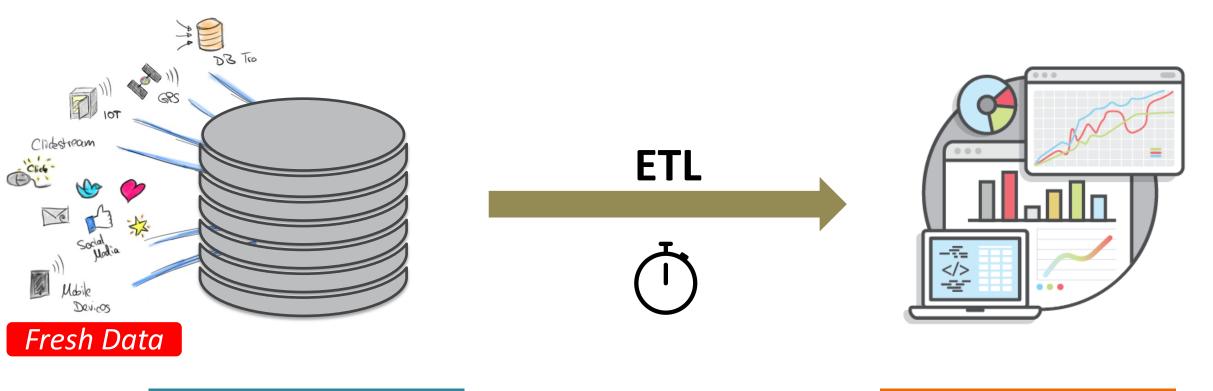
Efficient data veracity

Correct ALL errors on ALL data: costly and unnecessary



Clean useful data *adaptively* during analysis

Operations vs Analytics

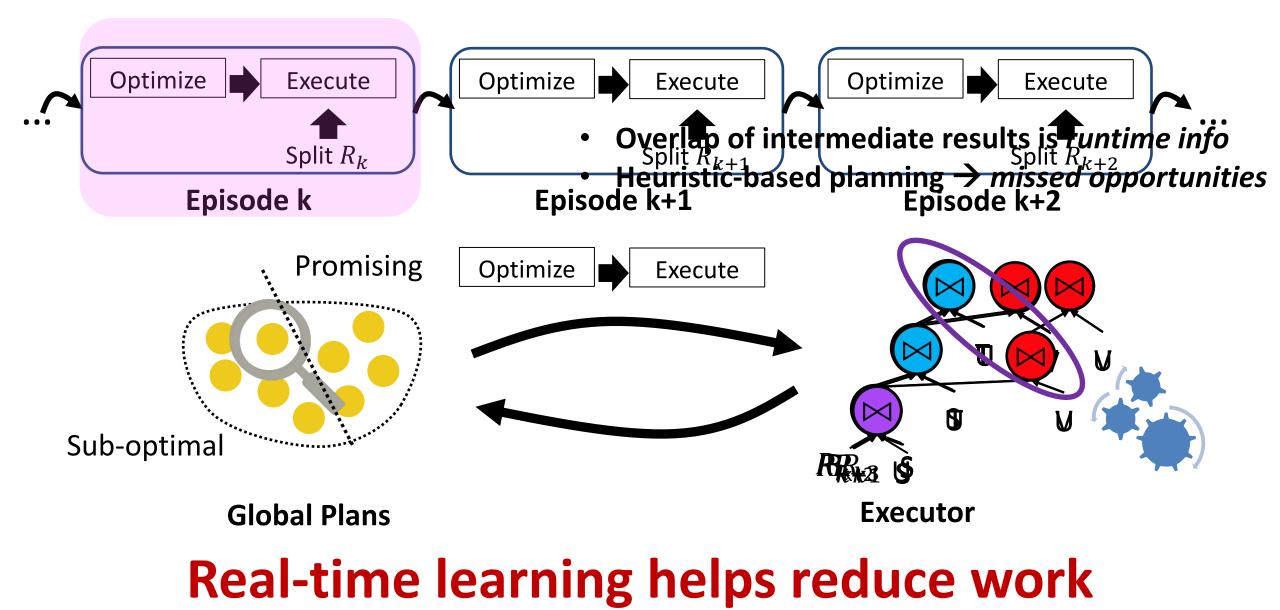


TRANSACTIONS

ANALYTICS

Ever-increasing number of concurrent queries Data freshness bounded by ETL latency

~10K concurrent tasks



Hybrid Transactional and Analytical Processing

- Transactions: task-parallel
 - High rate of short-lived processes
 - Mostly "point accesses" (high data access locality)
- Analytics: data-parallel

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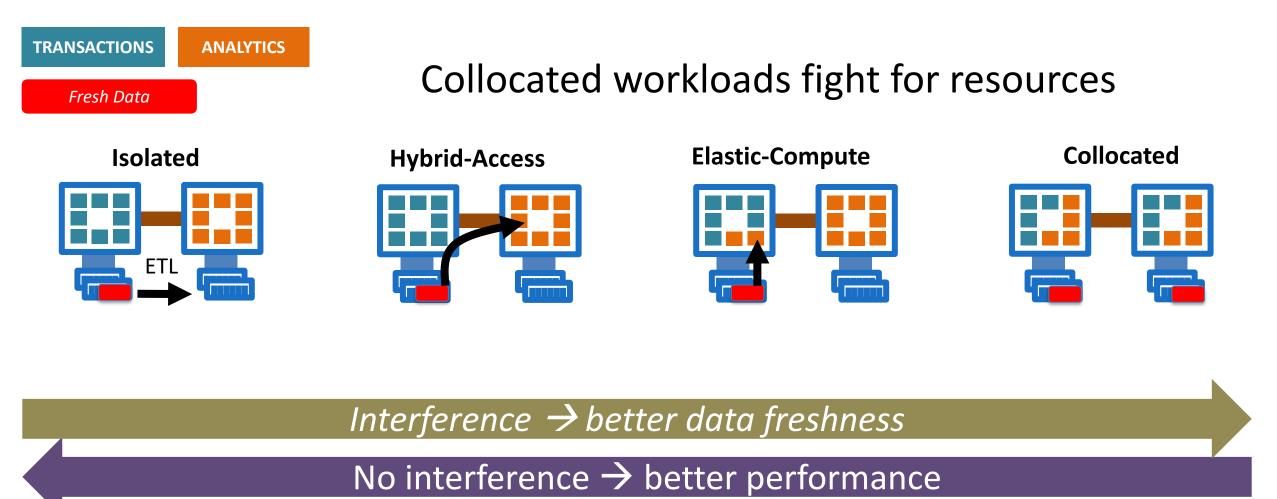


- Few, but long-running queries



Strong consistency is an invariant

Workload Isolation or Fresh Data?



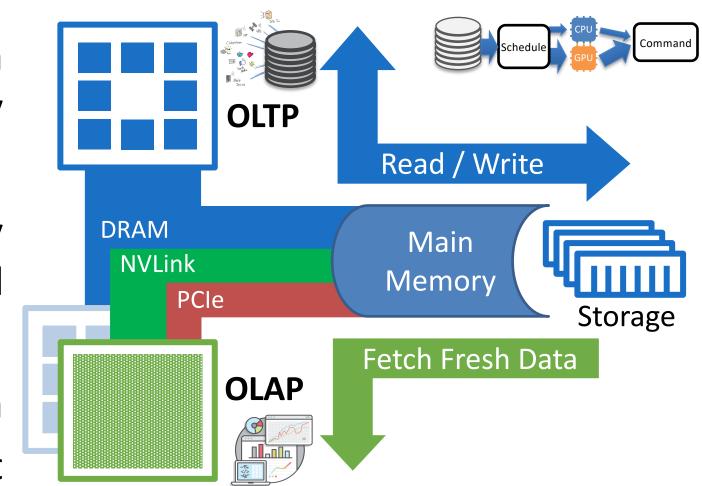
Viable hybrid alternatives

GPU acceleration for HTAP workloads

Transactions store fresh data on CPU Memory

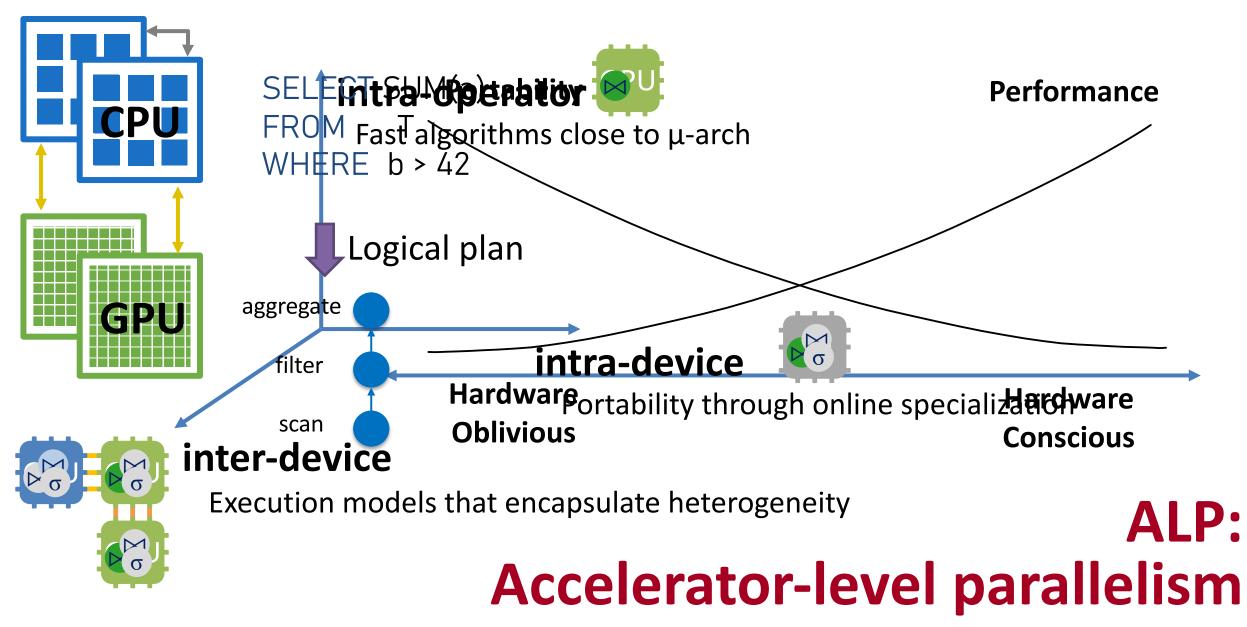
Data access protected by concurrency control

Analytics access fresh data through interconnect

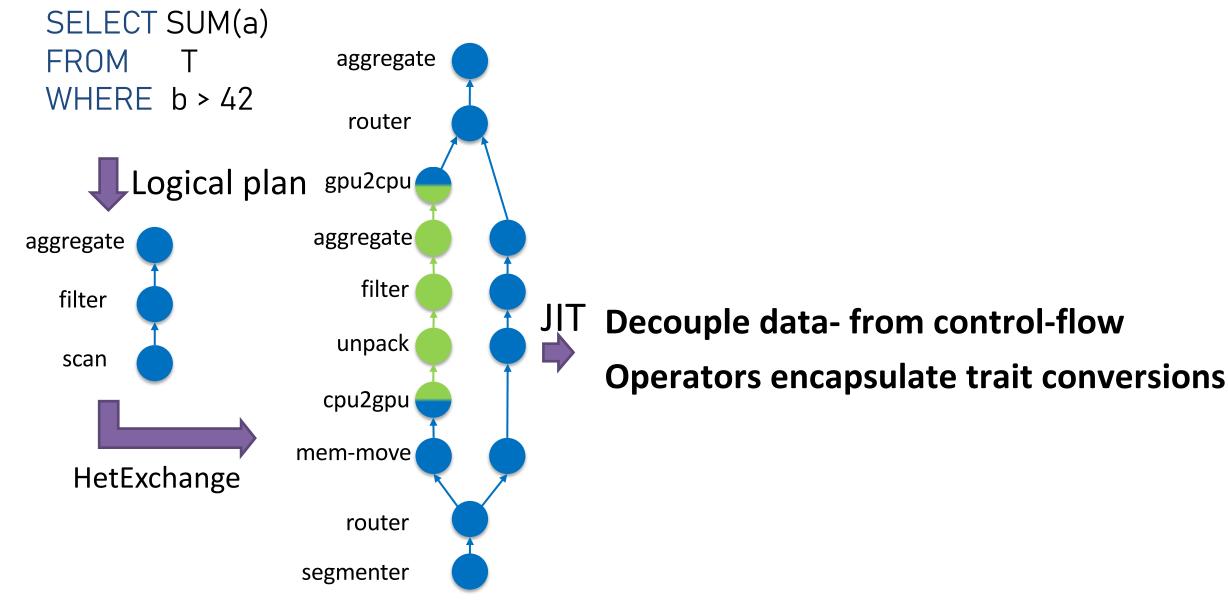


Interconnect speed is critical

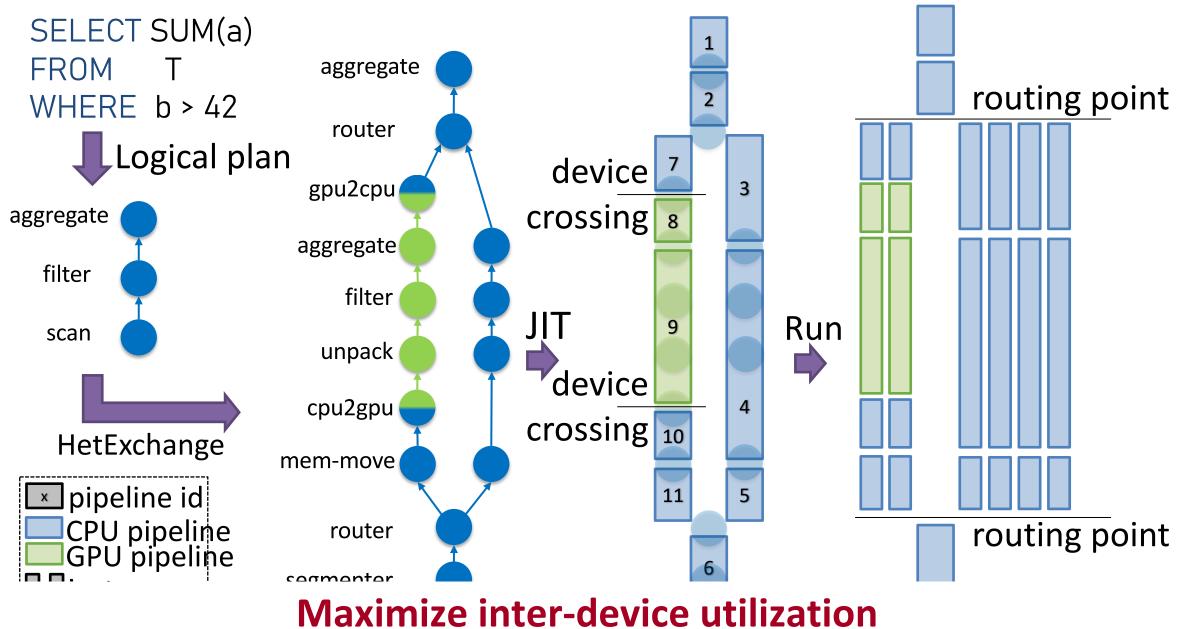
Device-conscious processing without regrets



HetExchange: Heterogeneity-aware plans

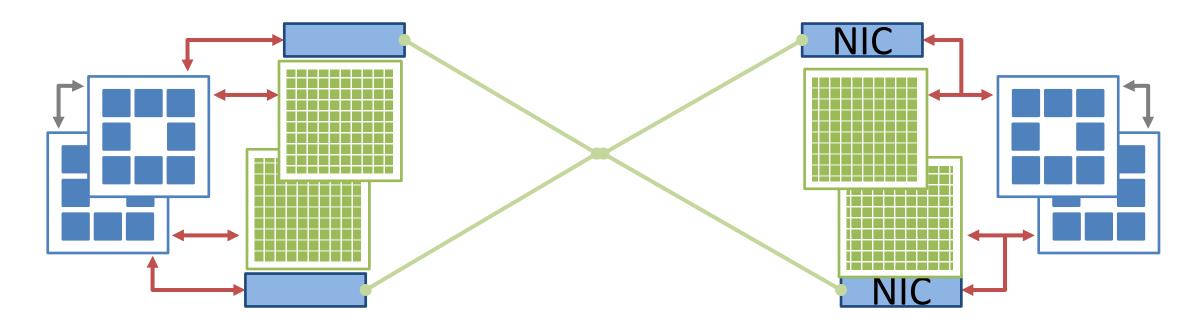


HetExchange in a JITed engine



Rack-scale analytics can use scaleup solutions

- Similar intra-/inter-server interconnect bandwidth
- Local memories and NUMA effects across devices
- CPU-GPU: Capacity-Throughput



Efficient use of heterogeneous interconnected devices

Ever-complex data pipelines

Diverse modern data problems

- IOT, OCR, ML, NLP, Medical, Mathematics etc...

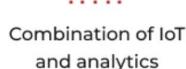
Database systems catch-up for popular functionality

- Human effort and big delays
- Oblivious to outside workflows

Vast resource of libraries

- Authored by domain experts, used by everybody
- Loose library-to-data-sources integration and optimization Conversational

Declarative programming for learning systems



Commercial AI/ML

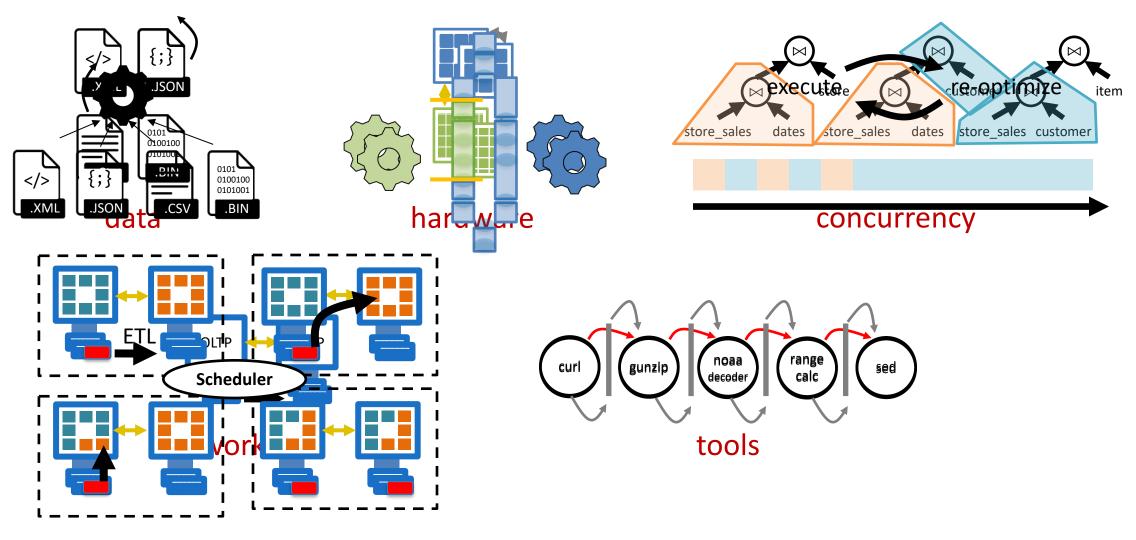


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analytics and NLP

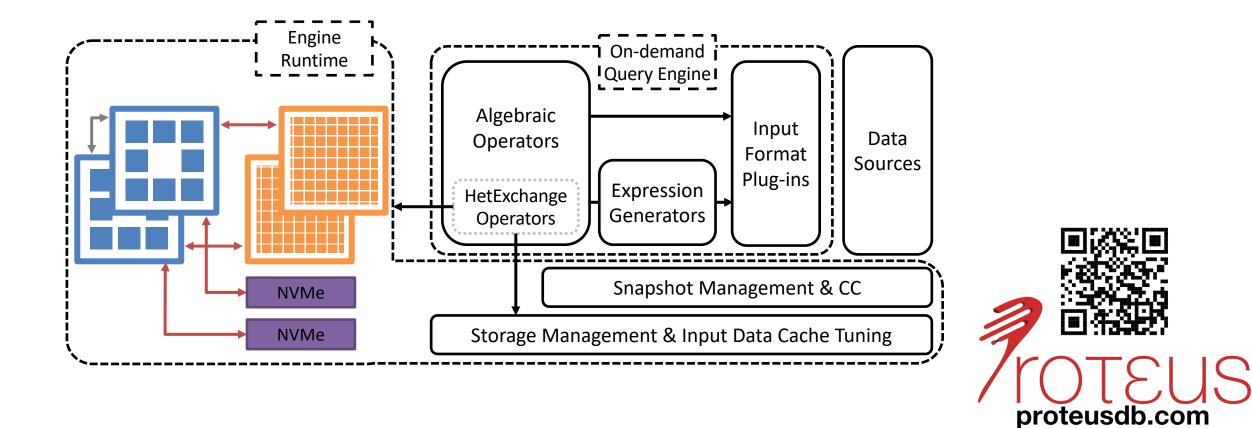


Heterogeneity in every aspect of data pipelines



Not ahead-of-time, JUST-IN-TIME!

Proteus: taming heterogeneity through adaptivity



JIT + no code = fast analytics on fresh data

Intelligent Real-time Systems

Street-smart engines incorporate change as a design principle, and react to surprises while learning.

