VectraFlow: An Al-Augmented Data-Flow System



Shu Chen, Alexander Lee, Duo Lu, Deepti Raghavan, Malte Schwarzkopf, Uğur Çetintemel

1. VectraFlow

A data-flow engine

that natively supports modern ML models with

an extended relational model for unstructured and

multi-modal data processing

Supports stream and batch processing

3. Data and Query Model

Classical **data-flow** architecture with an **extended** relational model:

- Data types
 - Vector (sparse and dense)
 - Unstructured (e.g., free-form text, images)
- Manipulation operators
 - E.g. convert data to vectors, cluster vectors
- Semantic relational operators
 Based on vectors, LLM prompts, and general ML models

2. Lighthouse Domain: Medical Data Lakes

(collaboration with the RI Hospital) **Example apps:**

- Medical data summarization
- Early warning system
- Compliance monitoring
- Automatic report generation

Key requirements:

- Integrate ML models (including LLMs)
- Support stream and batch oriented processing
- Ensure high reliability and scalability

4. Example Semantic Operators

- **iV-Filter()**: applies **embedding similarity** to **select** incoming tuples (Lu et al., 2025)
 - E.g., identify incoming patient records that are similar to historical patient records
- **P-Agg()**: prompts an **LLM** to **aggregate** over a window of tuples (Patel et al., 2024)
 - E.g., summarize over multiple medical documents
- M-Filter(): invokes a classifier to select tuples based on
- Retain general semantics of relational operators

5. iV-Filter (Lu et al., 2025)

Each base vector has a **radius**

iV-Filter: selects input vectors that fall within the radii of base vectors

and returns the corresponding base vector IDs

Centroid OPList (Overlapped Partition List):

- Insight: base vectors containing the incoming vector must overlap
- OPList: list of base vectors that overlap with the given base vector
- Centroid OPList: cluster base vectors and assign a radius + OPList to each centroid

Search: assign input vector to the nearest centroid and scan its OPList

Other optimizations: batching, sorting, bucketing, early stopping

6. Semantic Integrity Constraints

Problem: semantic operators may yield erroneous results

Solution: guardrails around semantic operators to enforce data consistency

User-specified predicates on output tuples

their attributes (Lu et al., 2025)

• E.g., identify abnormalities in medical imaging



Generated business report

Can apply constrained decoding for certain predicates Otherwise,	Inclusion/exclusion		doesn't contain undesirable language	
if tuple violates predicates, retry operator if specified retry threshold is reached, drop tuple	Grounding		Extracted test records are present in the original medical document	
	Check <predicate></predicate>		Evaluate arbitrary predicates (e.g., simple statements, UDFs)	
8. Enforcing Grounding Constraints				
 Want: attribute value is grounded in its source tuple(s) Recursively apply checks to all attributes in the attribute's lineage Check: output value is grounded in input value(s) Require different grounding semantics depending on the 	Semantics	Verification Mechanism		Use Case
	Match	Exact keyword match		Extractive
	Similarity	Similarity score		Abstractive
use case	Model	LLM evaluator		Extractive + abstractive
9. Acknowledgements				

Special thanks to other members of the team who contributed to this work: Justin Chan, Simeng Feng, Michael Fu, Nicolas Kim, Evan Li, Akshay Mehta, Weili Shi, Franco Solleza, Jonathan Zhou