

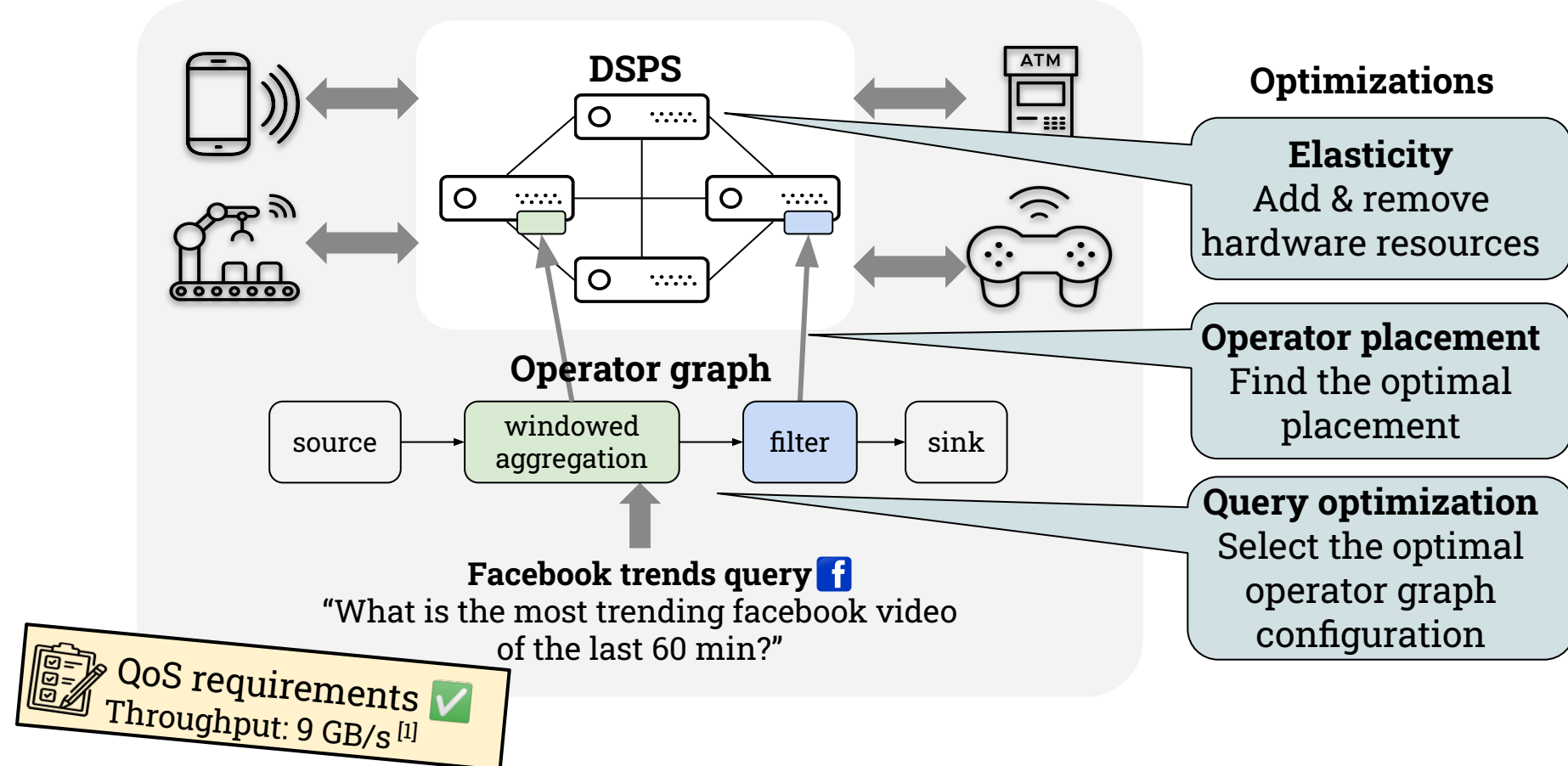
Zero-Shot Cost Models for Distributed Stream Processing

Roman Heinrich¹, Manisha Luthra², Harald Kornmayer¹, Carsten Binnig²

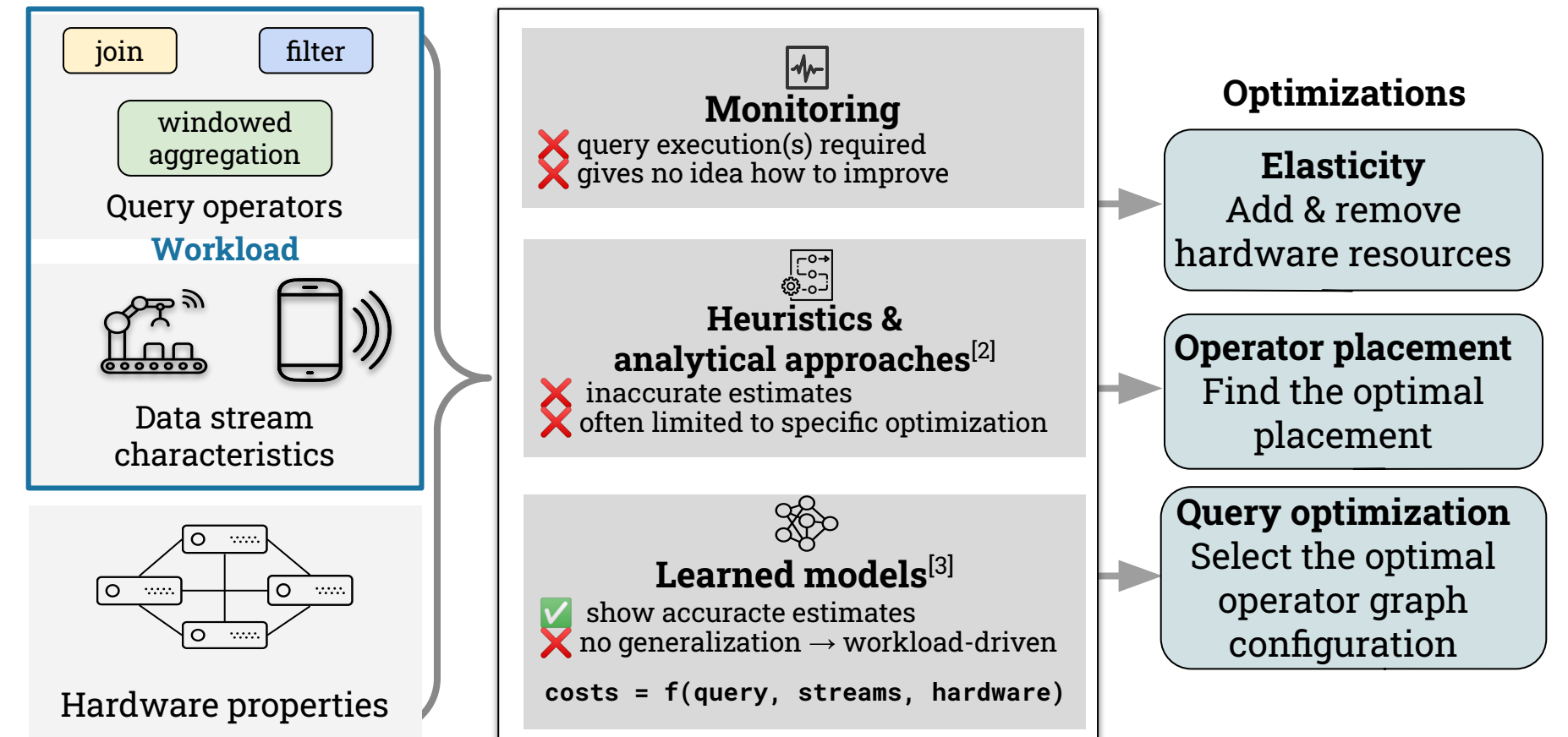
¹ DHBW Mannheim, ² TU Darmstadt

Distributed and Event-based Systems (DEBS 2022)

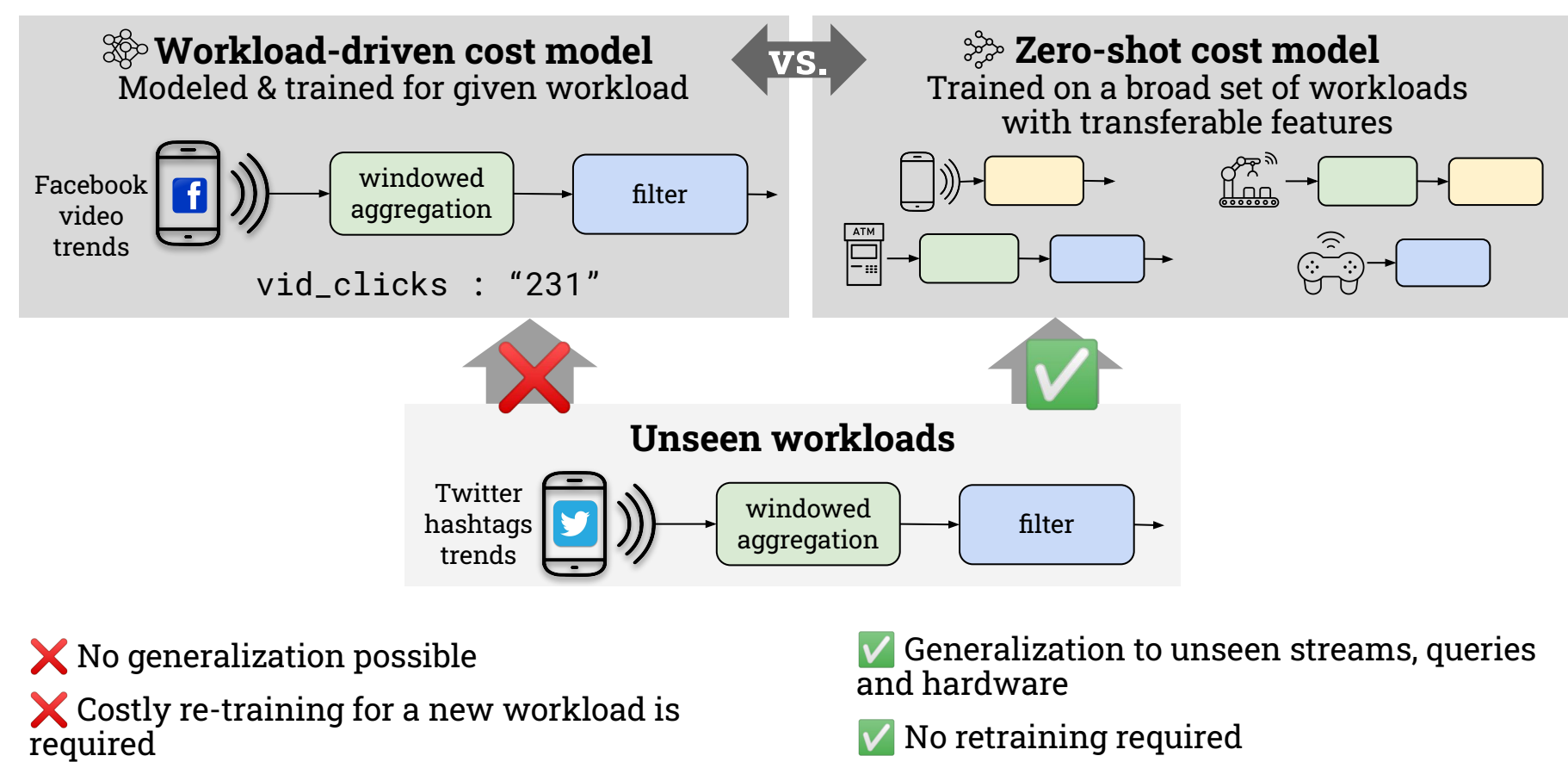
Cost models for DSPS



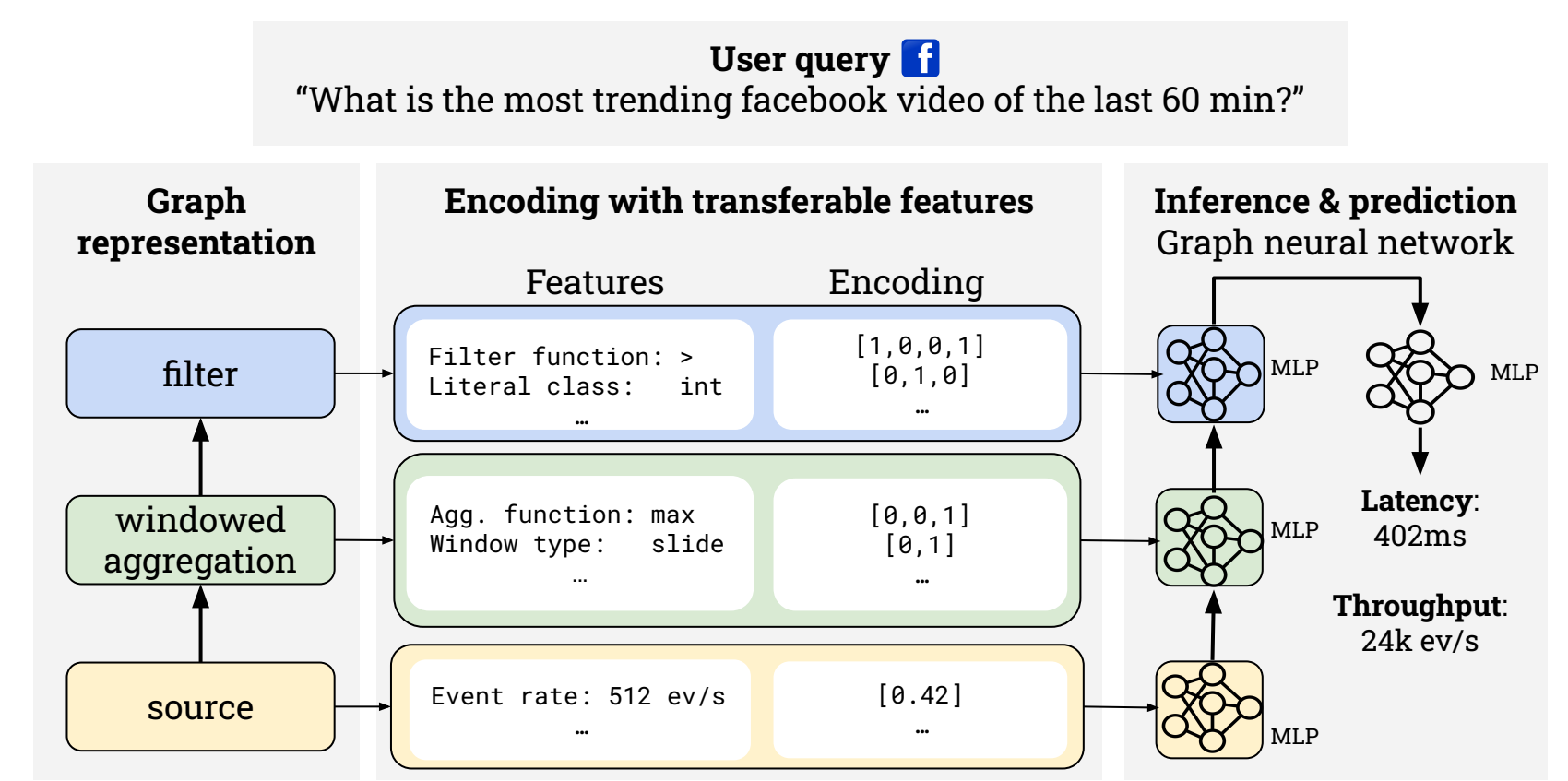
Limitations of existing approaches



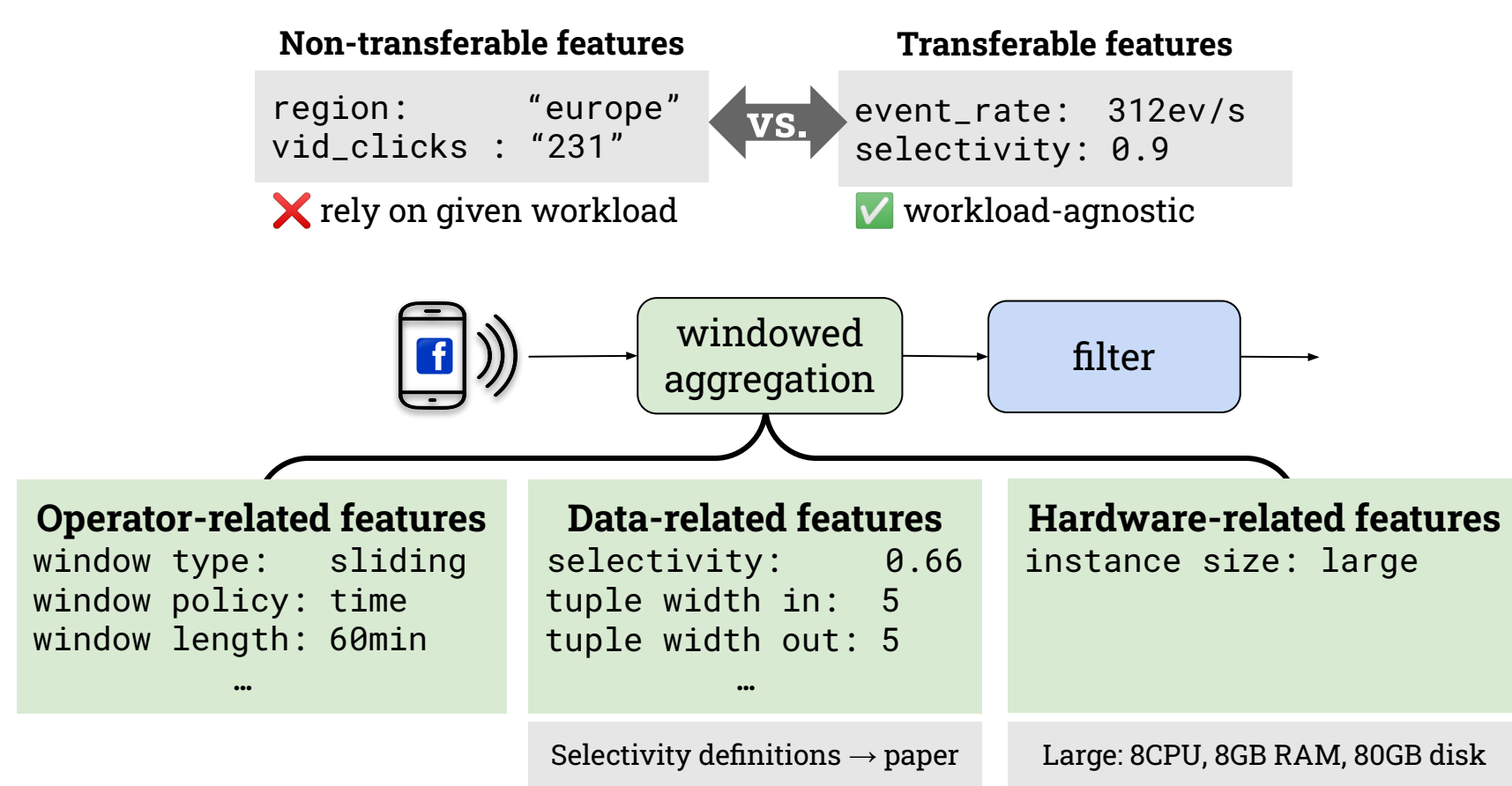
Workload-driven vs. zero-shot models



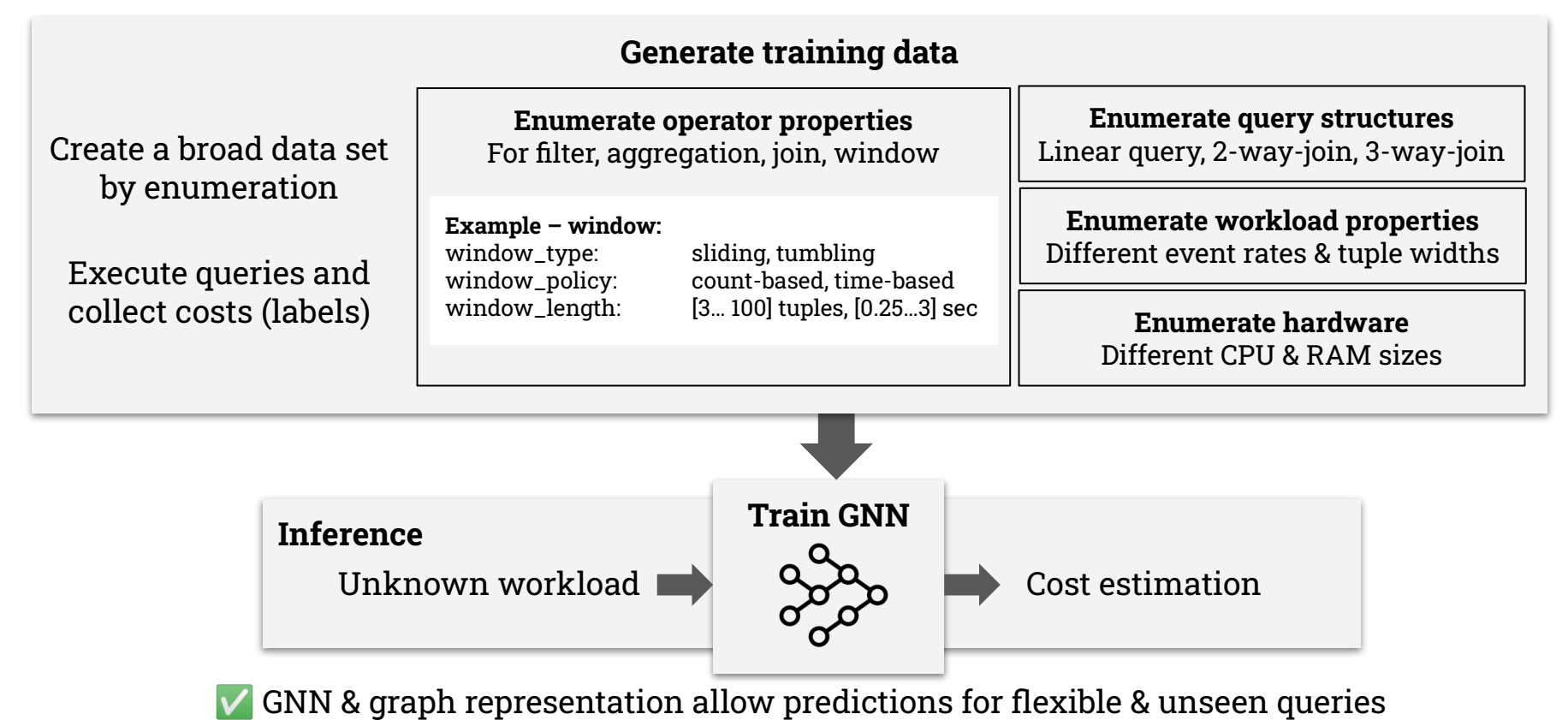
Zero-shot model architecture



Transferable features



Training & inference methodology



Evaluation on zero-shot models

- Set-up:** 10 clusters (each 10 nodes) with Apache Storm v2.2.0
- Cost metrics:** end-to-end latency & throughput
- Metric:** q-error
 $q(c, \hat{c}) = \max(c/\hat{c}, \hat{c}/c)$
 - reporting median and 95-percentile
 - q=1: perfect estimate
- Interpolation for workloads & placements**
 Latency:

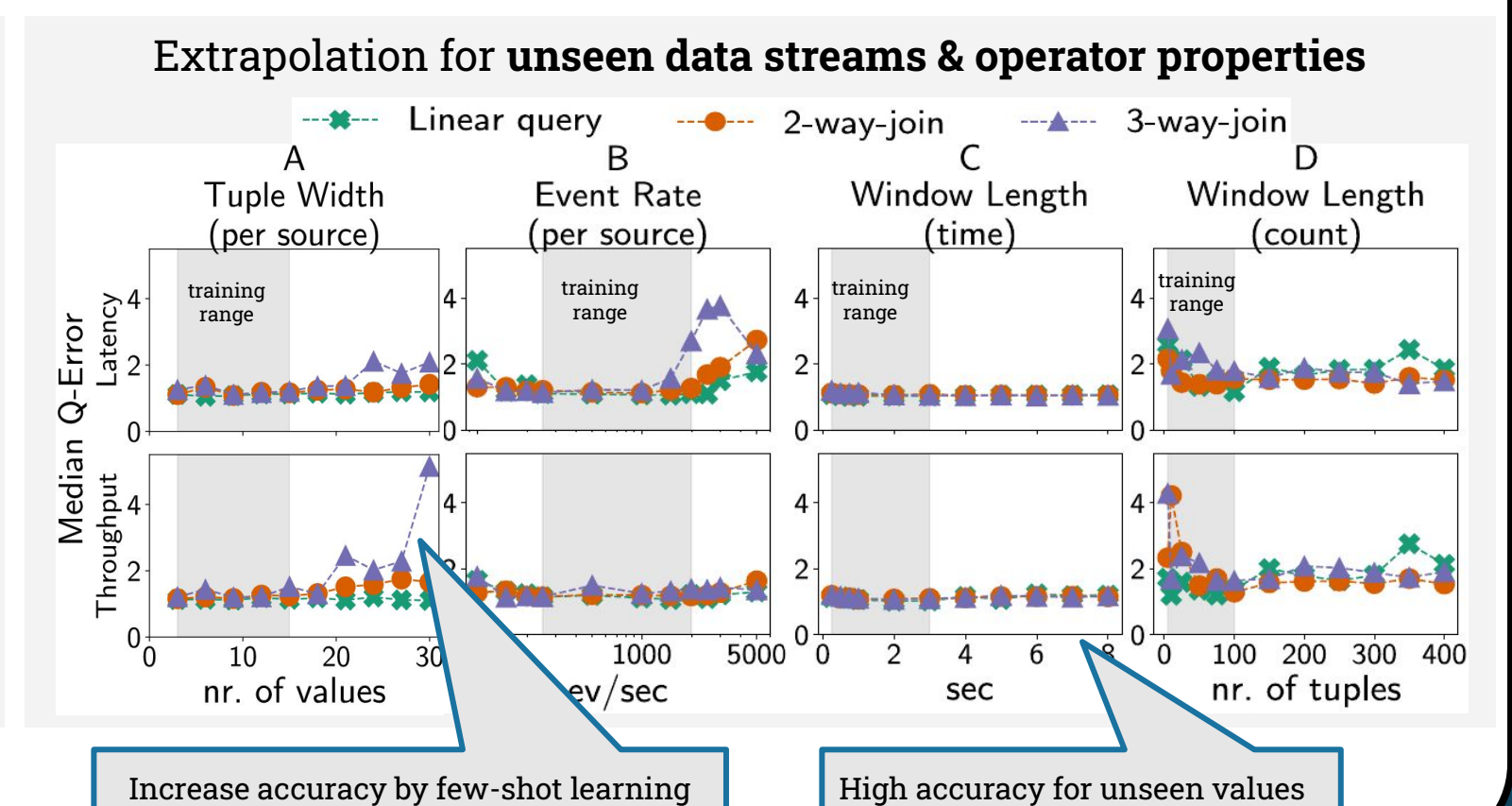
| | | | |
|-------------|------|------|------|
| median | 1.13 | 95th | 3.19 |
| Throughput: | 1.16 | | 3.50 |

Extrapolation for **unseen benchmarks** (DSPBench^[4])

| Benchmark | Latency | | Throughput | |
|------------------------|---------|------|------------|------|
| | median | 95th | median | 95th |
| Advertisement (clicks) | 1.51 | 1.53 | 1.38 | 1.39 |
| Advertisement (imp.) | 1.51 | 1.52 | 1.38 | 1.39 |
| Advertisement (join) | 1.99 | 2.06 | 1.55 | 2.16 |
| Spike Detection | 1.01 | 1.04 | 1.73 | 1.94 |
| Smart Grid (local) | 1.21 | 1.23 | 1.92 | 1.92 |
| Smart Grid (global) | 1.20 | 1.66 | 1.91 | 1.91 |

Extrapolation for **unseen query structures**

| | Latency | | Throughput | |
|----------------|---------|-------|------------|-------|
| | median | 95th | median | 95th |
| 2-filter chain | 1.14 | 2.41 | 1.59 | 3.65 |
| 3-filter chain | 2.67 | 46.34 | 2.82 | 27.78 |
| 4-filter chain | 7.33 | 54.68 | 3.94 | 59.73 |
| 4-way joins | 1.95 | 24.30 | 1.33 | 20.79 |
| 5-way joins | 1.91 | 26.67 | 1.35 | 21.87 |



Conclusion & outlook

Our zero-shot cost model...

- ...is generalizable and workload independent
- ...requires an one-time training effort
- ...predicts accurately and robustly for seen & unseen workloads
- ✓ can be used as a main building block in DSPS optimization tasks

Open questions on zero-shot models:

- How to model hardware properties more precisely?
- How to featurize co-location of operators?
- How to make use of the cost model in specific optimization tasks like providing elasticity?

[1] Z. Shao, "Real-time analytics at facebook," XDB, 2011.

[2] L. Eskandari, J. Mair, Z. Huang, and D. Eysers, "I-scheduler: Iterative scheduling for distributed stream processing systems," Future Generation Computing Systems, vol. 117, pp. 219–233, 2021.

[3] T. Li, Z. Xu, J. Tang, and Y. Wang, "Model-free control for distributed stream data processing using deep reinforcement learning," PVLDB, vol. 11, no. 6, p. 705–718, 2018.

[4] M. V. Bordin, D. Griebler, G. Mencagli, C. F. R. Geyer, and L. G. L. Fernandes, "DSPBench: A suite of benchmark applications for distributed data stream processing systems," IEEE Access, vol. 8, pp. 222 900–222 917, 2020.