

Advanced Database Systems for Big Data - Challenges

Introduction to Stream Processing

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Report Series on Database Research

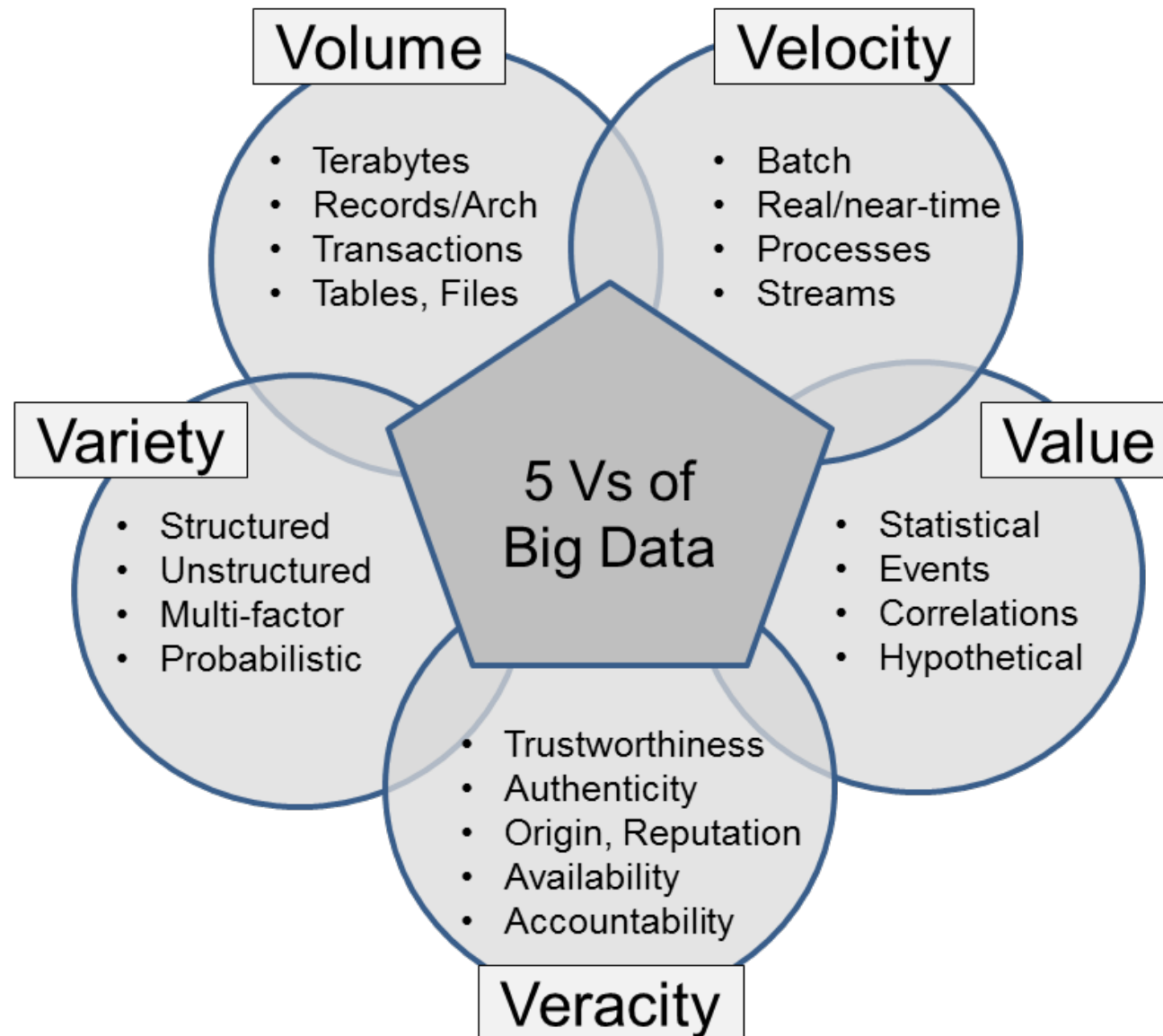
- Leading (30+) DB researchers & professors
- 10th meeting: 1989, 1990, 1995, 1996, 1998, 2003, 2008, 2013, 2018, 2023, ...
- 3 newest reports focus on AI/ML & Big Data:
 - The Cambridge Report on Database Research, A Ailamaki, S Madden, D Abadi, G Alonso, S Amer-Yahia, M Balazinska, PA Bernstein..., April 2025, arXiv preprint arXiv:2504.11259
 - Abadi, D., et al.: The Seattle Report on Database Research, Communications of the ACM, August 2022, Vol. 65, No. 8, pp. 72-79
 - Abadi, D., et al.: The Beckman Report on Database Research, Communications of the ACM, February 2016, Vol. 59, No. 2, pp. 92-99

Big Data – Defining Challenge

[Beckman Report 2016]

- Distribution
- Integration
- Heterogeneity
- In-memory processing
- Large-scale systems
- Data analysis
- ...

Big Data – Main Characteristics



Big Data – 5 Related Challenges

- Scalable big/fast data infrastructures
- Coping with diversity in data management
- End-to-end data-to-knowledge pipeline
- Cloud services
- Role of people in data life cycle

Scalable Big/Fast Data Infrastructures

- Parallel and distributed processing
- Query processing and optimization
- New hardware
- Cost-efficient storage
- High-speed data streams
- Late-bound schemas
- Consistency
- Metrics and benchmarks

Diversity in Data Management

- No one-size-fits-all
- Cross-platform integration
- Programming models
- Data processing workflows

End-to-End Processing of Data

- Data-to-knowledge pipeline
- Tool diversity
- Tool customizability
- Open source
- Understanding data
- Knowledge bases

Cloud Services

- IaaS, PaaS, SaaS
- Elasticity (SLA)
- Data replication
- System administration and tuning
- Multitenancy (VMs)
- Data sharing
- Hybrid clouds

Roles of Humans in the Data Life Cycle

- Data producers
- Data curators (crowdsourcing)
- Data consumers
- Online communities

DB Research - New Challenges

[Beckman Report 2016] -> [Seattle Report 2022] - 1

- Correctly identified Big Data as a big theme
 - now morphed into data science, which poses big challenges
- Missed the AI/ML trend
- Promoted five directions
 - scalable data infrastructure
 - diversity in data management
 - end-to-end processing and understanding of data
 - cloud services
 - roles of humans in the data life cycle
- Made good progress, also branched out
 - e.g., into AI/ML

DB Research - New Challenges

[Beckman Report 2016] -> [Seattle Report 2022] - 2

- Beckman predicted the rise of a **data-driven world**
- Correctly observed that this gives us unprecedented opportunities & challenges:
 - Increasing amount and use of personal data -> data governance, ethical and fair use of data
 - Managed cloud data systems -> serverless systems, data lakes, ETL (extract, transform, load) jobs
 - Industrial Internet-of-Things (IoT)
 - Significant changes in hardware, esp. for ML/Deep Learning: FPGAs, GPUs, ASICs, ...
- All of these have been true, but there are deep concerns that DBS community have failed to exploit this wealth of opportunities
 - while other communities have moved much faster.

DBS and Data Science ?

- **Data Science**: combines data cleaning and transformation, statistical analysis, data visualization, and ML techniques.
- Data Science [NSF CISE 2017]: “the processes and systems that enable the extraction of knowledge or insights from data in various forms, either structured or unstructured.”
- **DBS technology** plays a major role in **Data Science**: pipeline from raw input data to insights that requires use of data cleaning and transformation, data analytic techniques, and data visualization.
 - Data to insights pipeline
 - Data context and provenance
 - Data exploration at scale and data profiling
 - Declarative programming
 - Metadata management

Data Governance

- Data use policy -> GDPR, auditing
- Data privacy, e.g. differential privacy
- Ethical data science ->
responsible data management

Cloud Services

- Serverless data services
- Disaggregation
- Multitenancy
- Edge and cloud
- Hybrid cloud and multi-cloud
- Auto-tuning
- SaaS cloud DB applications

Database Engines

- Heterogeneous computing
- Distributed transactions
- Data lakes
- Approximation in query answering
- Machine Learning (ML) workloads
- ML for reimaging data platform components
- Benchmarking and reproducibility

DB Research - New Challenges

[Seattle Report 2022] -> [Cambridge Report 2025]

- **Core Data Systems**
 - Big Data everywhere, Cloud-based data systems, emerging new hardware, scalability, usability
- **Human Centric Systems and Data Science**
 - Big Data everywhere, data governance, NL-based querying & analysis interfaces
- **ML and AI for Data Systems**
 - AI/ML everywhere -> generative AI, LLMs, ...
- **Responsible Data Management**

Core Data Systems

- Big Data everywhere
 - > usability, DB at massive scales
- Cloud-native architectures
- Disaggregated storage and compute
 - > high degree of scalability and flexibility
- Evolving hardware landscape
 - > resource-hungry AI, spec. AI accelerators

Human Centric Systems & Data Science

- Data sharing and collaboration
 - > break down data silos/lakes, enable cross-organizational analytics
- Privacy, governance, query processing across distributed datasets
- End-to-end data pipeline and workflow systems
 - > data discovery, explanations, preparation, integration and cleaning, metadata and log mgnt., versioning, analysis and visualization

ML and AI for Data Systems

- Query optimization, cost models based on ML
- Cardinality estimation, high-dimensional correlations in data distributions
- Reinforcement learning to improve physical data organization, predictive I/O
- Cloud resource management
- ML models for serverless VM management

Responsible Data Management

- Prevalence of AI models for interpreting data and making complex decisions
- Integrating data management research into responsible AI
- Decisions made during data collection and preparation impact accuracy, fairness, robustness, interpretability, legal compliance of AI systems

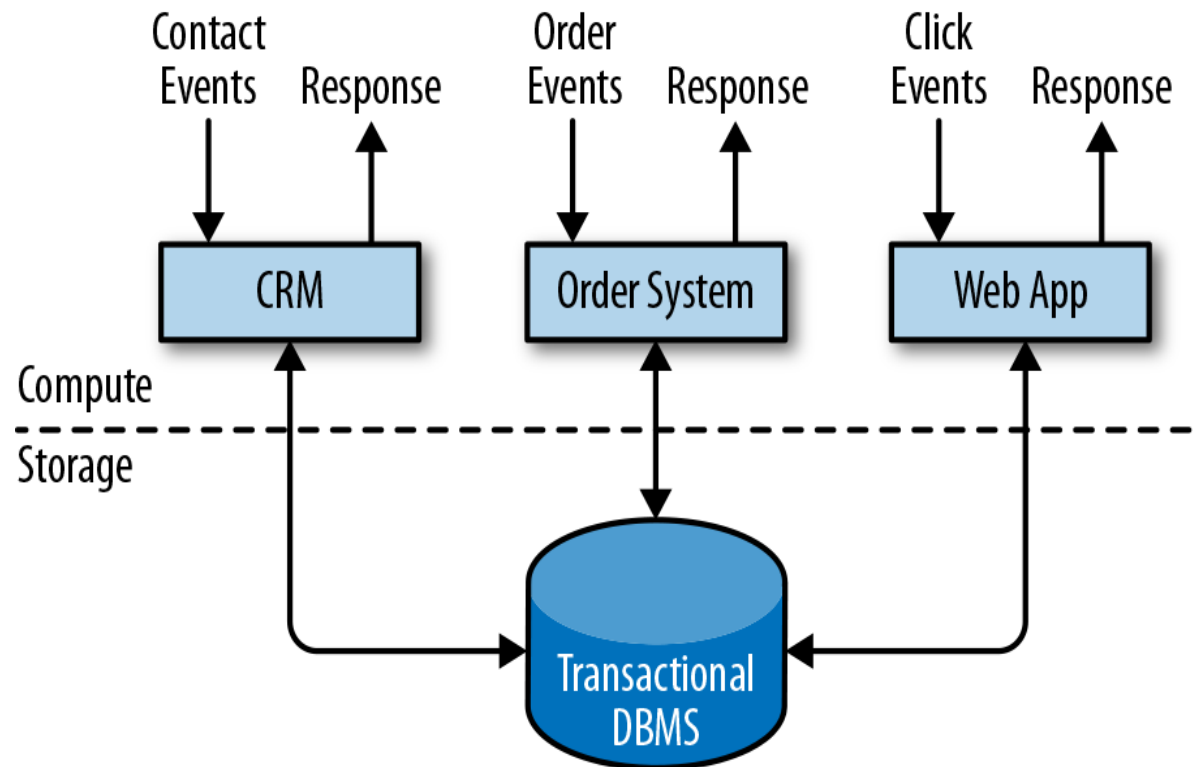
Content Overview

- Data Stream Processing (Part 1)
- Data Stream Processing (Part 2)
- Distributed Database Systems
- Heterogeneous Multi-Database Systems
- Web & XML Data Management
- Knowledge Discovery & Data Warehouses
- Machine Learning in Medicine *
- Scalable & Cloud Data Management
- Performance Analysis & Large DBS *

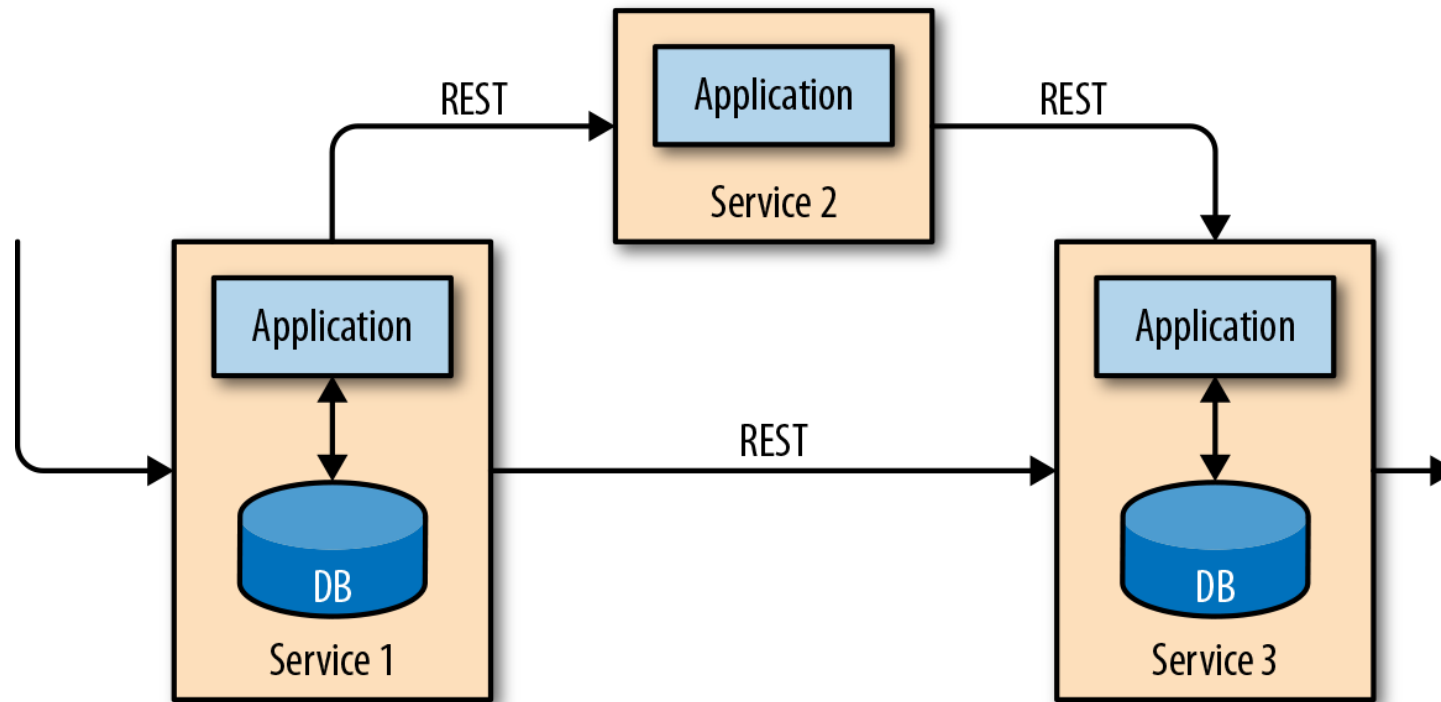
*Guest Lectures

From traditional DBMS to
heterogenous data processing
infrastructure

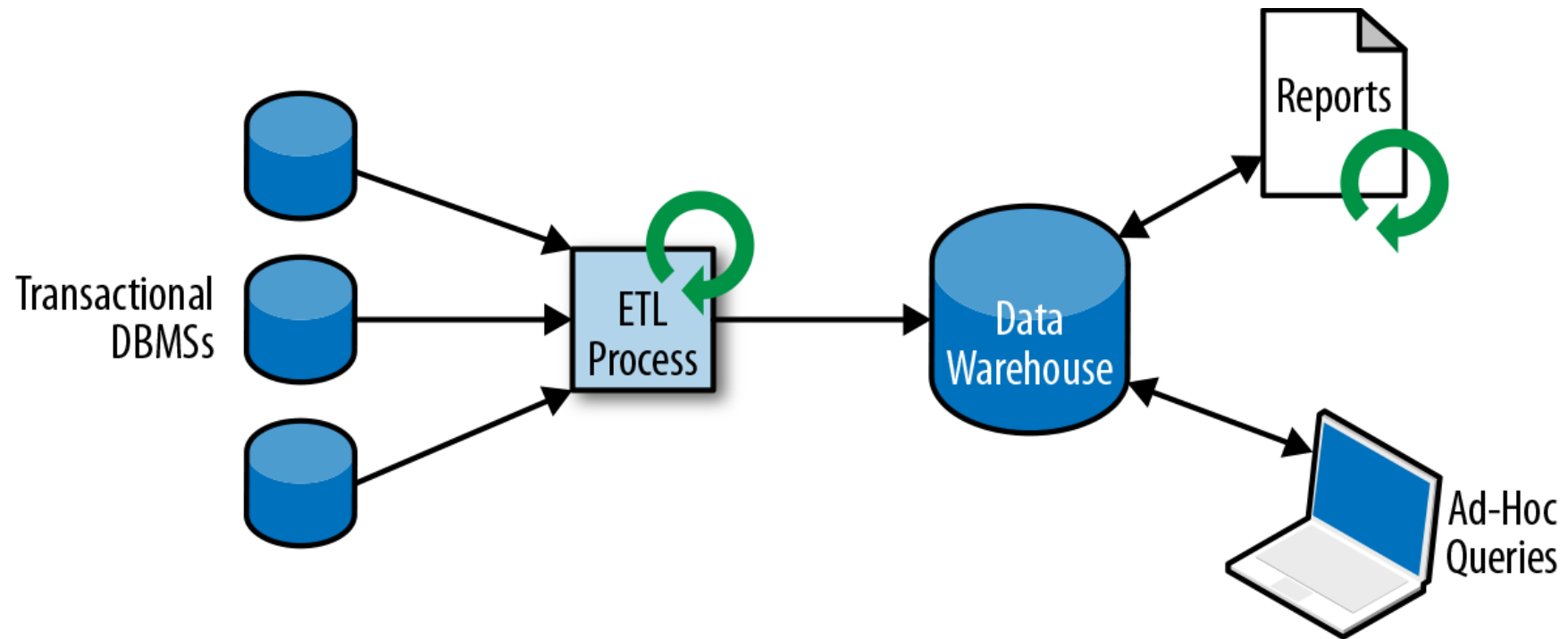
Transactional Processing - Traditional



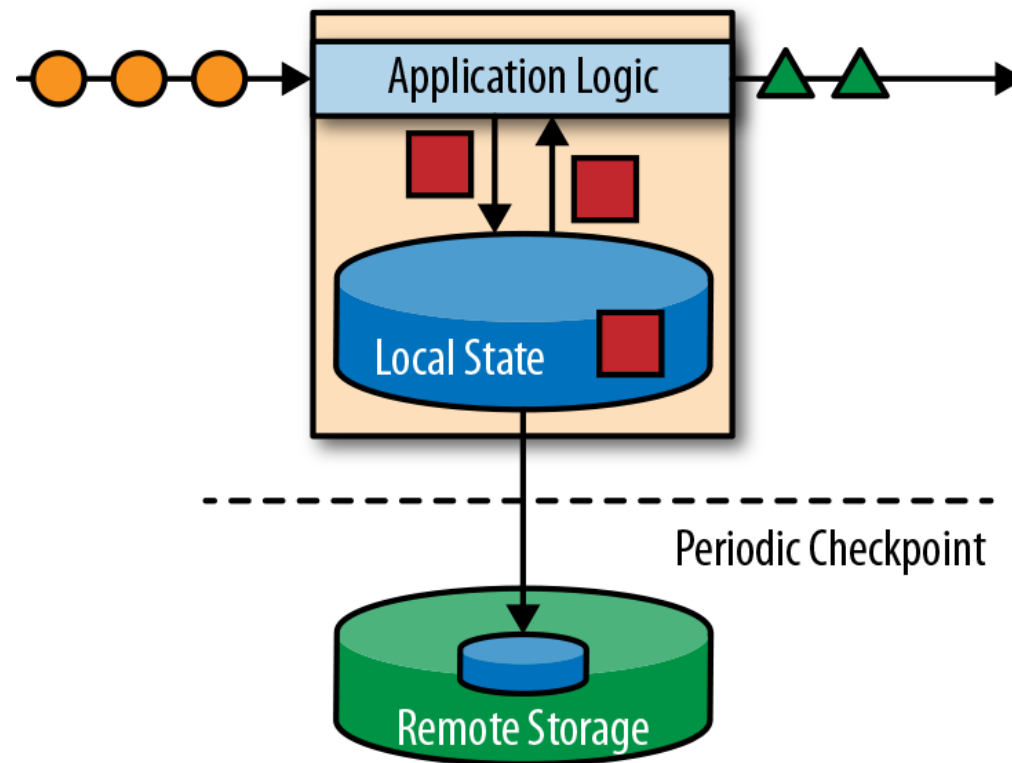
Transactional Processing – Micro Services



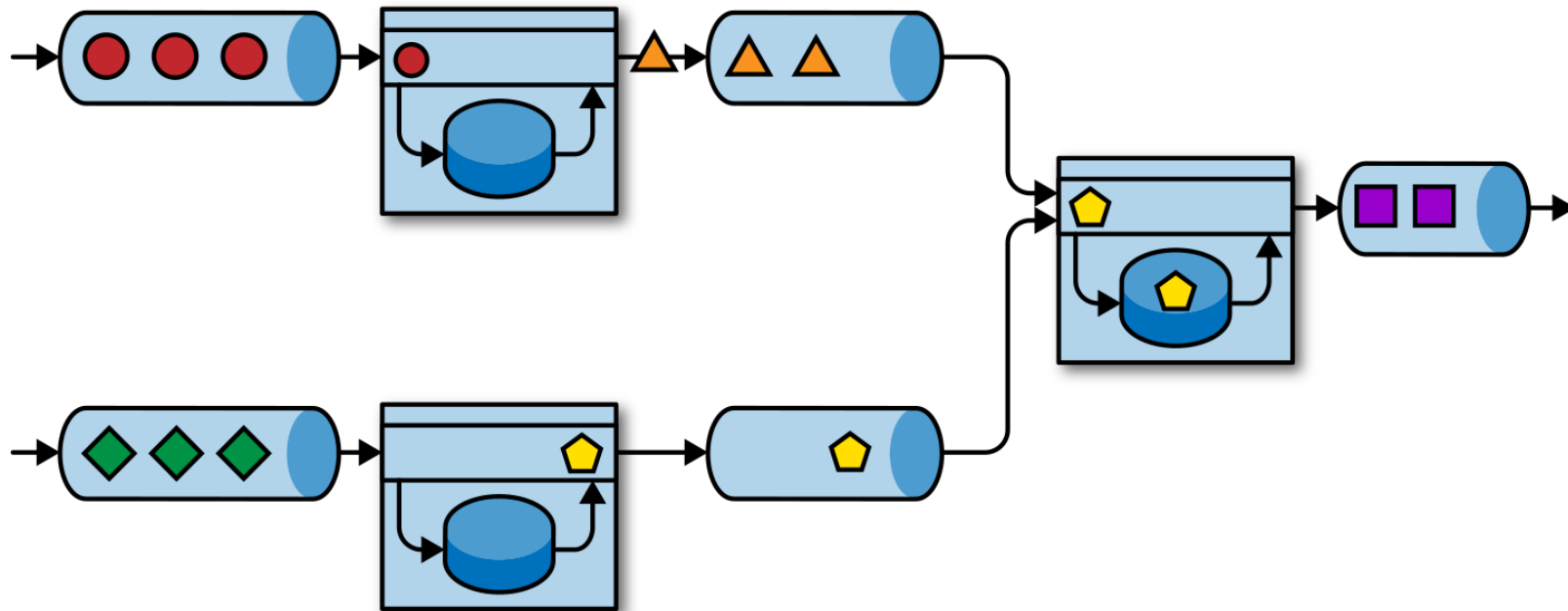
Analytical Processing – Data Warehouse



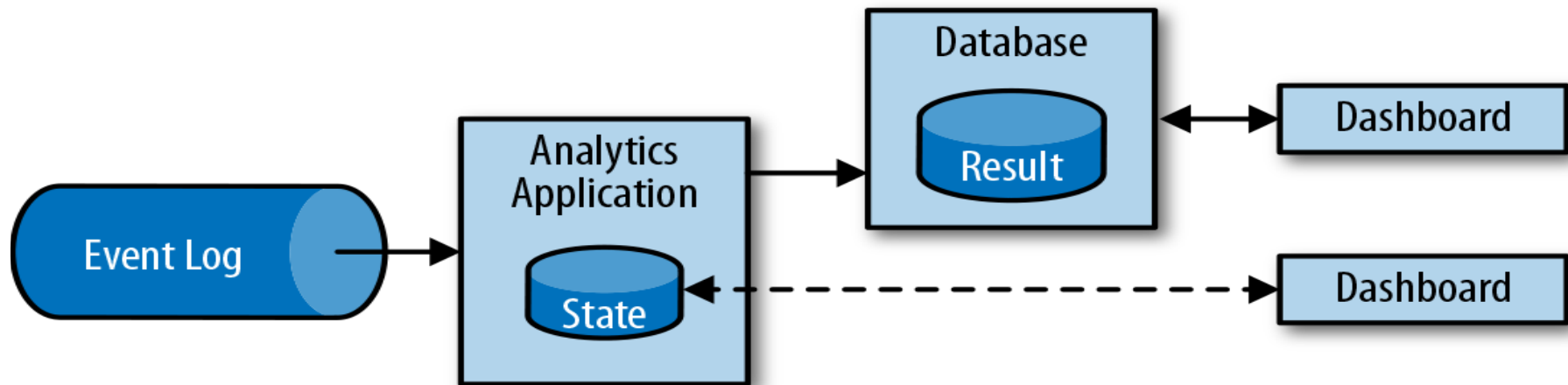
Analytical Processing – Stateful Stream Processing



Analytical Processing – Event Driven Architecture



Streaming Analytics



Looking for Master thesis topics?

- We are part of the DKM group at Ifi:
<https://www.mn.uio.no/ifi/english/research/groups/dkm/>
- Respire project:
Responsible Explainable Machine Learning for Sleep-related Respiratory Disorders
<https://www.mn.uio.no/ifi/english/research/projects/respire/index.html>
- Parrot project:
Privacy Engineering for Real-Time Analytics in Human-Centered Internet-of-Things
<https://www.mn.uio.no/ifi/english/research/projects/parrot/index.html>
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