Systems Infrastructure for Data Science

Web Science Group

Uni Freiburg

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Data Stream Processing

Today's Topic

- Stream Processing
 - Model Issues
 - System Issues
 - Distributed Processing Issues

Distributed Stream Processing Motivation

- Distributed data sources
- Performance and Scalability
- High availability and Fault tolerance

Design Options for Distributed DSMS

- Almost same split as with distributed databases vs cloud databases
- Currently, most of the work is on fairly tightly coupled, strongly maintained distributed DSMS
- We will study a number of general/traditional approaches for most of the lecture, look at some ideas for cloud-based streaming
- As usual, distributed processing is about tradeoffs!

Distributed Stream Processing Borealis Example

End-point Applications Borealis Sources Data Push-based Aurora

Distributed Stream Processing Major Problem Areas

- Load distribution and balancing
 - Dynamic / Correlation-based techniques
 - Static / Load-resilient techniques
 - (Network-aware techniques)
- Distributed load shedding
- High availability and Fault tolerance
 - Handling node failures
 - Handling link failures (esp. network partitions)

Load Distribution

- Goal: to distribute a given set of continuous query operators onto multiple stream processing server nodes
- What makes an operator distribution good?
 - Load balance across nodes
 - Resiliency to load variations
 - Low operator migration overhead
 - Low network bandwidth usage

Correlation-based Techniques

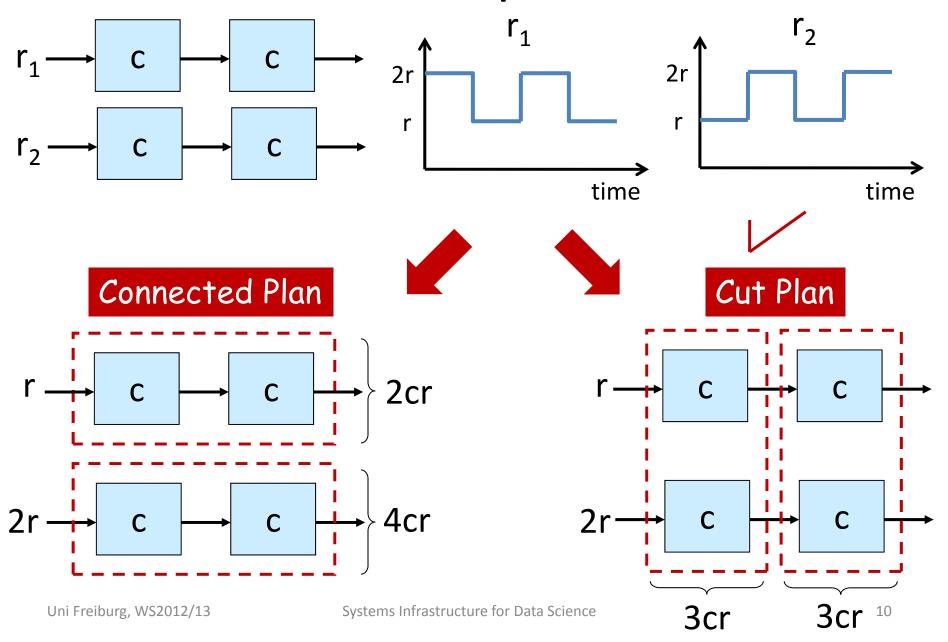
Goals:

- to minimize end-to-end query processing latency
- to balance load across nodes to avoid overload

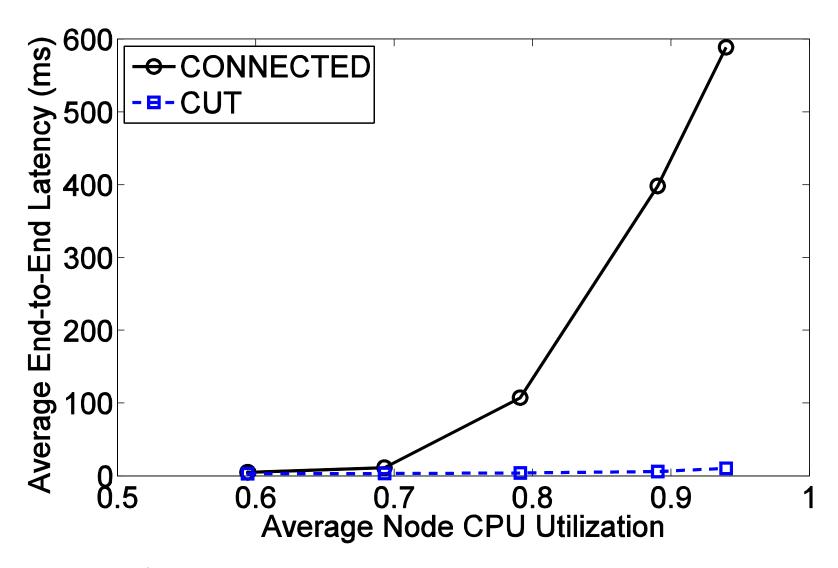
Key ideas:

- Group boxes with small load correlation together
 - ⇒ helps minimize the overall load variance on that node
 - ⇒ keeps the node load steady as input rates change
- Maximize load correlation among nodes
 - ⇒ helps minimize the need for load migration

Example



Example: Cut Plan beats the Connect Plan



Formal Problem Definition

- n: number of server nodes
- X_i: load time series of node N_i
- ρ_{ij} : correlation coefficient of X_i and X_j , $1 \le i, j \le n$
- Find a plan that maps operators to nodes with the following properties:

$$\triangleright$$
 EX₁ \approx EX₂ \approx ... \approx EX_n

- > $\frac{1}{n} \sum_{i=1}^{n} \text{var } X_i \text{ is minimized, or}$
- $\triangleright \sum_{1 \le i < j \le n} \rho_{ij}$ is maximized.

Dynamic Load Distribution Algorithms

- Periodically repeat:
 - 1. Collect load statistics from all nodes.
 - 2. Order nodes by their average load.
 - 3. Pair the ith node with the (n-i+1)th node.
 - If there exists a pair (A, B) such that |A.load B.load|
 threshold, then move operators between them to balance their average load and to minimize their average load variance.
- Two load movement algorithms for pairs in Step 4:
 - One-way
 - Two-way

One-way Algorithm

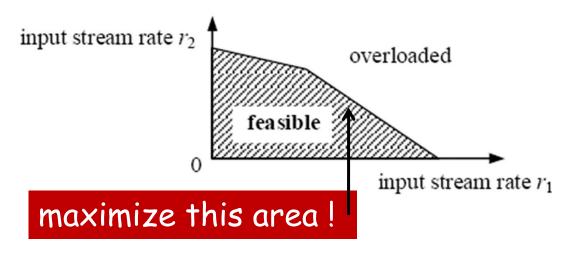
- Given a pair (A, B) that must move load, the node with the higher load (say A) offloads half of its excess load to the other node (B).
- Operators of A are ordered based on a score, and the operator with the largest score is moved to B until balance is achieved.
- Score of an operator O is computed as follows: correlation_coefficient(O, other operators at A)
 - correlation_coefficient(O, other operators at B)

Two-way Algorithm

- All operators in a given pair can be moved in both ways.
- Assume both nodes are initially empty.
- Score all the operators.
- Select the largest score operator and place it at the less loaded node.
- Continue until all operators are placed.
- Two-way algorithm could results in a better placement.
- But, load migration cost would be higher.

Load-resilient Techniques

- Goal: to tolerate as many load conditions as possible without the need for operator migration.
- Resilient Operator Distribution (ROD)
 - ROD does not become overloaded easily in the face of fluctuating input rates.
 - Key idea:



Comparison of Approaches

Correlation-based

- Dynamic
- Medium-to-long term load variations
- Periodic operator movement

Load-resilient

- Static
- Short-term load fluctuations
- No operator movement

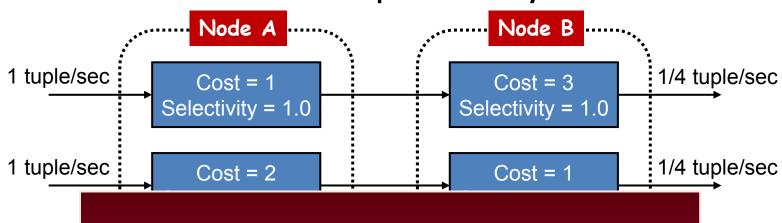
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Distributed Load Shedding

- Problem: One or more servers can be overloaded.
- Goal: Remove excess load from all of them with minimal quality loss at query end-points.
- There is a load dependency among the servers.
- To keep quality under control, servers must coordinate in their load shedding decisions.

Distributed Load Shedding Load Dependency



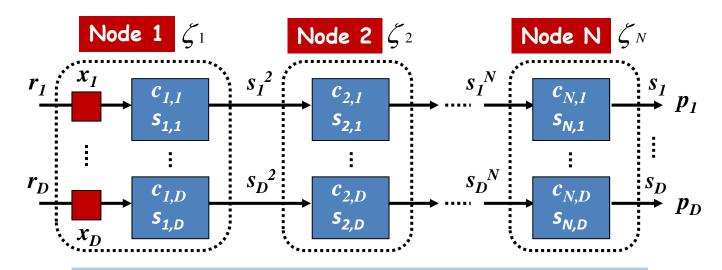
Server nodes must coordinate

	Plan	Rates at A	A.load	A.throughput	B.load	B.throughput
optimal	Q	to ach		101/3/1/3/11		114/4, 1/4
		I U UCI	IEVE		y	
for A	1	1, 0	1	1, 0	3	1/3,0
	2	0 1/2	1	0 1/2	1/2	0 1/2
feasible	_	0, 1/2	4	U, 1/2	1/6	0, 1/2
for both	3	1/5, 2/5	1	1/5, 2/5	1	1/5, 2/5
\longrightarrow	_	_, _, _, _,	_	-, -, -, -,	_	-, -, -, -





Distributed Load Shedding as a Linear Optimization Problem



Find x_j such that for all nodes $0 < i \le N$:

$$\sum_{j=1}^{D} r_j \times x_j \times s_j^i \times c_i, j \leq \zeta_i$$

$$0 \leq x_j \leq 1$$

$$\sum_{j=1}^{D} r_j \times x_j \times s_j \times p_j \text{ is maximized.}$$

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High Availability and Fault Tolerance Overview

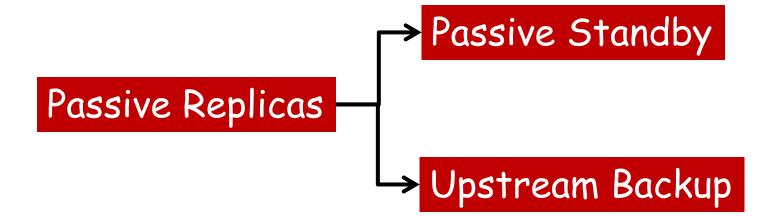
- Problem: node failures and network link failures
 - Query execution stalls
 - ➤ Queries produce incorrect results
- Requirements:
 - Consistency -> Avoid lost, duplicate, or out of order data
 - Performance -> Avoid overhead during normal processing+ overhead during failure recovery
- Major tasks:
 - Failure preparation -> Replication of volatile processing state
 - Failure detection -> Timeouts
 - Failure recovery -> Replica coordination upon failure

High Availability and Fault Tolerance General Approach

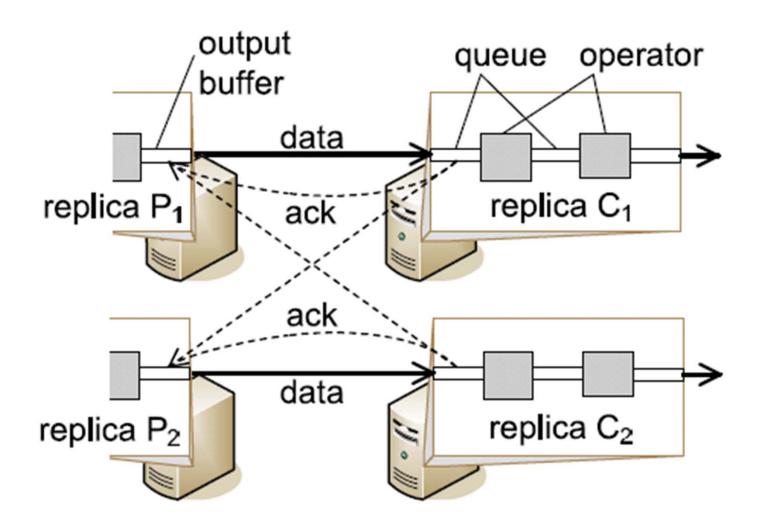
- Adapt traditional approaches to stream processing
- Two general approaches:
 - State-machine approach
 - Replicate the processing on multiple nodes
 - Send all the nodes the same input in the same order
 - Advantage: Fast fail-over
 - Disadvantage: High resource requirements
 - Rollback recovery approach
 - Periodically check-point processing state to other nodes
 - Log input between check-points
 - Advantage: Low run-time overhead
 - Disadvantage: High recovery time
- Different trade-offs can be made among:
 - Availability, Run-time overhead, and Consistency

Handling Node Failures

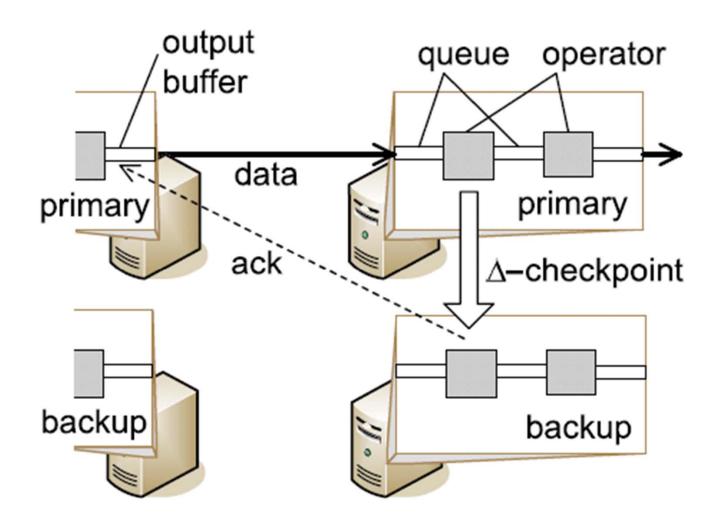
Active Replicas



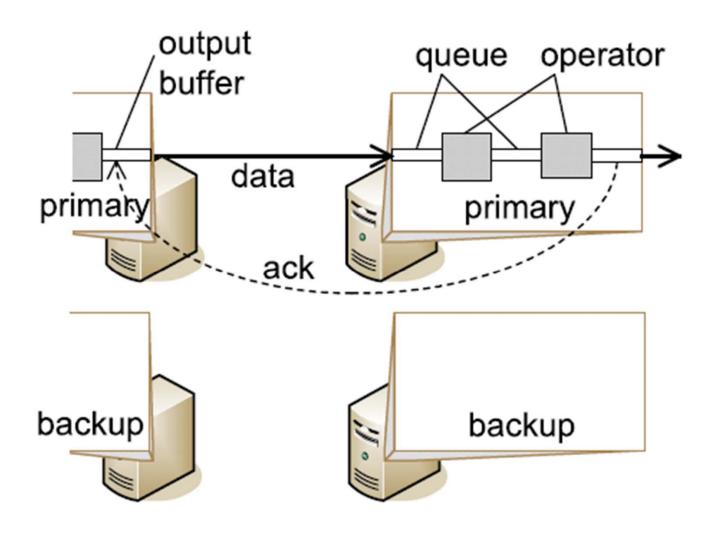
Active Replicas



Passive Standby



Upstream Backup



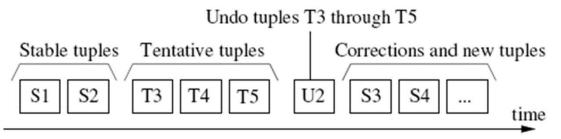
Run-time Overhead vs. Recovery Time Trade-off

Active Replicas:

- High run-time overhead
- Fast fail-over (i.e., low recovery time)
- Passive Standby:
 - Check-point interval can be flexibly adjusted
- Upstream Backup:
 - Low run-time overhead
 - Recovery time is proportional to the size of the upstream buffers

Handling Network Partitions

- "Network Partitions" occur when data sources, processing nodes, and clients are split into disconnected partitions due to network failures.
- Two general options:
 - Suspend processing to avoid inconsistency.
 - Continue processing to avoid unavailability.
- Delay-Process-Correct (DPC) Protocol
 - Adjust the trade-off btw consistency and availability using maximum tolerable latency threshold and tentative tuples.



Other Advanced HA Techniques

- Cooperative and Self-configuring HA [Borealis]
 - Each server node is backed up by multiple servers in a cooperative fashion, which can take over processing in parallel.
 - Backup assignment dynamically changes to balance HA load.
 - Wide-area extensions
- Integrating Fault Tolerance with Load Balancing [Flux]
 - Fine-granularity dataflow partitions
 - Rebalance load after failure recovery