Stream Processing

Bounded data - enumeratable/iteratable upon dataset

Unbounded data - only snapshot iteratable; no size property

Batch processing - apply algorithm on bounded dataset; produce single result

Stream processing - algorithm on continuously updating data; continuous results

Cases for stream processing

- intrusion and fraud detection
- algorithmic trading
- process monitoring
- traffic monitoring

Unix stream processing:

- tail/pipe streaming data acquisition
- pipe intermediate storage
- applying functions on streaming data
- no windowing (streams into batches)
- no triggers (recomputing when new batch)

Stream processing - techniques and systems that process time stamped events

- component -> acquire event from producer and forward to consumer
- component -> event processor

both components should be scalable, distributable and fault-tolerant messaging systems - connecting producers to consumers

Unix: tail -f log.txt I wc -I

tail - producer; wc - consumer

pipe - messaging system

- read data from producer and buffer
- block producer when buffer is full
- notify consumer for available data
- publish/subscribe 1 producer to 1 consumer

publish/subscribe system - connect multiple producers to multiple consumers direct messaging system - simple network communication (UDP) to broadcast message brokers/queues - centralised system, reliable message delivery

Broker-based messaging

Producer message modes:

- Fire and forget broke acks message immediately
- Transaction-based broker writes message to permanent storage before ack

type=topic *.orange.*

..rabbit

lazy.#

10 9 8 7 6 5 4 3 2 1 partition

8 7 6 5 4 3 2 1

10 9 8 7 6 5 4 3 2 1

8 7 6 5 4 3 2 1 partition 2

▶ 5 4 3 2 1 partition 1

consumer 1

Offsets

A.1 : 4 B.1 : 4

Consumer 2



- buffer messages, spill to disk
- route messages to queues
- notify consumers

Consumers:

- subscribe to queue
- ack message receipt

Messaging patterns:

- competing worker multiple consumers on single queue
 fan out each consumer with replicated queue
- Tan out caon consumer with replicated queue
- message routing keys to msg metadata; topic queues specified by key

producer 1

producer 2

Drawback - after message is received, it disappears

no message reprocessingno proof of message delivery

Log-based messaging

log - append-only data structure ' on disk

- Producer appends message to log
- All consumers connect to log and pull messages
- Broker partitions and distributes log to cluster of machines
- Broker keeps track of message offset for consumer per partition

Programming models

Event sourcing and Command Query Segregation (CQS)

- capture all changes to application state as sequence of events
- event causing mutation on application state in immutable log
- specialised systems for scaling writes and reads, stateless app
- separated continuously updated views of app state
- regenerate application state by reprocessing events

Reactive programming

- declarative programming paradigm data streams and propagation of change
- event sources as infinite collections, observers subscribe to receive events

Dataflow model

Processing time - time of observation of event in system

Event time - time of event occurrence

t - processing (wall-clock) time

skew = t - s, s - time stamp of latest event processed

lag = t - s, s - actual time stamp of event

4 dimensions of stream processing:

Processing from Shew Ideal

What

- results computed; operations on streams

Element-wise - apply function on individual message

Stream[A].map(x: A -> B): Stream[B]

Stream[A].filter(x: A -> Boolean): Stream[A]

Stream[A].merge(b: Stream[B>:A]): Stream[B]

Stream[A].flatMap(f: A -> Stream[B]): Stream[B]

Stream[A].keyBy(f: A -> K): Stream[(K, Stream[A])]

Stream[A].join(b: Stream[B], kl: $A \Rightarrow K$, kr: $B \Rightarrow K$, rs: $(A,B) \Rightarrow R$): Stream[R]

Aggregations - group events together to apply reduction

Where

- event time computation; streaming windows

Window - static size or time-length batches of data

Session window - dynamically sized, aggregate batches of user activity

When

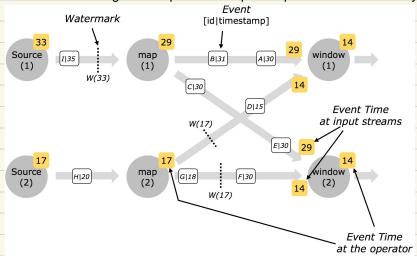
- processing time materialisation/processing

Triggers:

- per-record trigger fire after a number of records
- aligned delay fire after a number of seconds across all active windows
- unaligned delay fire after a number of seconds after first event in window

Watermarks - declaration that all events before this time stamp have arrived

- allow late messages to be processed up to a specific amount of delay



How

earlier results relation to later refinements
 triggers + watermarks = multiple materialisations per window
 discard, accumulate or accumulate + retract