Adding Back-pressure to Spark Streaming

Dean Wampler, Typesafe

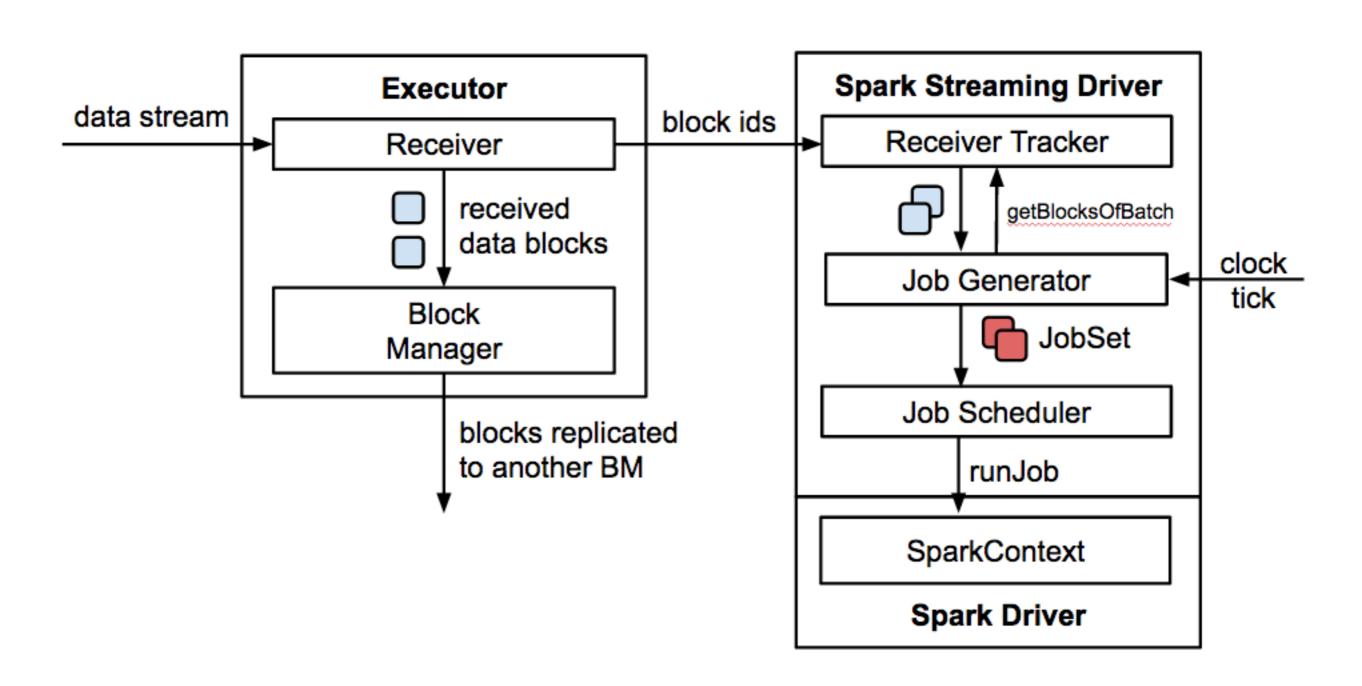
Typesafe and Spark

- Spark aligned with our passion for distributed, streaming ("fast data") systems... and Scala!
- We're contributing to Spark's use of Scala,
 Streaming, and Mesos integration.
- Commercial support for Spark on Mesos.
 - Mesosphere Infinity.

The problem

Data may arrive faster than Spark can process it, leading to an unstable system (delays, memory spills). Worst-case, you run out of memory!

Spark Streaming Architecture



If the batch interval is too short, data accumulates in the system

Rate Limiting in Spark 1.4

- You could set a **static** rate limit
 - Had to be conservative
 - difficult to find the right limit (depends on cluster size)
 - one limit to rule all streams for all time!

Back-pressure



Back-pressure

- (aka, flow control)
- a slow consumer should slow down the producer
 - Classic example, TCP
 - New example: Reactive Streams
 - reactive-streams.org

Back-pressure in Spark 1.5

- A dynamic rate limiter
 - Estimate the number of elements that can be safely processed by the system in the *next* batch interval
 - 2. Communicate the limit to Receivers
 - 3. Rely on TCP to slow down producers on the other side of the channel

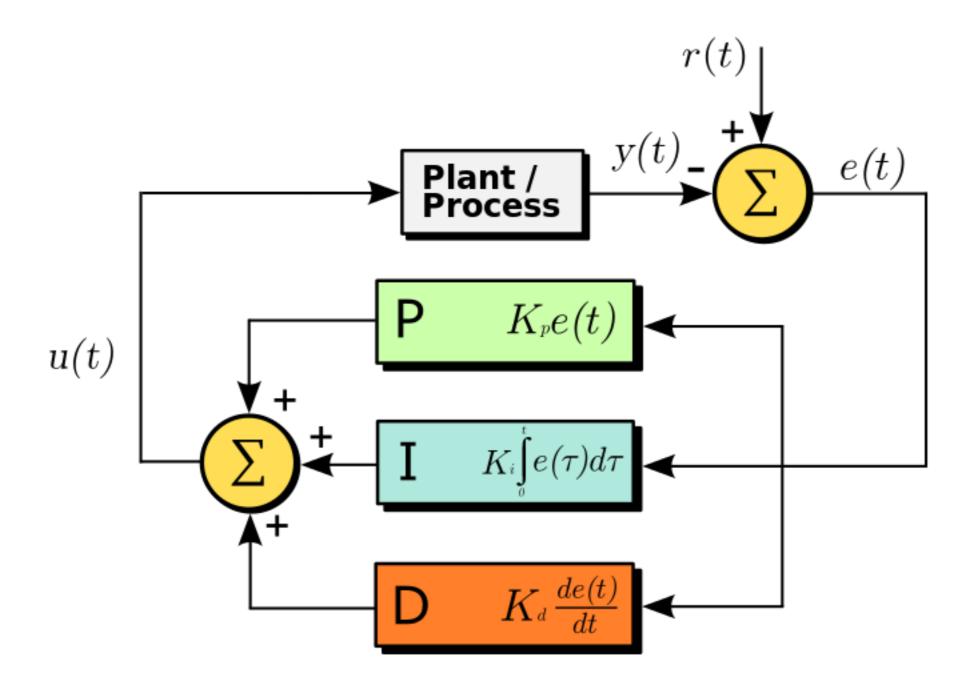
Note: What's Missing?

- What about other distributed streams?
- Do we want to rely solely on TCP backpressure?
 - We'll come back to this...

More Details...

- Each BatchCompleted event contains
 - processing delay, scheduling delay
 - number of elements in each InputStream
- the rate is elements/processingDelay (roughly)
 - but what about accumulated data?

- Proportional, Integrative, Derivative
 Controller
- https://en.wikipedia.org/wiki/PID_controller



- Goal: Keep the processing time close to batch interval
- Proportional term: Use the error:
 - batchInterval processingTime/batch
- Integral component: schedulingDelay
- Derived component: error previousError

- Not all 3 terms required.
- Does not require process knowledge (black box)
- P, I, D constants can change convergence, over-shooting, oscillations

To see how the constants affect convergence:

https://upload.wikimedia.org/wikipedia/commons/3/33/ PID_Compensation_Animated.gif

Back pressure in Spark 1.5

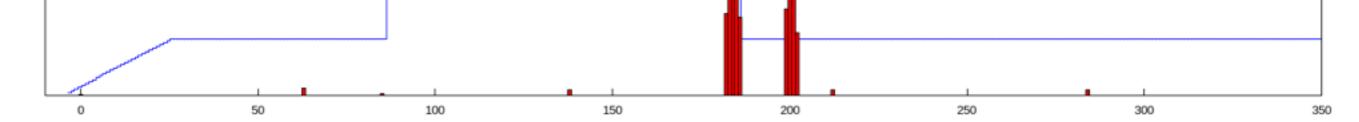
- Each input stream gets its own estimator
- Works with Receiver based inputs, and also KafkaDirectInputStream
- To enable, set the following properties:

```
spark.streaming.backpressure.enabled true
spark.streaming.backpressure.minRate R
```

What it does in practice.

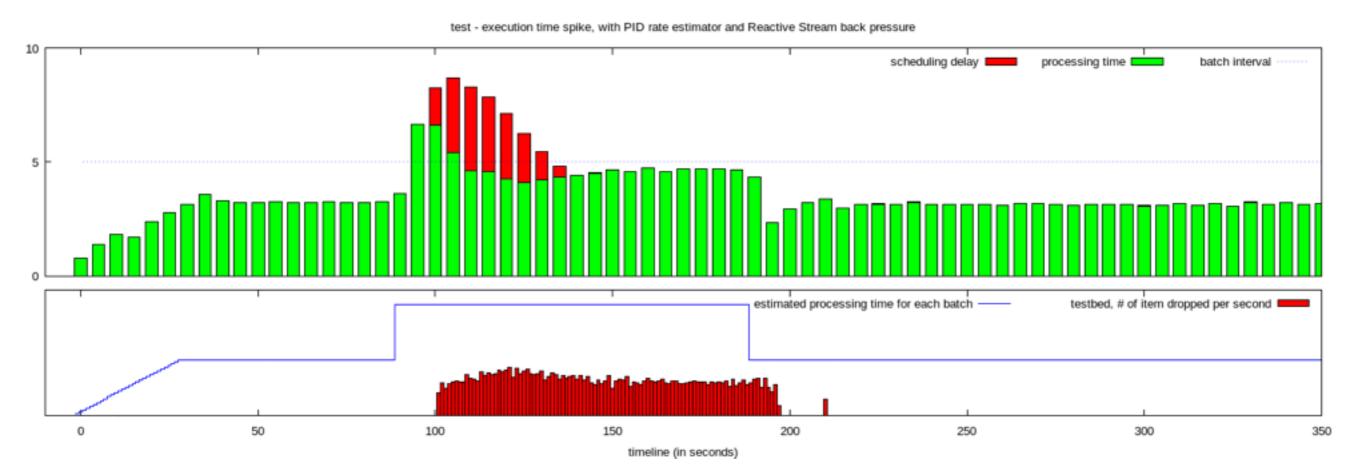
Before...

after



estimated processing time for each batch

testbed, # of item dropped per second



Limitations

- Doesn't account for records of different sizes
- Linearity assumption not accurate
- Back pressure accumulates data in the TCP channel; what happens to the data congesting at the source end??

A Peek at Reactive Streams

Reactive Streams

- reactive-streams.org
- A standard for asynchronous stream processing
- Consumer controls rate by asking for elements from producers

Reactive Streams

- A push model is used when the consumer can keep up.
- Switches to a pull model when rate limiting required.
- Dynamically switches back and forth...

Reactive Streams... in Spark?

- Hoped to be ready for v1.6, but TBD:
- JIRA 10420 (Reactive Streams Receiver)

Credits

- Typesafe: Luc Bourlier, Iulian Dragos,
 Nilanjan Raychaudhuri, Dean Wampler
- (formerly Typesafe) François Garillot
- Databricks: Tathagata Das (TD)