Throttling Detection and Reactive Worker Downscaling

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- Problem statement
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- Throttling metrics
- SDK implementations
- Runner implementations



Problem statement

Executing Beam pipelines on Dataflow exposes some rough edges. Dataflow autoscaling can spin up a large number of workers that can easily **overload** services that IO connector involves...



Beam 2.56, BigQuery DIRECT_READ TPC-DS_1T (tpc.org/tpcds)



How does parallelism get determined?

- Available parallelism
 - Parallelism of source / Number of keys
 - Whether source supports liquid sharding, extent of granularity

Type of sources by ability to parallelize

- No parallelism (e.g. JdbcIO.read)
- Static parallelism (e.g. JdbcIO.readWithPartitions)
- Bounded Dynamic parallelism (e.g. KafkaUoundedSource)
- Liquid sharding (e.g. FileIO readRange, BigQueryIO DIRECT READ)



How does parallelism get determined?

- Job signal (up/back/two-way pressure)
 - CPU usage
 - [back pressure] Batch: no upscaling if CPU usage < 5% [1]
 - [two-way pressure] Streaming: utilization hint [2]
 - User supplied metrics
 - [up pressure] Streaming: sources' BacklogBytes
 - [back pressure] Client side throttling metrics

[1] https://cloud.google.com/dataflow/docs/horizontal-autoscaling#batch

[2] cloud.google.com/dataflow/docs/guides/tune-horizontal-autoscaling#update-hint



Throttling metrics

A Beam counter metrics under dedicated name

(THROTTLE_TIME_COUNTER_NAME = throttling-msecs)

To setup such metrics

• Java SDK:

Counter throttlingMsecs = Metrics.counter(Metrics.THROTTLE_TIME_NAMESPACE, Metrics.THROTTLE_TIME_COUNTER_NAME);

• Python SDK:

THROTTLE_COUNTER = Metrics.counter(__name__, 'cumulativeThrottlingSeconds')



Throttling metrics

Notes:

- Under standard Beam metrics API, **NO** dedicated Beam model (proto) level API
- *throttling-msecs* counter long existing in a few Beam IOs (Datastore, GcsIO, BigQuery *STREAM_INSERT*) and works for Dataflow legacy worker.
- Recently generalized (<u>#31924</u>, 2.59.0) and extended into more IOs (BigQuery Storage API, Bigtable write), works with Dataflow runner v2.
- Up to runner to react on the throttling metrics as of scaling decision.



- Goal: extract the throttled time spent on interacting with external resources
 - wait time, latency
 - retry backoff, etc

Depends on where the throttled time incurred





 Case 1: Server returns HTTP 429 immediately, and **Beam** retry with backoff (e.g. BigQueryIO Storage Write API)

Approach: Set FluentBackoff.withThrottledTimeCounter (#29098, 2.53.0)



 Case 2: Client library involves retry, and backoff time tracked in retry callback (e.g. GcsUtil, BigQueryIO Storage Read API)
Caveat: callback thread may not have metrics container initialized*

Approach: incrementing pending throttling time and report back in DoFn (#31096)



*Warning: Reporting metrics are not supported in the current execution environment

 Case 3: Client library involves retry, and backoff time not tracked (e.g. BigtableIO write)

Bigtable client has built-in rate limiting mechanism, and associated configs, e.g.

- setThrottlingTargetMs
- setFlowControl



• Case 4: Server latency spikes, time out

Exact throttled time not known, same as Case 3.

Approach (for both Case 3&4):

- a configurable targeted latency time on API call
- if latency>set time, excessive amount is reported as throttling time



Throttling metrics: Runner implementations

• Dataflow runner scaling decision

Fraction of throttled time:

r=recent throttle time / recent done time

The targeted parallelism overwritten as

p = (1-r)p'

(p': targeted parallelism without considering the throttled time)

Downscale decision issued after p<current parallelism for 3 minutes.



Example: BigQuery Storage Read API (<u>#31404</u>)





upscale to 270 worker aggressively, before job cancel

After:



scales down until finish (autoscaler won't kill active worker until current work item finish)



Throttling metrics: Runner implementations

- Metrics defined in SDK side, and it is IO connector's responsibility to report
- Runner can react on the throttling metrics, and
- Decision on how runner reacts is runner dependent
 - Currently supported by Dataflow
 - Interface for other runners' support is open



Thank you!

Questions?

Resources:

Dataflow: Tune Horizontal autoscaling https://cloud.google.com/dataflow/doc s/guides/tune-horizontal-autoscaling

