



How to benchmark your Beam pipelines for performance and cost

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About me



- Cloud Solutions Architect @ Google Cloud
 - Team: Data Analytics, Solutions Engineering
 - Industries
 - Financial services, Healthcare & Life Sciences, Manufacturing & Retail
 - Use cases
 - Data movement, log analytics, security analytics
 - Products
 - BigQuery, Dataflow, Pub/Sub



Why benchmark thy pipeline

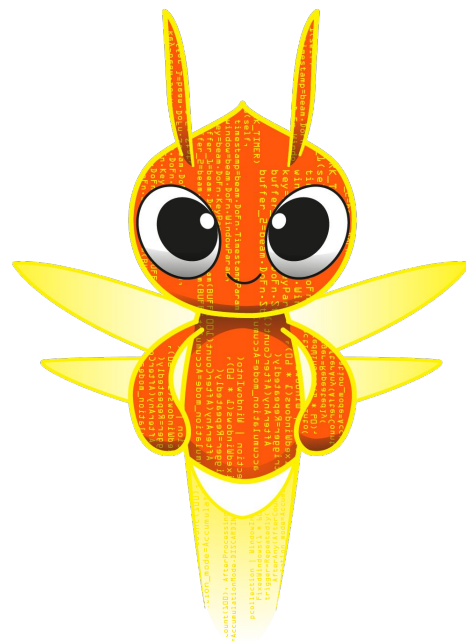
- Will my pipeline meet expected SLOs?
 - Job/event e2e latency, Event throughput (EPS)
- Will my pipeline operate with optimum performance/cost ratio
 - Maximize resource utilization
- Is my pipeline properly sized **and** configured?



Disclaimers



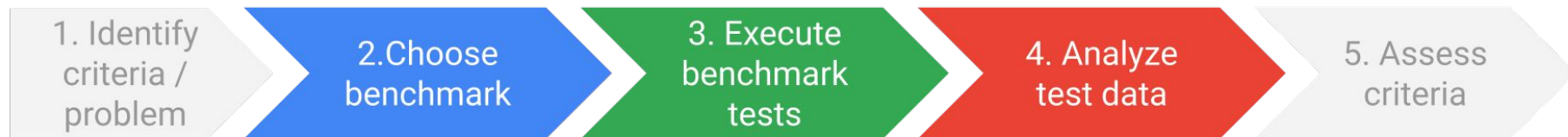
- Results presented are for demo purposes only.
 - No performance or cost guarantees.
 - Your mileage may vary.
 - Test your pipeline using your own real data and environment
- No pipeline harmed in this benchmarking process





Why benchmarking is so hard

- Heterogeneous data types, pipeline types, stages, cloud providers
- Selecting appropriate benchmark tests
- Configuring nontrivial environments
- Variety of Dataflow user-facing performance metrics
- Variability in performance due to number of moving parts
- Analyze data and get actionable insights



Benchmarking toolset



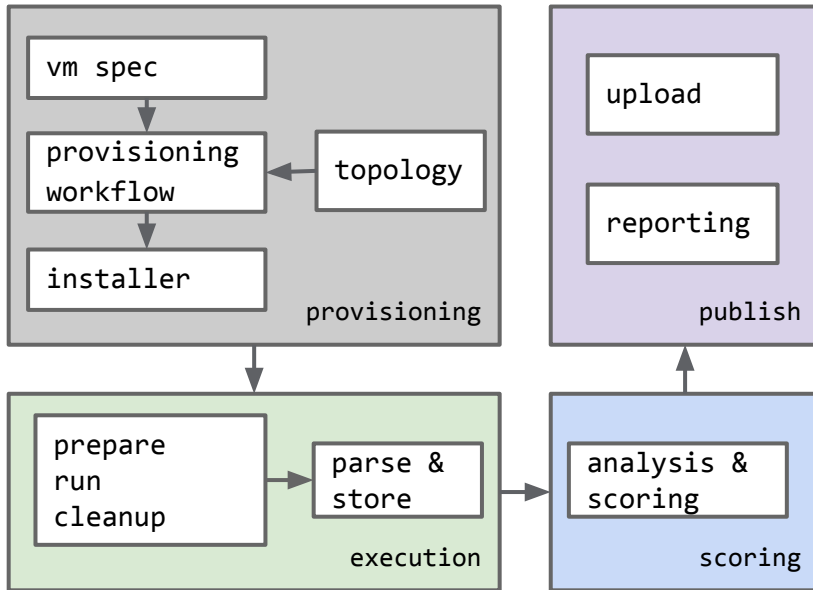
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PerfKit Benchmarkmarker

Open Source Benchmark Framework



- A **canonical** set of **public** benchmarks to measure and compare cloud offerings.
- PKB takes care all the work from provisioning, executing, cleaning up and publishing results through CLI.
- All benchmarks are running with **default** settings. (Not tuned to in favor of any providers).
- Supported providers: **GCP, AWS, Azure, Digital Ocean, Rackspace, OpenStack**
- Contribution from Stanford, MIT, CloudHarmony, CloudSpectator, etc.

Benchmark Example 1

Beam Wordcount





Example 1: Wordcount

- Apache Beam [WordCount example](#)
- PKB [wordcount benchmark](#)
- Running on different machine types

```
flags: &myflags
  dpb_service_zone: us-central1-a
  dpb_job_jarfile: ./word-count-beam/target/word-count-beam-bundled-0.1.jar
  dpb_job_classname: org.apache.beam.examples.WordCount

benchmarks:
- dpb_wordcount_benchmark: {
  dpb_service: { service_type: dataflow, worker_count: 1, worker_group: *eight_core },
  flags: *myflags
}
- dpb_wordcount_benchmark: {
  dpb_service: { service_type: dataflow, worker_count: 1, worker_group: *four_core },
  flags: *myflags
}
- dpb_wordcount_benchmark: {
  dpb_service: { service_type: dataflow, worker_count: 1, worker_group: *two_core },
  flags: *myflags
}
```

Demo



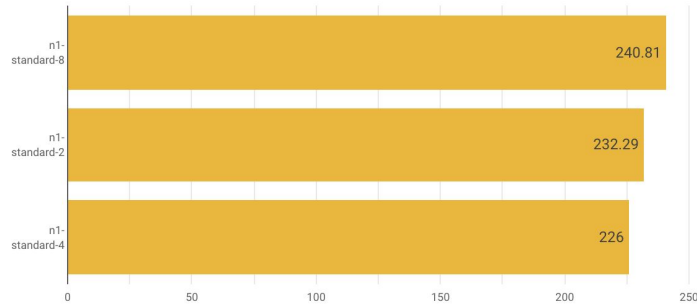
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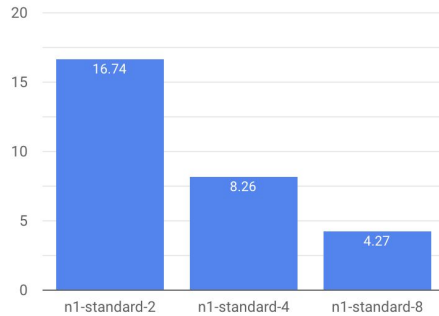


Example 1: Wordcount perf results

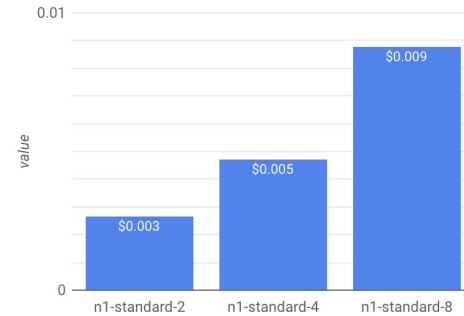
Run Time



Average CPU Utilization (%)



Total Cost



Benchmark Example 2

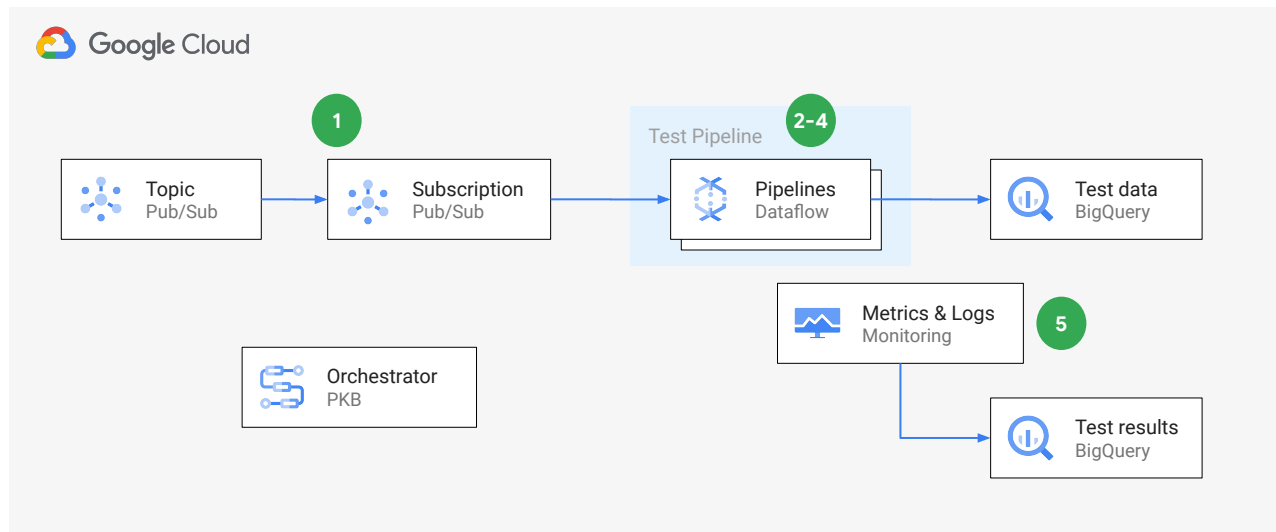
Custom or Google-provided
Dataflow template





Example 2: Dataflow Template

Pub/Sub Subscription to BigQuery



*Orchestrator state machine:

1. Deploy test bed and seed input data
2. Restore PubSub subscription snapshot
3. Run pipeline with configuration A
4. Wait until all messages are processed
5. Collect stats into BQ: pipeline utilization, cost, latency, and throughput
6. Repeat steps 2-5 for N different pipeline configurations
7. Synthesize all tests results into **optimal unit worker size** (for perf/\$) and **horizontal scaling guideline** for pipeline to meet SLO



Example 2: Dataflow Template

- Dataflow template: [Pub/Sub Subscription to BigQuery](#)
- PKB dataflow template benchmark
- Run across different recent versions

```
dpb_df_template_benchmark:  
  description: Run Dataflow template  
  dpb_service:  
    service_type: dataflow  
    worker_count: 1  
    worker_group: *four_core  
  flag_matrix: recent_versions  
  flag_matrix_defs:  
    recent_versions:  
      dpb_df_template_gcs_location:  
        - gs://dataflow-templates/2022-07-04-00_RC00/PubSub_Subscription_to_BigQuery  
        - gs://dataflow-templates/2022-06-27-00_RC00/PubSub_Subscription_to_BigQuery  
        - gs://dataflow-templates/2022-06-21-00_RC00/PubSub_Subscription_to_BigQuery  
        - gs://dataflow-templates/2022-06-06-00_RC00/PubSub_Subscription_to_BigQuery  
        - gs://dataflow-templates/2022-05-30-00_RC00/PubSub_Subscription_to_BigQuery
```



Example 2: Dataflow Template

- Dataflow template: [Pub/Sub Subscription to BigQuery](#)
- PKB dataflow template benchmark
- Run across different input sizes

```
dpb_df_template_benchmark:  
  description: Run Dataflow template  
  dpb_service:  
    service_type: dataflow  
    worker_count: 1  
    worker_group: *four_core  
  flag_matrix: input_sizes  
  flag_matrix_defs:  
    input_sizes:  
      dpb_df_template_input_subscription:  
        - projects/[MY_PROJECT]/subscriptions/perftest-pubsub-input-sub-100k  
        - projects/[MY_PROJECT]/subscriptions/perftest-pubsub-input-sub-1M  
        - projects/[MY_PROJECT]/subscriptions/perftest-pubsub-input-sub-10M
```

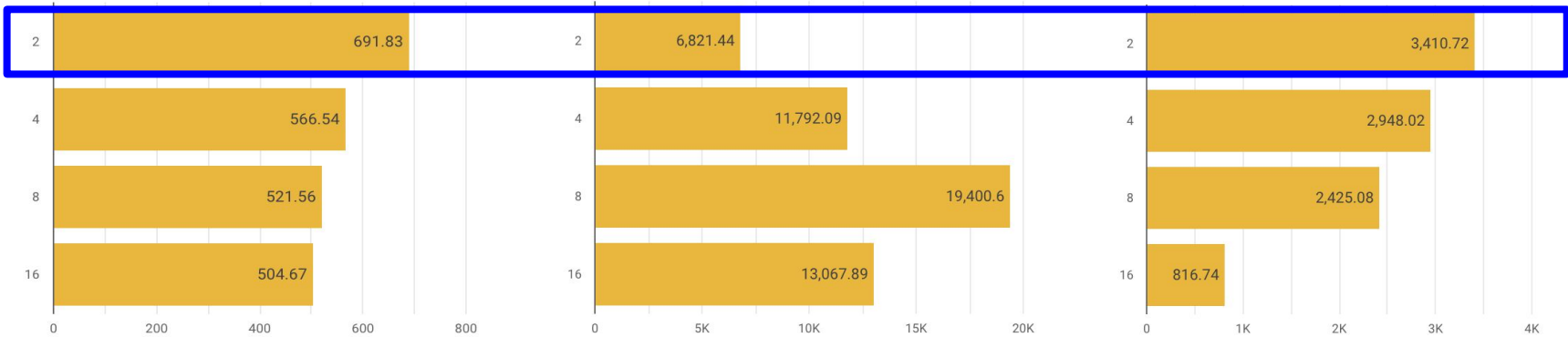
Example 2: Pipeline perf results



Run Time

Max Throughput (EPS)

Max Throughput per vCPU (EPS/core)



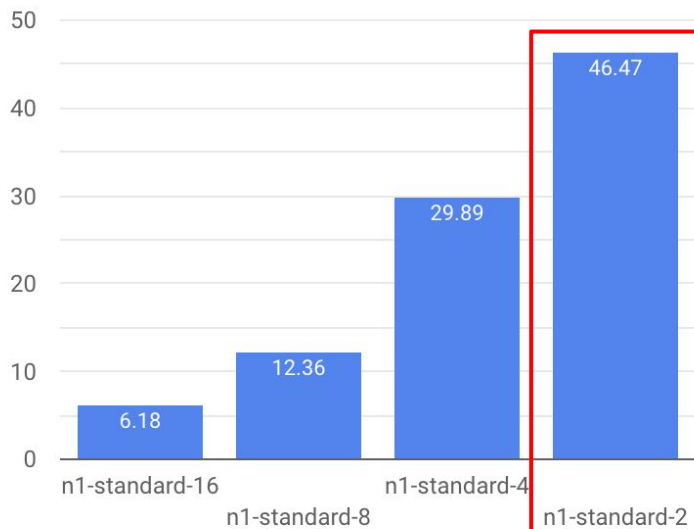
	vCPUs*	Mem (GB)*	Max egress (Gbps)*
n1-standard-2	2	7.5	10
n1-standard-4	4	15	10
n1-standard-8	8	30	16
n1-standard-16	16	60	32

*See [Compute Engine docs](#) for latest specs and factors

Example 2: Utilization & cost results



Average CPU Utilization (%)



Total Cost



Caveats and next steps



1. Variability in pipeline performance

- Run time, spin up/down time, failures
 - **Iterate each test configuration N times**
 - **Pre-warm pipeline before running each test**
 - **Mark test complete when all events written to sink**



2. Dependency on testbed setup, and source/sink characteristics

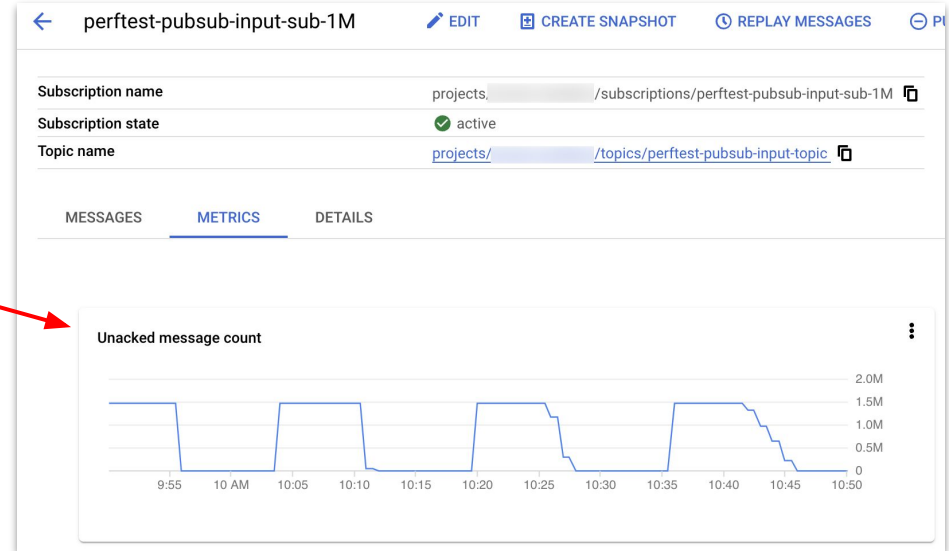
- Networking, Cloud NAT
- Limited to upstream and downstream performance
 - e.g. BQ write throughput quota per project

Caveats and next steps



3. Sensitivity to input workload

- Realistic benchmark and stress testing
 - **Backlog**
 - **Steady-state**
 - **Steady-state with bursts**
 - **Steady-state with step function**
- Realistic data sets
 - **Run different types of data**





Your mileage may vary..
based on your data, environment and workload profile



Call to Action

- **Benchmark** your own pipelines
 - Estimate costs, plan capacity to meet SLOs, and avoid performance regressions
- **Use** battle-tested PerfKit Benchmarker
- **Share** feedback and suggestions
- **Extend** PerfKit Benchmarker for benefit of Dataflow & Beam community

Wiki: <https://googlecloudplatform.github.io/PerfKitBenchmarker/>

GitHub: <https://github.com/rarsan/PerfKitBenchmarker/> (to be merged upstream)

Acknowledgements



- **Diego Orellana**, Software Engineer @ Google, PerfKit Benchmarking
- **Sergei Lilichenko**, Cloud Solutions Architect @ Google, Data Analytics
- **Rodd Zurcher**, Cloud Solutions Architect @ Google, App/Infra Modernization

Questions?

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