# Caching in Dataflow using Beam SDK

By Zeeshan Khan | Google



### What is a cache



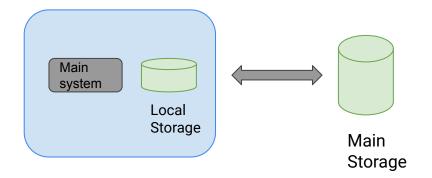
"software component that stores data so that future requests for that data can be served faster; the data stored in a cache might be the result of an earlier computation or a copy of data stored elsewhere" - Wikipedia

#### Why have a cache?

- Calls to external systems are expensive
- Lower latency
- Better throughput

#### Trade offs:

- Stale data





# Options in Beam SDK



### In memory caches

- Side Inputs
- Shared class

Stateful DoFn

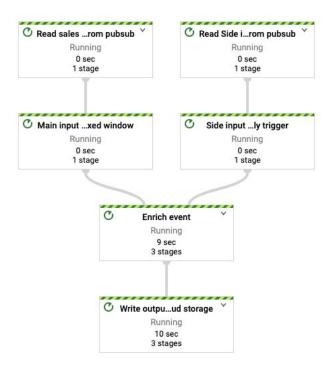


# In memory cache: Side Input



"A side input is an additional input that your DoFn can access each time it processes an element in the input PCollection"

This cache is a copy of data store elsewhere for example BigQuery, text files on GCS etc.







```
main_input = (
    pipeline
    | 'MpImpulse' >> beam.Create(sample_main_input_elements)
    | 'MapMpToTimestamped' >> beam.Map(lambda src: TimestampedValue(src, src))
    | 'WindowMpInto' >> beam.WindowInto(window.FixedWindows(main_input_windowing_interval)))
```

5





```
input = (
    pipeline
    | 'MpImpulse' >> beam.Create(sample_main_input_elements)
    | 'MapMpToTimestamped' >> beam.Map(lambda src: TimestampedValue(src, src))
    | 'WindowMpInto' >> beam.WindowInto(window.FixedWindows(main_input_windowing_interval)))

output = (
    input
    | 'ApplySomeFunction' >> beam.FlatMap(some_function))
```





```
main_input = (
    pipeline
    | 'MpImpulse' >> beam.Create(sample_main_input_elements)
    | 'MapMpToTimestamped' >> beam.Map(lambda src: TimestampedValue(src, src))
    | 'WindowMpInto' >> beam.WindowInto(window.FixedWindows(main_input_windowing_interval)))

side_input = (pipeline
    | 'PeriodicImpulse' >> PeriodicImpulse( first_timestamp, last_timestamp, interval, True)
    | 'MapToFileName' >> beam.Map(lambda x: src_file_pattern + str(x))
    | 'ReadFromFile' >> beam.io.ReadAllFromText())
```





```
main input = (
    pipeline
      'MpImpulse' >> beam.Create(sample_main_input_elements)
     'MapMpToTimestamped' >> beam.Map(lambda src: TimestampedValue(src, src))
      'WindowMpInto' >> beam.WindowInto(window.FixedWindows(main input windowing interval)))
side input = (pipeline
      'PeriodicImpulse' >> PeriodicImpulse( first timestamp, last timestamp, interval, True)
      'MapToFileName' >> beam.Map(lambda x: src_file_pattern + str(x))
      'ReadFromFile' >> beam.io.ReadAllFromText())
output = (
    main input
    | 'ApplySomeFunction' >> beam.FlatMap(some function, rights=beam.pvalue.AsIter(side_input))))
```

# In memory cache: Side Input



"A side input is an additional input that your DoFn can access each time it processes an element in the input PCollection"

#### Can be passed as:

- Singleton
- List
- Iterable
- Map
- Multimap



# Technical consideration on Dataflow



- Streaming jobs without Streaming Engine store side input in memory. For Java pipelines, there is 1 copy per worker, while for Python there is 1 copy per vCPU.
- Streaming jobs using Streaming Engine have a limit of 80MB as max size of side input
- For best performance, side inputs should be small (less than 1GB)



# Beam's Shared Class

apache\_beam.utils.shared.Shared



### Shared class



- Provides a way to sharing in-memory data object across multiple threads/DoFn with a process to improve space and access efficiency.

#### **DoFn lifecycle**

- setup Invoked after creation of an instance.
- Repeatedly process bundles:
  - start\_bundle Invoked before processing of each bundle.
  - Repeatedly process elements
    - **process** method
  - **finish\_bundle** Invoked after processing of each bundle
- **teardown** Invoked before instance is discarded and used for any clean up



# Shared class



13

- There can be hundreds of instances on Dofn on a Dataflow worker

	Batch job	Streaming job	
		Without SE	With SE
Java	1 DoFn/vCPU	300 DoFn/vCPU	500 DoFn/vCPU
Python	1 DoFn/vCPU	12 DoFn/vCPU	12 DoFn/vCPU



Austin, 2022

### Shared class



- Beam provides the Shared class (apache\_beam.utils.shared.Shared) to share object across threads. One object per class

```
class GetNthStringFn(beam.DoFn):
    def __init__(self, shared_handle):
        self._shared_handle = shared_handle

    def process(self, element):
        def initialize_list():
            # Build the giant initial list.
            return [str(i) for i in range(1000000)]

        giant_list = self._shared_handle.acquire(initialize_list)
        yield giant_list[element]

p = beam.Pipeline()
shared_handle = shared.Shared()
(p | beam.Create([2, 4, 6, 8])
            | beam.ParDo(GetNthStringFn(shared_handle)))
```

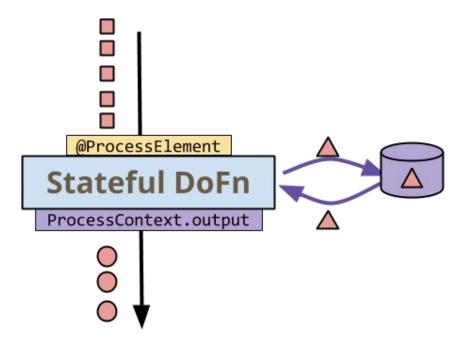
# Stateful DoFn



## Stateful DoFn

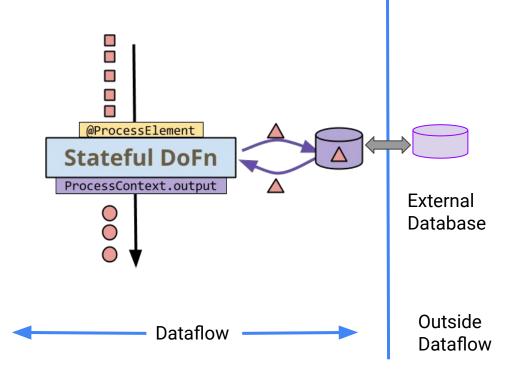


- 1
- With state, a DoFn has the ability to access persistent mutable state while processing each input element.
- State is persisted per key per window



## Stateful DoFn

- 1
- With state, a DoFn has the ability to access persistent mutable state while processing each input element.
- State is persisted per key per window



Austin, 2022

# Technical consideration on Dataflow



18

- For Batch jobs, state is stored in worker memory.
- For Streaming jobs using Streaming Engine its stored in Streaming Engine.
- For Streaming jobs without Streaming Engine, all the state is stored on the worker's disk locally. If the state size exceeds the disk capacity you may encounter a "No space left on device error"



# External Cache

Memorystore



# External Cache



- Set up an external cache to Dataflow
- Use service like Memorystore : Managed Redis and Memcached version



# Acknowledgement



- Prathap Kumar Parvathareddy, Cloud Data Engineer, Google PSO



# Questions?

