

# **Dataflow Templates**

Akvelon

https://akvelon.com/





## Agenda

**Dataflow Templates** 

Use Case - Data Protection Using Tokenization

Solution Architecture & Technical Highlights

Demo

Contributing to open source <a href="DataflowTemplates">DataflowTemplates</a>



















# Beam Pipeline



## **Dataflow Templates**

Template - a way to package and stage pipelines

Template types: Classic and Flex

Flex template benefits

- Dynamic DAG
- Eliminated need for ValueProvider
- Expanded templates flexibility

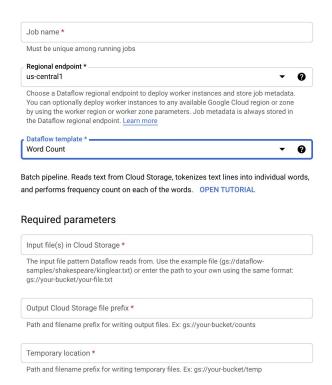


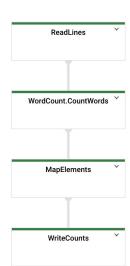
=	Google Cloud Platform	DataTokenization ▼ Q D Ø ♠ : Q
∅	Dataflow	← Create job from template   ⇒ LEARN
≣	Jobs	Job name * Demo
0	Snapshots	Must be unique among running jobs
E É	Notebooks	Regional endpoint * us-central1
		Choose a Dataflow regional endpoint to deploy worker instances and store job metadata. You can optionally deploy worker instances to any available Google Cloud region or zone by using the worker region or worker zone parameters. Job metadata is always stored in the Dataflow regional endpoint. Learn more
		Dataflow template *
		Custom Template   • ②
		gs:// tokenization_test/templates/databkenization_test.json BROWSE  Path to your template file stored in Cloud Storage  Tokenizes structured plain text data from GCS or Pub/Sub input source using an external API calls to Protegrity DSG and ingests date to GCS, Bigtable or BigQuery sink  Required parameters
		GCS location of data schema *
		Path to data schema file located on GCS. BigQuery compatible JSON format data schema required
		GCS file pattern for input files
		GCS file pattern for files in the source bucket
		File format of input files
		JSON, CSV or Avro input file format
		CSV file(s) contain headers: true or false
	Release Notes	'true' if CSV file(s) in the input bucket contain headers, and 'false' otherwise



- Implement pipeline
- Create metadata
- Build template

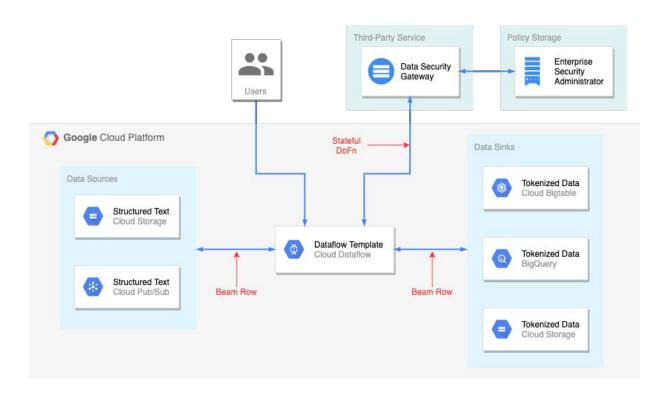






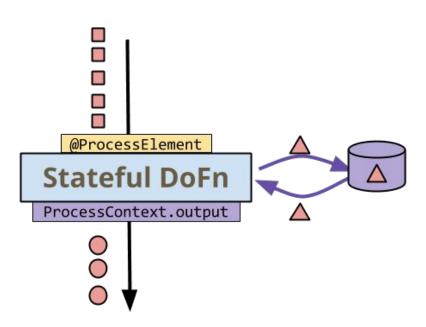


## **Data Protection Using Tokenization**





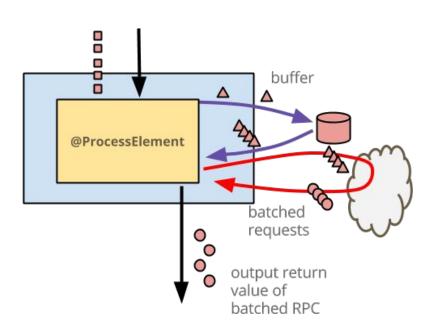
## Stateful Processing



More info <a href="https://beam.apache.org/blog/stateful-processing/">https://beam.apache.org/blog/stateful-processing/</a>



#### Stateful DoFn

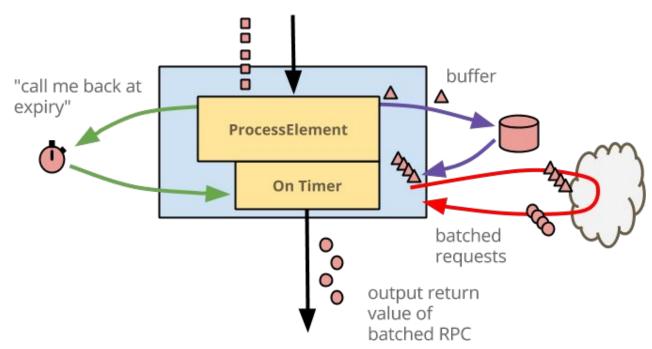


More info <a href="https://beam.apache.org/blog/timely-processing/">https://beam.apache.org/blog/timely-processing/</a>

```
new DoFn<Event, EnrichedEvent>() {
private static final int MAX_BUFFER_SIZE = 500;
@StateId("buffer")
private final StateSpec<BagState<Event>> bufferedEvents = StateSpecs.bag();
@StateId("count")
private final StateSpec<ValueState<Integer>> countState = StateSpecs.value();
@ProcessElement
public void process(
    ProcessContext context,
    @StateId("buffer") BagState<Event> bufferState,
    @StateId("count") ValueState<Integer> countState) {
  int count = firstNonNull(countState.read(), 0);
  count = count + 1;
  countState.write(count);
  bufferState.add(context.element());
  if (count >= MAX_BUFFER_SIZE) {
    for (EnrichedEvent enrichedEvent : enrichEvents(bufferState.read())) {
       context.output(enrichedEvent);
    countState.clear();
```



# Timely Stateful Processing



More info <a href="https://beam.apache.org/blog/timely-processing/">https://beam.apache.org/blog/timely-processing/</a>



## Stateful DoFn with Timer - Implementation

```
new DoFn<Event, EnrichedEvent>() {
  @TimerId("expiry")
 private final TimerSpec expirySpec = TimerSpecs.timer(TimeDomain.EVENT_TIME);
  @ProcessElement
  public void process(
      ProcessContext context,
      BoundedWindow window.
     @StateId("buffer") BagState<Event> bufferState,
     @StateId("count") ValueState<Integer> countState,
     @TimerId("expiry") Timer expiryTimer) {
    expiryTimer.set(window.maxTimestamp().plus(allowedLateness));
  @OnTimer("expiry")
  public void onExpiry(
      OnTimerContext context,
     @StateId("buffer") BagState<Event> bufferState) {
    if (!bufferState.isEmpty().read()) {
      for (EnrichedEvent enrichedEvent : enrichEvents(bufferState.read())) {
        context.output(enrichedEvent);
      bufferState.clear();
```



### GroupIntoBatches

- Groups your data into batches
- Implemented using Stateful DoFn
- Has several optimizations
  - Prefetches data
  - Autosharding in Dataflow

```
import apache beam as beam
with beam.Pipeline() as pipeline:
  batches with keys = (
      pipeline
       'Create produce' >> beam.Create([
         ('spring', '*),
         ('spring', '/"'),
         ('spring', '\'),
         ('spring', ''),
         ('summer', '/"'),
         ('summer', ''),
         ('summer', '%\'),
         ('fall', '/"'),
         ('fall', ''),
         ('winter', '\'),
        'Group into batches' >> beam.GroupIntoBatches(3)
       beam.Map(print))
```

#### Output:

```
('spring', ['d', '/', '\'])
('summer', ['/', 'd', '\'])
('spring', ['d'])
('fall', ['/', 'd'])
('winter', ['\'])
```



## Processing with GroupIntoBatches

```
public PCollectionTuple expand(PCollection<KV<Integer, Row>> inputRows) {
 FailsafeElementCoder<Row, Row> coder =
     FailsafeElementCoder.of(RowCoder.of(schema()), RowCoder.of(schema()));
 Duration maxBuffering = Duration.millis(MAX_BUFFERING);
 PCollectionTuple pCollectionTuple =
      inputRows
          .apply(
              name: "GroupRowsIntoBatches",
             GroupIntoBatches.<Integer, Row>ofSize(batchSize())
                  .withMaxBufferingDuration(maxBuffering))
          apply(
              name: "Tokenize",
             ParDo.of(new TokenizationFn(schema(), rpcURI(), failureTag()))
                  .withOutputTags(successTag(), TupleTagList.of(failureTag())));
 return PCollectionTuple.of(
         successTag(), pCollectionTuple.get(successTag()).setRowSchema(schema()))
      .and(failureTag(), pCollectionTuple.get(failureTag()).setCoder(coder));
```



#### **Use Case Recommendations**

#### Stateful DoFn

Customization of stateful processing

- Deduplication
- Arbitrary-but-consistent indexing

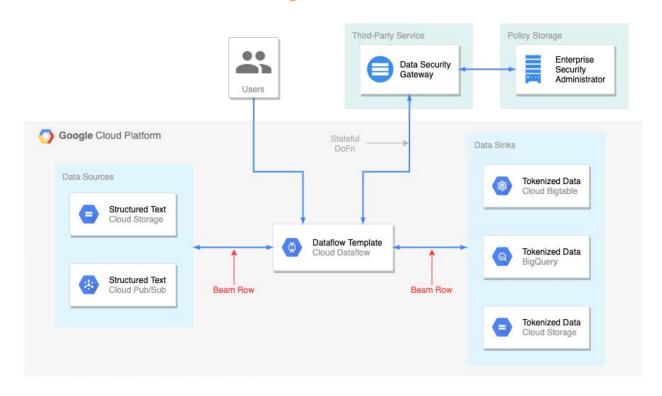
#### **GroupIntoBatches**

Optimized out-of-the-box transform

- External API call
- Data preparation for ML model



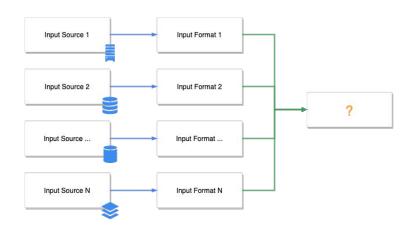
## **Data Protection Using Tokenization**





### How to Represent Data in Pipeline Code?

- Common abstractions for data representation
- Avoid writing transformation again and again
- Common interface to all IO transforms
- Easy and effective to serialize





## Common Approaches

#### Text based formats

- XML
- JSON
- CSV
- YAML

#### Binary based formats

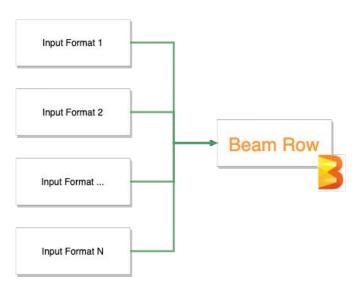
- Apache Avro
- Protobuf
- MessagePack





#### Beam Row!

- A schema which supports primitives
- Ordering same with schema
- Quite effective serialization using RowCoder
- Generic for json specified schema
- Available for Beam SQL





### How We Use Beam Row?





#### Demo

- Template overview
- Add a new input source Parquet IO
- Implement Transform Parquet to Row
- Build and Run template



#### Demo







**Template** Overview

Java Template structure

Metadata file

Schemas

**IO** transforms

Add a New Input **Source Parquet IO** 



Read .parquet files

**Implement Transform** Parquet to Row



Transform from Parquet to Beam Row



Work with schemas





Template building



Ways to run template

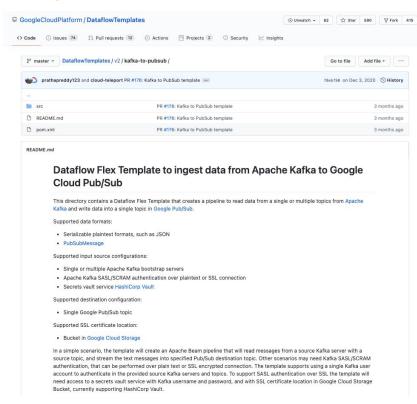


### Contributing to Dataflow Templates

https://github.com/GoogleCloudPlatform/

#### **DataflowTemplates**

- 1. Fork the repository to develop your template
- 2. Follow style guides and best practices
- 3. Sign CLA
- 4. Create a PR
- LGTM code review!





## Summary

- Dataflow Flex templates package pipelines into containers
- Stateful processing in Apache Beam
- BeamRow provides flexible abstraction for data representation
- <u>DataflowTemplates</u> repository Google and community contributed templates and utilities



### Thank You!



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