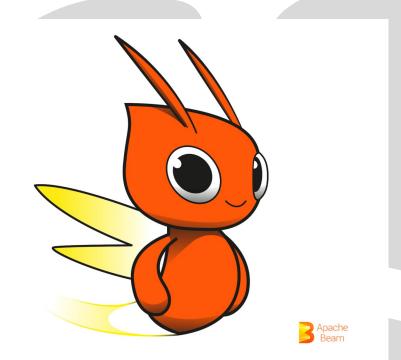
Building an streaming cloud production data pipeline with Apache Beam

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### About me



- Software Engineer and consultant
- VC Investor.
- Co-Organizer of PyCon Colombia Conference https://pycon.co/
- Co-Organizer of Python Medellin / Colombia Community
- Born and base in Colombia, Medellin

## Agenda

- Introduction.
  - Introduction to Beam.
  - Beam model and concepts.
  - Overview Beam io packages.
  - Beam supported languages.
- Architecture
  - A common ETL flow
  - Streaming data pipeline architecture overview.
  - Tips and tricks
  - Debugging streaming pipeline locally (Demo live).
- Deploy
  - Deploying the data pipeline in cloud runner (Demo Live).
- Q/A session.

### Introduction to Apache beam

#### What is Apache Beam?

- An advanced unified programming model.
- Implement batch and streaming data jobs agnostic that run in cloud, locally and many others runners/execution engines.

#### Why Beam?

- Write the code one time in (Java, Python, Go) and run anywhere.
- It's an agnostic technology which means you can run it in cloud runners like Google Dataflow. Or you can run it in your own infrastructure/machine using spark, flink, samza.
- It's open source, portable, flexible and extensible.

## Beam Model and Concepts

#### **PCollections**

The PCollection abstraction represents a potentially distributed, multi-element data set.

For example if you are reading a csv file like this.

```
id,firstname,lastname,email,profession
100,Micheline,Hartnett,Micheline.Hartnett@yopmail.com,police officer
101,Heddie,Georas,Heddie.Georas@yopmail.com,police officer
102,Corene,Genna,Corene.Genna@yopmail.com,developer
103,Chickie,Bivins,Chickie.Bivins@yopmail.com,developer
104,Sharai,Muriel,Sharai.Muriel@yopmail.com,developer
105,Gerianna,Isacco,Gerianna.Isacco@yopmail.com,worker
106,Jolyn,Khorma,Jolyn.Khorma@yopmail.com,worker
107,Vere,Orelee,Vere.Orelee@yopmail.com,police officer
108,Devina,Belanger,Devina.Belanger@yopmail.com,firefighter
109,Florie,Helfand,Florie.Helfand@yopmail.com,firefighter
```

The dataset above is our PCollection.

#### **PTransforms**

Transforms are the operations in your pipeline. You provide processing logic in the form of a function object (colloquially referred to as "user code"), and your user code is applied to each element of an input PCollection

#### For example

```
def get_columns(data):
    id, firstname, lastname, email, profession = data.split(",")

return [{
        "id": id,
        "firstname": firstname,
        "lastname": lastname,
        "email": email,
        "profession": profession
}]
```

The function above is a PTransform which is applied to get the data over each element from our previos PCollection and parse them in a dictionary object.

## Beam io packages

Apache beam support most of the know well industry sources and destinations technologies. Here is some of them

Cloud storage (Google storage GCS, AWS S3, Azure blobstorage...). Queue brokers (Kafka, Google pub/sub, AWS SQS...) Files (Plain text, csv, parquet...)
Databases (Mongo, JDBC, Snowflake, Bigquery, Cassandra...)

You can consult the entire list according the language here.

- https://beam.apache.org/releases/pydoc/2.32.0/apache\_beam.io.html
- https://pkg.go.dev/github.com/apache/beam/sdks/go/pkg/beam?utm\_source
   =godoc
- https://beam.apache.org/releases/javadoc/2.32.0/

## Beam sdk supported languages.



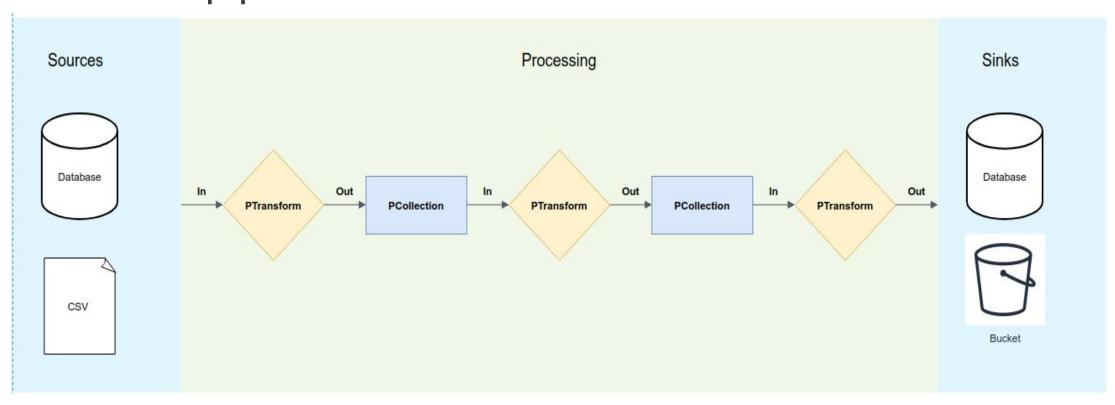






### A common ETL flow

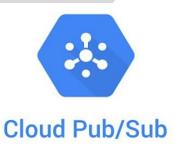
Understanding the things mentions before, most of the time our model pipeline looks like.



#### (Source)

- We need a queue broker like rabbitmq, kafka, AWS SQS or google pub/sub when we are building streaming pipelines.
  - For batch pipelines the source could be a file text, csv or database.
- Streaming common sources









Batch common sources





#### (Processing)

- An ETL model we wrote before needs a runner. The supported runners are Google Dataflow, Flink, Spark, samza and more.
- Most known runners supported by Beam











#### (Sink/Destination)

A common good destination sources are PostgreSQL, S3, cloud storage, MongoDB, Google Bigquery, Amazon RDS.

#### Sinks/Destination







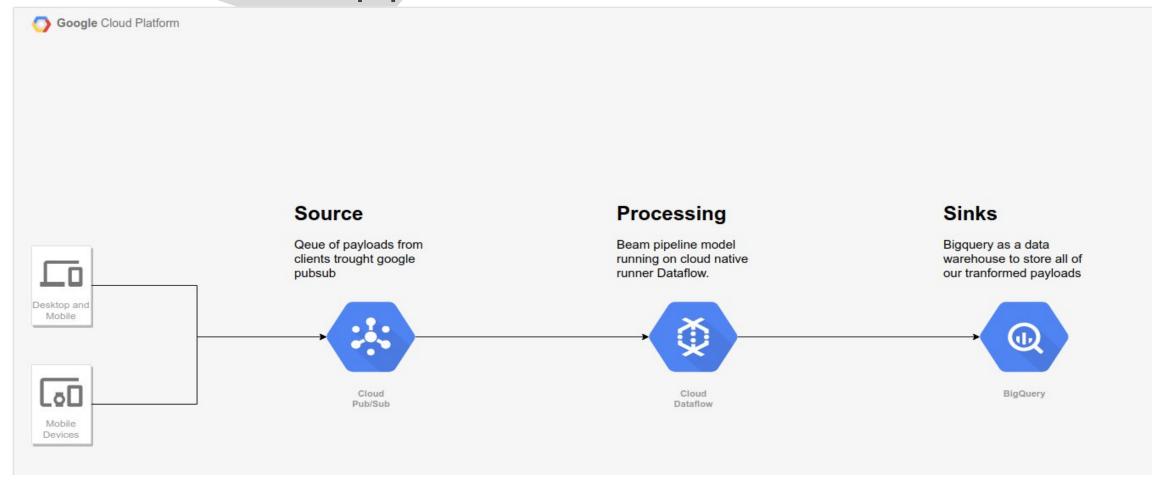






## Architecture for streaming data pipeline.

Our architecture pipeline looks like



## Architecture for streaming data pipeline.

#### Stages:

- We are going to use google pub/sub because is serverless, auto scalable and not ops config needed (Source).
- We are going to use dataflow, because is auto scalable and we don't need a lot of configuration or ops knowledge (Processing).
- We are goin to use Google bigquery beacause again here is serverless, auto scalable, without ops config and cheaper only pays for the query executed (Sink).

## Architecture Performance tips for streaming data pipeline.

 Avoid to make request to another api or external network resource, this can delay or block your pipeline.

```
Events_payload = requests.get('https://api.github.com/events') # tooks 100ms-
200ms for each requests
```

#If we have 1.000.000 of payloads so  $100ms-200ms \times 1.000.000$ , this could be broke the pipeline or have a delay issue

# Architecture Performance tips for streaming data pipeline.

 Avoid to make validations againts database or any other resource outside of the logic pipe.

user = UserModel.find(id=1234)

```
If user.active:
#Do something here
Elif user.name == "Joe":
# Another logic here
# request a database with queries and validate logic that depends of external resources like this is not recommended for streaming pipelines. The database could have a heavy load and the pipeline can be broke or stop due database response.
```

## Architecture for streaming data pipeline.

This is our beam processing model for demo



# Tips and tricks debugging streaming pipeline locally

Apache Beam provide us a DirectRunner which we can use to debug the pipeline in our favorite IDE. (Demo Live)

```
class GetUserAgent(beam.DoFn):
48
49
          def parse user agent(self, user agent):
50
51
                  ua parse = parse(user agent)
52
                  ua properties = str(ua parse).split("/")
53
                           {'name': 'Cody King', 'company': 'Davis-Peters', 'msg': 'Decision show attack w...
54
55
                  browse > special variables
56
                           > function variables
57
                            'name': 'Cody King'
                  return
58
                            'company': 'Davis-Peters'
59
                             'msg': 'Decision show attack want.'
60
                             'remote ip': '220.30.17.100'
61
                             'user agent': 'Mozilla/5.0 (Macintosh; U; PPC Mac OS X 10 6 3 rv:3.0; tcy-IN) Ap
62
              except Exc
                             'date': '2021-10-04'
63
                  return
                           > 'ip info': {'continent': 'Asia', 'country': 'Japan', 'city': 'Tokyo'}
64
65
          def process(se > 'device': {'device': 'PC', 'os': 'Mac OS X', 'browser': 'Safari'}
              user agent
67
              data "devi
68
                           Hold Alt key to switch to editor language hover
              return data
```

# Deploying the data pipeline in cloud runner (Demo Live).

