drivy



How we use and operate Apache Airflow at Drivy

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drivy



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Data @ Drivy

- A 3 nodes Redshift WH (Snowflake soon)
- 6TB of data (including tracking)
- 2M users
- 8 employees (6 analysts + 2 engs.)

We are hiring in the data team. Join us!



Summary

- How do we manage Airflow at Drivy?
- How do we use it?



How do we manage airflow at Drivy?

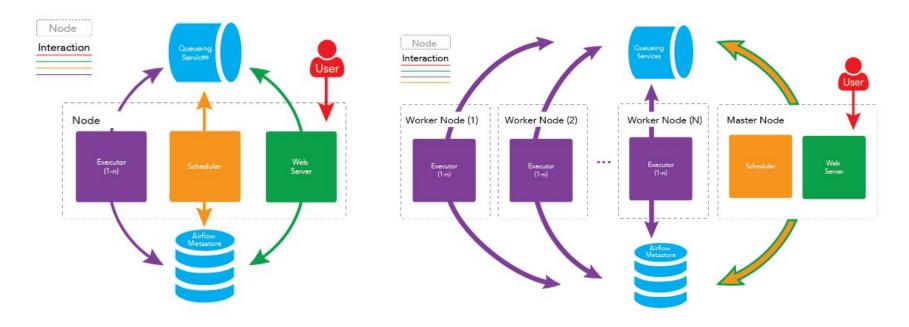
Airflow Recap

"Airflow is a platform to programmatically author, schedule and monitor data pipelines."

Airflow Components

- Metadata Database: stores information regarding tasks state.
- Scheduler: decides which tasks need to be executed.
- Web server: accepts HTTP requests and allows user to interact with Airflow.
- Worker: executes tasks.
- A queuing service: hold information about next task to execute.

Airflow Architecture



Local Executor

Celery Executor

Single-Node vs Multi-Nodes Architecture

Single-Node Architecture

- Easy to set up
- Single point of failure
- Hard to scale

Multi-Node Architecture

- Higher availability
- Dedicated workers for specific tasks
- Scaling horizontally

Airflow 6 months ago

- Single-node architecture
- 1 ec2 instance manually set up
- Crontable deployment tool

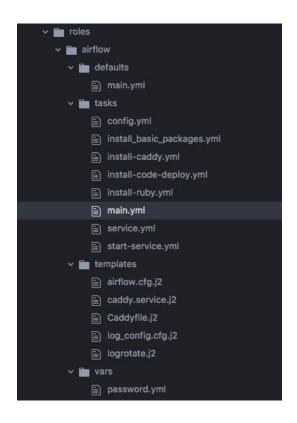


Airflow Architecture now

- Multi-node architecture
- 4 ec2 instances installed through ansible
- Code deploy as a deployment tool



Airflow Installation



```
- name: Install Airflow
  import_tasks: install_basic_packages.yml
- name: Set up caddy
  import_tasks: install-caddy.yml
- name: Install ruby on workers
  import_tasks: install-ruby.yml
 when: inventory_hostname in groups['airflow-workers']
- name: Configure Airflow
  import_tasks: config.yml
- name: Setting up Systemd
  import tasks: service.yml
- name: Start services
 import_tasks: start-service.yml
- name: Deploy code-deploy agents
 import_tasks: install-code-deploy.yml
```

Airflow Installation

```
- name: Download airflow zip
  get url:
    url: "https://codeload.github.com/apache/incubator-airflow/zip/{{ airflow version }}"
    dest: /tmp/airflow.zip
- name : Create Airflow tmp directory
  file:
    path: /tmp/airflow
    state: directory
    mode: 0755
- name: Extract airflow.zip into /tmp/airflow
  unarchive:
    remote src: yes
                                               [airflow:children]
    src: /tmp/airflow.zip
                                               airflow-master
    dest: /tmp/airflow
                                               airflow-workers
                                               [airflow-master]
                                               master ansible_host=##.###.### ansible_port=22 ansible_user=ec2-user ansible_ssh_private_key_file=~/.secret_file
                                               [airflow-workers]
```

worker-1 ansible_host=##.###.### ansible_port=22 ansible_user=ec2-user ansible_ssh_private_key_file=~/.secret_file worker-2 ansible_host=##.###.### ansible_port=22 ansible_user=ec2-user ansible_ssh_private_key_file=~/.secret_file worker-3 ansible_host=##.###.### ansible_port=22 ansible_user=ec2-user ansible_ssh_private_key_file=~/.secret_file

Airflow Migration pain points

- Migrate dag by dag to make sure they are celery executor compliant
- Airflow logs on S3
- Two dag folders
- Need to create symlinks for shared files (can't specify severals dag folders)
- Maintain two Airflows



Airflow Monitoring Grafana Airflow DB Telegraf InfluxDB **AWS** API Cloudwatch W1 W2 M

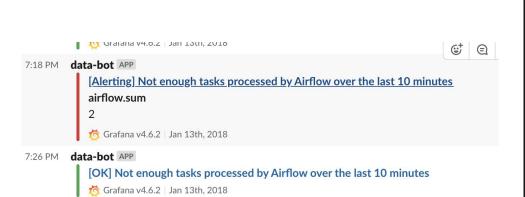
Airflow Monitoring

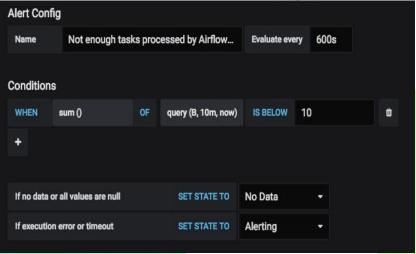
```
3 require_relative '../rb-helpers/mysql'
   require_relative '../rb-helpers/exec'
6 dbs = ['airflow', 'airflow2']
8 dbs.each do |db|
     mysql_client(db.to_sym).query(%Q{
       SELECT state, count(1) count
       FROM task instance
       WHERE end_date BETWEEN (now() - INTERVAL 1 MINUTE) AND now()
       GROUP BY state
     }, symbolize_keys: true).each do |row|
       output(
        prefix: db,
       tags: { task_state: row[:state] },
        values: {
          count: row[:count]
```

```
[[inputs.exec]]
  interval = "1m"
  commands = [
    "bundle exec exec/mysql/airflow_tasks.rb"
  ]
  timeout = "10s"
  data_format = "influx"
```

Airflow Alerting







Airflow tomorrow

- Improve Airflow deployment
- Test managed services (Cloud Composer, Astronomer)
- Create a test environment



Infrastructure



How do we use Airflow at Drivy?

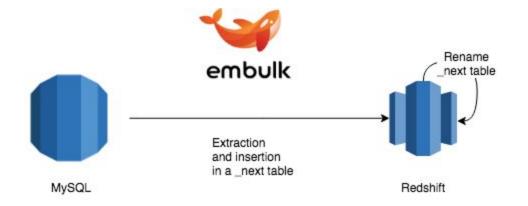
MySQL dump and enrichment

Three kinds of ingestions:

- → <u>Full dump</u>: tables where records can be updated in the past without being timestamped, or having low volumes.
- → Incremental append-only: tables where only new records are added.
- → <u>Incremental UPSERT</u>: tables where new records are added and updated ones are timestamped.

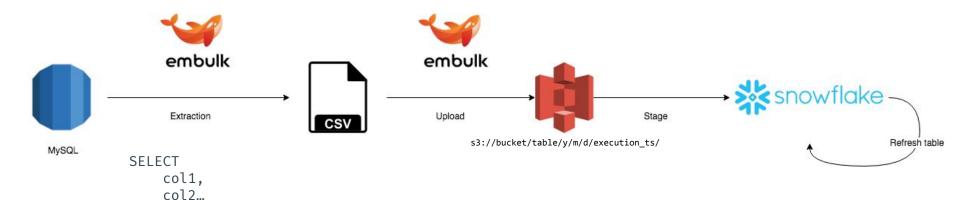
Each of those can be URGENT (every 2h30 to 6h) or NON URGENT (every 6 to 12h)

Full dump imports, the old way



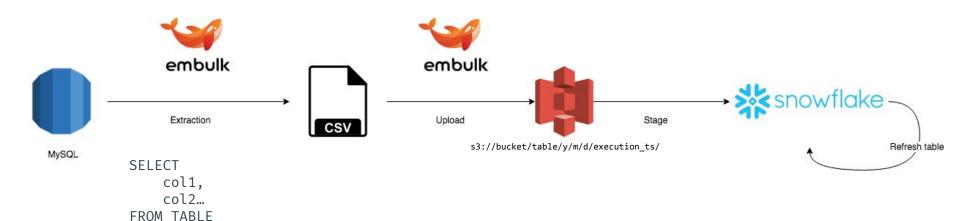
Full dump imports, the new way

FROM TABLE;



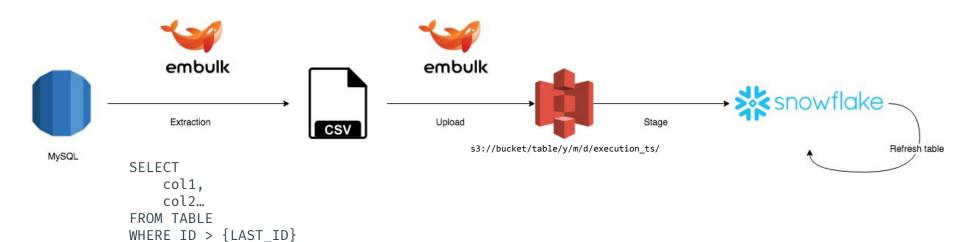
Incremental append

WHERE ID > {LAST_ID};

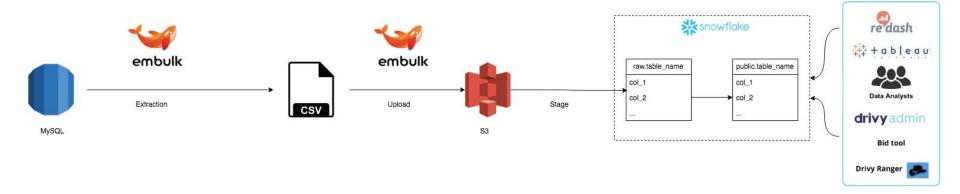


Incremental upsert

OR (UPDATED_AT <> {LAST_UPDATED_AT};



Raw / Transformed isolation



Captur

"Captur is our homemade cross-device tracking framework that mimics Segment architecture"

Captur - Usage

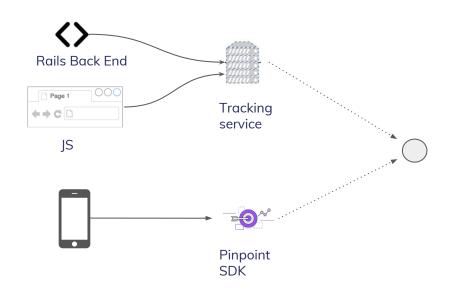
The goal of our backend engineers was to make sending events simple, for every developer.

```
analyticsSendView(Tracking.Agreement.Mobile.Checkin.confirmation, objectId: rentalId) SWITT

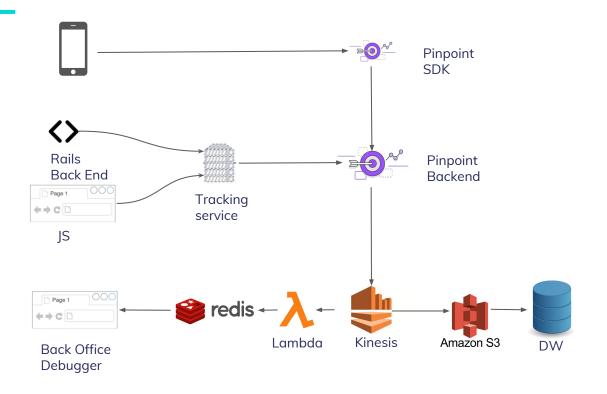
AnalyticsUtils.sendViewWithValues(AnalyticsEvents.DEEPLINK_FACEBOOK, values);
```

Captur - Data collection

```
"type" : "event",
"attributes" : {
 "name" : "car-preview",
 "car_id" : "3082",
 "source" : "car_card".
 "context" : "instant_bookable",
 "title" : "Renault Clio",
 "url" : "https://staging.drivy.com/location-voiture/paris/r
 "path": "/location-voiture/paris/renault-clio-3082",
 "referrer": "https://staging.drivy.com/",
 "user_agent" : "Mozilla/5.0 (Macintosh; Intel Mac OS X 10_1
 "remote_encrypted_ip" : "2a7c3399d5db02c55996365b99fd38d47a
 "browser_width" : "1440",
 "browser_height": "710",
 "browser_language" : "en",
 "browser_timezone_offset" : "+01:00",
 "browser_persistent_storage" : "true",
 "anonymous_id": "59a718e6-6dab-44b7-98d5-27148cc43439".
 "user_id" : "1252",
 "tracking_source": "frontend/beacon"
```

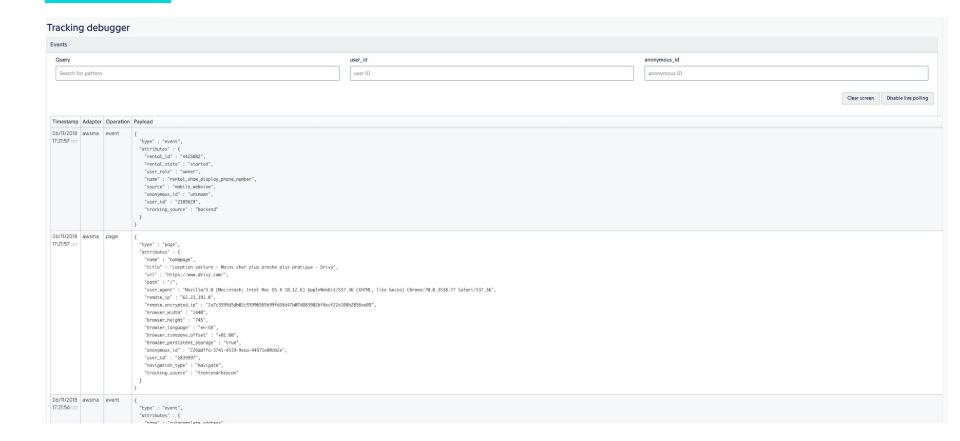


Captur - Data collection

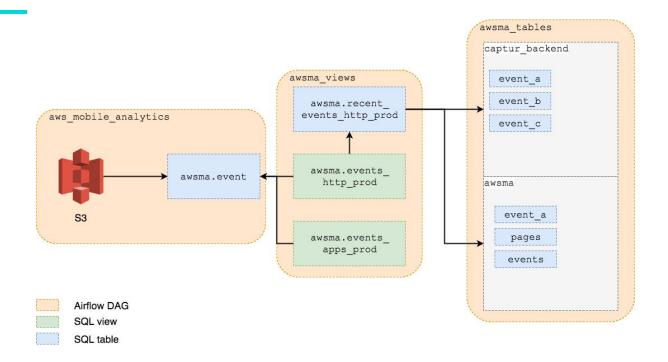


Infrastructure

Captur - Debug



Captur: ETL side



Usage example: Tableau

Web Analytics

Ex: Booking Form Funnel

Usage example: Fraud Prevention



Rules engine that can detect various types of fraud in minutes

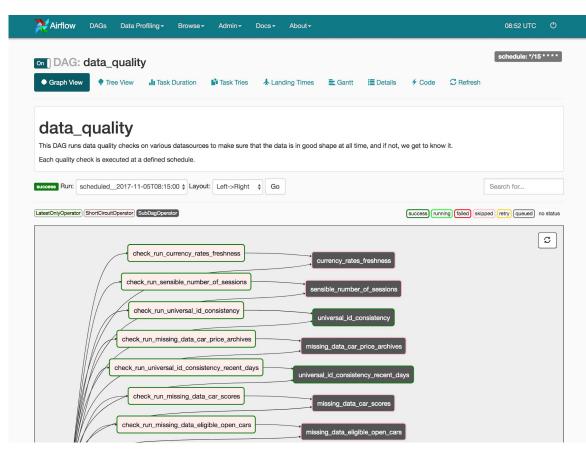
- → >50 alerts in place (theft, insurance fraud, bad usage, phishing)
- → >100 todos / day
- → >40 users blacklisted / day
- → Alerts added instantaneously in prod
- → Escalation process per alerts (slack, pager duty...)
- → Operated by 3 analysts 24/7

Data Quality Checkers

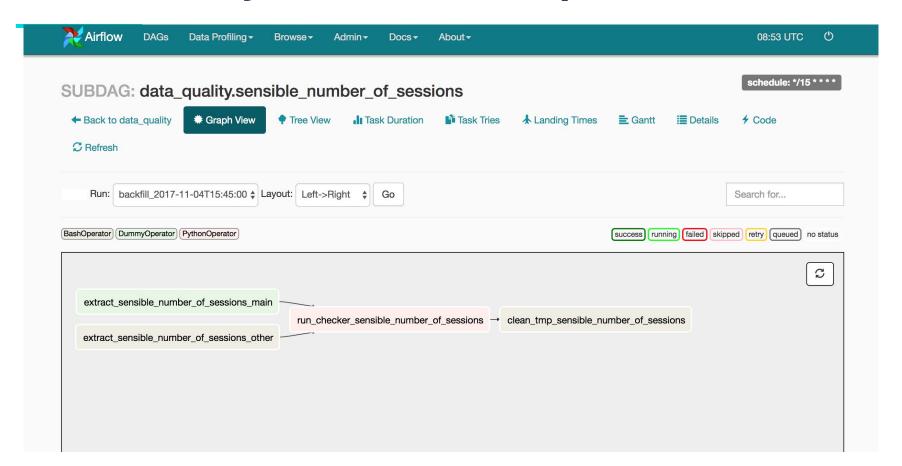
"Data quality checks are performed at a specified interval on one or multiple datasets that are coming from various datasources, using predicates we define. Checks have a tolerance and trigger alerts on alert destinations with an alert level defined by the severity of the found errors."

Data Quality Checkers

- ShortCircuitOperator: determine if the checker should be executed or not.
- **SubDAGOperator**: actually runs the checks.



Data Quality Checker example



Questions?



Thank you!

