Apache Airflow



Agenda (Часть 1)

- 1. Что такое Apache Airflow?
- 2. Что такое Workflow Manager (orchestrator) и что такое ETL
- 3. Серверные компоненты и базовая установка, cli
- 4. DAG, basic DAG params & DAGFile, Tasks
- 5. Экземпляр запуска и задачи DAG
- 6. Operators, Sensors
- 7. Schedule interval & catch up & execution date
- 8. Junja2 Шаблоны,
- 9. Task Statuses

Agenda (Часть 2)

- 1. Macros, User Defined Marcos, Xcom
- 2. SLAs, Alerts, Retries
- 3. BranchOperator, TriggerRules
- 4. Hooks, Connections
- 5. Executors
- 6. Configuration (let's add Celery Executor & PostgreSQL)
- 7. Workers & Flower
- 8. Variables, Run DAG with Params
- 9. Customization: UI plugins
- 10. Airflow in clouds: Google Compose (Airflow in GCP), Astronomer.io

Что такое Apache Airflow?

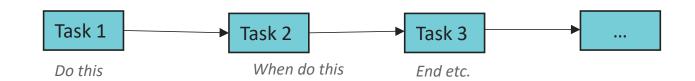


Что такое Apache Airflow?



Это менеджер рабочего процесса! (или оркестратор)

Что такое рабочий процесс, конвейер или DAG (направленный ациклический граф)



workflow or pipeline or DAG (Direct Acyclic Graph)

Alternatives (Orchestrators)











For DS & ML





Similar things for Ops/DevOps/non data workflows





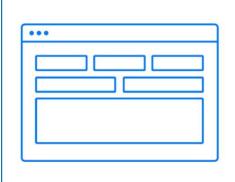
and etc.









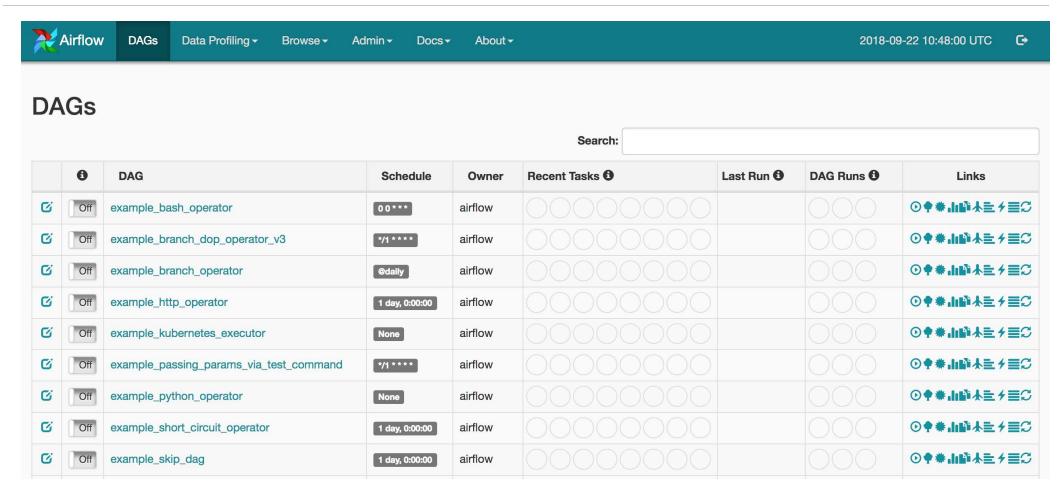


Useful UI

Monitor, schedule and manage your workflows via a robust and modern web application. No need to learn old, cron-like interfaces. You always have full insight into the status and logs of completed and ongoing tasks.



UI Screens – DAGs List



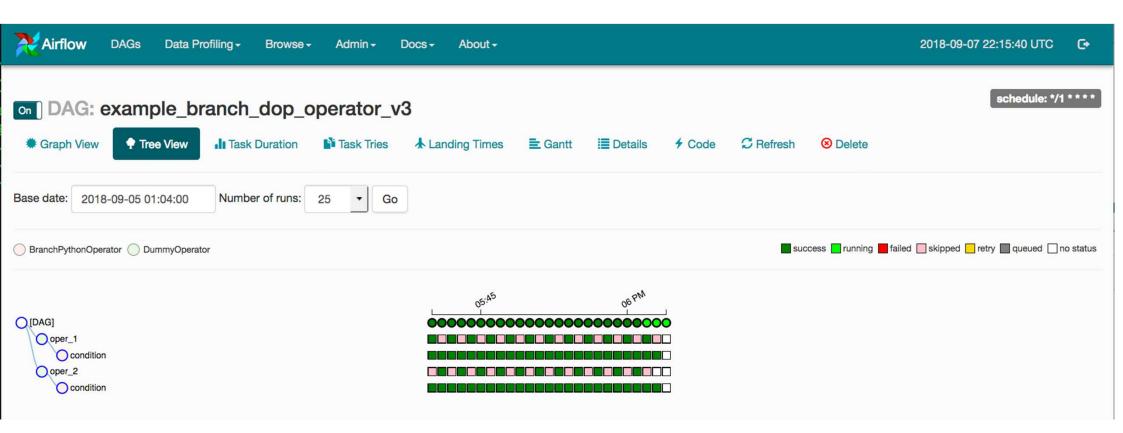


UI Screens – DAGs View

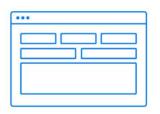




UI Screens – DAGs Runs, Tree View

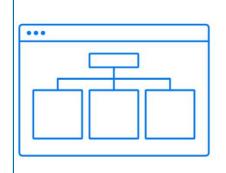






Useful UI

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Robust Integrations

Airflow provides many plug-and-play operators that are ready to execute your tasks on Google Cloud Platform, Amazon Web Services, Microsoft Azure and many other third-party services. This makes Airflow easy to apply to current infrastructure and extend to next-gen technologies.



Integrations from the box (Operators, Sensors, Connectors & Hooks)













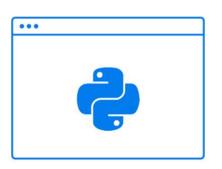








https://airflow.apache.org/docs/apache-airflow/stable/api/airflow/operators/index.html https://airflow.apache.org/docs/apache-airflow/stable/api/airflow/contrib/operators/



Pure Python

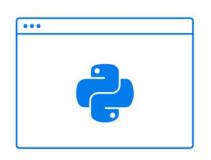
No more command-line or XML black-magic! Use standard Python features to create your workflows, including date time formats for scheduling and loops to dynamically generate tasks. This allows you to maintain full flexibility when building your workflows.



DAG Code Example

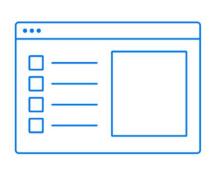
```
import uuid
from datetime import datetime
from airflow import DAG
from airflow.utils.trigger rule import TriggerRule
from airflow.operators.postgres operator import PostgresOperator
dag params = {
    'dag id': 'PostgresOperator dag',
    'start date': datetime(2019, 10, 7),
    'schedule interval': None
with DAG(**dag params) as dag:
    create table = PostgresOperator(
        task id='create table',
        sql='''CREATE TABLE new table(
            custom id integer NOT NULL, timestamp TIMESTAMP NOT NULL, user id VARCHAR (50) NOT NULL
    insert row = PostgresOperator(
       task_id='insert_row',
        sql='INSERT INTO new_table VALUES(%s, %s, %s)',
        trigger_rule=TriggerRule.ALL_DONE,
        parameters=(uuid.uuid4().int % 123456789, datetime.now(), uuid.uuid4().hex[:10])
    create_table >> insert_row
```





Pure Python

No more command-line or XML black-magic! Use standard Python features to create your workflows, including date time formats for scheduling and loops to dynamically generate tasks. This allows you to maintain full flexibility when building your workflows.



Easy to Use

Anyone with Python knowledge can deploy a workflow. Apache Airflow does not limit the scope of your pipelines; you can use it to build ML models, transfer data, manage your infrastructure, and more.

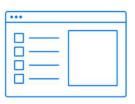


Main Features



Pure Python

No more command-line or XML black-magic! Use standard Python features to create your workflows, including date time formats for scheduling and loops to dynamically generate tasks. This allows you to maintain full flexibility when building your workflows.



Easy to Use

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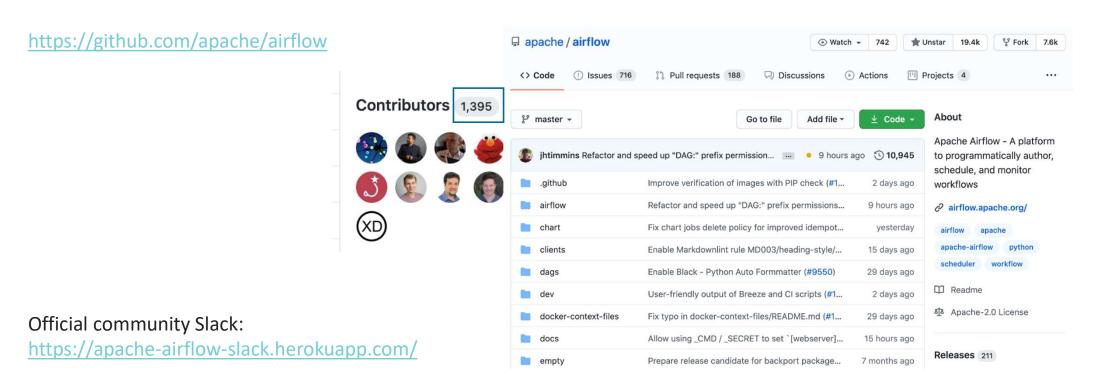


Open Source

Wherever you want to share your improvement you can do this by opening a PR. It's simple as that, no barriers, no prolonged procedures. Airflow has many active users who willingly share their experiences. Have any questions? Check out our buzzing slack.



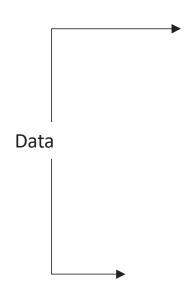
Apache Airflow Community

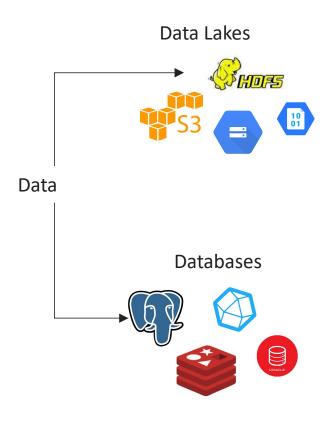


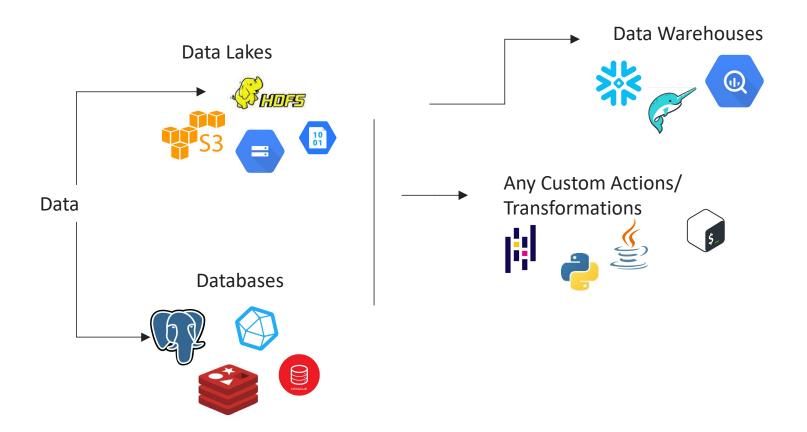
List of committers (maintainers):

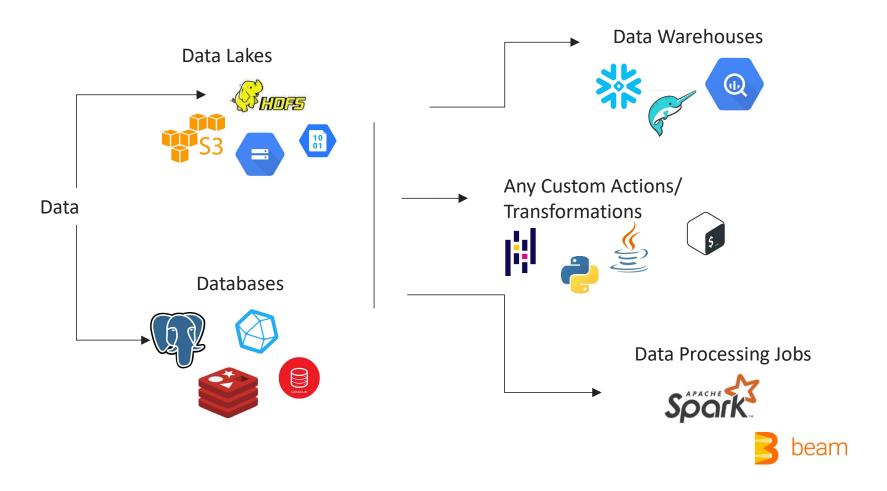
https://people.apache.org/committers-by-project.html#airflow (about 40 people)

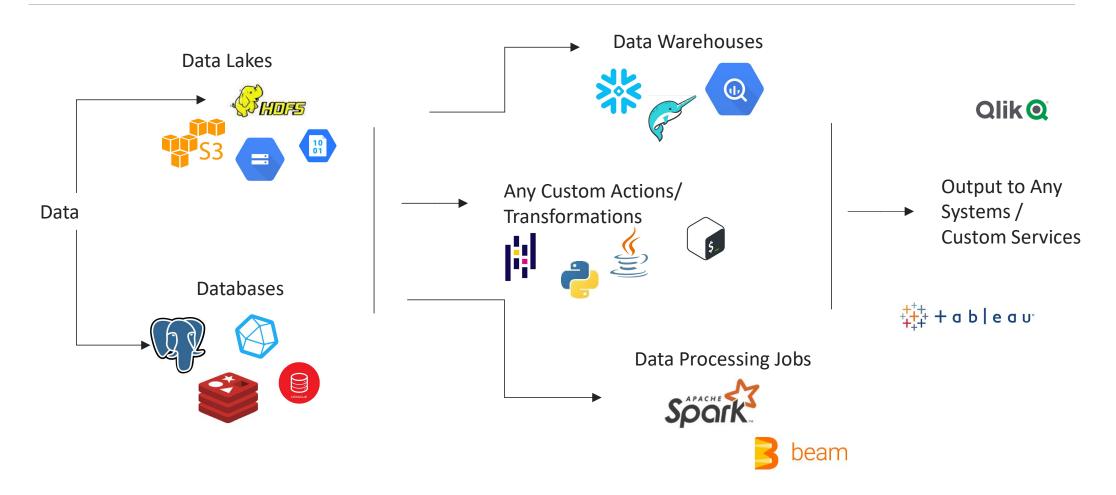
Что такое менеджер рабочего процесса? (или оркестратор)











Оркестратор или менеджер рабочих процессов Позволяет создавать конвейеры данных & describe all steps of your Data Flow:

откуда куда, что, когда и как— многозадачность в любой последовательности (не только классическая ETL)

Extract, transform, load

Extract

Data Lakes









Databases









Transform

Any Custom Actions/ Transformations









Data Processing Jobs





Load

Data Warehouses



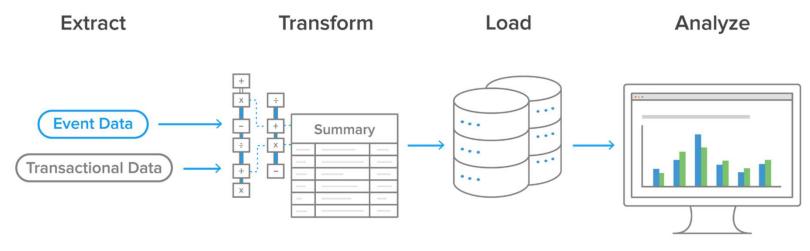
Output to Any Systems / Custom Services





Возможности рабочего процесса, которые нам нужны

- 1. Monitoring Dashboard (что происходит с нашим конвейером?)
- 2. Alerts (если что-то не так Я должен знать об этом быстро)
- 3. SLAs (если у нас нет данных за день у нас проблемы?))
- 4. Way to make customization



Staging Area / OLAP Summaries

Pipeline Example in Words

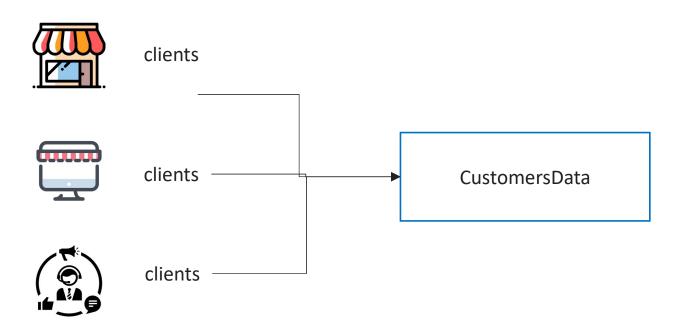
Мы работаем в команде Data Engineering в магазине канцелярских товаров (ручки, бумага и т. д.). У нас около 1500 офлайн-магазинов, интернет-магазин и прямые продажи.

Мы работаем над конвейером данных, который предполагает получение информации о наших клиентах из разных источников.

Pipeline Example in Words

Мы работаем в команде Data Engineering в магазине канцелярских товаров (ручки, бумага и т. д.). У нас около 1500 офлайн-магазинов, интернет-магазин и прямые продажи.

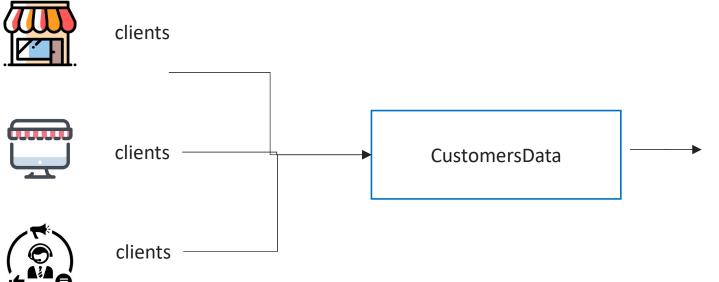
Мы работаем над конвейером данных, который предполагает получение информации о наших клиентах из разных источников.



Pipeline Example in Words

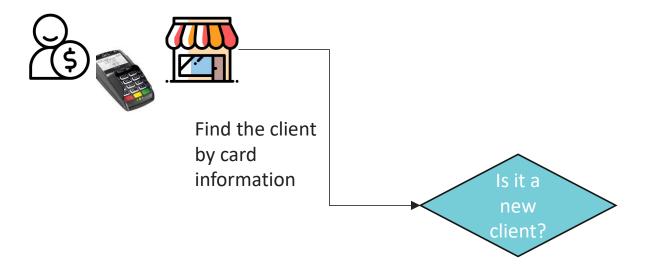
Мы работаем в команде Data Engineering в магазине канцелярских товаров (ручки, бумага и т. д.). У нас около 1500 офлайн-магазинов, интернет-магазин и прямые продажи.

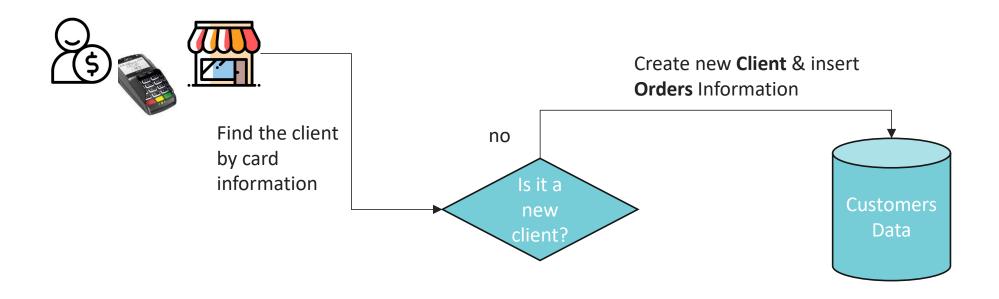
Мы работаем над конвейером данных, который предполагает получение информации о наших клиентах из разных источников.

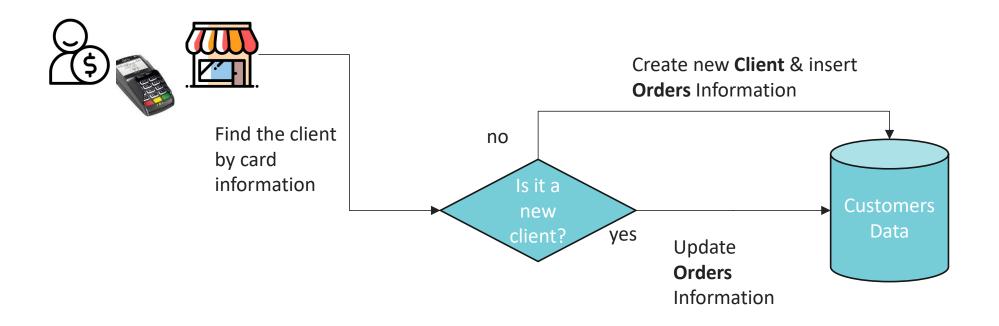


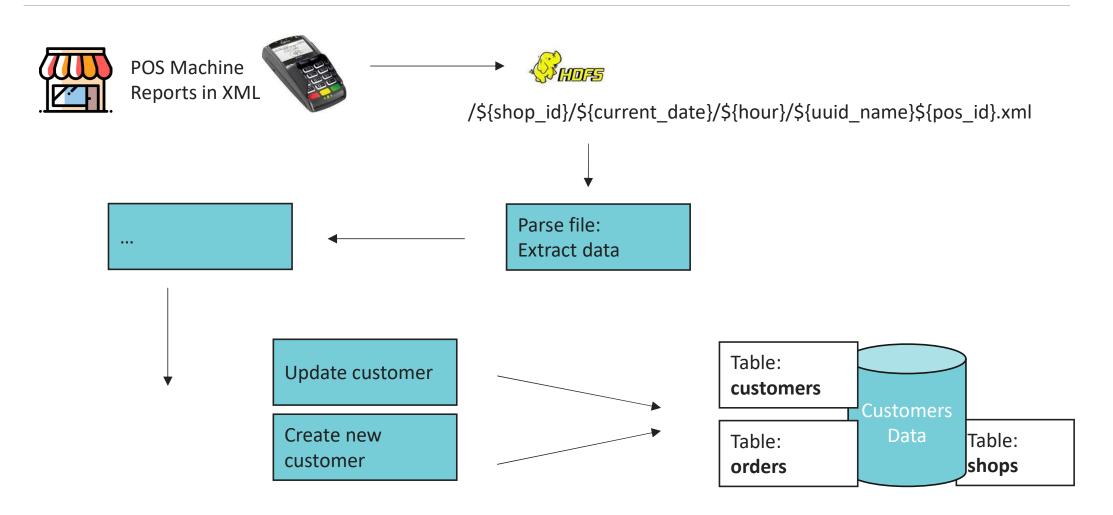
- 1. Это новый клиент?
- 2. Что мы уже знаем об этом клиенте?
- 3. Попробуйте сопоставить клиента по некоторым «критериям» на основе уже существующей информации.
- 4. Обновление данных (данные изменены на временной шкале заказы, маркетинговая деятельность...)



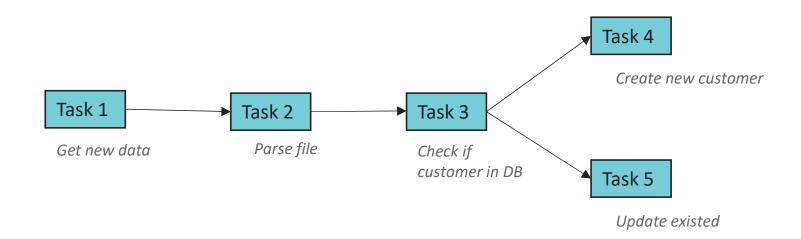








Abstract visualization



Мы только что описали рабочий процесс или конвейер...

Or DAG – main concept of Apache Airflow

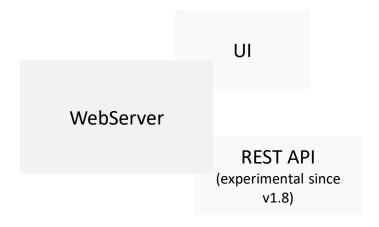


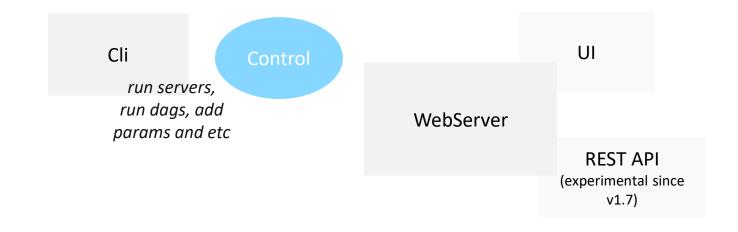
Some key characteristic of Pipelines

- 1. Расписание: они выполняются вовремя с разным расписанием, продолжительностью и т. д.
- 2. Триггеры: конвейеры могут иметь триггеры, вызывающие необходимость запуска конвейера.
- 3. Сбой: конвейеры могут выйти из строя. Нам нужно
 - 1. Как можно скорее узнать об этом
 - 2. запустить с момента сбоя ваша задача должна быть атомарной и простой.
- **4. Повторная обработка:** иногда вам нужно повторно обрабатывать данные за целые длительные периоды в прошлом.

Прежде чем мы создадим первую DAG — запустим Apache Airflow Server

вся информация относительно Apache Airflow актуально для версии 1.10.12 и может отличаться от версии 2.0+



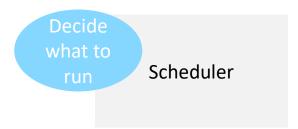


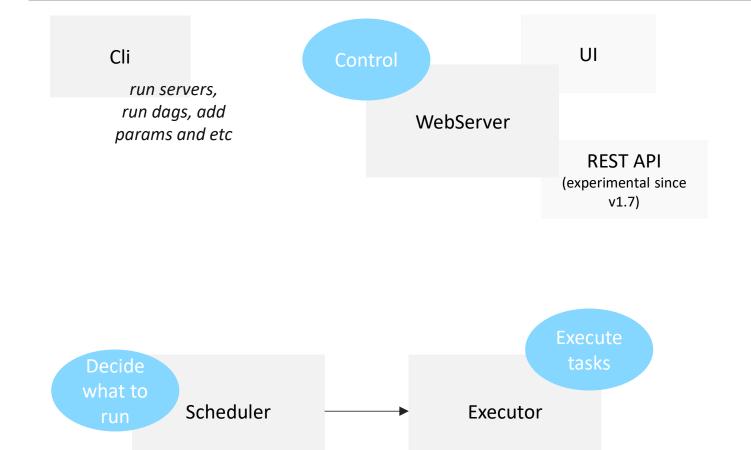
https://airflow.apache.org/docs/apache-airflow/stable/cli-ref

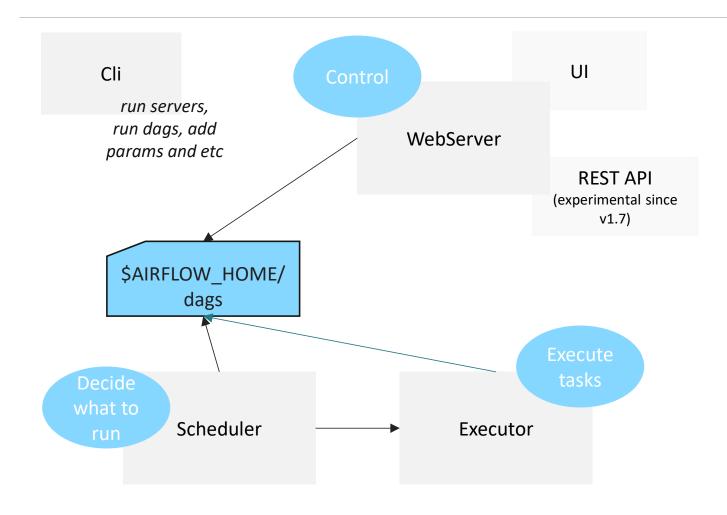
Server commands:

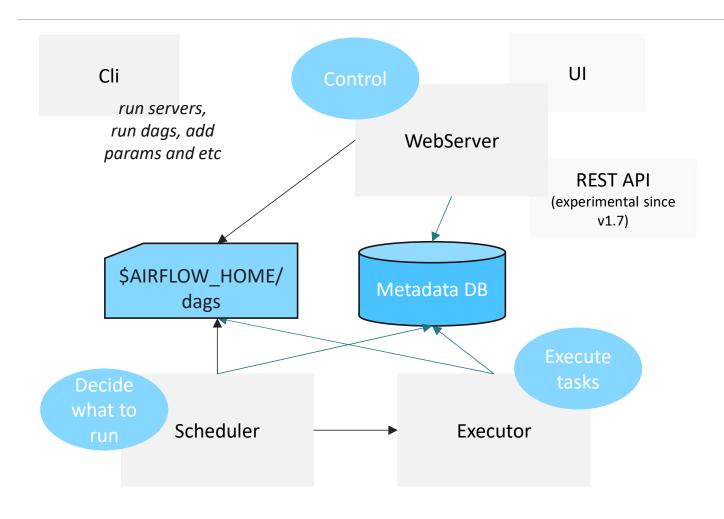
airflow initdb airflow webserver airflow scheduler and etc.



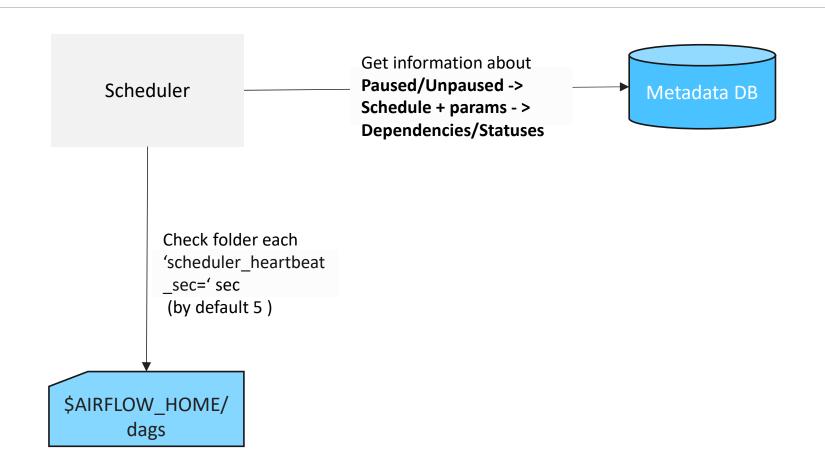




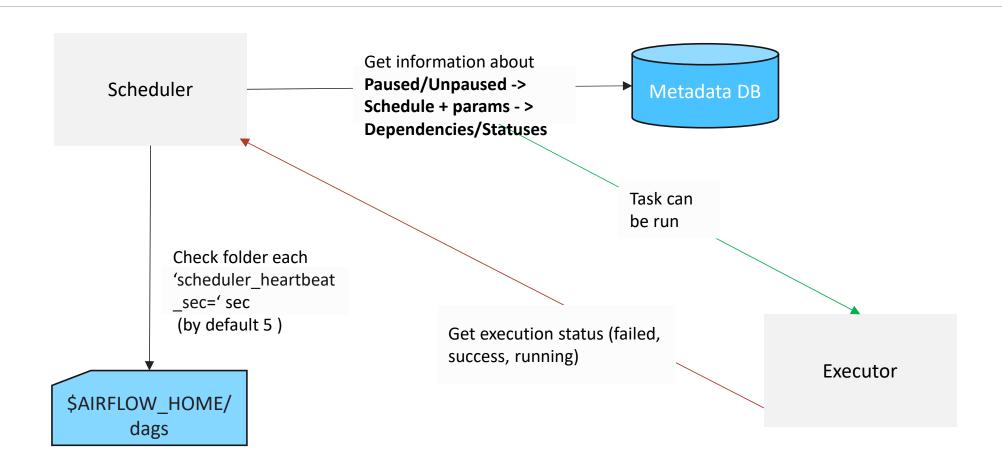


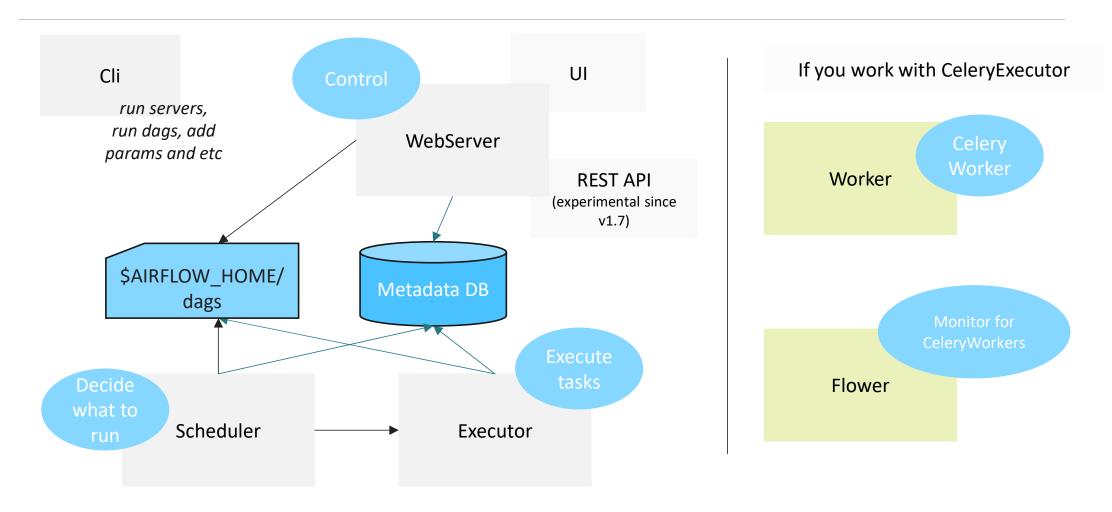


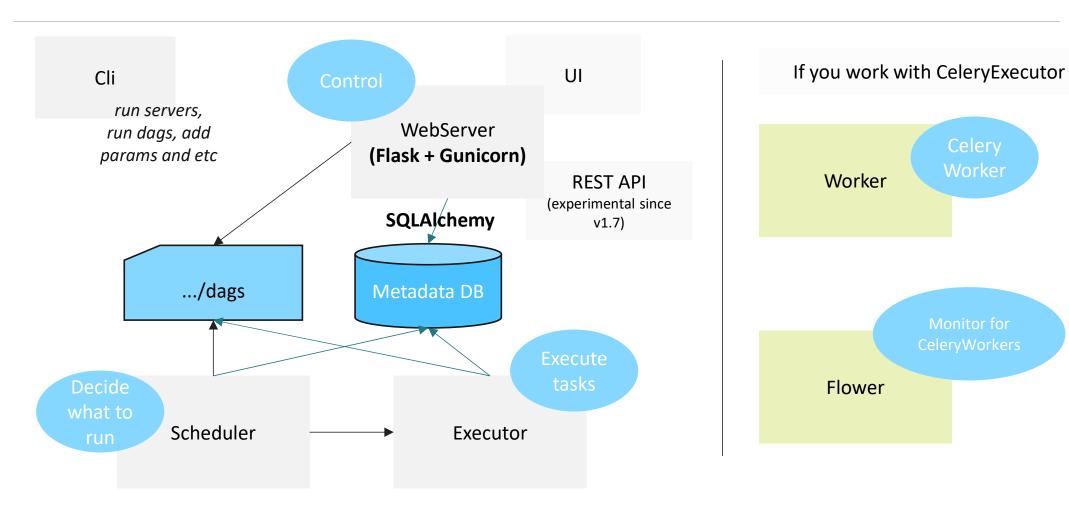
Process of DAG execution



Process of DAG execution







Quick Start

https://airflow.apache.org/docs/stable/start.html#quick-start

install from pypi using pip

pip install apache-airflow

initialize the database (create all needed tables)

airflow initdb

start the web server, default port is 8080 airflow webserver -p 8080 # start the scheduler airflow scheduler # airflow needs a home, ~/airflow is the default, # but you can lay foundation somewhere else if you prefer # (optional)

export AIRFLOW_HOME=~/airflow

Errors

In **November 2020** after install apache-airflow==1.10.12 if you will try to run 'airflow initdb' you will get an error:

from attr import fields, resolve_types
ImportError: cannot import name 'resolve_types' from 'attr'

To solve it you need install cattrs==1.1.0:

\$ pip install cattrs==1.1.0

Remove DAG examples

1. Set in config option "load_examples = False" before airflow initdb

If you already did 'airflow initdb' and want to remove example DAGs

- 1. Set in config option "load_examples = False"
- 2. Run "airflow resetdb"

Airflow by default

executor = SequentialExecutor

sql_alchemy_conn = sqlite:////Users/iuliia_volkova2/airflow/airflow.db - only 1 connection

Extra packages in Installation:

https://airflow.apache.org/docs/apache-airflow/stable/installation.html#extra-packages

Let's define our first DAG

Create a DAGFile in \$AIRFLOW_HOME/dags directory

DAGFile - file with .py that contains words 'airflow' and 'DAG'

If you don't want Apache Airflow to parse your files: add it to .airflowignore in DAGs folder

Let's define our first DAG

```
from datetime import datetime
        from airflow import DAG
        from airflow.operators.dummy_operator import
        DummyOperator
        with DAG(
           dag id="consume new data from pos",
           start_date=datetime(2020, 12, 1),
           schedule interval=None
           ) as dag:
dag_id - unique dag_id (dag name)
start_date - date from that we start process the date
schedule interval – schedule how we plan to run DAG (daily, hourly and etc)
```

Let's define our first DAG

```
from datetime import datetime
from airflow import DAG
from airflow.operators.dummy_operator import
DummyOperator

with DAG(
    dag_id="consume_new_data_from_pos",
    start_date=datetime(2020, 12, 1),
    schedule_interval=None
    ) as dag:
```

dag_id - unique dag_id (dag name)
start_date - date from that we start process the date
schedule_interval - schedule how we plan to run DAG (daily, hourly and etc)

Add tasks to the DAG

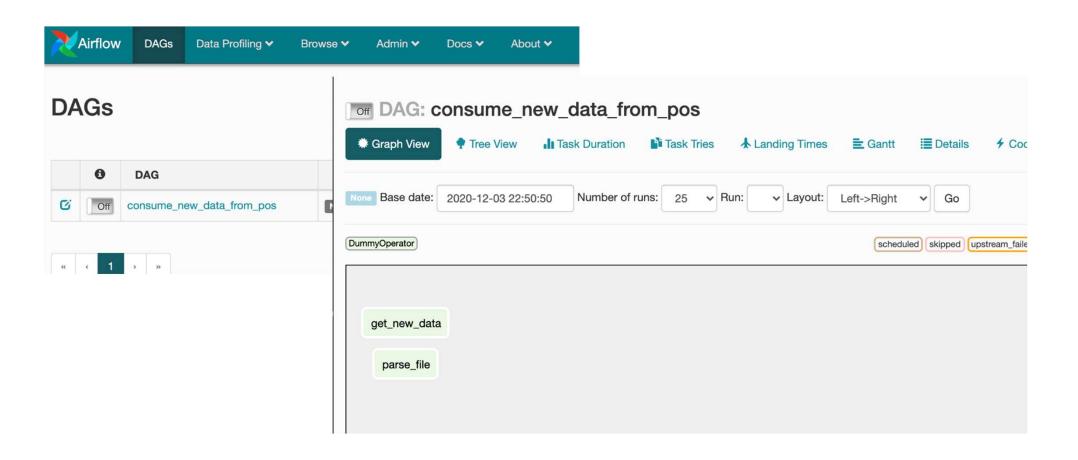
```
from datetime import datetime
from airflow import DAG
from airflow.operators.dummy_operator import DummyOperator

with DAG(
    dag_id="consume_new_data_from_pos",
    start_date=datetime(2020, 12, 1),
    schedule_interval=None
    ) as dag:

    get_new_data = DummyOperator(task_id="get_new_data")
    parse_file = DummyOperator(task_id="parse_file")
```

task_id - unique task_id, mandatory to all Operators
DummyOperator - operator that does nothing (useful to prototype pipeline)

Let's check the UI

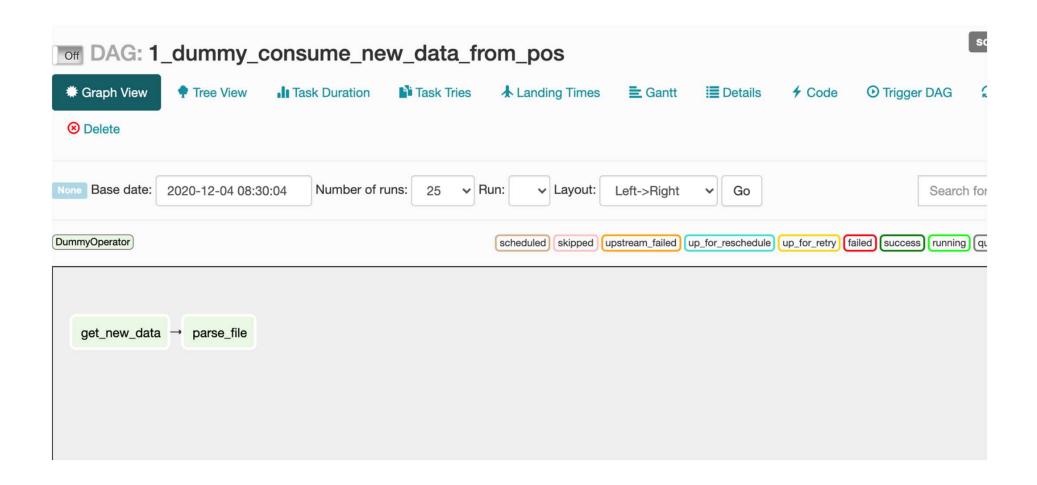


Define a sequence of tasks

```
from datetime import datetime
from airflow import DAG
from airflow.operators.dummy_operator import DummyOperator
with DAG(
  dag_id="consume_new_data_from_pos",
  start_date=datetime(2020, 12, 1),
  schedule_interval=None
  ) as dag:
  get_new_data = DummyOperator(task_id="get_new_data")
  parse file = DummyOperator(task id="parse file")
  get_new_data >> parse_file
```

set_downstream	>>
set_upstream	<<

Define a sequence of tasks



Define a sequence of tasks

```
[task1, task2, task3] >> task4 - allowed
task4 >> [task1, task2, task3] - allowed
task5 >> [task1, task2, task3]

[task1, task2, task3] >> [task4, task5] - not allowed

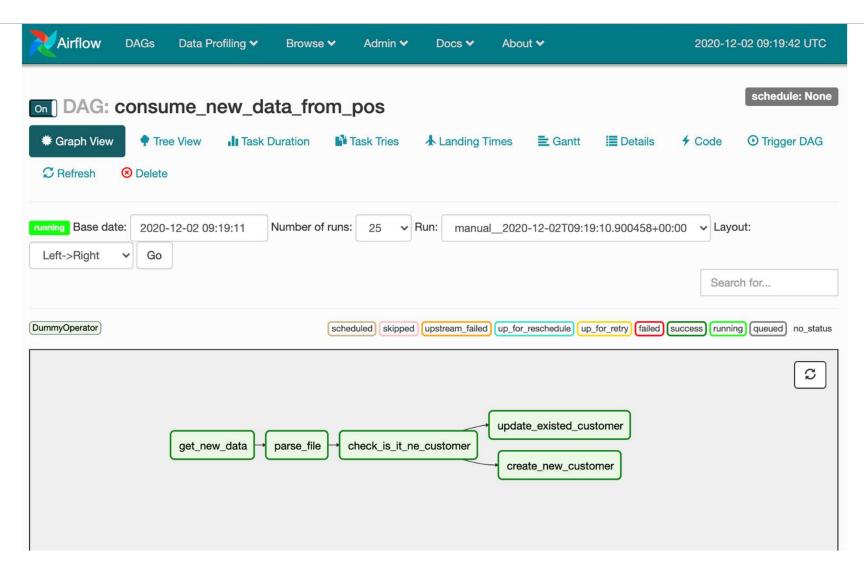
[task4, task5] >> [task1, task2, task3] - not allowed

unsupported operand type(s) for >>: 'list' and 'list'
```

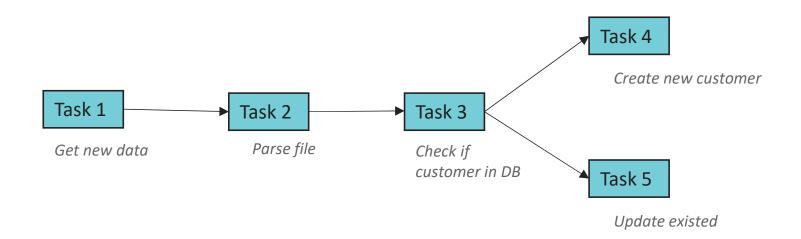
Let's define the full DAG

```
from datetime import datetime
from airflow import DAG
from airflow.operators.dummy operator import DummyOperator
with DAG(
 dag id="consume new data from pos",
 start date=datetime(2020, 12, 1),
  schedule interval=None
  ) as dag:
 get new_data = DummyOperator(task_id="get_new_data")
 parse file = DummyOperator(task id="parse file")
 check is it ne customer =
DummyOperator(task_id="check_is_it_ne_customer")
 create new customer = DummyOperator(task id="create new customer")
 update existed customer =
DummyOperator(task id="update existed customer")
 get new data >> parse file >> check is it ne customer >>
[create new customer, update existed customer]
```

Apache Airflow UI



DAG - Directed Acyclic Graph



What can be a Task?

Operators

Просто СДЕЛАЙТЕ прямо сейчас - отчет о завершении

Sensors

Poke (ожидание) условие пока не выполнено

What can be a Task?

Operators

Просто СДЕЛАЙТЕ прямо сейчас - отчет о завершении

Examples:

- FileToGoogleCloudStorageOperator
- MySqlOperator
- AWSAthenaOperator

•••

Sensors

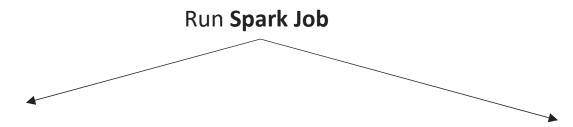
Poke (ожидание) условие пока не выполнено

Examples:

- HdfsSensor
- HttpSensor
- SqlSensor

•••

Moment of Task Completion



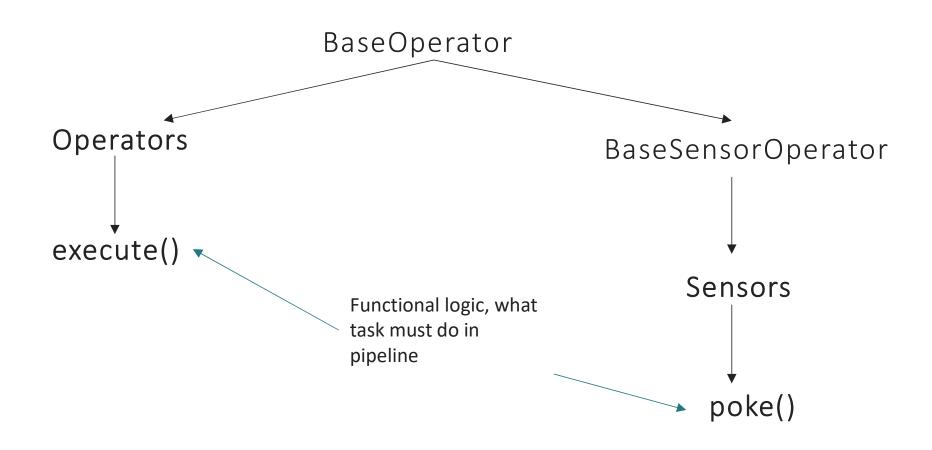
- run as background process
- By ssh in another server
- By REST

Run as a java commandin current server, wait until it finish

(Task is 'success' after send a command to run job)

(Task in 'running' status until complete)

What can be a Task?



Let's define our primitive Operator

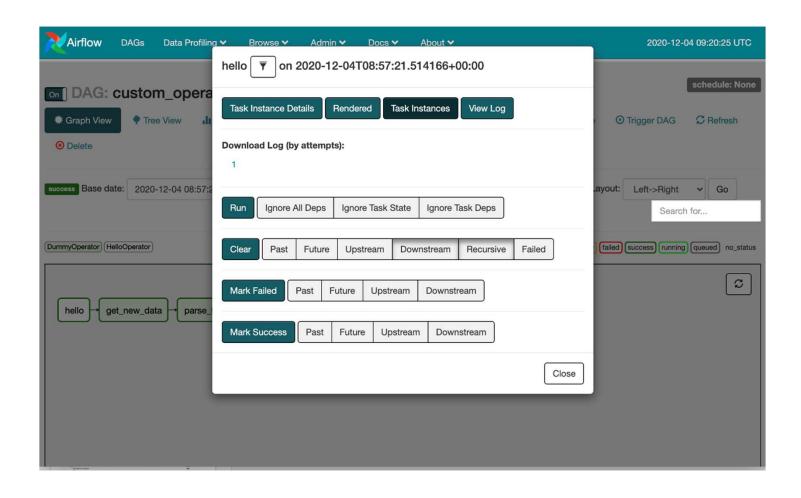
```
from typing import Union, Iterable, Dict
from airflow.models import BaseOperator, SkipMixin

class HelloOperator(BaseOperator, SkipMixin):
    def execute(self, context):
        self.logger.info("Hello, World!")
```

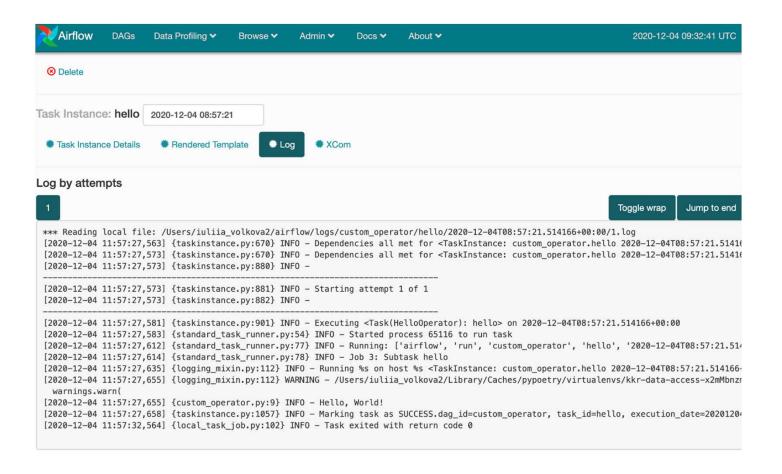
And put module with it to \$AIRFLOW_HOME/dags directory

Airflow add \$AIRFLOW_HOME/dags to PYTHONPATH so everything inside it you can use with import

Check the UI that all works good



Check the UI that all works good



Check the UI that all works good

