# What's new in Airflow 2

Apache Airflow Online Summit 8th of July 2020



### Who are we?



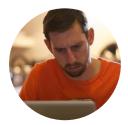
Tomek Urbaszek Committer, PMC Member Software Engineer @ Polidea



Jarek Potiuk
Committer, PMC member
Principal Software Engineer @ Polidea



Kamil Breguła
Committer, PMC member
Software Engineer @ Polidea



Ash Berlin-Taylor
Committer, PMC member
Airflow Engineering Lead @ Astronomer



Daniel Imberman
Committer, PMC Member
Senior Data Engineer @ Astronomer



Kaxil Naik
Committer, PMC member
Senior Data Engineer @ Astronomer

### Announcements

New PMC members



Tomek Urbaszek Committer, PMC Member Software Engineer @ Polidea



Daniel Imberman
Committer, PMC Member
Senior Data Engineer @ Astronomer



Kamil Breguła
Committer, PMC member
Software Engineer @ Polidea

New committer



QP Hou Committer Senior Engineer @ Scribd

Talk: Teaching an old DAG new tricks Friday July 10 th, 5 am UTC

### "Ask Me Anything" session with Airflow PMCs

- Asia friendly time-zone
- Thursday 11 pm PDT / Friday 6 am UTC
- Hosted by Bangalore Meetup
- BYO questions

### High Availability



### Scheduler High Availability

#### Goals:

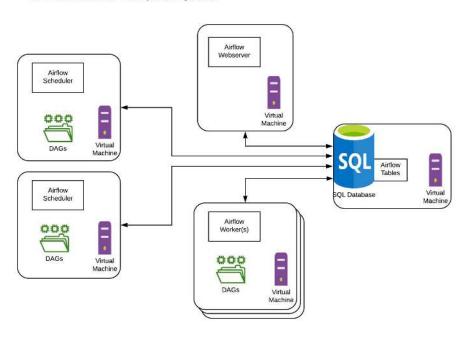
- Performance reduce task-to-task schedule "lag"
- Scalability increase task throughput by horizontal scaling
- Resiliency kill a scheduler and have tasks continue to be scheduled

### Scheduler High Availability: Design

- Active-active model. Each scheduler does everything
- Uses existing database no new components needed, no extra operational burden
- Plan to use row-level-locks in the DB (SELECT ... FOR UPDATE)
- Will re-evaluate if performance/stress testing show the need

### Example HA configuration

Airflow Schedulers running in High Availability on virtual machines - example configuration



### Scheduler High Availability: Tasks

Separate DAG parsing from DAG scheduling

This removes the tie between parsing and scheduling that is still present

Run a mini scheduler in the worker after each task is completed



A.K.A. "fast follow". Look at immediate down stream tasks of what just finished and see what we can schedule

Test it to destruction - In progress

This is a big architectural change, we need to be sure it works well.

### Measuring Performance

Key performance we define as "Scheduler lag":

- Amount of "wasted" time not running tasks
- ti.state\_date max(t.end\_date for t in upstream\_tis)
- Zero is the goal (we'll never get to 0.)
- Tasks are "echo true" -- tiny but still executing

### Preliminary performance results

Case: 100 DAG files | 1 DAG per file | 10 Tasks per DAG | 1 run per DAG

Workers: 4 | Parallelism: 64

**1.10.10:** 54.17s (σ19.38) Total runtime: 22m22s

**HA** branch - 1 scheduler: 4.39s (σ1.40) 1m10s

**HA** branch - 3 schedulers: 1.96s (σ0.51) Total runtime: 48s

### Preliminary performance results

Case: 1 Dag File | 1 Dag Per File | 20 Tasks per DAG | 1000 runs per DAG

Workers: 30 | Parallelism: 40960 | Default pool size 40960

**1.10.10:** 42.14s (σ7.06) Total runtime: 1h 30m 14s

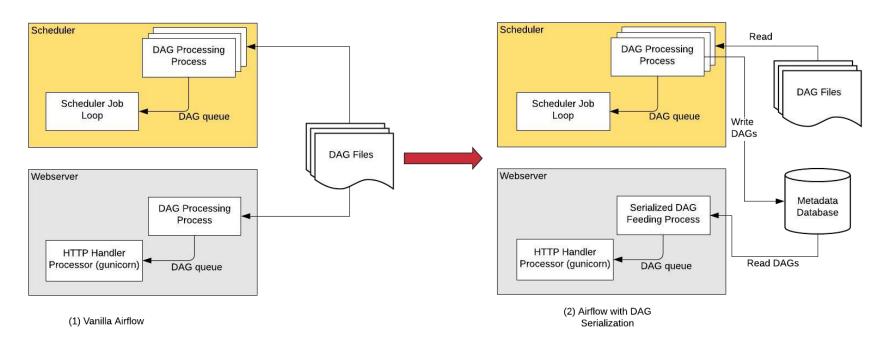
**HA branch - 1 scheduler:** 0.68s (σ0.19) Total runtime: 18m 51s

**HA** branch - 3 schedulers\*: 1.54s (σ1.79) Total runtime: 12m 52s

### **DAG** Serialization



### Dag Serialization



### Dag Serialization (Tasks Completed)

- Stateless Webserver: Scheduler parses the DAG files, serializes them in JSON format & saves them in the Metadata DB.
- Lazy Loading of DAGs: Instead of loading an entire DagBag when the Webserver starts we only load each DAG on demand. This helps reduce Webserver startup time and memory. This reduction in time is notable with large number of DAGs.
- Deploying new DAGs to Airflow no longer requires long restarts of webserver (if DAGs are baked in Docker image)
- Feature to use the "JSON" library of choice for Serialization (default is inbuilt 'json' library)
- Paves way for DAG Versioning & Scheduler HA

### Dag Serialization (Tasks In-Progress for Airflow 2.0)

- Decouple DAG Parsing and Serializing from the scheduling loop.
- Scheduler will fetch DAGs from DB
- DAG will be parsed, serialized and saved to DB by a separate component "Serializer"/ "Dag Parser"
- This should reduce the delay in Scheduling tasks when the number of DAGs are large

### **DAG** Versioning



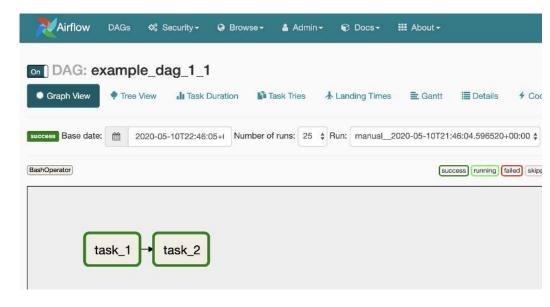
### Dag Versioning

#### **Current Problem:**

- Change in DAG structure affects viewing previous DagRuns too
- Not possible to view the code associated with a specific DagRun
- Checking logs of a deleted task in the UI is not straight-forward

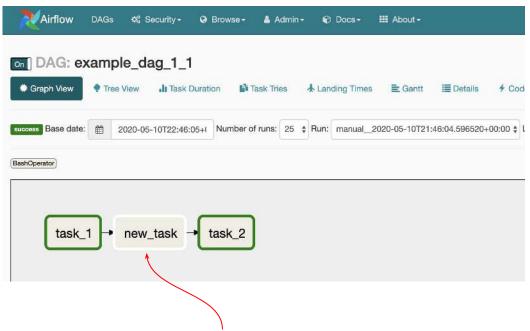
### Dag Versioning (Current Problem)

```
. .
from airflow.models.dag import DAG
from airflow.operators.bash operator import BashOperator
from datetime import datetime
with DAG('example_dag_1_1', schedule_interval=None,
        start_date=datetime(2020, 4, 25)) as example_dag_1_1:
   task 1 = BashOperator(
       task id='task 1'.
       bash command='echo hello',
       task id='task 2',
       bash command='echo hello',
```



### Dag Versioning (Current Problem)

```
. .
from airflow.models.dag import DAG
from airflow.operators.bash_operator import BashOperator
from datetime import datetime
with DAG('example_dag_1_1', schedule_interval=None,
         start date=datetime(2020, 4, 25)) as example dag 1 1:
   task 1 = BashOperator(
       task id='task 1',
       bash command='echo hello',
       task_id='new_task',
       bash_command='echo hello',
   task 2 = BashOperator(
       task id='task 2'.
       bash command='echo hello',
```



New task is shown in Graph View for older DAG Runs too with "no status".

### Dag Versioning

#### **Current Problem:**

- Change in DAG structure affects viewing previous DagRuns too
- Not possible to view the code associated with a specific DagRun
- Checking logs of a deleted task in the UI is not straight-forward

#### Goal:

- Support for storing multiple versions of Serialized DAGs
- Baked-In Maintenance DAGs to cleanup old DagRuns & associated Serialized DAGs
- Graph View shows the DAG associated with that DagRun

### Performance Improvements

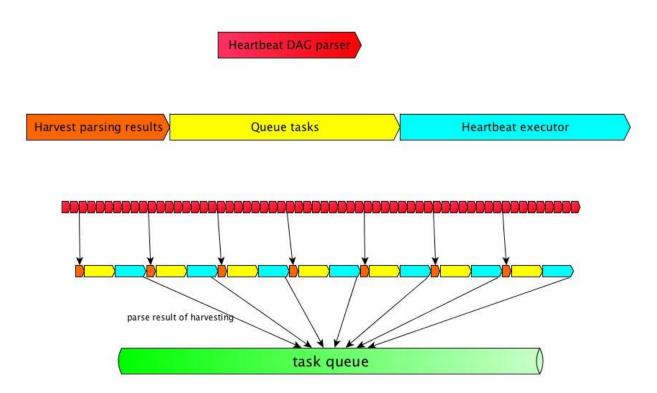


### Components performance improvements

- Focus on the current code
  - Reviews each components in turn
- Tools supporting performance tests perf\_kit

```
@timing()
def test_dag_sync():
    with count_queries():
    DAG.bulk_sync_to_db()
```

### Avoid loading DAGs in the main scheduler loop



### Limit queries count

#### DagFileProcessor:

When we have one DAG file with 200 DAGs, each DAG with 5 tasks:

	Before	After	Diff
Average time:	8080.246 ms	628.801 ms	-7452 ms (92%)
Queries count:	2692	5	-2687 (99%)

#### **Celery Executor:**

When we have one DAG file with 200 DAGs, each DAG with 5 tasks:

	Postgres		Redis	
	Before	After	Before	After
Average time	3.1 s	27.825 ms	778.557 ms	3.417 ms
Queries count	5000	1	5000	1

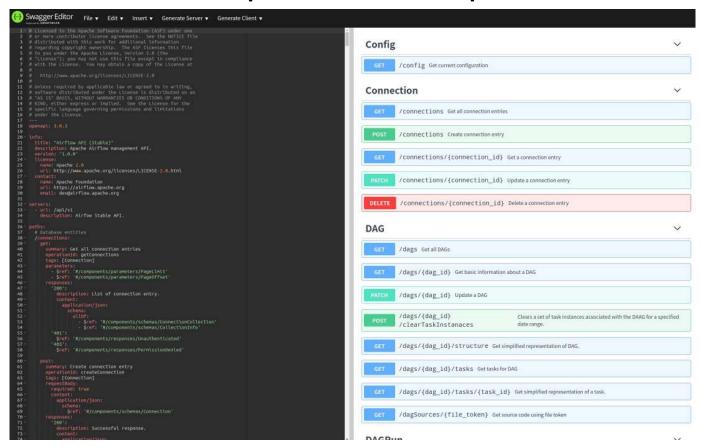
### How to avoid regression?

```
with assert_queries_count(3):
    DAG.bulk_sync_to_db(dags)
```

### **REST API**



### API: follows Open API 3.0 specification



#### Outreachy interns



Ephraim Anierobi



Omair Khan

### API development progress

#### AIP-32 - Airflow REST API

Updated 5 days ago



### Dev/CI environment



### CI environment

- Moved to GitHub Actions
  - Kubernetes Tests
  - Easier way to test Kubernetes Tests locally
- Quarantined tests
  - Fixing the Quarantined tests
- Thinning CI image
  - Moved integrations out of the image
- Future: Automated System Tests (AIP-21)

### Dev environment

- Breeze
  - unit testing
  - o package building 🇸
  - release preparation 🇸
  - kubernetes tests
  - refreshed videos
- Code Spaces / VSCode

```
Usage: breeze [FLAGS] [COMMAND] -- <EXTRA ARGS>
 default the script enters IT environment and drops you to bash shell, but you can choose one
  the commands to run specific actions instead. Add --help after each command to see details:
ommands without arguments:
                                           [Default] Enters interactive shell in the container
 shell
 build-docs
                                           Builds documentation in the container
                                           Builds CI or Production docker image
                                           Cleans up the container image created
                                           Execs into running breeze container in new terminal
                                           Generates pinned requirements for pip dependencies
 push-image
                                           Pushes images to registry
                                           Initializes local virtualenv
 setup-autocomplete
                                           Sets up autocomplete for breeze
                                           Stops the docker-compose environment
                                           Stops the docker-compose environment including DB cleanup
 toggle-suppress-cheatsheet
                                           Toggles on/off cheatsheet
 toggle-suppress-ascliart
                                           Toggles on/off ascilart
ommands with arguments:
                                           Executes specified docker-compose command
 docker-compose
                                           Manages KinD cluster on the host
 prepare-backport-readme
                               <ARG>
                                           Prepares backport packages readme files
                                           Prepares backport packages
 prepare-backport-packages
                                           Performs selected static check for changed files
                               <ARG>
                                           Runs selected tests in the container
Help commands:
                                           Shows all breeze's flags
                                          Shows this help message
 help-all
                                           Shows detailed help for all commands and flags
```

### Backport Packages V

- Bring Airflow 2.0 providers to 1.10.\* 🗸
- Packages per-provider V
- 58 packages (!) V
- Python 3.6+ only(!)
- Automatically tested on CI V
- **Future** 
  - Automated System Tests (AIP-4)
  - Split Airflow (AIP-8)?

Talk: Migration to Airflow backport providers, Anita Fronczak Thursday July 16th, 4 am UTC

* *	Prepare & test backport packages
1880	
1881	Prepared backporting package jdbc
1882	======================================
1883	Preparing backporting package jenkins
1884	
1885	Prepared backporting package jenkins
1886	
1887 1888	Preparing backporting package jira
1889	Prepared backporting package jira
1890	======================================
1891	Preparing backporting package microsoft.azure
1892	
1893	Prepared backporting package microsoft.azure
1894	
1895	Preparing backporting package microsoft.mssql
1896 1897	Prepared backporting package microsoft.mssql
1898	
1899	Preparing backporting package microsoft.winrm
1900	
1901	Prepared backporting package microsoft.winrm
1902	
4000	- B - 100 -

v 🗸	Prepare & test backport packages
	Installing apache-airflow-backport-providers-microsoft-mssql
	Installed apache-airflow-backport-providers-microsoft-mssql
	Uninstalling apache-airflow-backport-providers-microsoft-mssql
	Uninstalled apache-airflow-backport-providers-microsoft-mssql
	Airflow version after installation 1.10.10
	Installing apache-airflow-backport-providers-microsoft-winrm
	Installed apache-airflow-backport-providers-microsoft-winrm
	Uninstalling apache-airflow-backport-providers-microsoft-winrm
	Uninstalled apache-airflow-backport-providers-microsoft-winrm
	Airflow version after installation 1.10.10
2519	

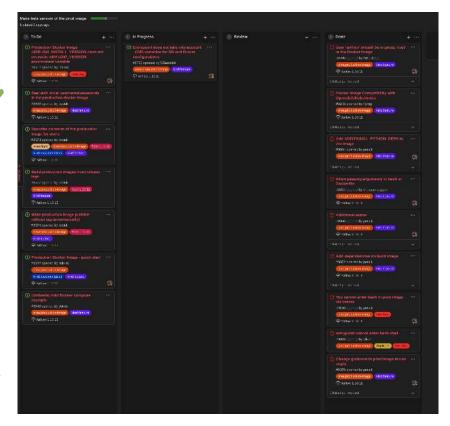
# Support for Production Deployments



### **Production Image**

- Beta quality image is nearly ready
- Started with "bare image"
- Listened to use cases from users
- Integration with Helm Chart
- Implemented feedback
- Docker Compose

Talk, Production Docker image for Apache Airflow Jarek Potiuk, Tuesday July 14th, 5 am UTC

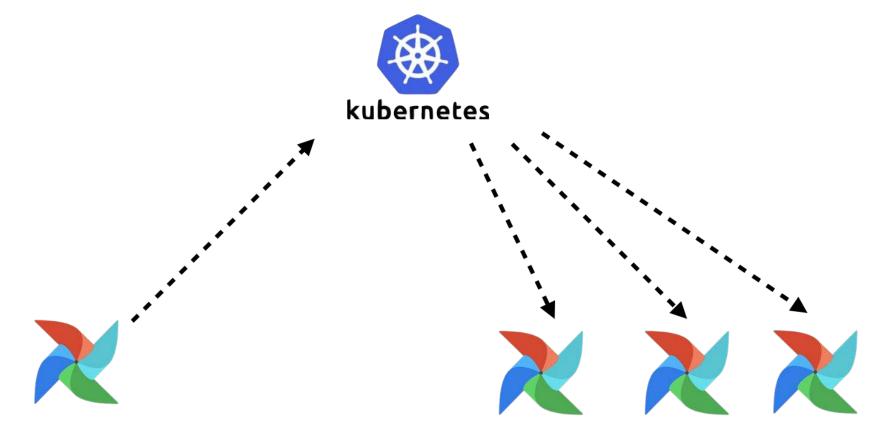


## What's new in Airflow + Kubernetes

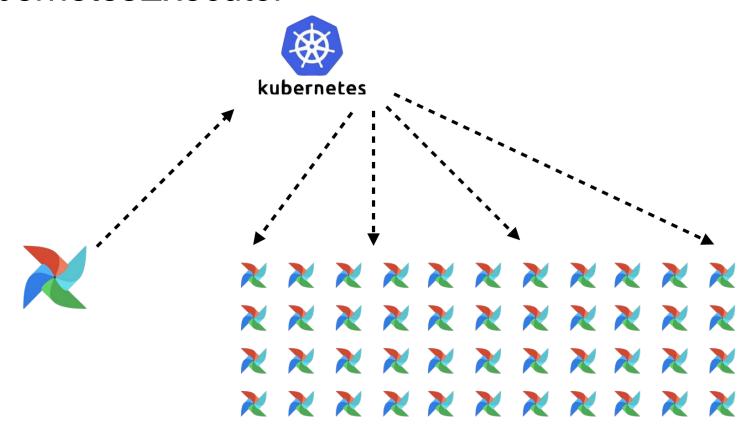




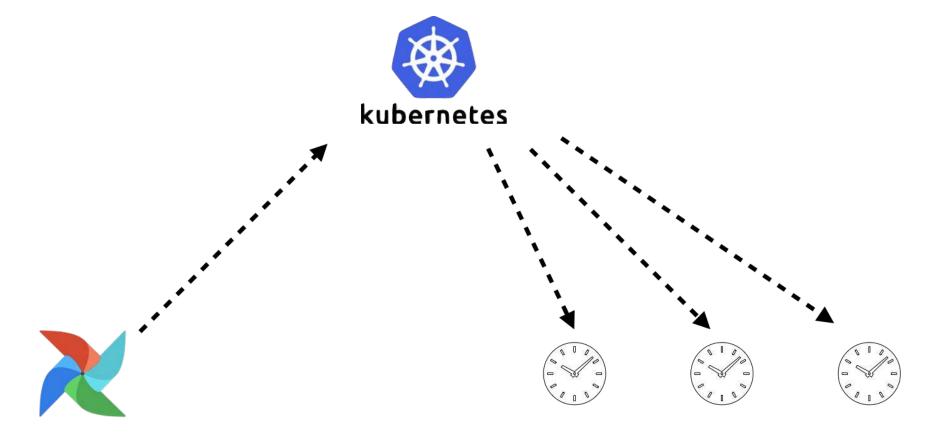
## KubernetesExecutor



## KubernetesExecutor



## KubernetesExecutor



## KubernetesExecutor vs. CeleryExecutor

#### KubernetesExecutor

- Dynamic Allocation
- executor\_config

#### CeleryExecutor

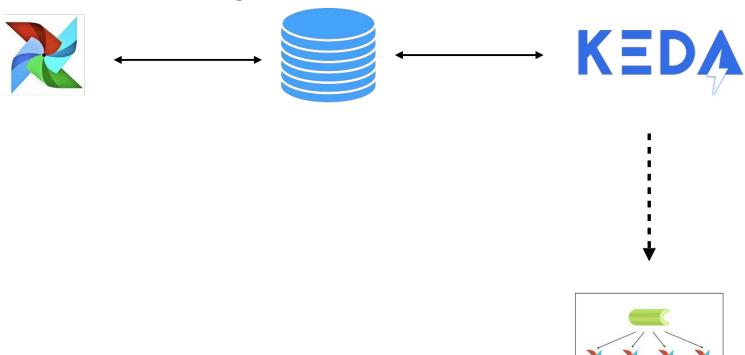
- Immediate SLAs
- Multiple tasks per-worker

# KEDA

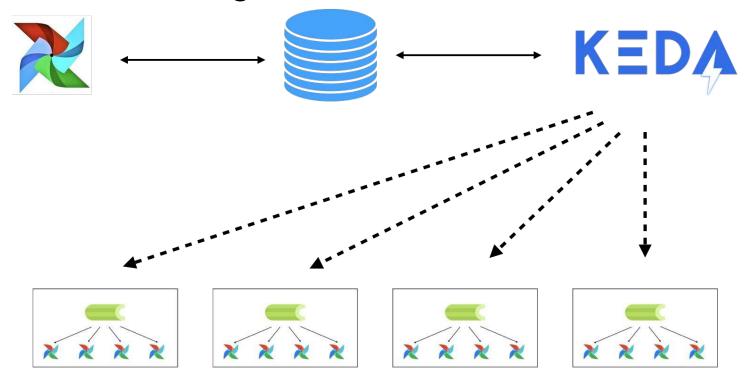
- Kubernetes Event-driven Autoscaler
- Scales based on # of RUNNING and QUEUED tasks in PostgreSQL backend



CEIL((0 RUNNING + 0 QUEUED)/16) = 0 workers



CEIL((0 RUNNING + 1 QUEUED)/16) = 1 workers



CEIL((20 RUNNING + 20 QUEUED)/16) = 4 workers

## **KEDA Queues**

- Historically Queues were expensive and hard to allocate
- With KEDA, queues are free! (can have 100 queues)
- KEDA works with k8s deployments so any customization you can make in a k8s pod, you can make in a k8s queue (worker size, GPU, secrets, etc.)

KubernetesExecutor
Pod Templating from
YAML/JSON



## KubernetesExecutor Pod Templating

- In the K8sExecutor currently, users can modify certain parts of the pod, but many features of the k8s API are abstracted away
- We did this because at the time the airflow community was not well acquainted with the k8s API
- We want to enable users to modify their worker pods to better match their use-cases

## KubernetesExecutor Pod Templating

- Users can now set the pod\_template\_file config in their airflow.cfg
- Given a path, the KubernetesExecutor will now parse the yaml file when launching a worker pod
- Huge thank you to @davlum for this feature

## Official Airflow Helm Chart



#### Helm Chart

- Donated by astronomer.io.
- This is the official helm chart that we have used both in our enterprise and in our cloud offerings (thousands of deployments of varying sizes)
- Helm 3 compliant
- Users can turn on KEDA autoscaling through helm variables
- "helm install apache/airflow"

#### Helm Chart

- Chart will cut new releases with each airflow release
- Will be tested on official docker image
- Significantly simplifies airflow onboarding process for Kubernetes users



```
def get_cat_pictures(num: int) -> List[Dict]:
    response = requests.get("https://cat_pictures.com", params={"num": num})
    return response.json()["cats"]

def save_cats(list_of_cats: List[Dict]) -> None:
    for cat in list_of_cats:
        save_it_somehow(cat)

with DAG("cat_fetcher"):
    get_task = PythonOperator(
        task_id="get_task", python_callable=get_cat_pictures, op_args=[42]
    )
    cats = "{{ task_instance.xcom_pull('get_task') }}"
    save_task = PythonOperator(
        task_id="save_task", python_callable=save_cats, op_args=[cats]
    )
    get_task >> save_task
```

- → PythonOperator boilerplate code
- → Define separately:
  - order relation
  - data relation
- → Writing jinja strings by hand

```
0 0 0
                                                                                       . .
                                                                                       def get_cat_dictures(num: int) -> List[Dict]:
    response = requests.get("https://cat_pictures.com", params={"num": num})
                                                                                           recounse = requests.get("https://cat_pictures.com", params={"num": num})
                                                                                           return response.json()["cats"]
                                                                                       def save_cats(list_of_cats: List[Dict]) -> None:
        save it somehow(cat)
                                                                                           for cat in list_of_cats:
                                                                                               save_it_somehow(cat)
                                                                                       with DAG("cat_fetcher"):
      task_id="get_task", python_callable=get_cat_pictures, op_args=[42]
                                                                                           get_task = get_cat_pictures(42)
                                                                                           save_task = save_cats(get_task)
      task_id="save_task", python_callable=save_cats, op_args=[cats]
```

Data and order relationship are same!

And works for all operators

Data and order relationship are same!

#### And works for all operators

#### AIP-31: Airflow functional DAG definition

- → Easy way to convert a function to an operator
- → Simplified way of writing DAGs
- → Pluggable XCom Storage engine

#### Find out more:

AIP-31: Airflow functional DAG definition

by Gerard Casas Saez

10th of July

```
@task
def get_cat_victures(num: int) -> List[Dict]:
    recponse = requests.get("https://cat_pictures.com", params={"num": num})
    return response.json()["cats"]

@task
def save_cats(list_of_cats: List[Dict]) -> None:
    for cat in list_of_cats:
        save_it_somehow(cat)

with DAG("cat_fetcher"):
    get_task = get_cat_pictures(42)
    save_task = save_cats(get_task)
```

**Example**: store and retrieve DataFrames on GCS or S3 buckets without boilerplate code

# Smaller changes



## Other changes of note

Connection IDs now need to be unique (<u>#8608</u>)

It was often confusing, and there are better ways to do load balancing

Python 3 only

Python 2.7 unsupported upstream since Jan 1, 2020

■ "RBAC" UI is now the only UI

Was a config option before, now only option. Charts/data profiling removed due to security risks

## Road to Airflow 2.0



## When will Airflow 2.0 be available?



## Airflow 2.0 – deprecate, but (try) not to remove

- Breaking changes should be avoided where we can if upgrade is to difficult users will be left behind
- Release "backport providers" to make new code layout available "now":

 Before 2.0 we want to make sure we've fixed everything we want to remove or break.

## How to upgrade to 2.0 safely

- Install the latest 1.10 release
- Run airflow upgrade-check (doesn't exist, yet #8765)
- Fix any warnings
- Upgrade Airflow

Thank you!

Time for Q & A

