Airflow operators

INTRODUCTION TO AIRFLOW IN PYTHON



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Operators

- Represent a single task in a workflow.
- Run independently (usually).
- Generally do not share information.
- Various operators to perform different tasks.

```
DummyOperator(task_id='example', dag=dag)
```

BashOperator

```
BashOperator(
   task_id='bash_example',
   bash_command='echo "Example!"',
   dag=ml_dag)
```

```
BashOperator(
    task_id='bash_script_example',
    bash_command='runcleanup.sh',
    dag=ml_dag)
```

- Executes a given Bash command or script.
- Runs the command in a temporary directory.
- Can specify environment variables for the command.

BashOperator examples

```
bash_task = BashOperator(task_id='clean_addresses',
  bash_command='cat addresses.txt | awk "NF==10" > cleaned.txt',
  dag=dag)
```

Operator gotchas

- Not guaranteed to run in the same location / environment.
- May require extensive use of Environment variables.
- Can be difficult to run tasks with elevated privileges.

Let's practice!

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Airflow tasks

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Tasks

Tasks are:

- Instances of operators
- Usually assigned to a variable in Python

Referred to by the task_id within the Airflow tools

Task dependencies

- Define a given order of task completion
- Are not required for a given workflow, but usually present in most
- Are referred to as *upstream* or *downstream* tasks
- In Airflow 1.8 and later, are defined using the *bitshift* operators
 - >>, or the upstream operator
 - <<, or the downstream operator</p>

Upstream vs Downstream

Upstream means **before**

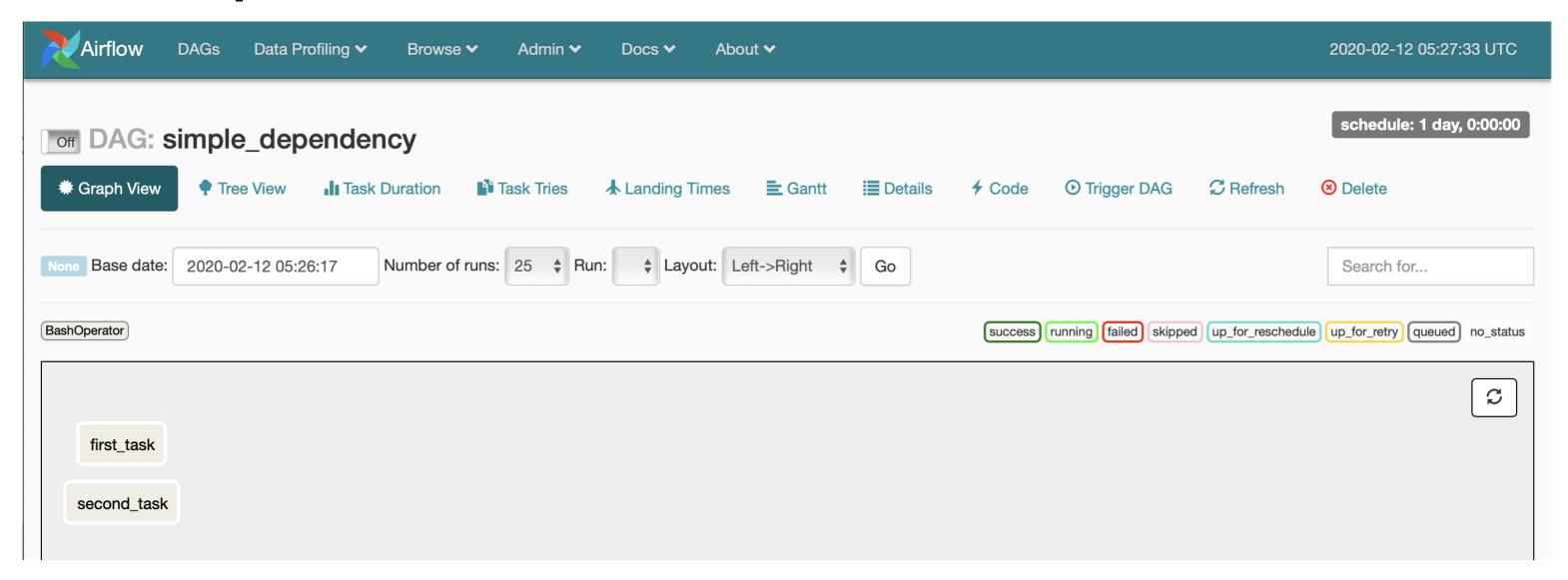
Downstream means after



Simple task dependency

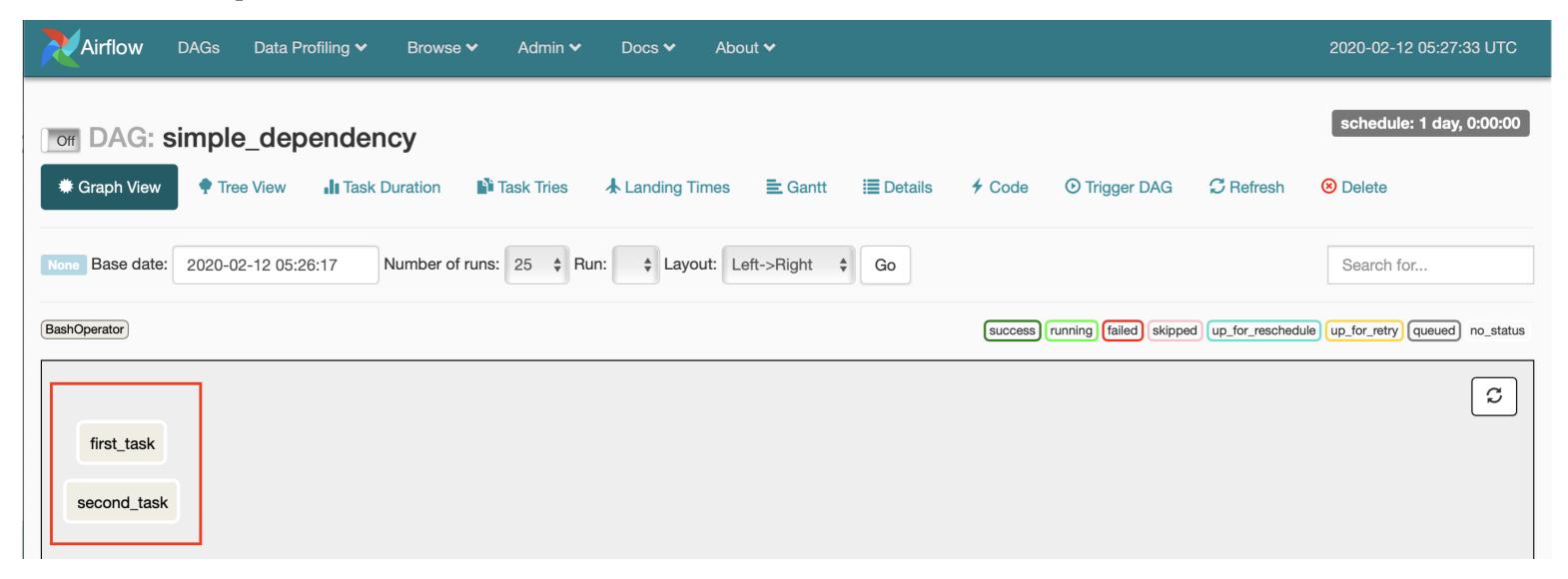
```
# Define the tasks
task1 = BashOperator(task_id='first_task',
                     bash_command='echo 1',
                     dag=example_dag)
task2 = BashOperator(task_id='second_task',
                     bash_command='echo 2',
                     dag=example_dag)
# Set first_task to run before second_task
task1 >> task2 # or task2 << task1
```

Task dependencies in the Airflow Ul



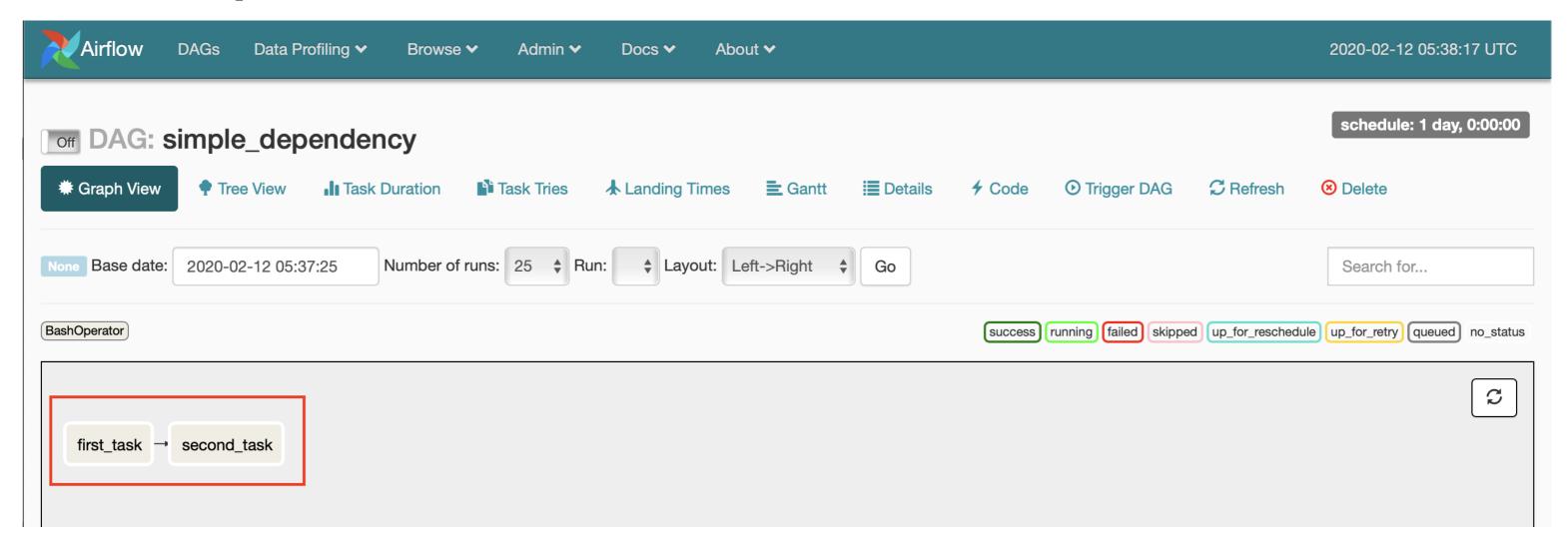


Task dependencies in the Airflow Ul





Task dependencies in the Airflow Ul



Multiple dependencies

Chained dependencies:

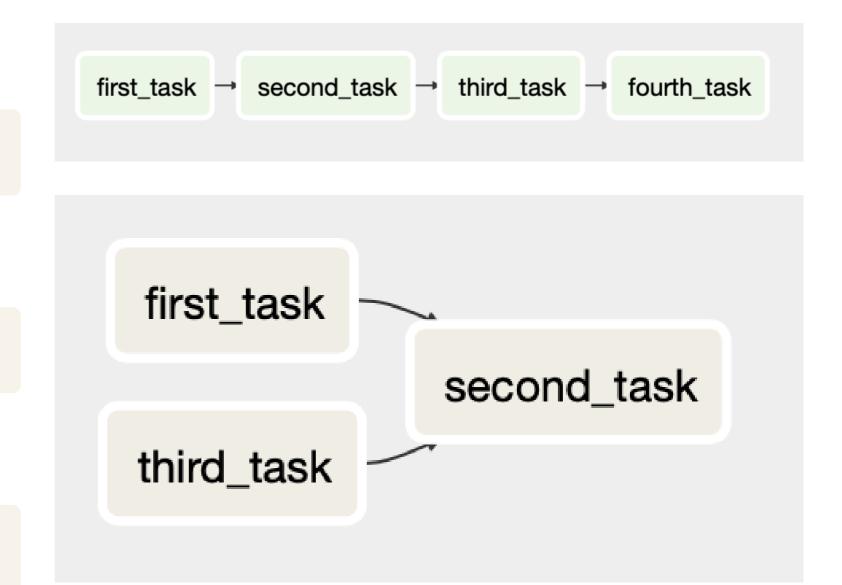
```
task1 >> task2 >> task3 >> task4
```

Mixed dependencies:

```
task1 >> task2 << task3
```

or:

```
task1 >> task2
task3 >> task2
```



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Additional operators

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PythonOperator

- Executes a Python function / callable
- Operates similarly to the BashOperator, with more options
- Can pass in arguments to the Python code

```
from airflow.operators.python_operator import PythonOperator

def printme():
    print("This goes in the logs!")

python_task = PythonOperator(
    task_id='simple_print',
    python_callable=printme,
    dag=example_dag
)
```

Arguments

- Supports arguments to tasks
 - Positional
 - Keyword
- Use the op_kwargs dictionary

op_kwargs example

```
def sleep(length_of_time):
  time.sleep(length_of_time)
sleep_task = PythonOperator(
    task_id='sleep',
    python_callable=sleep,
    op_kwargs={'length_of_time': 5}
    dag=example_dag
```

EmailOperator

- Found in the airflow.operators library
- Sends an email
- Can contain typical components
 - HTML content
 - Attachments
- Does require the Airflow system to be configured with email server details

EmailOperator example

```
from airflow.operators.email_operator import EmailOperator
email_task = EmailOperator(
    task_id='email_sales_report',
    to='sales_manager@example.com',
    subject='Automated Sales Report',
    html_content='Attached is the latest sales report',
    files='latest_sales.xlsx',
    dag=example_dag
```

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Airflow scheduling

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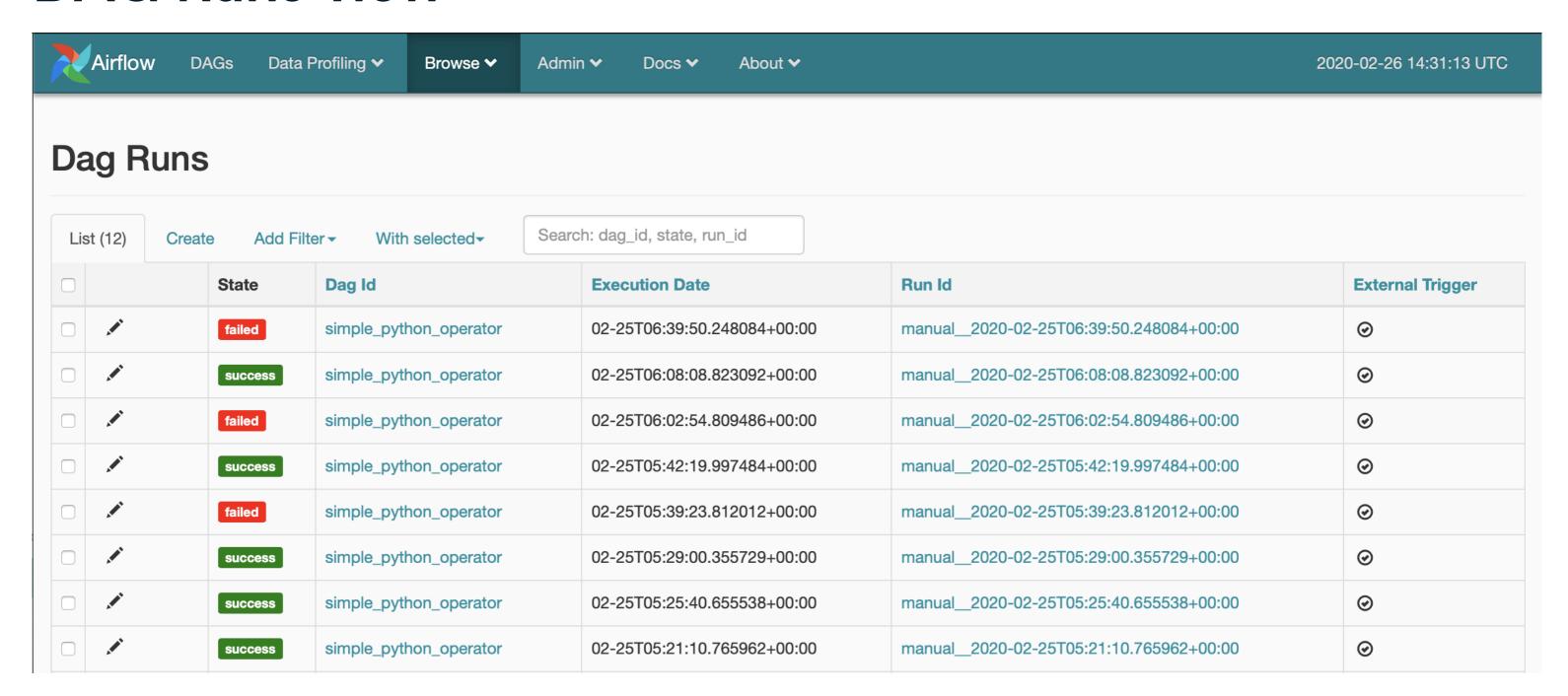
DAG Runs

- A specific instance of a workflow at a point in time
- Can be run manually or via schedule_interval
- Maintain state for each workflow and the tasks within
 - running
 - o failed
 - o success

¹ https://airflow.apache.org/docs/stable/scheduler.html

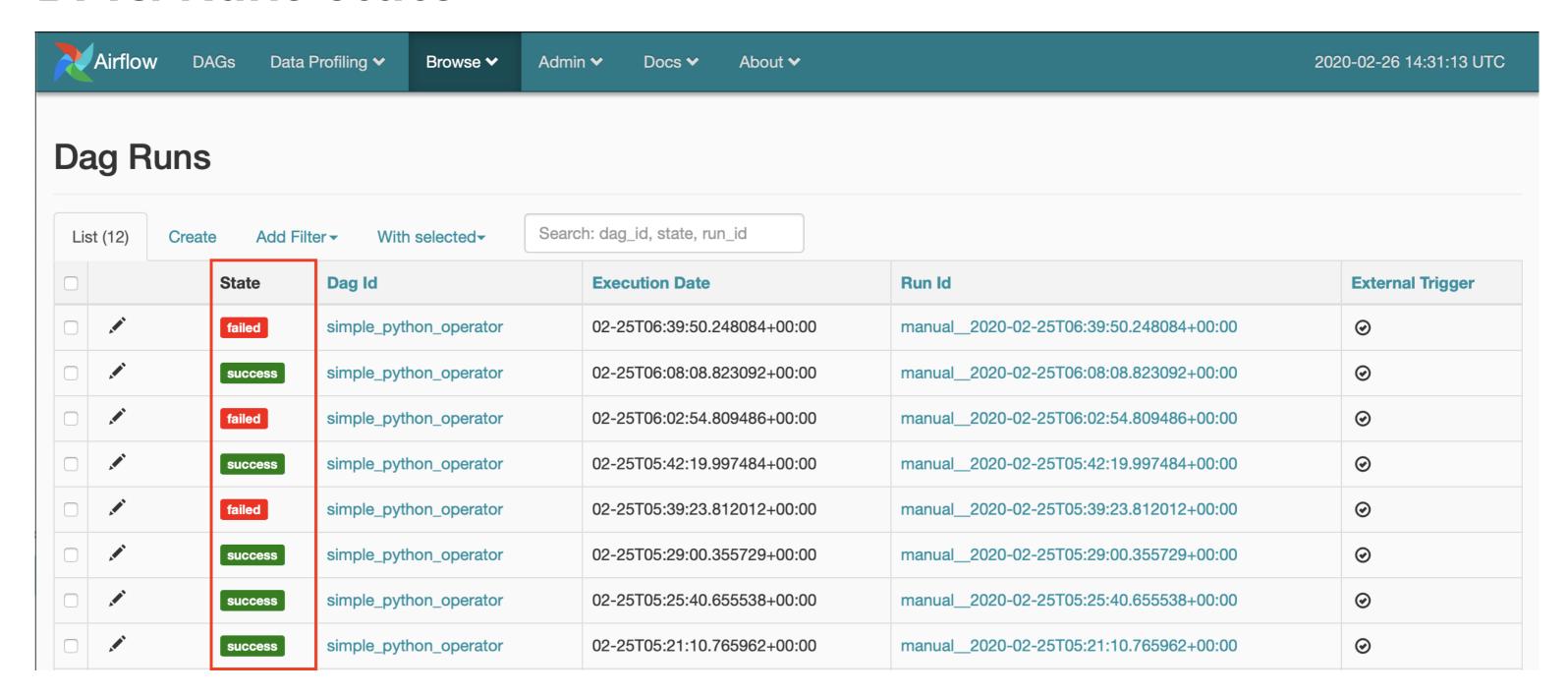


DAG Runs view





DAG Runs state





Schedule details

When scheduling a DAG, there are several attributes of note:

- start_date The date / time to initially schedule the DAG run
- end_date Optional attribute for when to stop running new DAG instances
- max_tries Optional attribute for how many attempts to make
- schedule_interval How often to run

Schedule interval

schedule_interval represents:

- How often to schedule the DAG
- Between the start_date and end_date
- Can be defined via cron style syntax or via built-in presets.

cron syntax

- Is pulled from the Unix cron format
- Consists of 5 fields separated by a space
- An asterisk * represents running for every interval (ie, every minute, every day, etc)
- Can be comma separated values in fields for a list of values

cron examples

0 12 * * * # Run daily at noon

* * 25 2 * # Run once per minute on February 25

0,15,30,45 * * * * # Run every 15 minutes

Airflow scheduler presets

Preset:

- @hourly
- @daily
- @weekly
- @monthly
- @yearly

cron equivalent:

¹ https://airflow.apache.org/docs/stable/scheduler.html

Special presets

Airflow has two special schedule_interval presets:

- None Don't schedule ever, used for manually triggered DAGs
- @once Schedule only once

schedule_interval issues

When scheduling a DAG, Airflow will:

- Use the start_date as the earliest possible value
- Schedule the task at start_date + schedule_interval

```
'start_date': datetime(2020, 2, 25),
'schedule_interval': @daily
```

This means the earliest starting time to run the DAG is on February 26th, 2020

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