

Who are we?



- Globally active bank, based in Amsterdam, the Netherlands
- 52.000 employees, 38.4 million customers
- Wholesale Banking Advanced Analytics (WBAA)
 - Works for the corporate clients, like Shell and Unilever
 - Consists of mostly Data Scientists(booo!) and Data Engineers(yeaaaah!)
 - Build data-driven algorithmic products

GO DATA DRIVEN

- Big Data and Data Science Consultancy, based in Amsterdam, The Netherlands
- 40 people, 2/3 data scientist, 1/3 data engineer
- Provide expertise in machine learning, big data, cloud, and scalable architectures
- Help organizations to become more datadriven
- Develop production-ready data applications



Real data sucks

Square peg, round hole

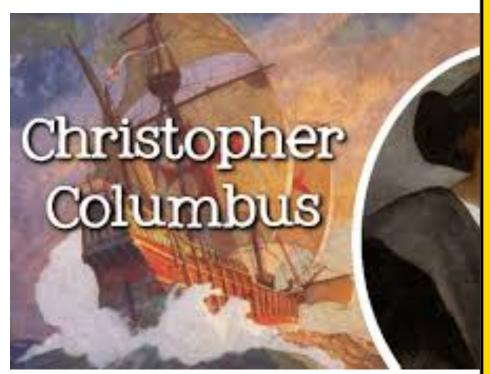
Test the pegs!

So, you thought it was square?





Data quality is a problem everywhere and always!



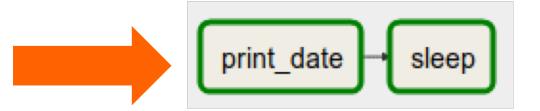




Airflow 101

Define your data pipeline in Python, as a (complex) sequence of tasks called a Directed Acyclic Graph (DAG) which can be scheduled

```
t1 = BashOperator(
  task_id='print_date',
  bash_command='date',
  dag=dag)
t2 = BashOperator(
  task_id='sleep',
  bash_command='sleep 5',
  retries=3,
  dag=dag)
t2.set_upstream(t1)
```



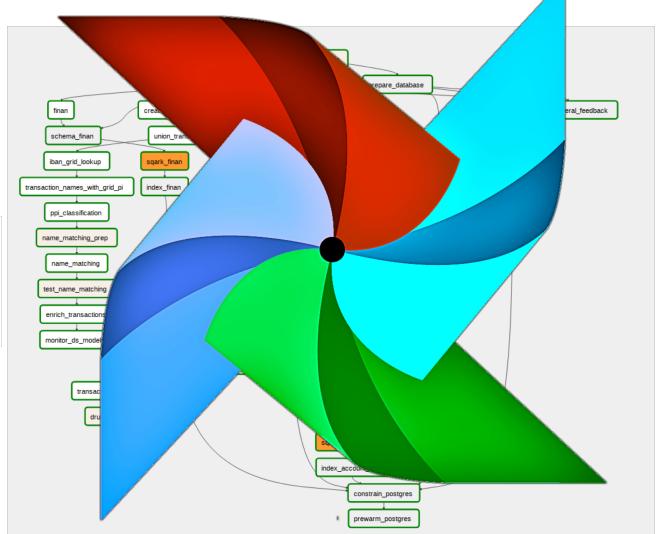


Our environment

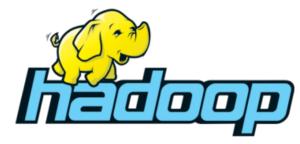
















Seems like a highway to hell?







DAG integrity test (Layer 1)



2 main use cases:

- Airflow version upgrade testing in Continuous Integration pipeline (CI)
- Sanity and typo checking of our DAGs

Broken DAG: [/usr/share/airflow/dags/finan_ingestion/airflowfile.py] No module named ssh_operator

ssh_execute_operator → ssh_operator



DAG integrity test (Layer 1)



```
task_a = BashOperator(task_id='task_a', ...)

test_task_a = BashOperator(task_id='test_task_a', ...)

task_b = BashOperator(task_id='task_a') ...)
```

airflow.exceptions.AirflowException: Cycle detected in DAG. Faulty task: <Task(BashOperator): templated>



DAG integrity test (Layer 1)



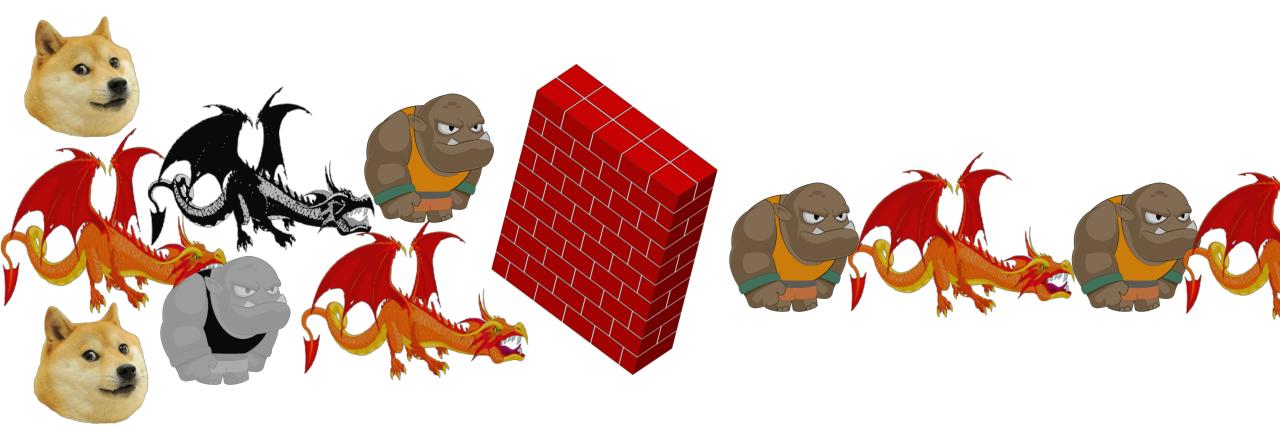
for all DAG.py in /dags/:
 assert all objects are valid Airflow DAGs

- We spend a lot of time fixing typos and logic errors in DAGs, after we upload it to Airflow to actually run
- You can now do this in your CI thanks to the guys and girls at CloverHealth





Split your data ingestion from your data deployment (Layer 2)





Split your data ingestion from your data deployment (Layer 2)

- Data comes in from many different sources
- Create an ingestion DAG per source
- Create an interface for systems that do the same thing i.e. payment transactions
- Let your data deployment pipeline for your project work with "clean" data
- Make sure you add a debugging column from your source, like a unique ID, if none exists. Also add a column indicating from which source it came





Data tests (Layer 3)

 After every action we take we have a test to check if that step has gone as expected

We split this up in testing ingestion of data sources and testing data

deployment steps

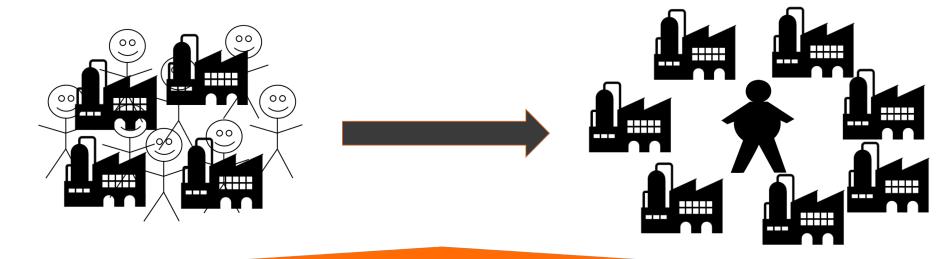


- Are there files available for ingestion?
- Did we get the columns that we expected?
- Are the rows that are in there valid? (Join it)
- Did the count only increase?



Data tests (Layer 3)

The data deployment pipeline also contains tests along the way

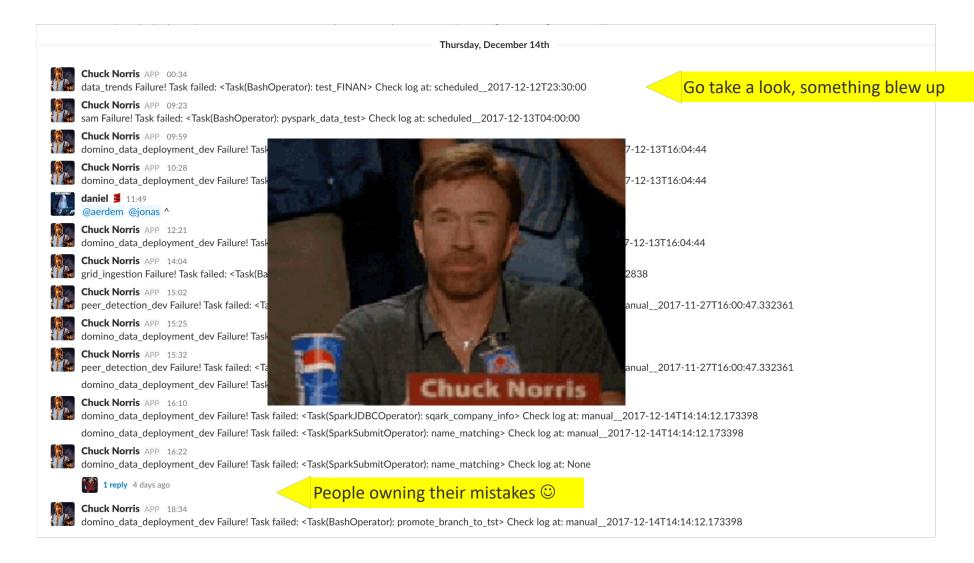


- Are the known private individuals filtered out?
- Are the known companies still in?
- Do all the output rows have a classification?
- Has the aggregation of PI's worked correctly?





Alerting by Chuck Norris (Layer 4)





Nuclear GIT (Layer 5)

if PRD≈DEV:







Nuclear GIT (Layer 5)

```
# this will hard reset all repos to the
version on master branch
# any local commits that have not been
pushed yet will be lost.
echo "Resetting "${dir%/*}
git fetch
git checkout -f master
git reset --hard origin/master
git clean -df
git submodule update ——recursive ——force
```

Don't copy paste this at home!





Mock pipeline tests (Layer 6)



Two variables; code and data



Control the exact data that goes in to your pipeline



Code is the variable; allowing you to test your logic



Mock pipeline tests (Layer 6)



Step 1: Create fake data that looks like your real data in a pytest fixture

```
PERSONS = [('name': 'Kim Yong Un', 'country': 'North Korea', 'iban': 'NK99NKBK000000666'), ...]
TRANSACTIONS= [('iban': 'NK99NKBK000000666', 'amount': 10 ), ...]
```

Step 2: Run your code in pytest

```
filter_data(spark, PERSONS, TRANSACTIONS)
```

Step 3: Check if your task returns the data that you expect:

```
assert spark.sql("""SELECT COUNT(*) ct FROM filtered_data WHERE
iban = 'NK99NKBK0000000666'""").first().ct == 0
```





DTAP (Layer 7)





DTAP (Layer 7)

DEV

- Quickly run your pipeline on a very small subset of your data
- In our case 0.0025% of all data
- Nothing will make sense, but it's a nice integration test

ACC

- Carbon copy of production
- You can check if you feel comfortable pushing to PRD
- Give access to a Product Owner for them to check

TST

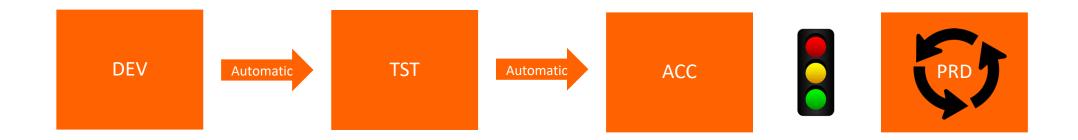
- Select a subset of your data for data that you know
- Immediately see if something is off
- Still quick to run

PRD

- Greenlight procedure for merging from ACC to PRD
- Manual operation



DTAP (Layer 7)



- 4 branches, dev, tst, acc, prd, each separately checked out in your /dags/ directory
- An environment.conf file outside of GIT in the corresponding directory
- Automatic promotion of code from dev to tst and tst to acc if everything went "green" in the DAG
- TriggerDagRunOperator to trigger the next DAG automatically for dev to tst and tst to acc



Local testing of Airflow with Whirl

Colleagues Bas Beelen and Kris Geusebroek made some very nice improvements after our time on this project.

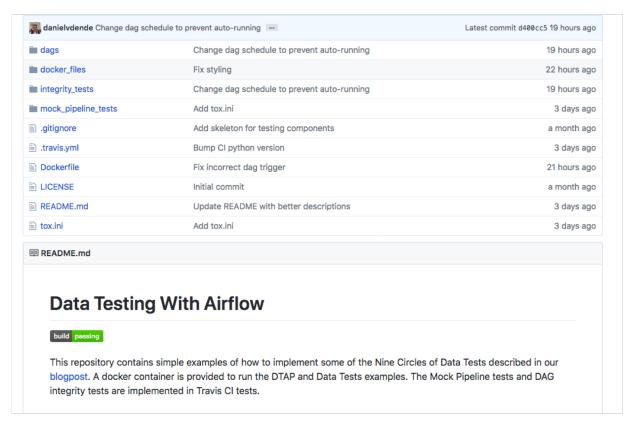
https://www.youtube.com/watch?v=jqK_HCOJ9Ak

High level overview:

- Data is confidential, can't take this local
- There are many different DAGs, some of which are very complex
- Whirl speeds up development by
 - Being able to reuse standard components of a DAG
 - Test your DAG locally end to end with fake data using Docker
- Open source code is in the pipeline with Bas and Kris ©



Now take your time to understand it all!



Blogpost: https://medium.com/@ingwbaa/datas-inferno-7-circles-of-data-testing-hell-with-airflow-cef4adff58d8

Github: https://github.com/danielvdende/data-testing-with-airflow



WH .gov

Thank you!





Questions?