Let's talk orchestration. An Intro to scheduling dbt with Dagster.

Sydney dbt meetup October 2021 Benoit Perigaud





Agenda

An intro to dagster and dbt

- 1 What are we going to talk about
- Why we might need orchestration tools
- 3 Dagster, Prefect and Airflow
- 4 Dagster core concepts
- 5 Demo time!



Benoit Perigaud

Me, myself and I

- Principal Analytics Engineer at Aginic
- Data consulting company, part of the list of dbt Preferred Consulting Providers





What this talk will and won't be about

Especially won't



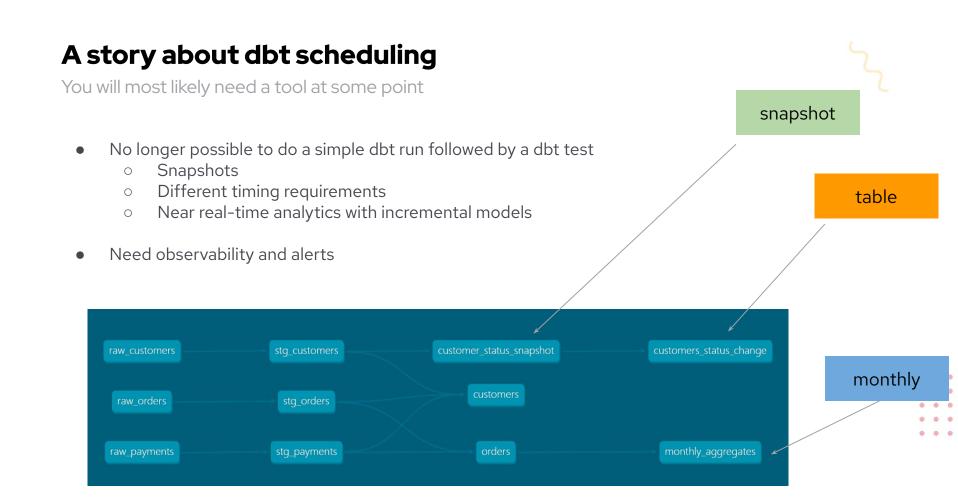
- An intro to why you might need an orchestration tool
- A brief overview of different options in the data space
- An overview of key concepts in Dagster
- A repo that you can clone and play with



- CI/CD orchestration of dbt
- An in-depth view of which orchestration tool is the best for you
- Ins and Outs of Dagster
- How to build a Dagster project using best practices
- How to deploy Dagster







Different options

But where to start

- dbt Cloud
- automation servers
 - As part of CI/CD
- cloud capabilities
 - Azure example: Azure Function App / Azure Data Factory
 - o ...
- cron (**)
- data orchestration tools
 - Airflow
 - Dasgter
 - Prefect



The 3 main players for data orchestration

Elevator pitch



"Airflow is a platform created by the community to programmatically author, schedule and monitor workflows."

- The OG
- Big community and numerous packages
- More difficult to host/manage (managed services exist)
- Local development not as easy
- Existing dbt examples



"Dagster is a data orchestrator for machine learning, analytics, and ETL"

- In the process of releasing a commercial hosted version, on top of the open source package
- Cares about the "what" on top of the "how"
- Configurable IO Managers
- Concept of data assets
- Easy local testing
- Integrates with Airflow



"Orchestrate the modern data stack"

- Commercial cloud offering as well as open source
- Talks to
 "negative/positive
 engineering"
- More modern, like Dagster





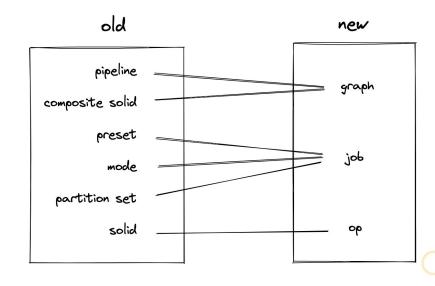
Dagster 101

Main concepts

The current way of working (as of oct 2021)

- solid
 - functional unit of work
 - read inputs, do an action, emit outputs
 - can materialise assets
- pipeline
 - links solids together
 - e.g.: solid_2(solid_1())
- preset/mode
 - run pipelines with specific configuration

This is changing





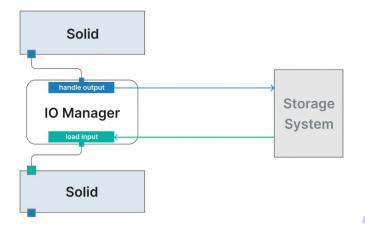


Other dagster concepts

Differentiators

IO Managers

Ability to switch between In Memory, Local, Blob or create your own. Allow partial re-runs



Asset Materialisations

"Asset" is Dagster's word for an entity, external to solids, that is mutated or created by a solid.

Accessible through the Asset Catalog

Asset-Oriented Ops



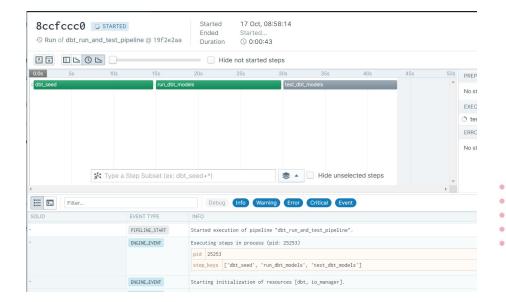
Authors ← Pipelines ← Assets



dbt deps | dbt run | dbt test

- dagit shows the results and progress of the different dbt commands
- Ability to look at past runs and re-run them

```
> @pipeline(...
)
def dbt_run_and_test_pipeline():
    dbt_seed_result = dbt_seed()
    run_result = run_dbt_models(dbt_seed_result)
    test_dbt_models(run_result)
```



dbt assets materialisation

- Out of the box support for Asset Materialisation of dbt models
- Reads and Parses the "run_results.json" artefact
- By default, no support for tests, but very easy to implement (see function generate materializations dbt test)

```
@solid(
    required_resource_keys={"dbt"},
    config_schema={"models": Noneable(str), "exclude": Noneable(str)},
)

def run_dbt_models(context, run_result: DbtCliOutput) -> DbtCliOutput:
    dbt_output = context.resources.dbt.run(
    context.solid_config["models"], context.solid_config["exclude"]
    )
    for materialization in generate_materializations(dbt_output):
        yield_materialization
    yield_Output(dbt_output)
```



setup presets and schedules

- presets
 - used to execute ad-hoc pipelines
 - can be loaded as YAML files
- schedules
 - cron-based
 - o able to "skip" runs based on filters

```
mode defs=[ModeDefinition(resource defs={"dbt": my dbt resource})],
   tags={"use-case": "dbt run"},
   preset_defs=[
       PresetDefinition(
           "daily schedule".
           run config={
                    "run_dbt_models": {"config": {"exclude": "tag:monthly+"}},
                    "test_dbt_models": {"config": {"exclude": "tag:monthly+"}},
       PresetDefinition(
           "monthly schedule",
           run_config={
                "solids":
                    "run_dbt_models": {"config": {"models": "tag:monthly+"}},
                    "test dbt models": {"config": {"models": "tag:monthly+"}},
def dbt run and test pipeline():
   dbt seed result = dbt seed()
   run_result = run_dbt_models(dbt_seed_result)
   test_dbt_models(run_result)
```

asset sensors

- trigger actions when an asset has been updated
- useful for dbt snapshots
- allows easier maintenance and less dependencies of cross-teams data pipelines



What next?

More configuration and going beyond dbt

- 🔹 Leverage "dbt build" 🎉
- Prevent concurrent runs of dbt
 - https://docs.dagster.io/deployment/run-coordinator
- Setup alerts and/or retries when the pipeline fails
- Maybe try to generate the full dbt DAG in dagster
 - Trying with Dynamic Mapping & Collect:
 https://docs.dagster.io/concepts/solids-pipelines/pipelines#dynamic-mapping--collect
- Make our DAG more than dbt
 - Pause/resume EL pipeline
 - Refresh downstream applications (e.g. Power BI dataset)
 - Send notifications in slack
- Use the GraphQL API
 - Query information
 - Trigger pipelines



Thanks everyone!

That's all folks

- This talk (slides and code)
 - https://github.com/Aginic/syd-dbt-meetup-dagster
- Dagster
 - dagster: https://dagster.io/
 - o dagster and dbt: https://docs.dagster.io/integrations/dbt
 - dagster slack: https://dagster-slackin.herokuapp.com/
- dbt
 - o dbt build: https://docs.getdbt.com/reference/commands/build



Questions?