Building a robust data pipeline with the "dAG" Stack: dbt, Airflow, and Great Expectations

ODSC East, April 2021 Sam Bail @spbail

Hi, I'm Sam!

I'm a "data person" and (former)
engineering director at Superconductive,
the team behind Great Expectations. I'm
from Germany, but currently based in NYC.*
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and at sambail.com



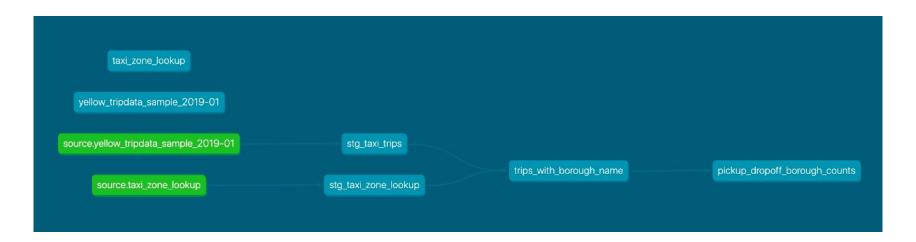
Welcome to the "dAG" Stack: dbt, Airflow, Great Expectations

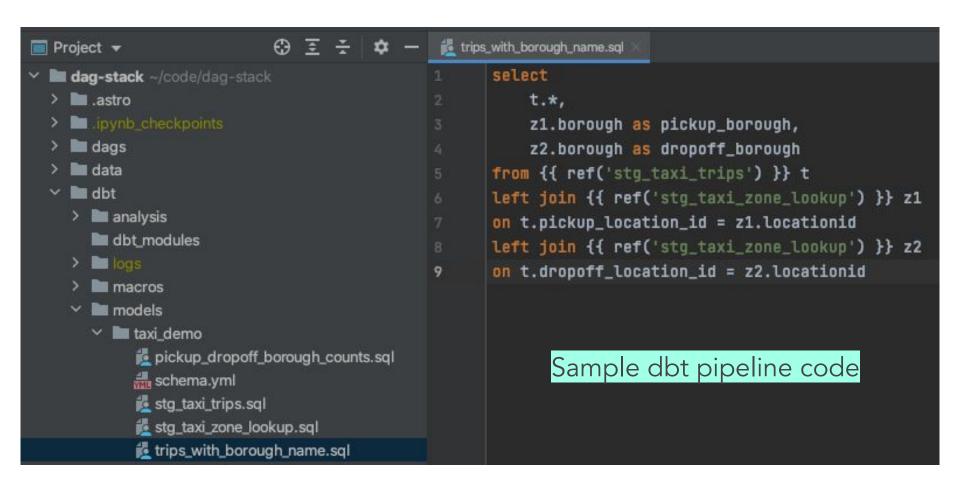
- 1. What are all these tools?
- 2. Let's talk about testing...
- 3. Ok, how does this all fit together?

1: dbt (data build tool) **Xdbt**



- "The T in FIT"
- Open source Python package
- Write a data pipeline as a series of templated SQL nodes



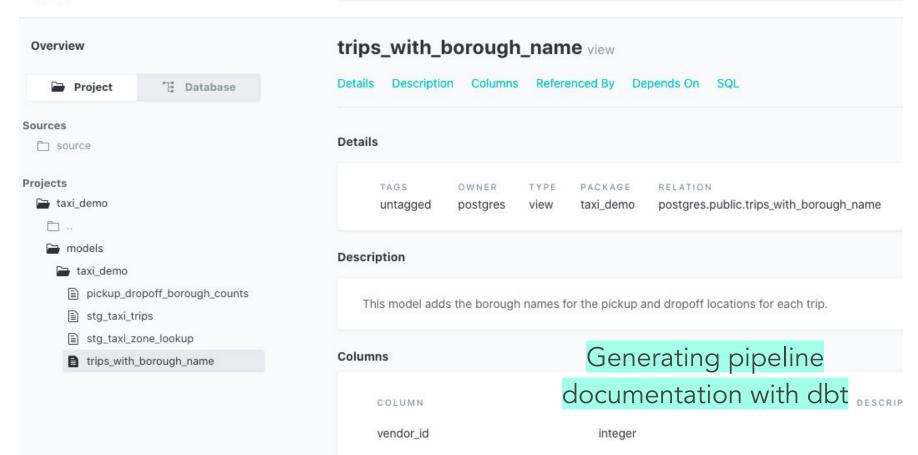


```
(dag_stack) sam@DeLorean dbt % dbt run --profiles-dir /Users/sam/code/dag-stack/ --target local_dev
Running with dbt=0.19.0
Found 4 models, 0 tests, 0 snapshots, 0 analyses, 138 macros, 0 operations, 2 seed files, 2 sources, 0 exposures
16:06:33 | Concurrency: 1 threads (target='local_dev')
16:06:33
16:06:33 | 2 of 4 OK created view model public.stg_taxi_zone_lookup..... [CREATE VIEW in 0.14s]
16:06:34 | 3 of 4 OK created view model public.trips_with_borough_name..... [CREATE VIEW in 0.09s]
16:06:34 | 4 of 4 START view model public.pickup_dropoff_borough_counts..... [RUN]
16:06:34 | 4 of 4 OK created view model public.pickup_dropoff_borough_counts.... [CREATE VIEW in 0.10s]
16:06:34
16:06:34 | Finished running 4 view models in 1.78s.
Completed successfully
Done. PASS=4 WARN=0 ERROR=0 SKIP=0 TOTAL=4
```

Running a dbt pipeline



Search for models...



```
(dag_stack) sam@DeLorean dbt % dbt test --profiles-dir /Users/sam/code/dag-stack/ --target local_dev
Running with dbt=0.19.0
Found 4 models, 5 tests, 0 snapshots, 0 analyses, 138 macros, 0 operations, 2 seed files, 2 sources, 0 exposures
16:25:31 | Concurrency: 1 threads (target='local_dev')
16:25:31 |
16:25:31 | 1 of 5 START test accepted_values_trips_with_borough_name_vendor_id__1_2_4 [RUN]
16:25:31 | 1 of 5 PASS accepted_values_trips_with_borough_name_vendor_id__1_2_4 [PASS in 0.05s]
16:25:31 | 2 of 5 START test not_null_trips_with_borough_name_dropoff_datetime.. [RUN]
16:25:31 | 2 of 5 PASS not_null_trips_with_borough_name_dropoff_datetime...... [PASS in 0.05s]
16:25:31 | 3 of 5 START test not_null_trips_with_borough_name_passenger_count... [RUN]
16:25:31 | 3 of 5 PASS not_null_trips_with_borough_name_passenger_count...... [PASS in 0.05s]
16:25:31 | 4 of 5 START test not_null_trips_with_borough_name_pickup_datetime... [RUN]
16:25:31 | 4 of 5 PASS not_null_trips_with_borough_name_pickup_datetime...... [PASS in 0.05s]
16:25:31 | 5 of 5 START test not_null_trips_with_borough_name_vendor_id...... [RUN]
16:25:31 |
16:25:31 | Finished running 5 tests in 1.17s.
```

Completed successfully

Adding tests to a dbt pipeline

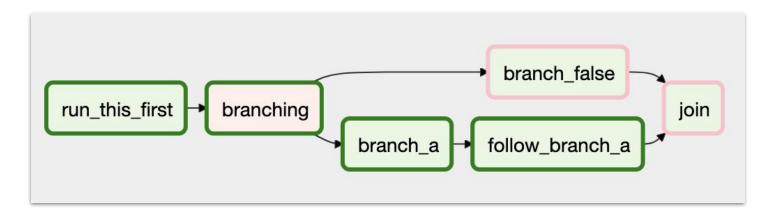
Done. PASS=5 WARN=0 ERROR=0 SKIP=0 TOTAL=5

dbt does not have any built-in scheduling functionality, or ability to run arbitrary non-SQL code

If you want to do that, you need... something else!

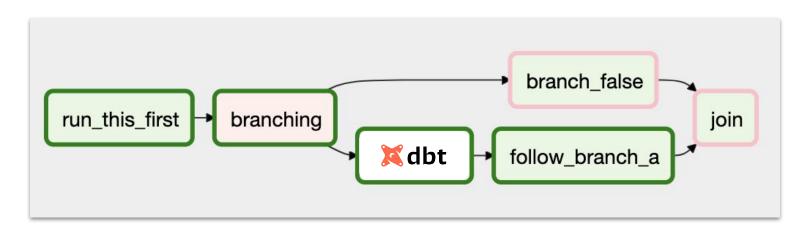


- Workflow orchestration tool
- Another open source Python package :)
- "Cron on steroids" scheduling and more



2: Apache Airflow

We can run a dbt pipeline as a task in an Airflow DAG:



```
dag = DAG(
   dag_id='dag_stack_dag',
    schedule_interval=None,
   default_args=default_args
# This runs the transformation steps in the dbt pipeline
dbt_run = DbtRunOperator(
   task_id='dbt_run',
    dir=DBT_PROJECT_DIR,
    profiles_dir=PROJECT_HOME,
    target=DBT_TARGET,
    dag=dag
```

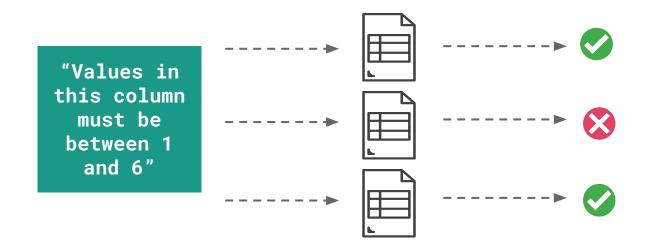
Running a dbt pipeline in an Airflow task*

^{*} See the "Additional resources" section in the demo repo for other patterns to run dbt in Airflow

3: Great Expectations



- Open source data validation and documentation tool
- Let's you express what you *expect* from your data (ha!)



What is an Expectation?

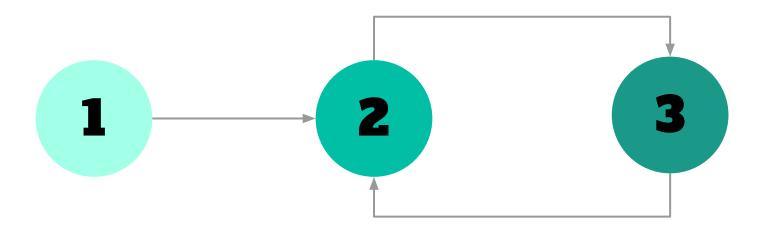
A statement about what we expect from our data, that can be expressed in code

That is stored in JSON

"Values in this column must be between 1 and 6"

And can be translated into a human-readable format

Typical Great Expectations workflow



Create an Expectation Suite

- `suite new` (from scratch)
- `suite scaffold` (profiled)
- Save the suite

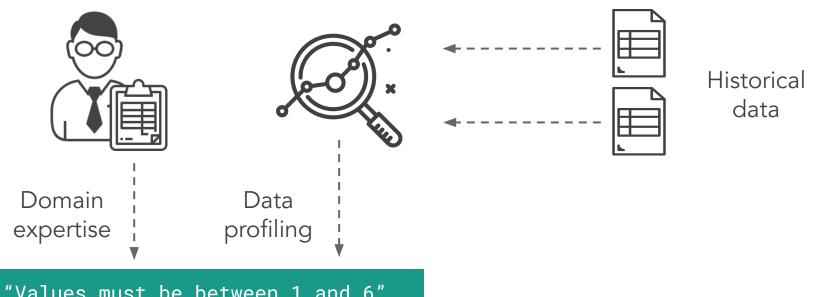
Validate incoming data

- Load new data
- Load the suite
- Run validation

Respond to validation results

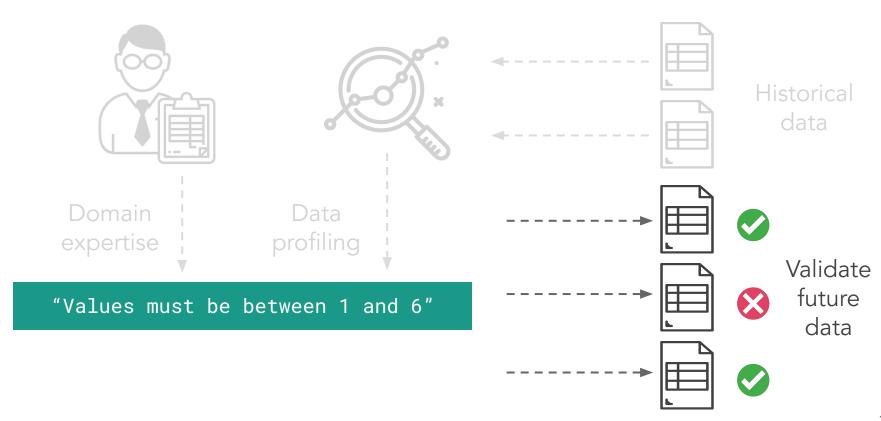
- Stop pipeline
- Send alerts
- Build & inspect Data Docs

Automated profiling to "scaffold" Expectations



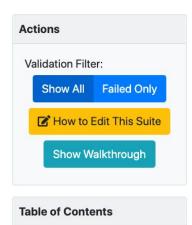
"Values must be between 1 and 6"

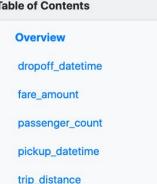
Validating your data

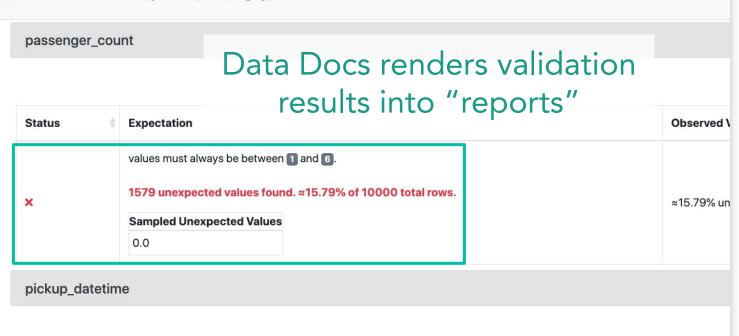




great_expectations Home / Validations / taxi.demo / yellow_tripdata_staging / 2020-08-04T16:38:46.210132+00:00







Status	Expectation	Observed Value
•	values must never be null.	100% not null

```
validate_source_data = GreatExpectationsOperator(
   task_id='validate_source_data',
   assets_to_validate = [
           'batch_kwargs': {
               'path': os.path.join(PROJECT_HOME, 'data', 'taxi_zone_lookup.csv'),
               'datasource': 'data_dir'
           },
           'expectation_suite_name': 'taxi_zone.source'
       },
           'batch_kwarqs': {
               'path': os.path.join(PROJECT_HOME, 'data', 'yellow_tripdata_sample_2019-01.csv'),
               'datasource': 'data_dir'
           'expectation_suite_name': 'taxi_trips.source'
       },
   1,
                                                  Running validation with Great
   data_context_root_dir=GE_ROOT_DIR,
                                                  Expectations in an Airflow task
   dag=dag
```

Ok, let's talk about testing!



1: You should test your data.

(No, really.)

Don't believe me?

"Our stakeholders would notice data issues before we did... which really eroded trust in the data and our team."

(A Great Expectations user)

"Re-running our pipelines after finding a data quality issue would incur actual costs for the compute environment."

(A Great Expectations user)

"Remember that one Thanksgiving where we worked all weekend to fix those data issues we only noticed at the last minute? Never again."

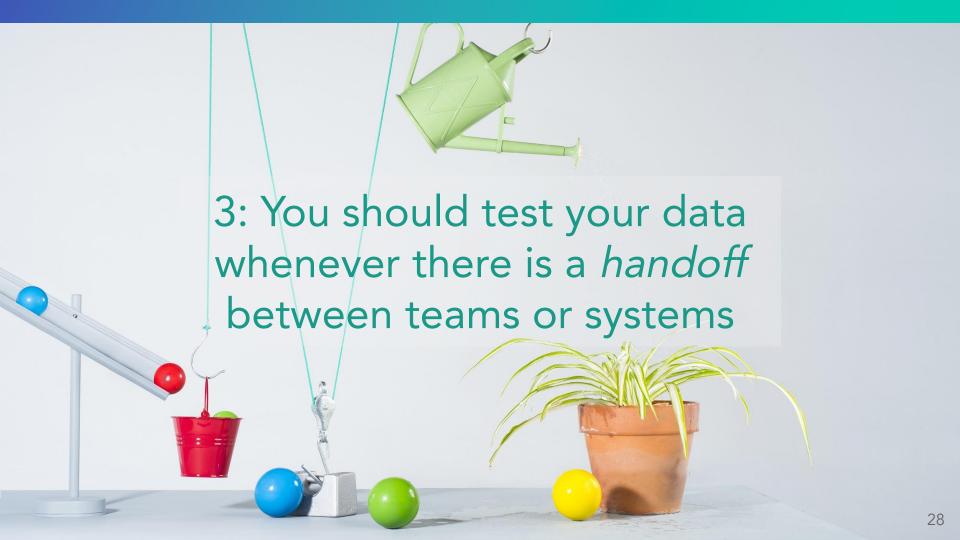
(That was me.True #datahorrorstory.)

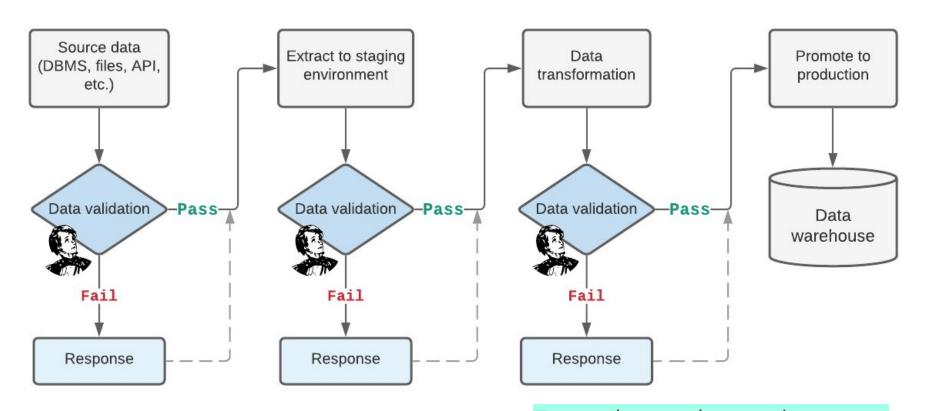
But... where do we start?

(Which tool should we use? How do we know we're testing the right thing? What do we do when tests fail? Who owns this? How do we keep them up to date? How do our stakeholders find out about the state of the data?)

2: Data testing is kinda hard.

(But I can show you how to get started...)



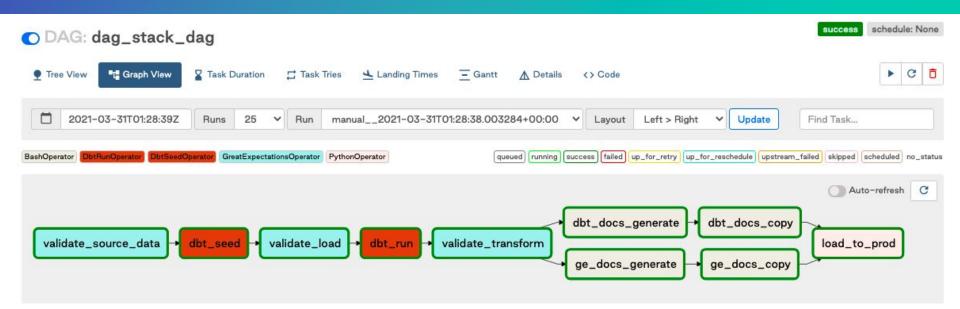


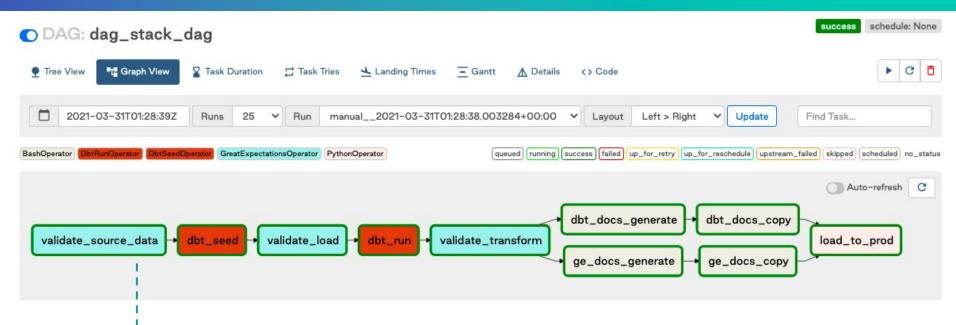
Example pipeline with tests at every relevant "handoff" point

Let's put this all together now.

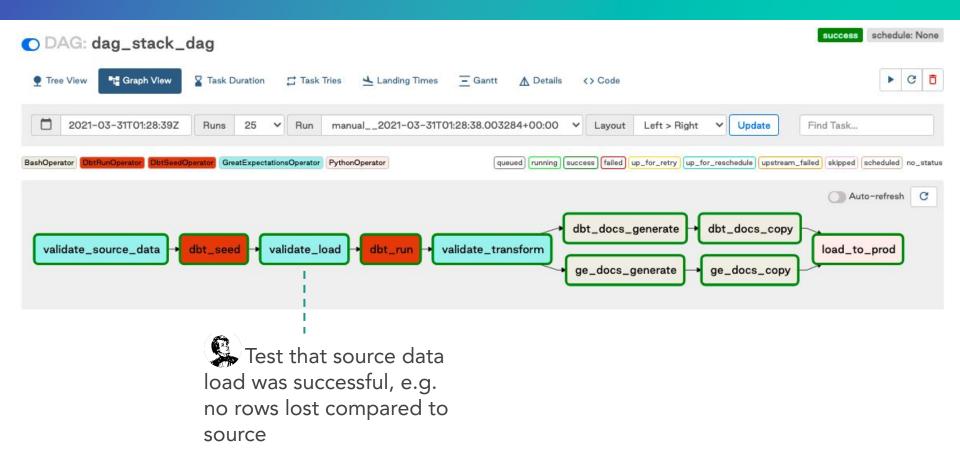
```
dag-stack ) dags ) & dag stack dag.py
                               dag_stack_dag.py
  ■ Project ▼
                                       dag-stack ~/code/dag-stack
    > astro
                                       dag = DAG(
    > ipynb checkpoints
                                          dag_id='dag_stack_dag',

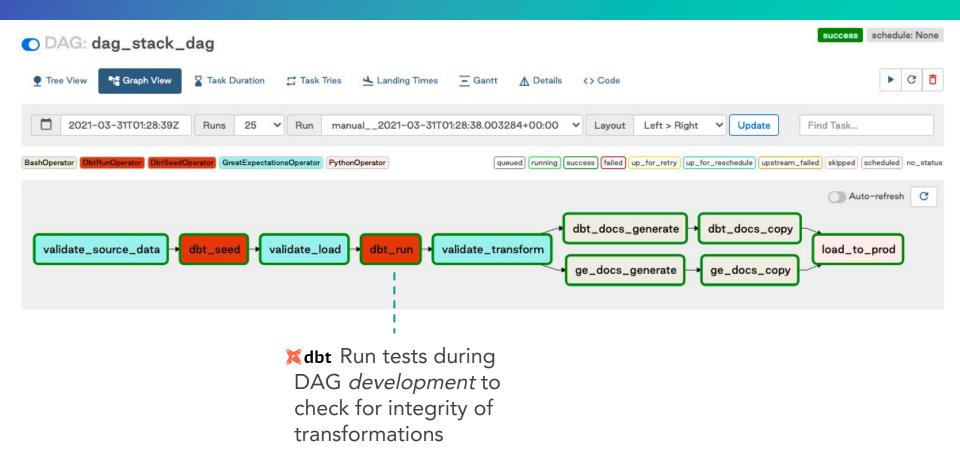
✓ ■ dags
                                          schedule_interval=None,
        dag_stack_dag.py
                                          default_args=default_args
   > data
   ∨ ■ dbt
      > analysis
       ■ dbt_modules
                                       validate_source_data = GreatExpectationsOperator(
     > logs
                                          task_id='validate_source_data',
      > macros
      > models
                                          assets_to_validate = [...],
                                          data_context_root_dir=GE_ROOT_DIR,
      > snapshots
      > larget
                                          dag=dag
      > tests
        .gitignore
       dbt_project.yml
                                       # The dbt seed command loads files in the data directory to the database
       # README.md
                                       dbt_seed = DbtSeedOperator(
   great_expectations
                                          task_id='dbt_seed',
       checkpoints
                                          dir=DBT_PROJECT_DIR,
      > expectations
                                          profiles_dir=PROJECT_HOME,
                                                                         Demo repo live on
      > notebooks
                                          target=DBT_TARGET,
      > plugins
                                          dag=dag
                                                                 github.com/spbail/dag-stack
      > uncommitted
        agitignore.
        great_expectations.yml
```

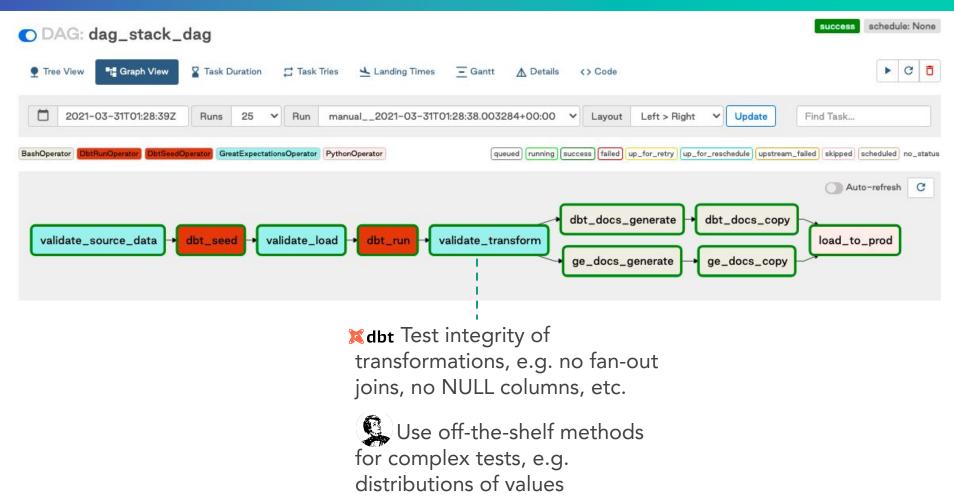




Test that source data (in this example: CSV files) matches expected format, e.g. correct number of columns, data types, row count "similar" to last month's, etc.







Wrap-up!

 Airflow, dbt, and Great Expectations = a modern data pipeline stack with built-in testing capabilities

 Test your data whenever there is a handoff between teams or systems

 Use dbt or Great Expectations depending on what data you test and the complexity of your tests

Thank you!

Ping me - Sam Bail, @spbail Repo: github.com/spbail/dag-stack

