

Building a robust data pipeline with dbt, Airflow, and Great Expectations



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Coalesce 2020

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...)

Data testing in the “dAG” Stack: dbt, Airflow*, Great Expectations

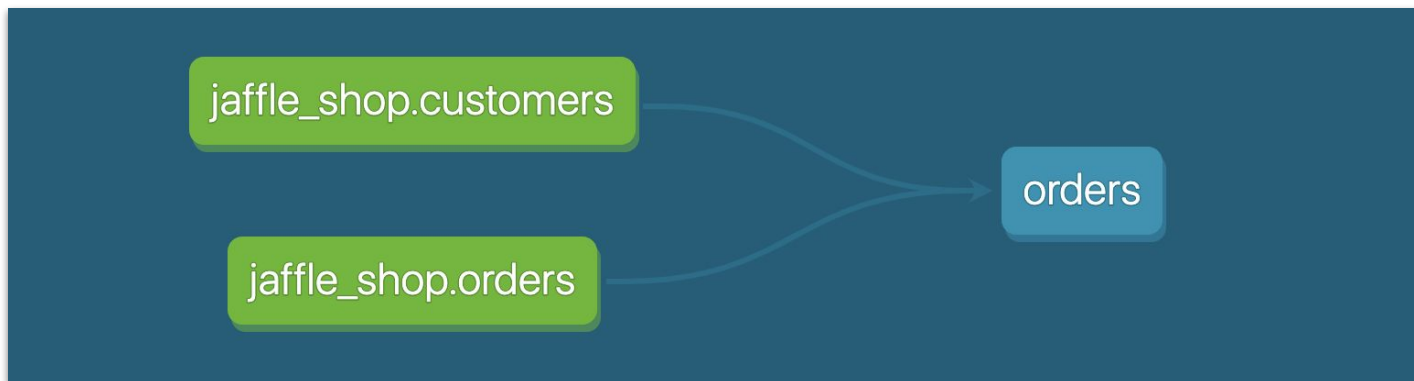
* This could be any other workflow orchestration tool,
in fact, basically all of them work with dbt and Great
Expectations! Hi @ dagster, Prefect, Kedro...





"The T in ELT"

... but you know this already :)

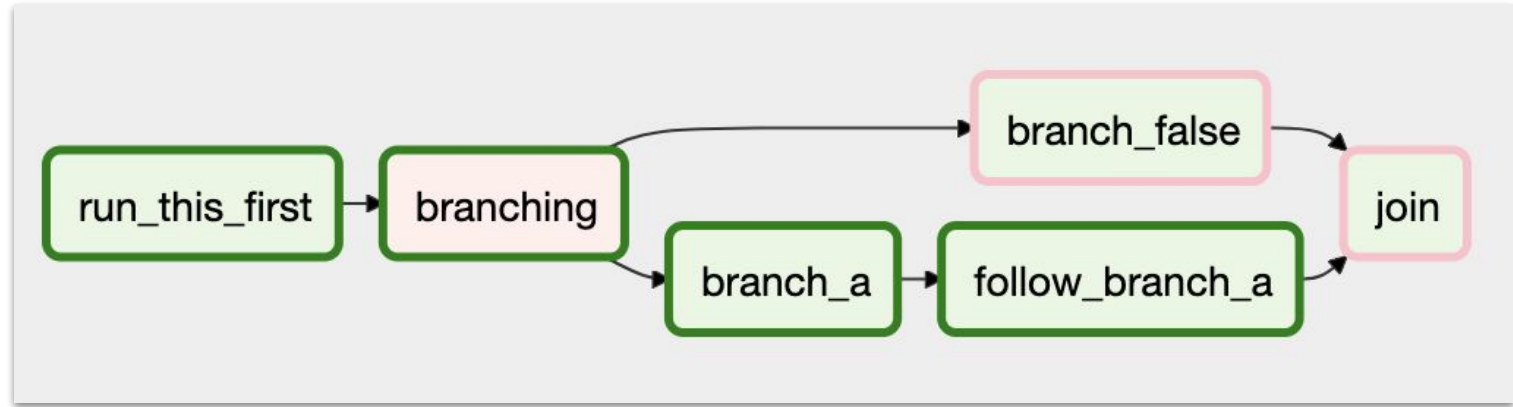


Apache Airflow



Workflow orchestration tool

"Cron on steroids" - scheduling and more

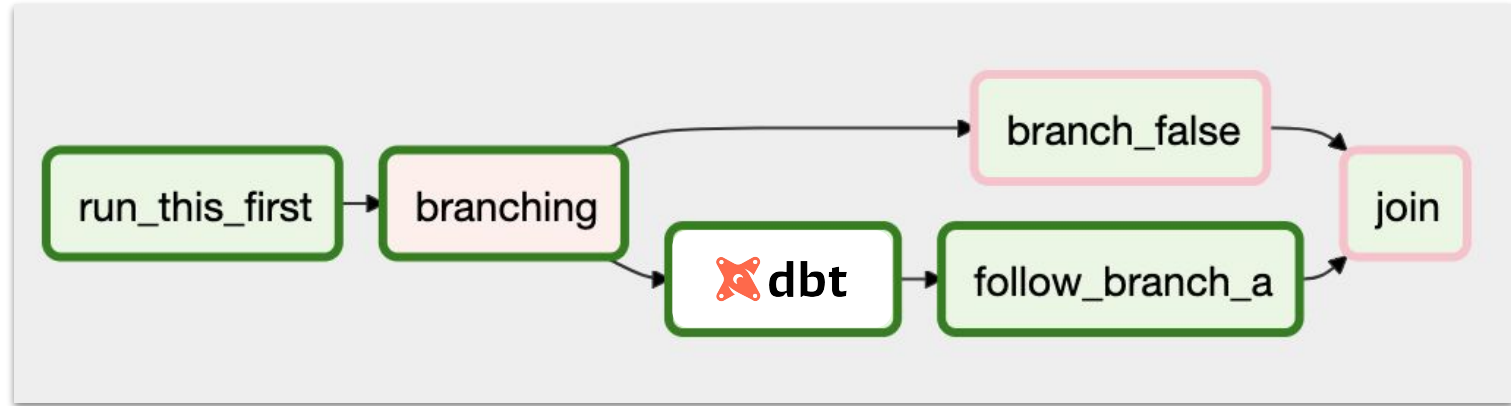


Apache Airflow



Workflow orchestration tool

"Cron on steroids" - scheduling and more

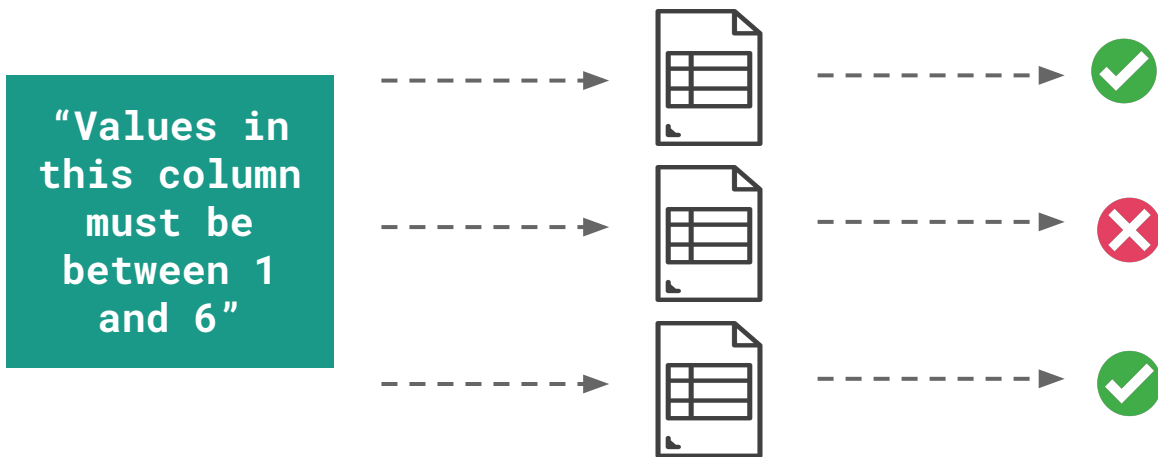


Great Expectations



Open source data validation and documentation tool

Let's you express what you **expect** from your data (ha!)



What is an Expectation?

```
expect_column_values_to_be_between(  
    column='passenger_count',  
    min_value=1,  
    max_value=6  
)
```

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

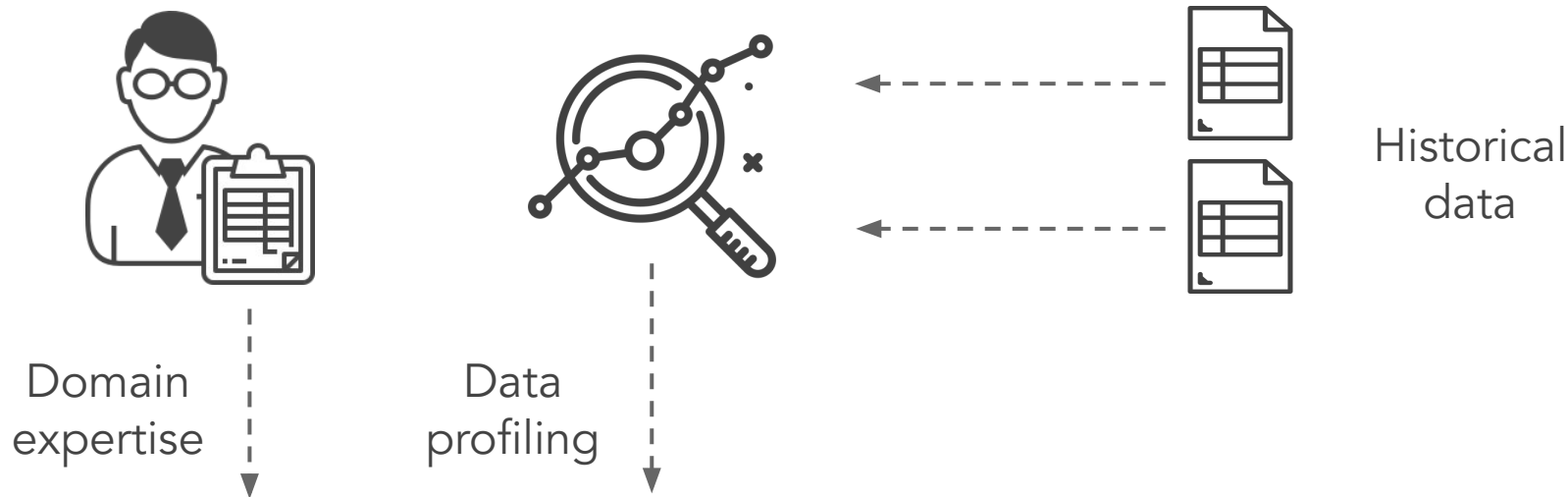
```
{  
  "expectation_type": "expect_column_values_to_be_between",  
  "kwargs": {  
    "column": "passenger_count",  
    "min_value": 1,  
    "max_value": 6  
  },  
}
```

That is stored in JSON

“Values in this column must be between 1 and 6”

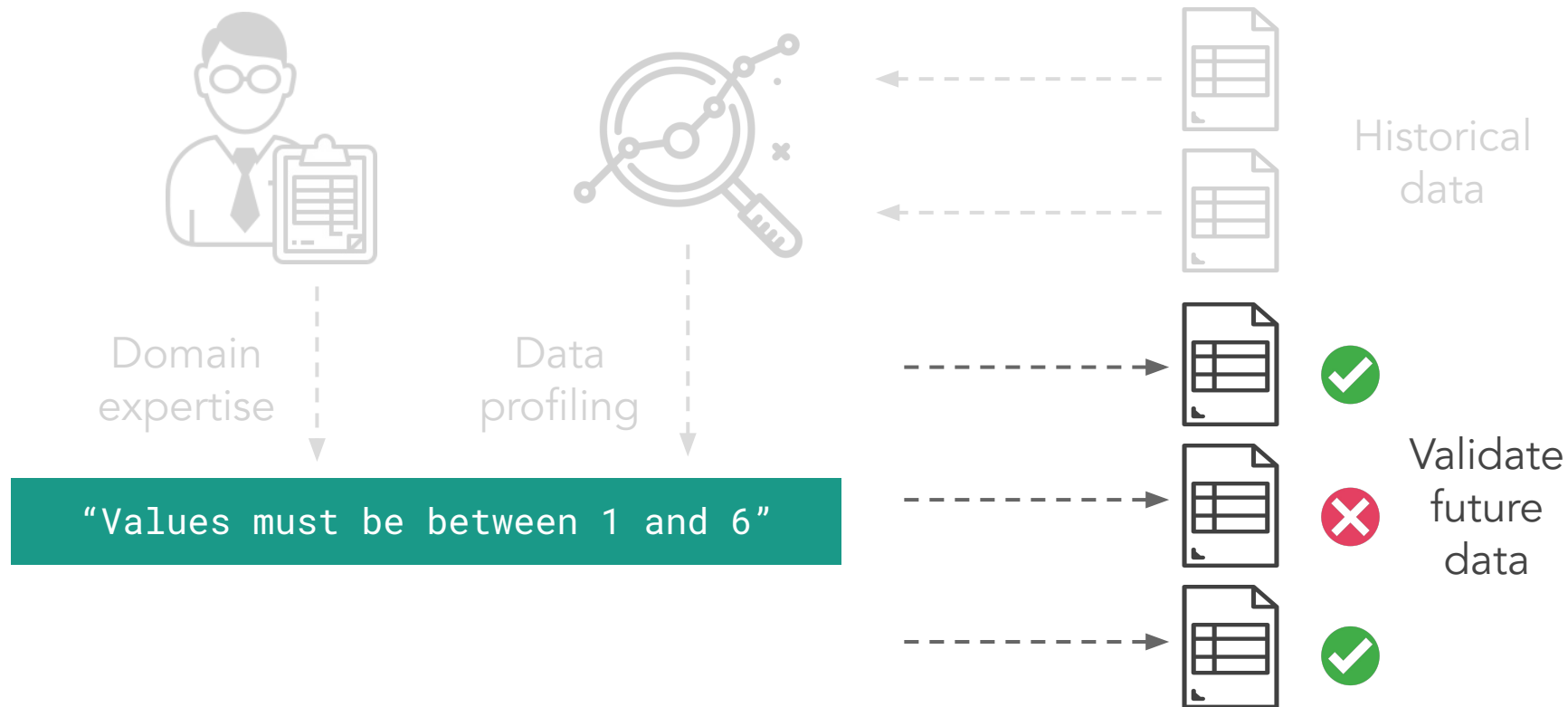
And can be translated into a human-readable format

Automated profiling to “scaffold” Expectations



"Values must be between 1 and 6"

Validating your data





Actions

Validation Filter:

[Show All](#) [Failed Only](#)[✎ How to Edit This Suite](#)[Show Walkthrough](#)

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[Overview](#)[dropoff_datetime](#)[fare_amount](#)[passenger_count](#)[pickup_datetime](#)[trip_distance](#)

passenger_count

Data Docs renders validation results into "reports"

Status	Expectation	Observed V
✖	<p>values must always be between 1 and 6.</p> <p>1579 unexpected values found. ≈15.79% of 10000 total rows.</p> <div>Sampled Unexpected Values</div> <div>0.0</div>	≈15.79% un

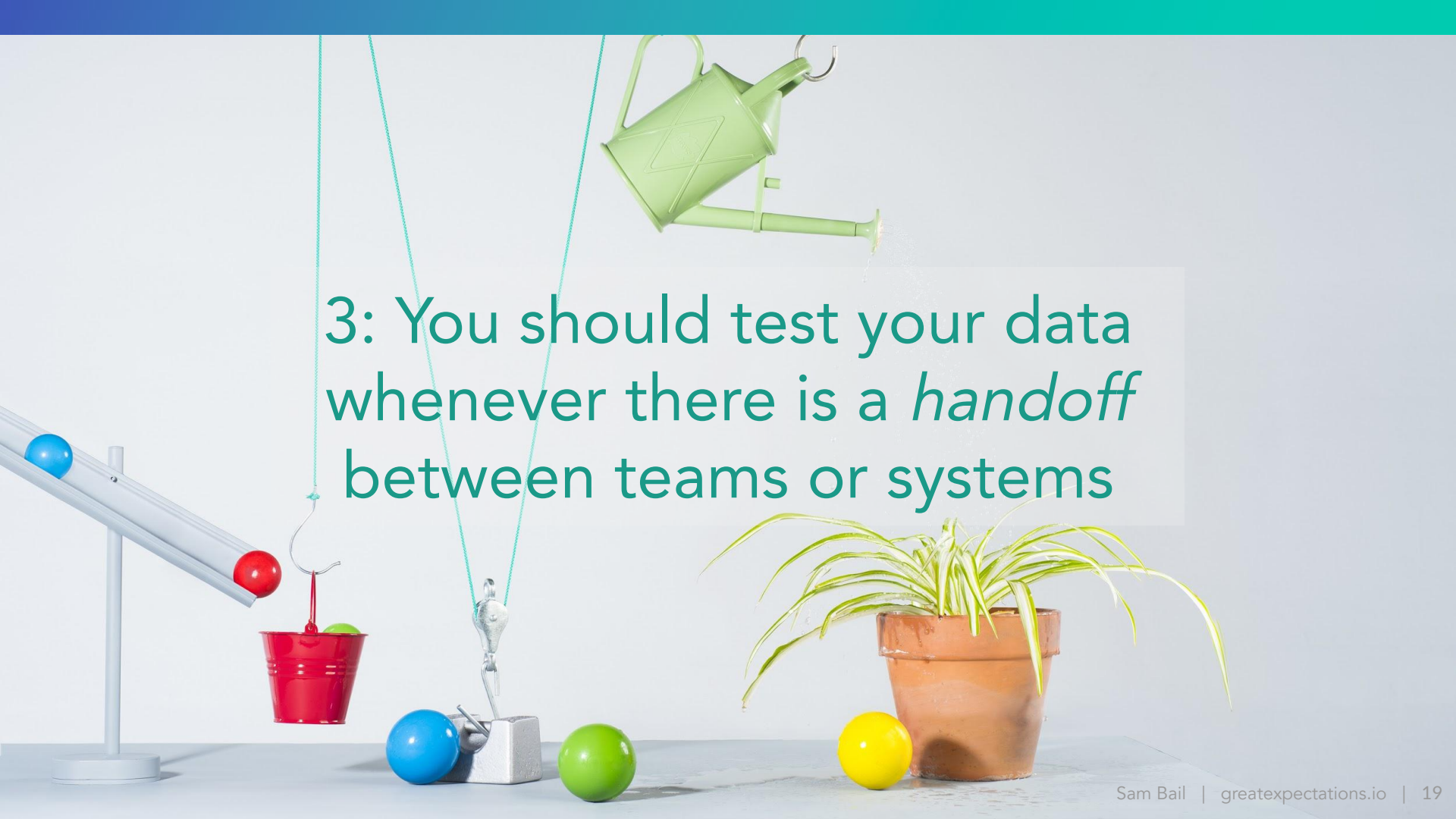
pickup_datetime

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

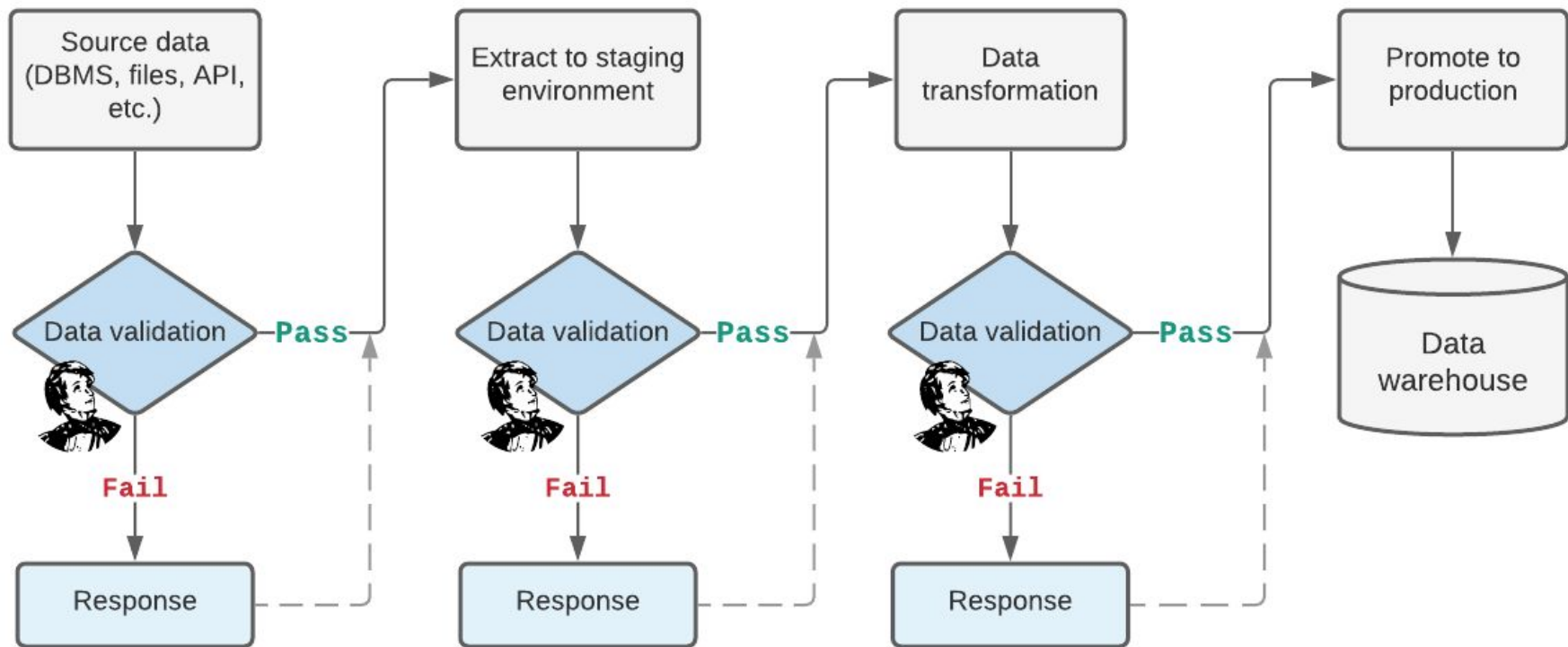


Ok, now back to our stack: How does this all fit together?



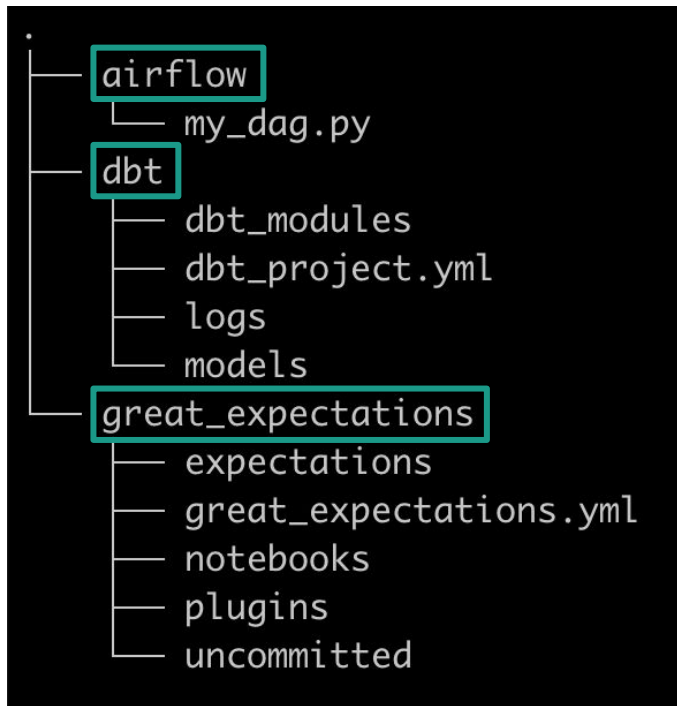
A Rube Goldberg-style contraption is shown against a light blue background. At the top, a green watering can is suspended by a string and is tipping over, pouring water. Below it, a red bucket is suspended by a string and is also tipping over, pouring water. A pulley system is visible in the center, with a string running over a pulley and a small metal piece attached to it. To the left, a ramp with a blue ball and a red ball is shown. To the right, a potted plant with long, thin leaves is shown. The text "3: You should test your data whenever there is a *handoff* between teams or systems" is overlaid on the image.

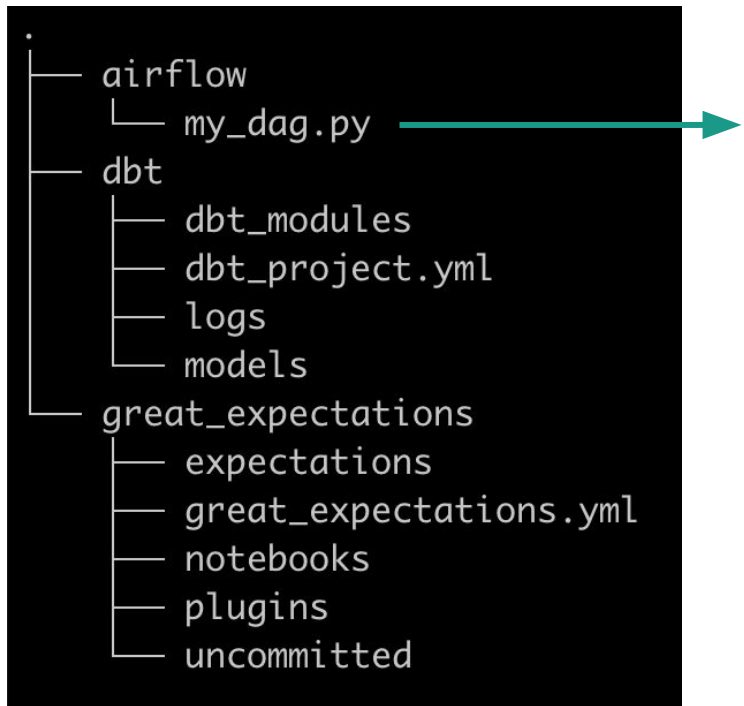
3: You should test your data
whenever there is a *handoff*
between teams or systems



Ok, *now* we're back to the stack.







```
task_validate_source_data = GreatExpectationsOperator(
    ...
)

task_load_source_data = PythonOperator(
    ...
)

task_validate_source_data_load = GreatExpectationsOperator(
    ...
)

task_run_dbt_dag = dbt_run = DbtRunOperator(
    ...
)

task_validate_analytical_output = GreatExpectationsOperator(
    ...
)

task_publish = PythonOperator(
    ...
)
```

DAG: example_ge_dbt_dag

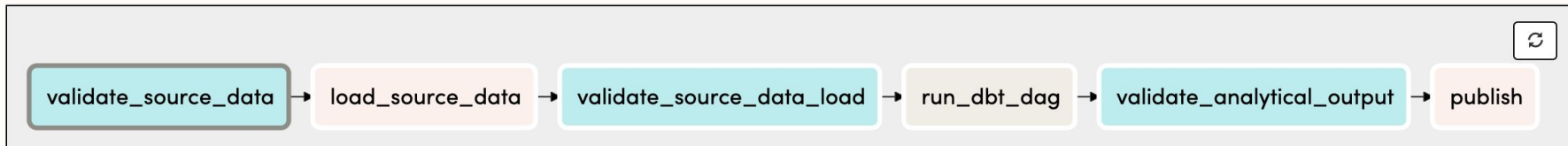
schedule: 1 day, 0:00:00

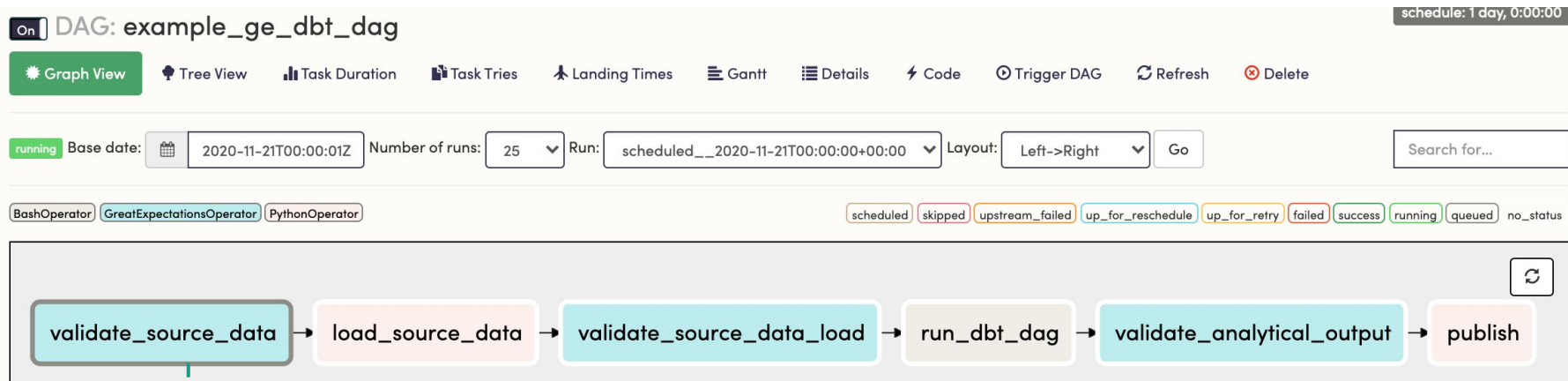
Graph View Tree View Task Duration Task Tries Landing Times Gantt Details Code Trigger DAG Refresh Delete

running Base date: 2020-11-21T00:00:01Z Number of runs: 25 Run: scheduled__2020-11-21T00:00:00+00:00 Layout: Left->Right Go Search for...

BashOperator GreatExpectationsOperator PythonOperator

scheduled skipped upstream_failed up_for_reschedule up_for_retry failed success running queued no_status





Test that source data matches expected format, e.g. correct number of columns, data types, row count “similar” to last month’s, etc.



On DAG: example_ge_dbt_dag

schedule: 1 day, 0:00:00

Graph View

Tree View

Task Duration

Task Tries

Landing Times

Gantt

Details

Code

Trigger DAG

Refresh

Delete

running

Base date:



2020-11-21T00:00:01Z

Number of runs:

25

Run:

scheduled__2020-11-21T00:00:00+00:00

Layout:

Left->Right

Go

Search for...

BashOperator GreatExpectationsOperator PythonOperator

scheduled skipped upstream_failed up_for_reschedule up_for_retry failed success running queued no_status

validate_source_data

load_source_data

validate_source_data_load

run_dbt_dag

validate_analytical_output

publish



Test that source data load was successful, e.g. no rows lost compared to source

On DAG: example_ge_dbt_dag schedule: 1 day, 0:00:00


Graph View Tree View Task Duration Task Tries Landing Times Gantt Details Code Trigger DAG Refresh Delete

running Base date: 2020-11-21T00:00:01Z Number of runs: 25 Run: scheduled__2020-11-21T00:00:00+00:00 Layout: Left->Right Go Search for...

BashOperator GreatExpectationsOperator PythonOperator

scheduled skipped upstream_failed up_for_reschedule up_for_retry failed success running queued no_status

```
graph LR; validate_source_data --> load_source_data; load_source_data --> validate_source_data_load; validate_source_data_load --> run_dbt_dag; run_dbt_dag --> validate_analytical_output; validate_analytical_output --> publish;
```

 **dbt** Run tests during DAG *development* to check for integrity of transformations

On DAG: example_ge_dbt_dag schedule: 1 day, 0:00:00


Graph View Tree View Task Duration Task Tries Landing Times Gantt Details Code Trigger DAG Refresh Delete


running Base date: 2020-11-21T00:00:01Z Number of runs: 25 Run: scheduled__2020-11-21T00:00:00+00:00 Layout: Left->Right Go Search for...

BashOperator GreatExpectationsOperator PythonOperator

scheduled skipped upstream_failed up_for_reschedule up_for_retry failed success running queued no_status

```
graph LR; validate_source_data --> load_source_data; load_source_data --> validate_source_data_load; validate_source_data_load --> run_dbt_dag; run_dbt_dag --> validate_analytical_output; validate_analytical_output --> publish;
```

 **dbt** Test integrity of transformations, e.g. no fan-out joins, no NULL columns, etc.

 Use off-the-shelf methods for complex tests, e.g. distributions of values - and generate Data Docs

Test your data 🙌
In multiple places 🙌
With different types of tests 🙌

Thank you!



Looking forward to chatting \o/