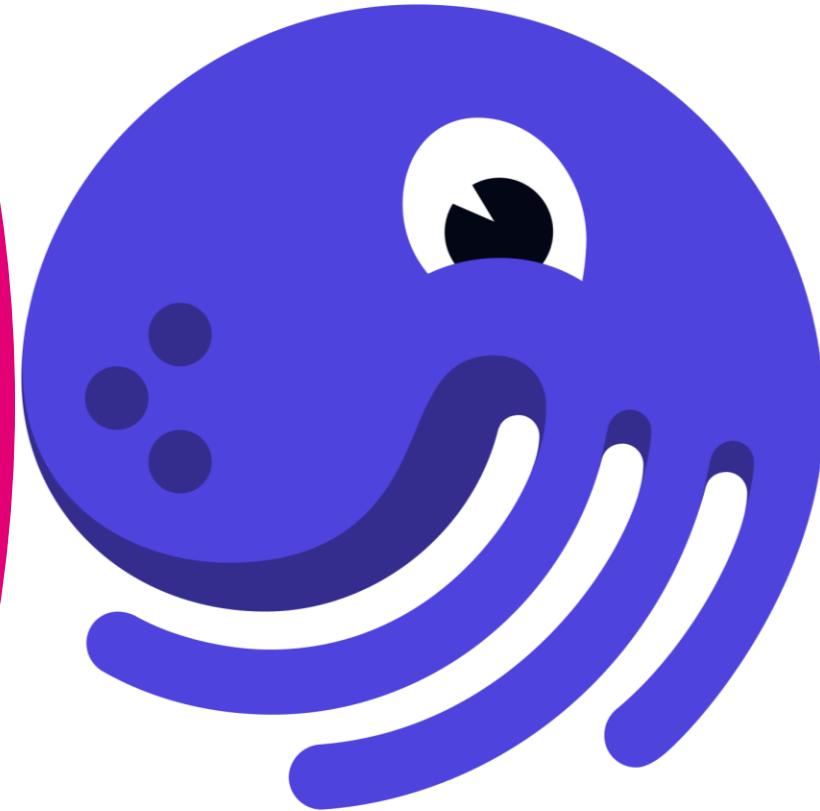


Introduction to using and scaling dagster

Aleksandar Milicevic
Georg Heiler



Agenda

- 1 Welcome and introduction to**
 - What is a data platform
 - What is the role of dagster
- 2 Crash course on Dagster concepts**
 - Asset based
 - Metadata-created pipelines
 - Resources and IO managers
- 3 Hands-on lab** (local or GitHub Codespace)
 - Spin up the environment
 - Tour Dagster UI and run first asset
 - Understand, run and extend the above Dagster concepts
- 4. Dagster @Magenta**
 - Open source implementation with local-data-stack

QA & Wrap-up

About us



Data expert in academia and industry Magenta Telecom

- meetup organizer and conference speaker
- data architecture, multimodal and complex data challenges

 geoheil

 geoheil

 @geoheil.com



Data engineer at Magenta Telecom

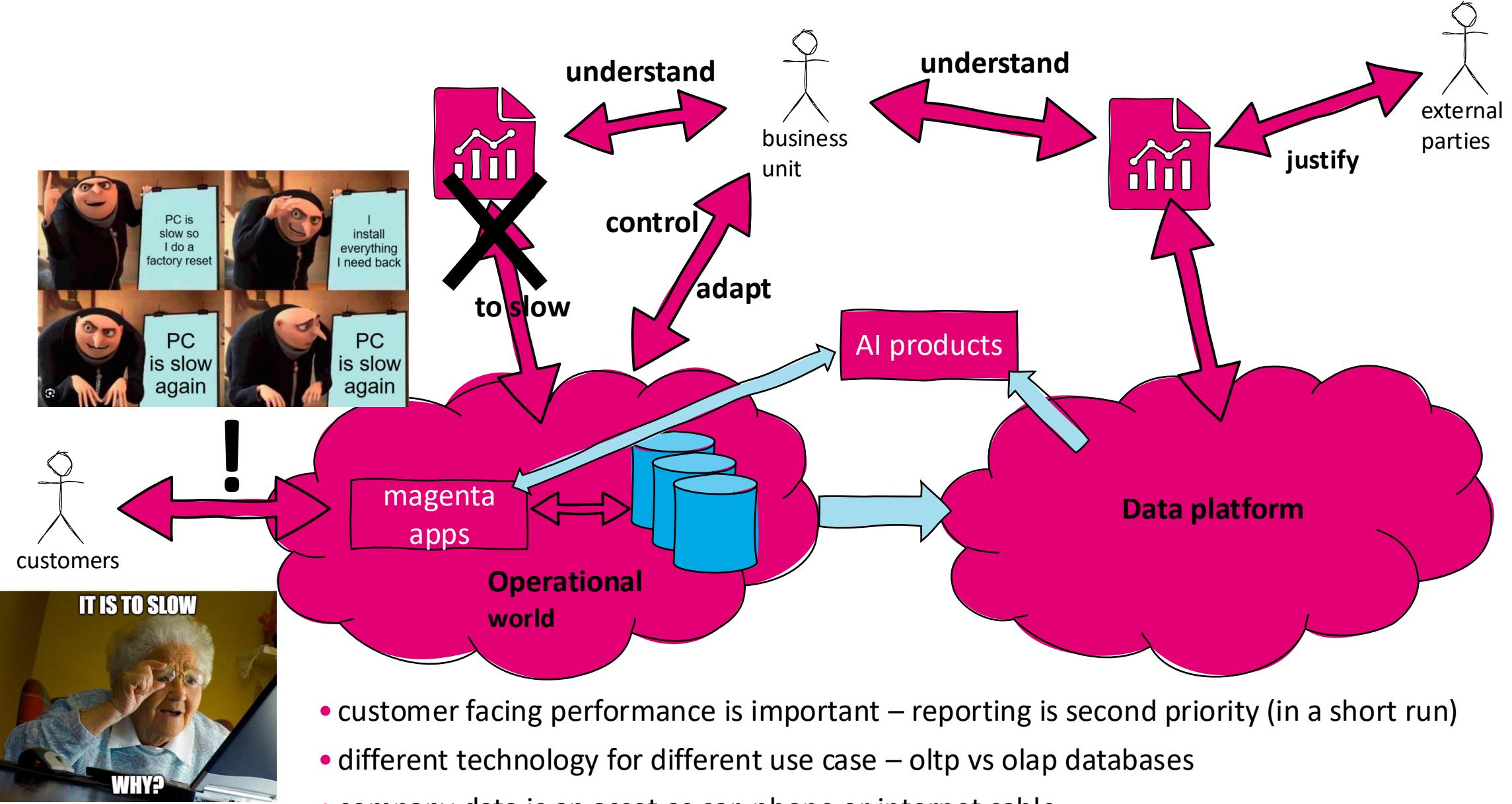
- database internals, data platform engineering

 milicevica23

 milicevica23

 @milicevica23.bsky.social

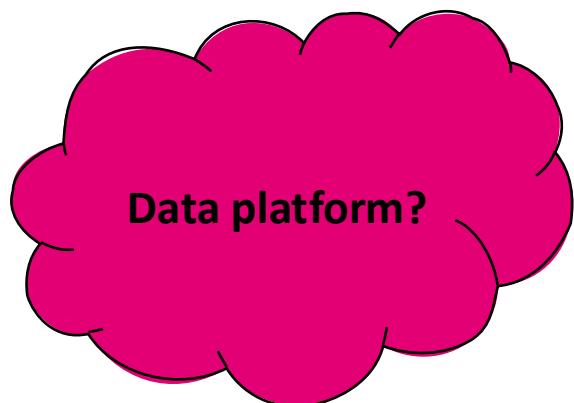
Data platform is there to help understand business process improve service



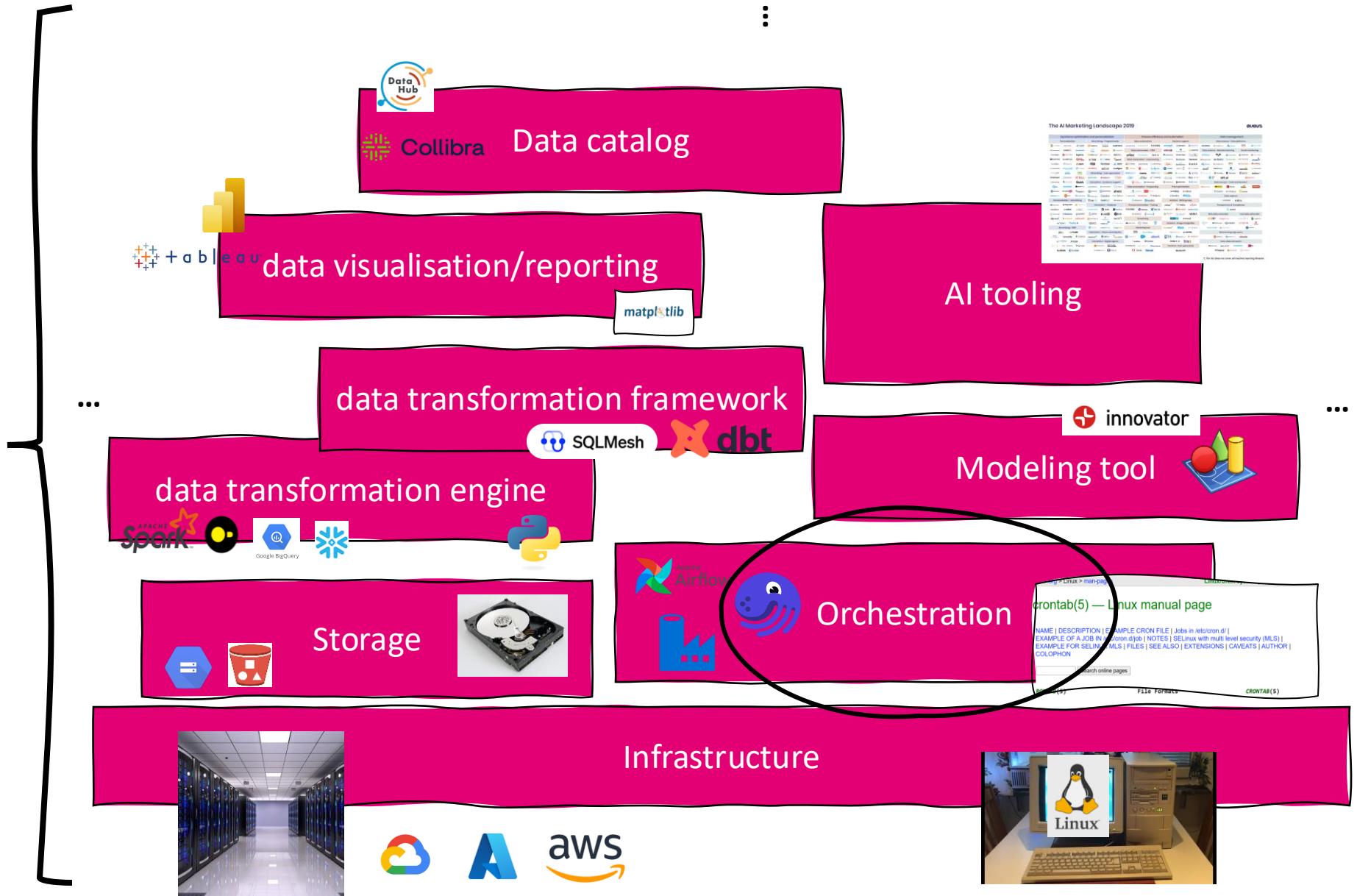
Data platform components

What do I need for a bare minimum E2E reporting?

How to make it last?
Modern tooling, governance, development...

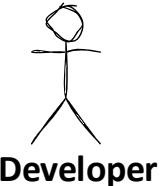


How to make it appealing?
AI...

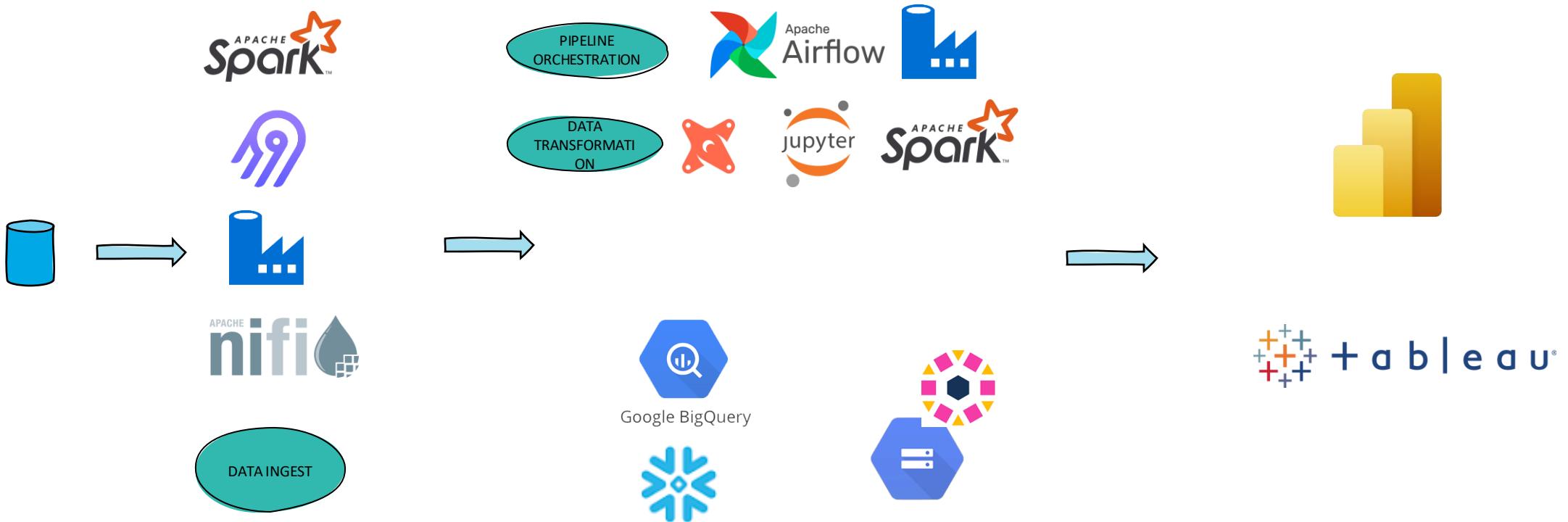
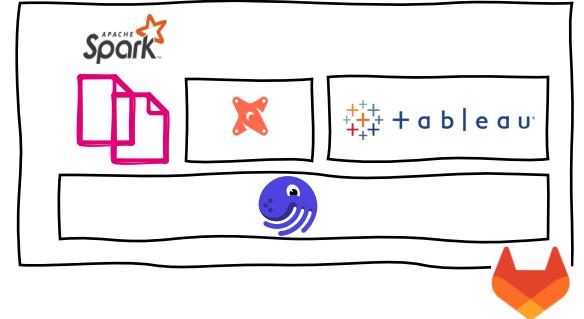
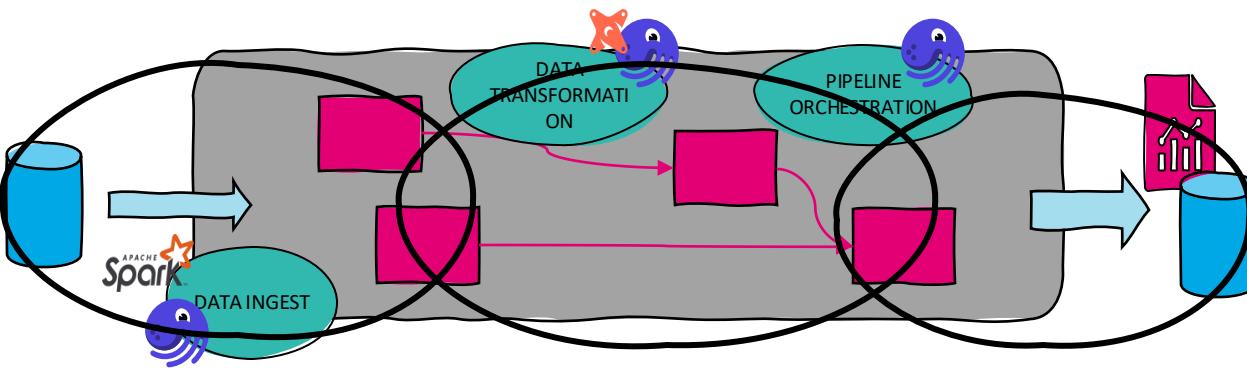


Understanding tool silos

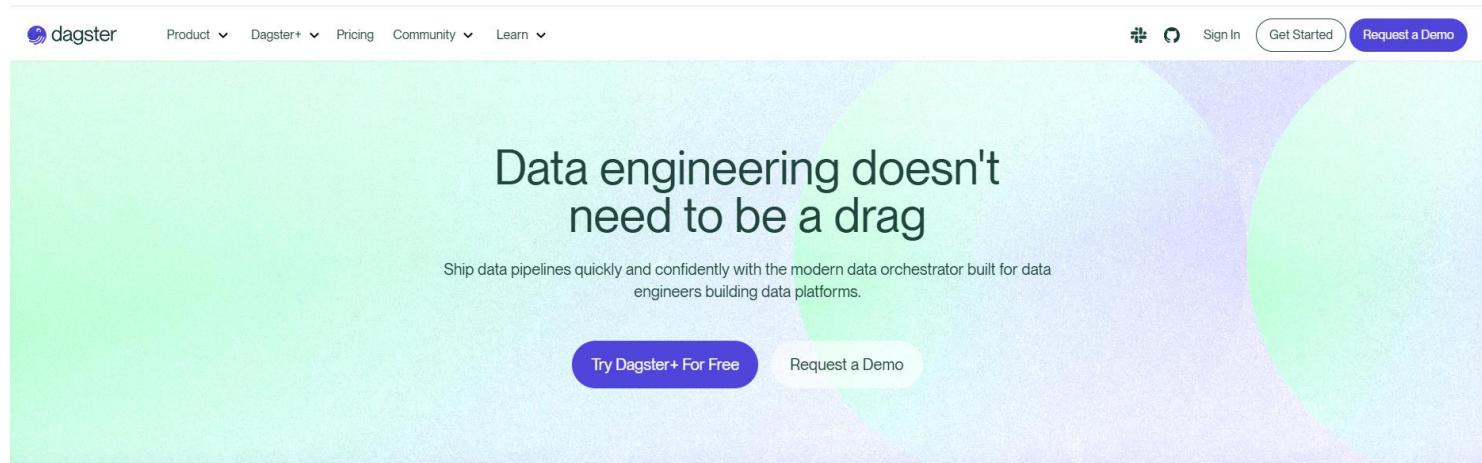
What should i do to get E2E reporting use case done?



Developer



Dagster as the core of the platform



Loved by data platform engineers at fast-growing companies

Read our users' success stories

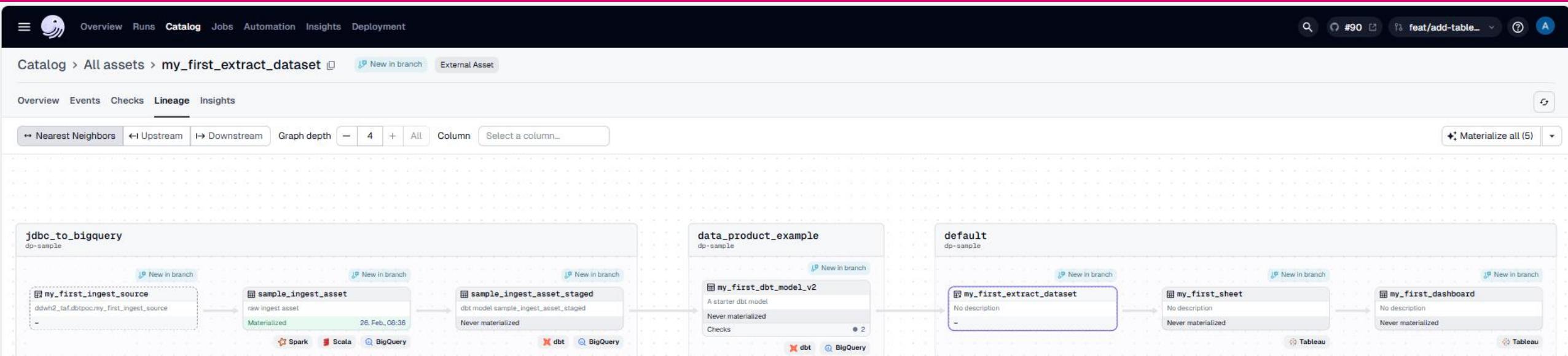


- at Magenta we decided to build around the orchestrator
 - hybrid deployment – controlplane SaaS – runtime in our k8s
 - software engineering best practices for project development and deployment
 - asset-based mindset for data flows (graph like a calculator for data dependencies)
- new concepts in orchestration...

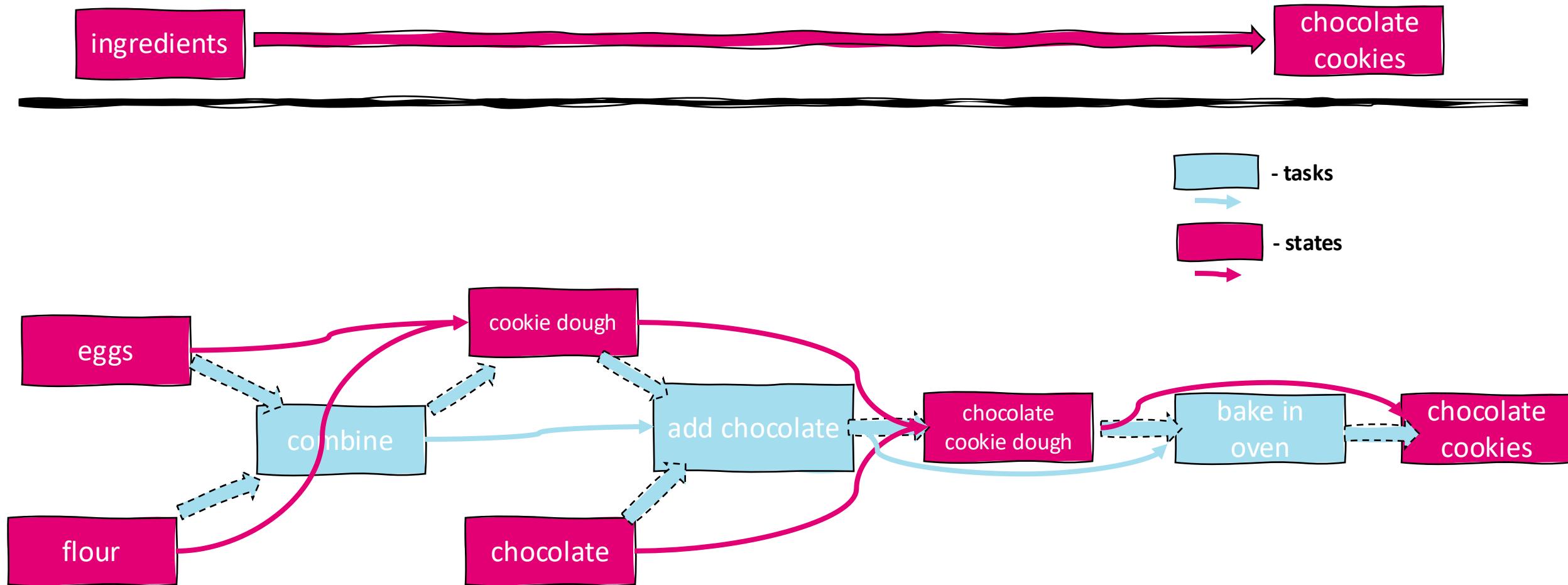
New enabled concepts

docs.dagster.io/getting-started/concepts

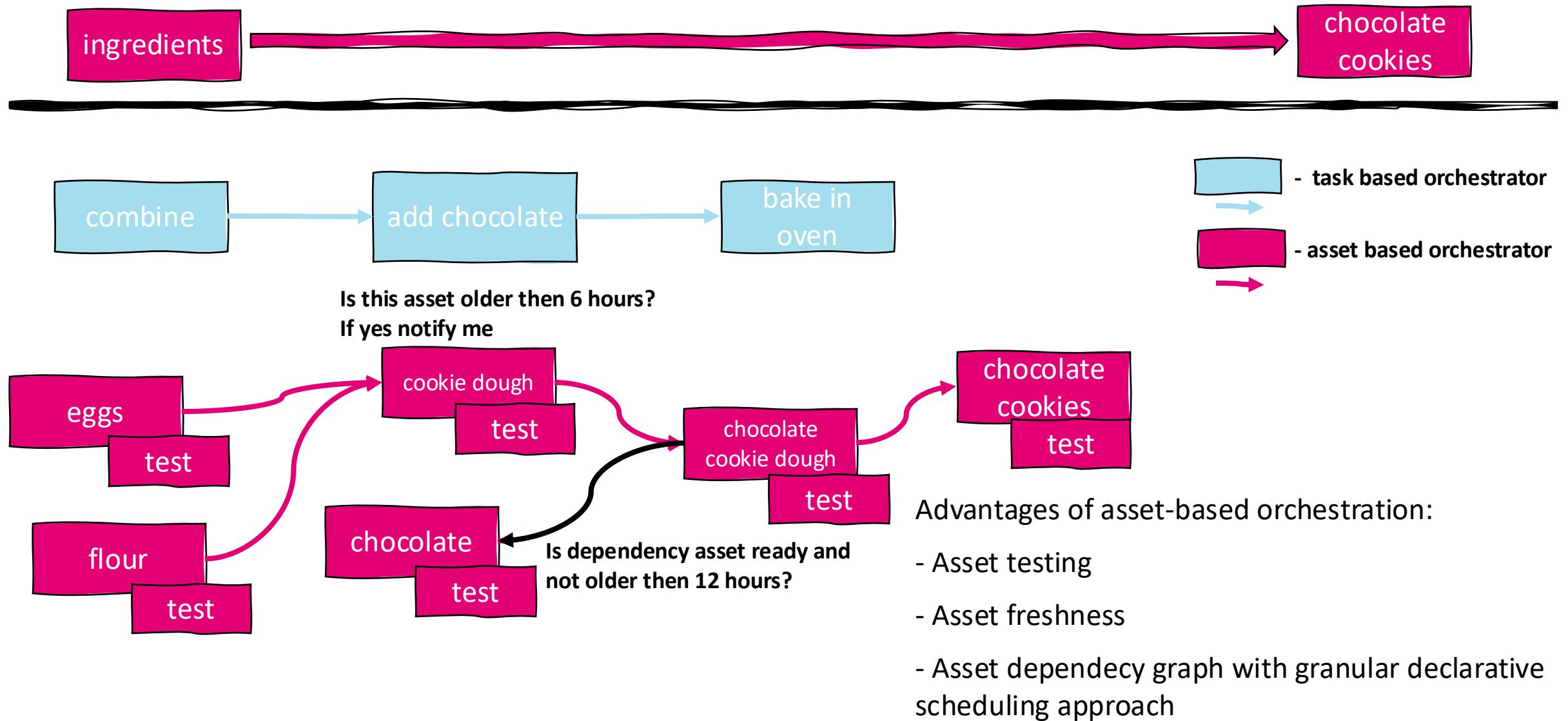
- Asset based graph
- Metadata driven pipeline creation
- Reusable objects (resource, io manager)



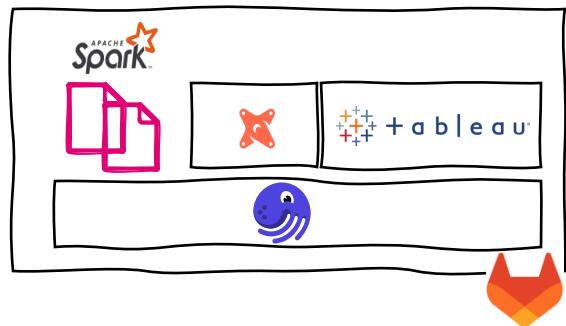
Asset and Task based orchestration: Chocolate cookie example



Asset based orchestration



Machine-readable metadata pipeline generation



```
● ● ●
1 metadata = collect_metadata()
2
3 list_of_assets = []
4 for each_metadata_node in metadata:
5     #.. use_metadata
6     @asset
7     def asset():
8         #..use_metadata
9
10    list_of_assets.append(asset)
11
12 Definitions(assets = list_of_assets)
```

```
Global Ass
1 config
2 list
3 for
4
5
6
7
8
9
10
11
12
13
14 Definitions
15
16
17
18
19
20
21
22
23 Definitions(assets = list_of_assets)

1 tableau_server = TableauServer(creds)
2 #send rest api call to tableau server and get information
3 tableau_metadata = tableau_server.get_metadata()
4 list_of_assets = []
5 for each_tableau_object in tableau_metadata:
6     tableau_object_deps = each_tableau_object.deps
7     tableau_object_name = each_tableau_object.deps
8     if each_tableau_object.type == extract_datasource:
9         @asset(
10             name = tableau_object_name,
11             deps = tableau_object_deps
12         )
13     def refresh_extract(tableau_server):
14         #send api call to refresh object
15         tableau_server.refresh_extract(tableau_object_name)
16         list_of_assets.append(run_dbt_asset)
17     else:
18         @asset(name = tableau_object_deps)
19         def asset():
20             pass
21         list_of_assets.append(asset)
22
23 Definitions(assets = list_of_assets)
```

Reusable components

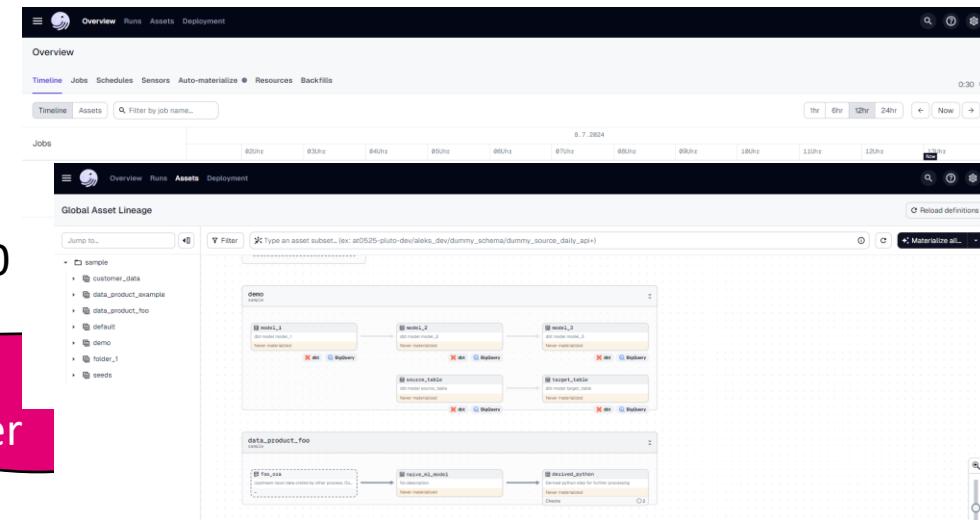
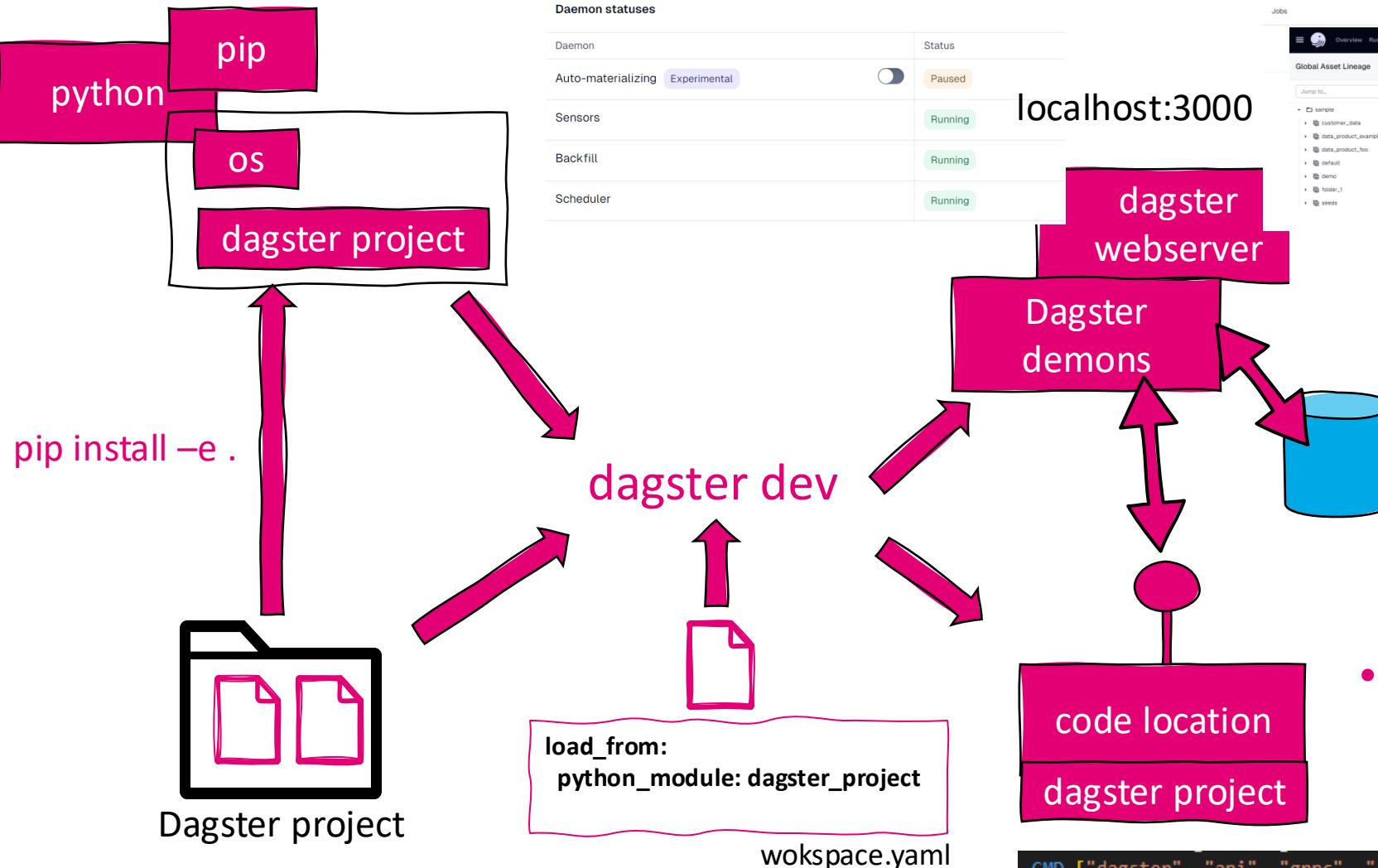
- Reusable objects
 - Resources → Encapsulate complex logic to interact with external systems
 - IO manager → Make complex IO interactions substitutable & testable
- Benefits
 - Dependency injection
 - Day 1 productivity: Scale the data pipeline down to a single laptop
 - Increase self-service: Business/DS focus not required to handle complex IO

```
2  @asset(
3      io_manager_key="bigquery_io_manager",
4  )
5  def awesome_ml_model(context, reference_addresses: pd.DataFrame, bigquery: BigQueryResource) -> pd.DataFrame:
6      # simple normal python code here
7      # IO is abstracted
8      context.log.info(f"from source: \n{reference_addresses.head()}")
9      # auth & complexity (imagine web API) is abstracted
10     with bigquery.get_client() as client:
11         job = client.query("select * from example.upstream")
12         query_result = job.result().to_dataframe()
13         context.log.info(f"direct query: \n{query_result.head()}")
14     return pd.DataFrame({"foo": [1,2,3]})
```

Lab - Demo

- basic asset and UI overview
- resources
- metadata pipeline creation
- components

dagster - pip install dagster



- storage for dagster runs and metadata
- sqlite/postgres

- grpc server serving the assets to the dagster deamon
- needs definition object from the module

```
CMD ["dagster", "api", "grpc", "-h", "0.0.0.0", "-p", "4000", "-f", "sample/__init__.py"]
```

Takeaways

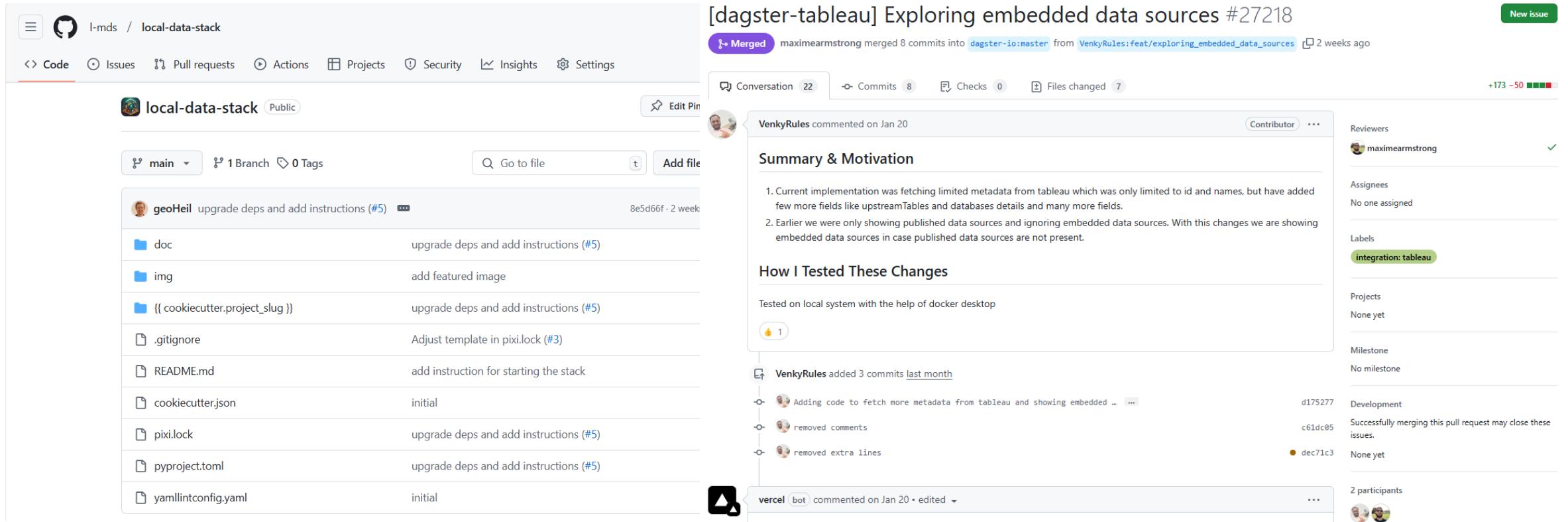
- Integrated asset-based graph is key (from ingest, transformation, reporting, tests – to AI)
 - Event driven connection
 - Better collaboration (scaling)
- Software engineering principles enable business self service
 - Blueprint
 - Automate all the things: CI/CD (stateful & stateless)
 - DRY: build tested foundation – dependency injection
 - Make business departments part of the key processes and pipelines
- Executable specification (metadata, contracts)
 - Interface Management
 - Preserve semantics
 - Preserve compliance (security classification, PII, retention)

Q&A

Dagster @ Magenta

Local Modern Data Stack

Open-source contributions: L-MDS + Tableau integration + VertexAI integration



The image shows a GitHub repository interface. On the left, the repository 'l-mds/local-data-stack' is displayed, showing a list of files and their changes. A pull request by 'geoHeil' is visible, with a commit message 'upgrade deps and add instructions (#5)' and a timestamp '8e5d66f · 2 weeks ago'. On the right, a merged pull request titled '[dagster-tableau] Exploring embedded data sources #27218' is shown. The pull request summary states: '1. Current implementation was fetching limited metadata from tableau which was only limited to id and names, but have added few more fields like upstreamTables and databases details and many more fields. 2. Earlier we were only showing published data sources and ignoring embedded data sources. With this changes we are showing embedded data sources in case published data sources are not present.' The pull request has 22 conversations, 8 commits, 0 checks, and 7 files changed. A comment from 'VenkyRules' on Jan 20 is highlighted, mentioning 'upgrade deps and add instructions (#5)'. A comment from 'vercel bot' on Jan 20 is also shown. The pull request has 173 additions and 50 deletions. The right sidebar shows details like Reviewers (maximearmstrong), Assignees (No one assigned), Labels (integration: tableau), Projects (None yet), and Milestone (None yet). The Vercel bot comment indicates 2 participants.

Building a data platform is team work

Further material

- public live stream about the basics: georgheiler.com/event/magenta-pixi-25
- n-depth Magenta implementation georgheiler.com/event/magenta-data-architecture-25
- <https://courses.dagster.io/>

The screenshot shows the homepage of Dagster University. At the top, there's a purple header with the text "Welcome to Dagster University" and a subtext "Learn how to leverage Dagster and build robust, production-ready data platforms with ease." Below the header, there's a "Courses" section with three cards:

- Dagster & ETL**: A course with 5.0 (1 review). Description: Learn how to ingest data to power your assets. You'll build custom pipelines and see how to use Embedded ETL and Dagster Components...
- Dagster Essentials**: A course with 4.8 (160 reviews). Description: In this course, learn the basics of Dagster, including how to represent a data pipeline as the data assets it produces and orchestrate a...
- Dagster & dbt**: A course with 4.9 (19 reviews). Description: Learn how to integrate and orchestrate dbt projects with Dagster. You'll load dbt models into Dagster as assets, build dependencies,...

The screenshot shows the "Libraries" section of the Dagster documentation. The left sidebar lists various libraries: Airbyte, Airlift, Anthropic, AWS, Azure Data Lake Storage Gen 2, Census, Chroma, Cube, Databricks, Datadog, dbt, Delta Lake, DingTalk, dlt, Docker, DuckDB, and Embedded ELT. The main content area shows cards for each library, each with an icon, the library name, and a brief description.

- Airbyte**: 3 items
- Airlift**: Airlift is a toolkit for integrating Dagster and Airflow.
- Anthropic**: Community. The Anthropic integration allows you to easily interact with the Anthropic REST API using the Anthropic Python API to...
- AWS**: 10 items
- Azure Data Lake Storage Gen 2**: Dagster helps you use Azure Storage Accounts as part of your data pipeline. Azure Data Lake Storage Gen 2 (ADLS2)...
- Census**: Community. With the Census integration you can execute a Census sync and poll until that sync completes, raising an error if it's...