



FROM FABRIC TO FANTASTIC

HOW DBT MAKES LAKEHOUSES AND
WAREHOUSES SHINE

Sam Debruyne

Fabric Global Online Conference
September 2024





Who am I?

Sam Debruyn

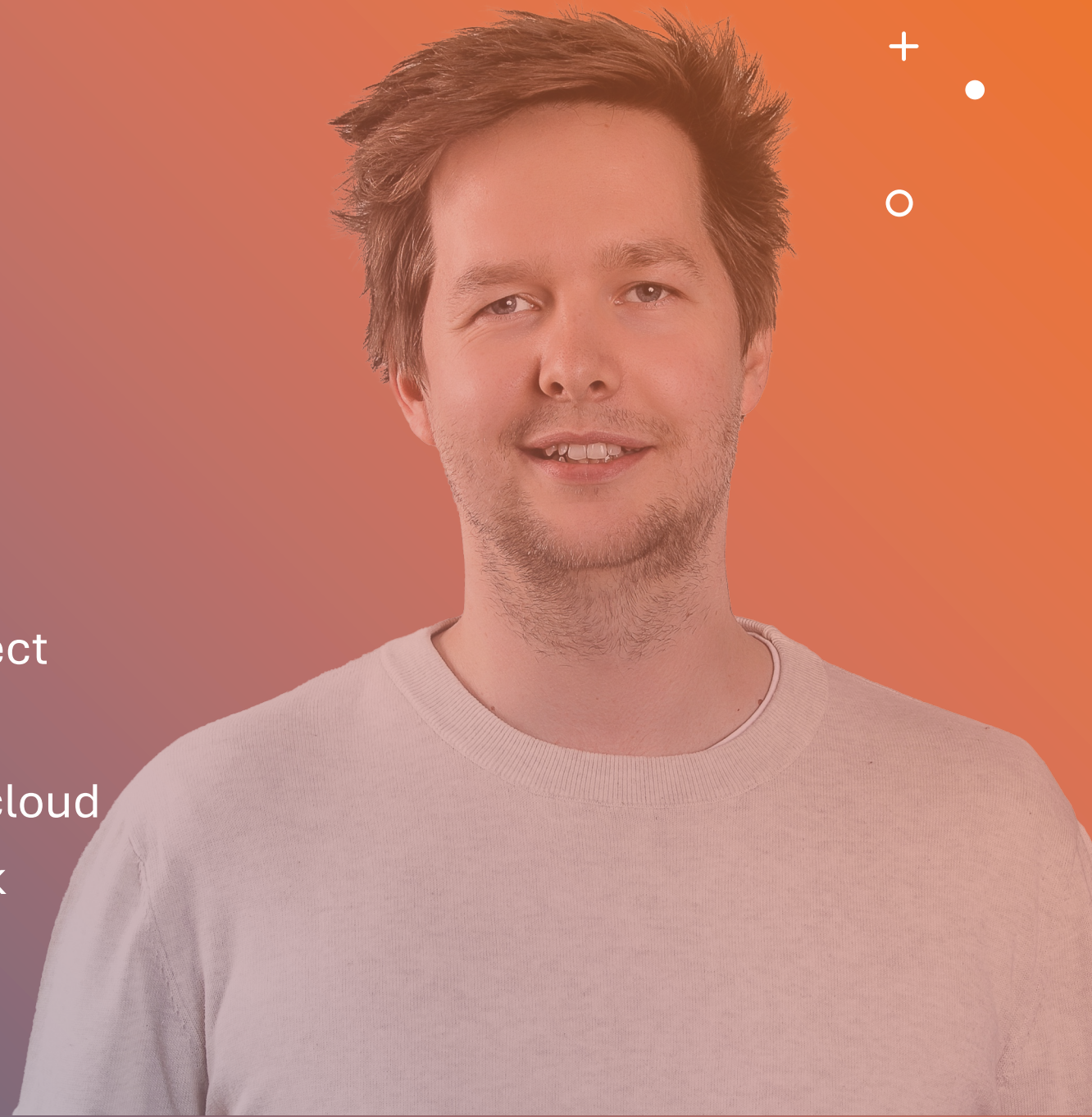
 Heist-op-den-Berg, BE

 Consultant / Data & Cloud Architect

 5 years in data

 10 years in software / architecture / cloud

 dbt, Microsoft, modern data stack

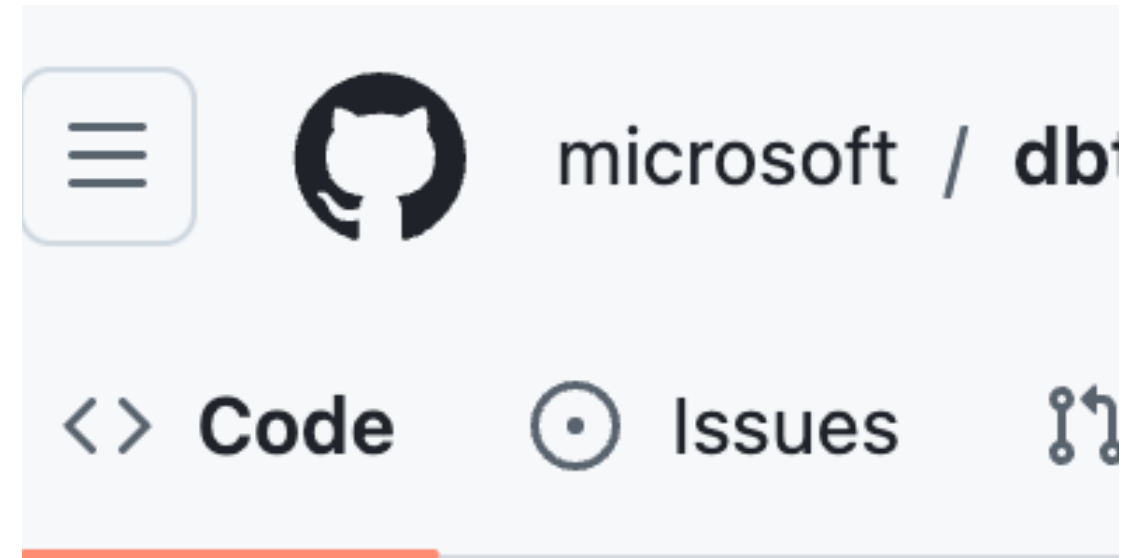


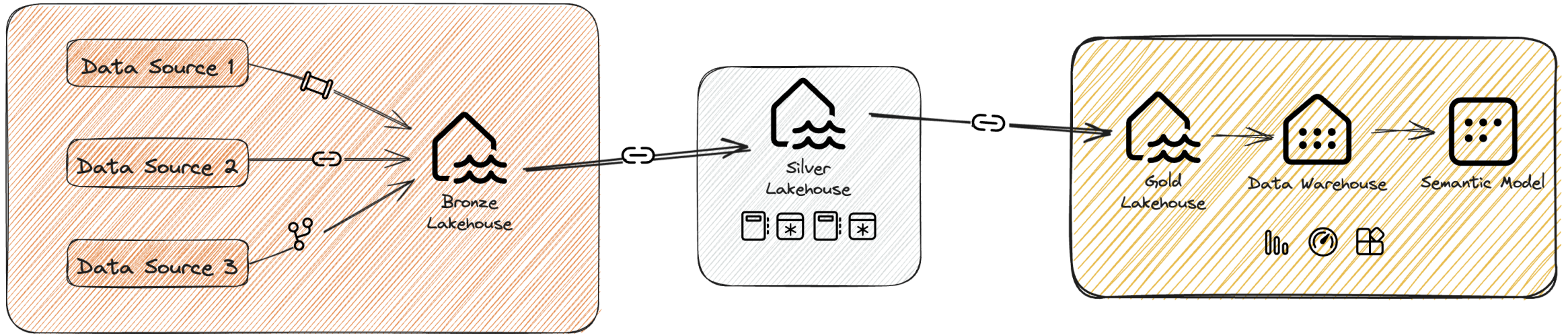
dbt-fabric: a quick lookback

OSS project (dbt adapter) created in 2019 to bring dbt support to SQL Server & Azure SQL

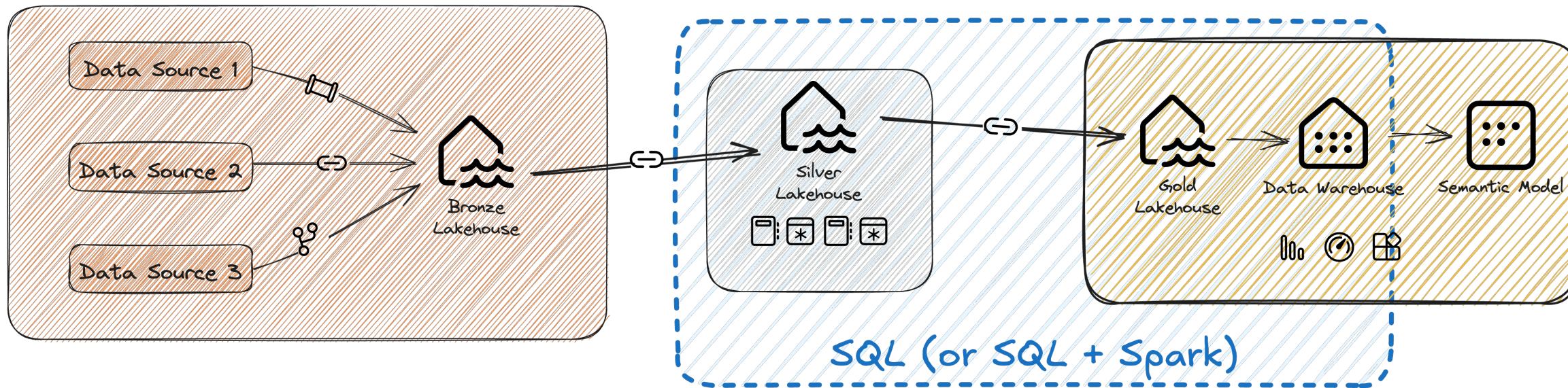
Led maintenance on adapters for SQL Server, Azure SQL, Synapse...

Worked with Microsoft to bring dbt to Fabric





Lakehouse/Warehouse



Where does SQL fit in?

Different ways to transform data



Programming languages

Python and Scala. High learning curves and often creates a boundary between business users and specialized engineers. Very powerful and easy to maintain.

Declarative languages

SQL, SAS, and the likes. Code is easy to write and understand but offers limited flexibility and can be hard to maintain (adopting software eng. best practices).

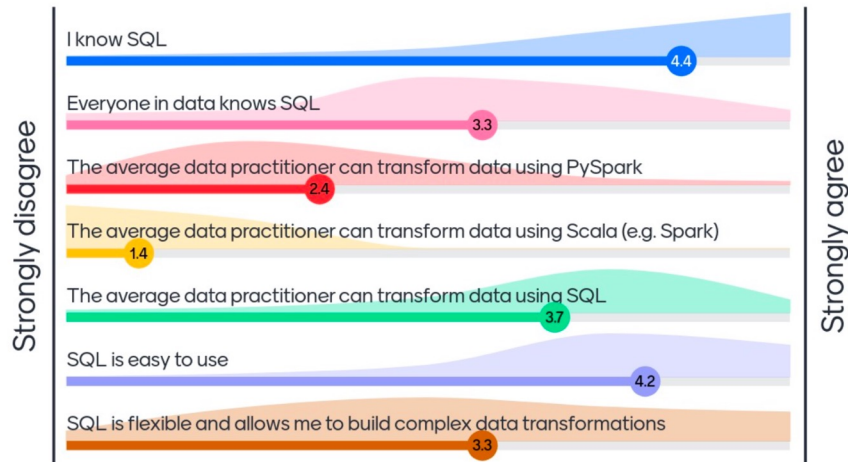
Low-code / UI-based

Easy to adopt, use, and achieve results. Very high vendor lock-in and limited flexibility and modularity.

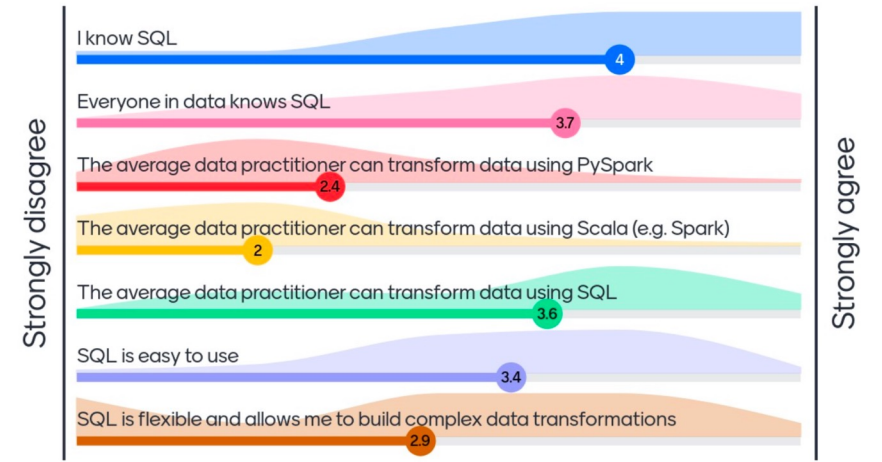
A quick survey done at local meetups



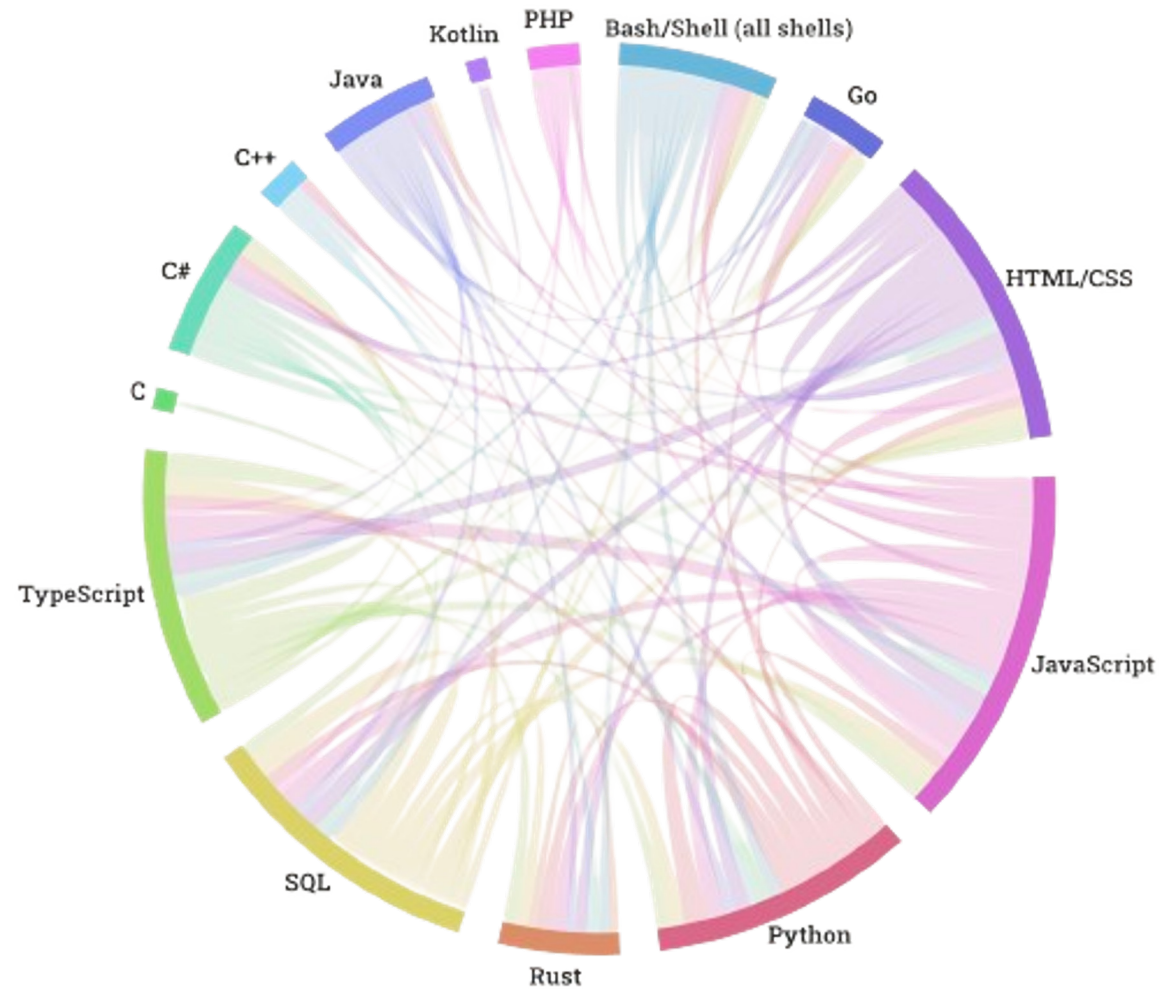
How would you rate these statements?



How would you rate these statements?



Programming languages 2023



source: Stack Overflow
worked with / wants to
work with

The common language of data transformations is **not** drag-and-drop

Data architects

ERWIN
Wherescape

Data engineers

Informatica
Matillion

Analytics engineers

Alteryx
Talend

BI developers

Tableau
Qlik

Analysts

Excel / Sheets
Power BI



The common language of data transformations is **SQL**

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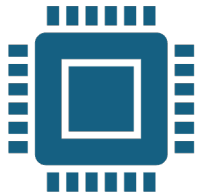
Introducing dbt

Open-source Python utility for building data transformations

Free/OSS version: dbt Core / version with all the bells & whistles included: dbt Cloud

The de facto default tool for analytics engineering

3 things to know



No compute

dbt requires a data warehouse to function, it only sends SQL queries



SQL with Jinja

dbt is built for SQL, in some cases you can also use Python

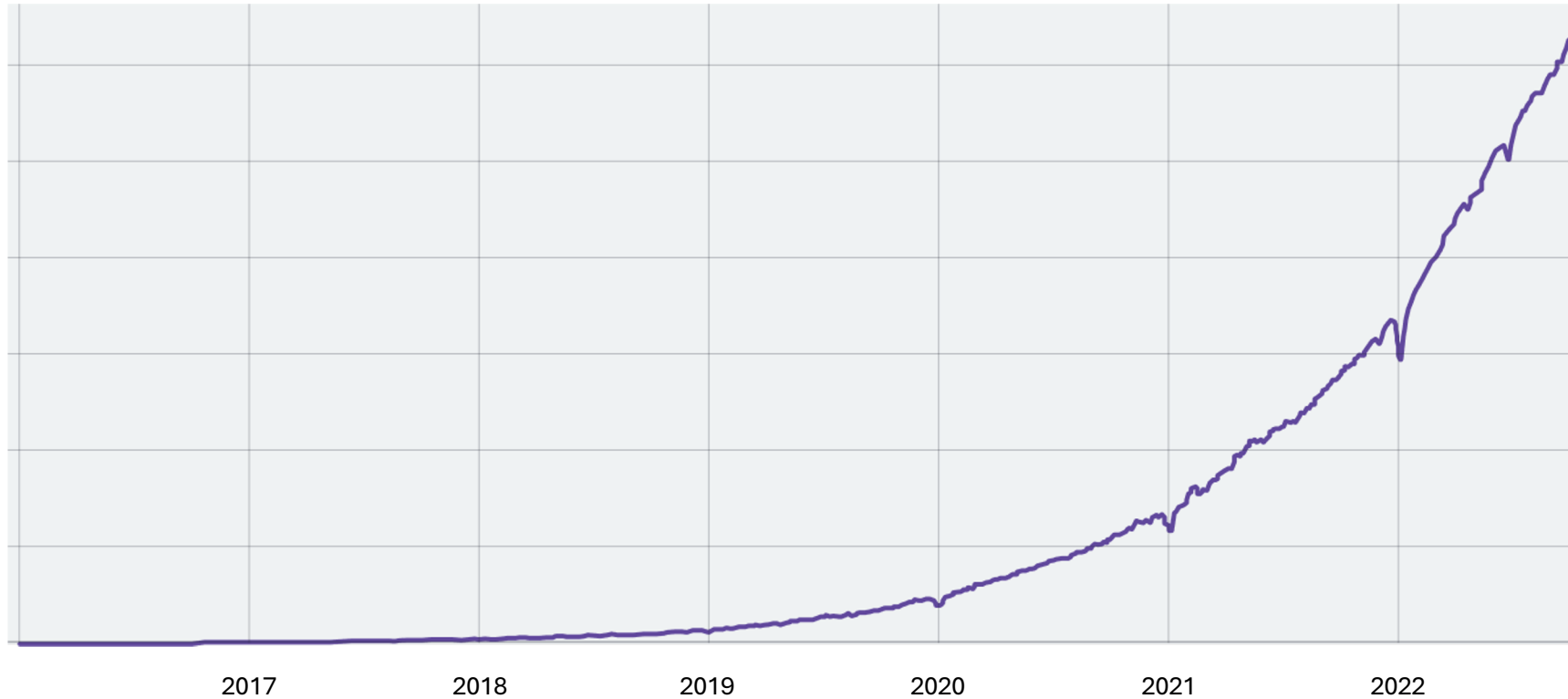


Free/self-hosted or cloud

dbt Core is free but requires "plumbing" (e.g. an orchestrator)

dbt Cloud is paid, but will be cheaper than building everything around it manually

dbt adoption past 6 years



October 2023: 30000+ weekly active projects


```
{% set item_types=["food", "drink"] %}

compute_booleans as (

select
  orders.*,

  order_items_summary.order_cost,
  order_items_summary.order_items_subtotal,
  order_items_summary.count_food_items,
  order_items_summary.count_drink_items,
  order_items_summary.count_order_items,
  {% for type in item_types %}
  case
    when order_items_summary.count_{{ type }}_ite
    else 0
  end as is_{{ type }}_order
  {% if not loop.last %},{% endif %}
  {% endfor %}

from orders

left join
  order_items_summary
  on orders.order_id = order_items_summary.order_id

),
```

Modular development

Write transformations in separate version-controlled files

SQL on steroids with Jinja: control logic, loops

Customize and parametrize with variables

Reusable code blocks with macros

Easy to follow DRY principles

Sources



__sources.yml



models / staging / __sources.yml

```
1  version: 2
2
3  sources:
4    - name: ecom
5      schema: raw
6      description: E-commerce data for the Jaffle Shop
7      freshness:
8        warn_after:
9          count: 24
10         period: hour
11     tables:
12         Generate model
13         - name: raw_customers
14           identifier: customers
15           description: One record per person who has purchased one or more items
16         Generate model
17         - name: raw_orders
18           identifier: orders
19           description: One record per order (consisting of one or more order items)
20           loaded_at_field: ordered_at
21         Generate model
22         - name: raw_items
23           identifier: items
24           description: Items included in an order
25         Generate model
26         - name: raw_stores
27           identifier: stores
28           loaded_at_field: opened_at
```

Manage data sources and monitor data freshness

Sources

stg_customers.sql

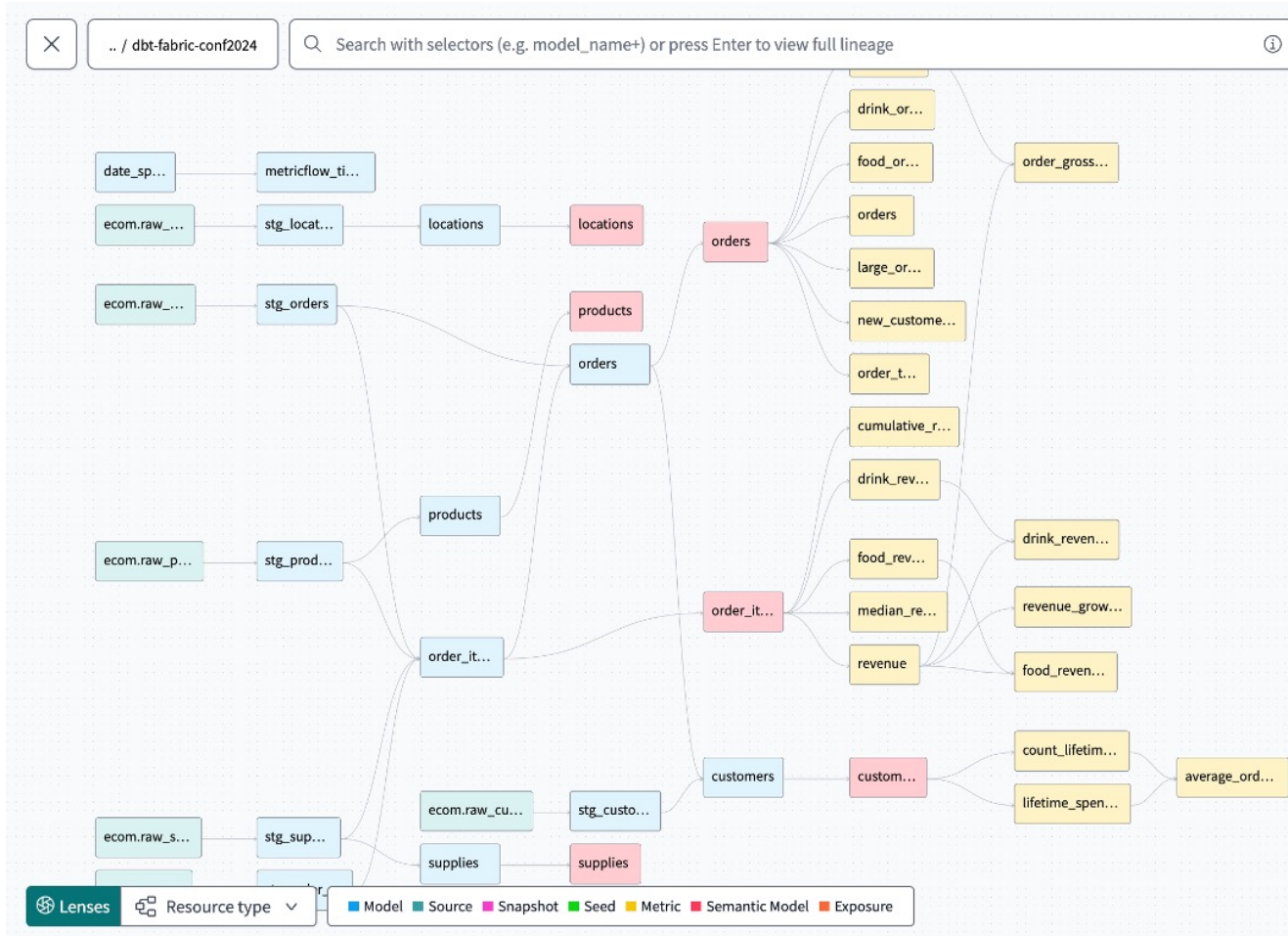
models / staging / stg_customers.sql

```
1  with source as (  
2    |   select *  
3    |   from {{ source('ecom', 'raw_customers') }}  
4  | ),  
5  
6  renamed as (  
7    |   select  
8    |     id as customer_id,  
9    |     name as customer_name  
10   |   from source  
11  | )  
12  
13  select * from renamed  
14  |
```

Dynamic schema
selection

Start tracking
lineage from the
source

Data lineage



Understand the flow of data

Impact of modifying a transformation

How a dimension/fact is constructed

Data lineage



Spot and
detect bad
data model
design

```
er_id ..... [PASS in 0.28s]
customer_id__customer_id__ref_stg_customers_ [RUN]
er_item_id ..... [PASS in 0.29s]
orders_order_items_subtotal_subtotal . [PASS in 24.56s]
mer_id ..... [PASS in 0.32s]
r_items_order_id__order_id__ref_stg_orders_ [RUN]
mer_id ..... [RUN]
er_item_id ..... [RUN]
ion_id ..... [PASS in 0.35s]
..... [RUN]
t_id ..... [PASS in 0.24s]
ustomer_id ..... [RUN]
..... [PASS in 0.16s]
ocation_id ..... [RUN]
y_uuid ..... [PASS in 0.31s]
_order_item_id ..... [RUN]
r_id ..... [PASS in 0.24s]
r_id ..... [RUN]
n_id ..... [PASS in 0.15s]
product_id ..... [RUN]
r_id__customer_id__ref_stg_customers_ [PASS in 0.50s]
pply_uuid ..... [RUN]
id ..... [PASS in 0.16s]
uid ..... [PASS in 0.18s]
..... [PASS in 1.34s]
m_id ..... [PASS in 1.37s]
der_id__order_id__ref_orders_ ..... [PASS in 1.48s]
s_order_id__order_id__ref_stg_orders_ [PASS in 1.43s]
..... [PASS in 1.32s]
_item_id ..... [PASS in 1.50s]
```

Data tests & unit tests

Automated testing for your code, as well as for your data

Tests can be integrated in other tooling to get a good view on your data quality

Simple YAML- or SQL-based syntax to define tests

Documentation and tests



customers.yml

models / marts / customers.yml

```
1  models:
2    - name: customers
3      description: Customer overview data mart, offering key details for each unique customer. One row per customer.
4      tests:
5        - dbt_utils.expression_is_true:
6          | expression: "lifetime_spend_pretax + lifetime_tax_paid = lifetime_spend"
7      columns:
8        - name: customer_id
9          description: The unique key of the customers mart.
10         tests:
11           - not_null
12           - unique
13         - name: customer_type
14           description: Options are 'new' or 'returning', indicating if a customer has ordered more than once or has only placed their first order to date.
15           tests:
16             - accepted_values:
17               | values: ["new", "returning"]
18
```

dbt docs



dbt-fabric-conf2024

Search for resources and columns

Project details

Overview

Performance

Recommendations

Resources File tree Database

Models 14

Sources 6

Tests 27

Exposures

Groups

Metrics 19

Semantic Models 6

Seeds

Macros 732

Snapshots

Dataroots (Partner) / dbt-fabric-conf2024 / Models / customers

customers

Open in IDE

Share

Last run Sep 8, 2024, 2:10 PM CEST View

General Code Columns NEW Performance Recommendations 0

Search for columns

customer_id VARCHAR

The unique key of the orders mart.

NOT NULL

UNIQUE

customer_name VARCHAR

Customers' full name.

count_lifetime_orders INT

Total number of orders a customer has ever placed.

first_ordered_at DATE

The timestamp when a customer placed their first order.

last_ordered_at DATE

The timestamp of a customer's most recent order.



Clear convention-based data documentation



Good step-up to a data catalog



dbt packages: don't reinvent the wheel

Similar to libraries in software development

Benefit from global knowledge by using pre-built common data transformations and data modelling techniques

Share publicly or privately within your organization

Can contain models (transformations), macros, tests, ...

Date dimension in 1 line



models / marts / date_dimension.sql



Save

```
1  {% set sql_stmt %}  
2  |   select {{ dateadd(datepart="year", interval=1, from_date_or_timestamp=current_timestamp()) }} as val  
3  | {% endset %}  
4  | {{ dbt_date.get_date_dimension('2017-01-01', dbt_utils.get_single_value(sql_stmt)) }}
```

Preview

Compile

Build



Format

Results

Code quality

Compiled code

Lineage

15.2s | Results limited to 500 rows. ⓘ

Change row display

Download CSV

date_day	prior_date_day	next_date_day	prior_year_date_day	prior_year_over_year...	day_of_week	day_of_week_iso	day_of_week
2017-01-01	2016-12-31	2017-01-02	2016-01-01	2016-01-03	1	7	Sunday
2017-01-02	2017-01-01	2017-01-03	2016-01-02	2016-01-04	2	1	Monday
2017-01-03	2017-01-02	2017-01-04	2016-01-03	2016-01-05	3	2	Tuesday
2017-01-04	2017-01-03	2017-01-05	2016-01-04	2016-01-06	4	3	Wednesday
2017-01-05	2017-01-04	2017-01-06	2016-01-05	2016-01-07	5	4	Thursday
2017-01-06	2017-01-05	2017-01-07	2016-01-06	2016-01-08	6	5	Friday

There is more

Implement SCD with *snapshots*

Incremental loads

Hooks & operations

Run *Python models* through Spark (coming soon on Fabric)

Manage access with *grants*

Track dataset usage in BI & ML with *exposures*

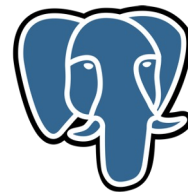
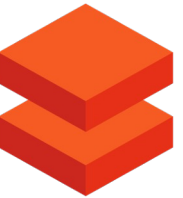
Data contracts

...





Compatibility





Accomplish great things

Version controlled and reproducible

- Collaboration within the team & other teams

Built-in docs & lineage

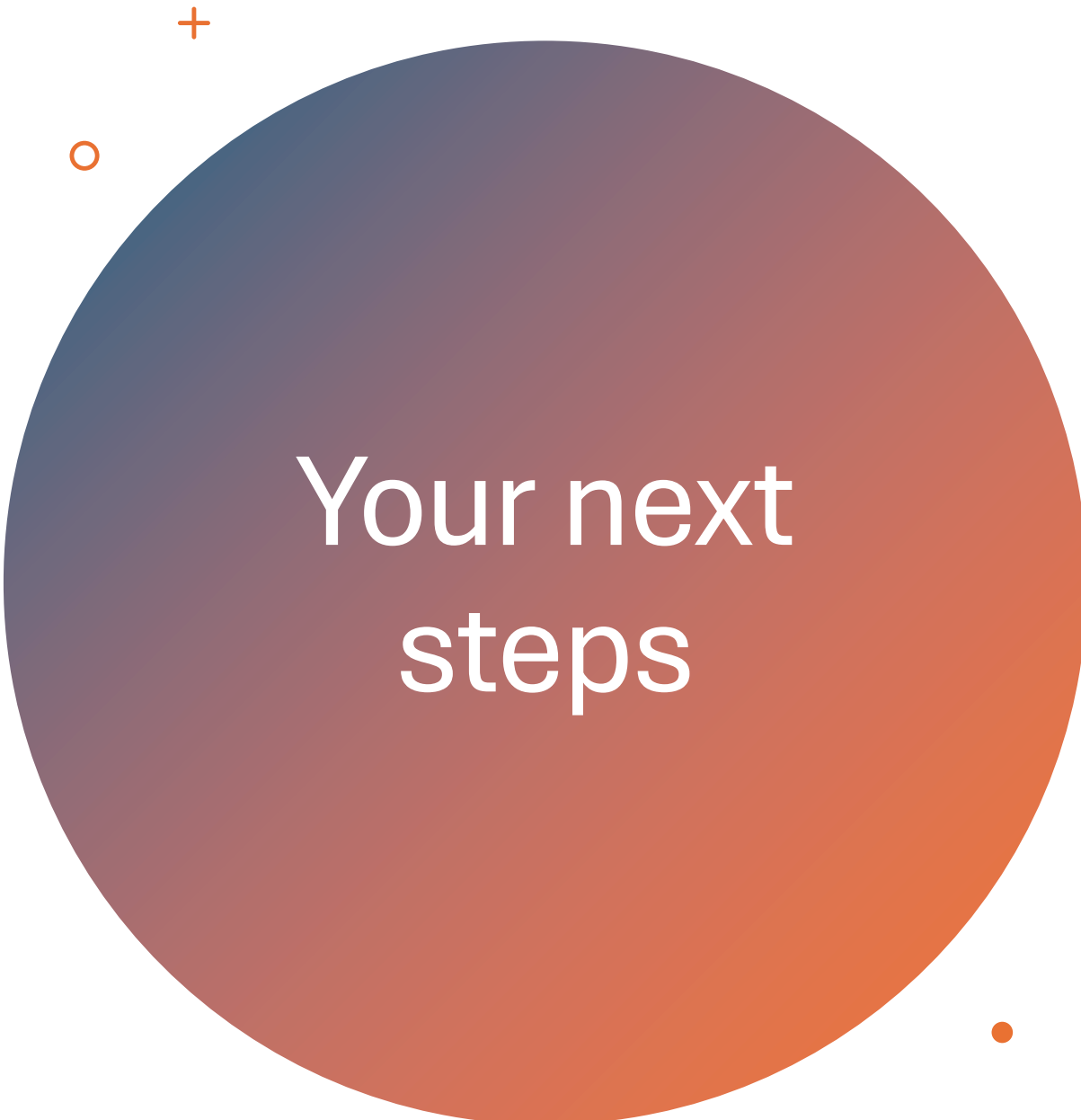
- Know and understand your data

Test code & data

- Deploy & run with confidence

Modular & easy to use

- Easy to extend and maintain
- 



Your next steps


dbt Community: over 100K
members

Active and helpful Slack channels

A lot of development in the open-
source space

Local meetups all over the world

learn.getdbt.com: free courses to
get started with dbt



Questions?



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