Big Data Integration Patterns

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12+ million

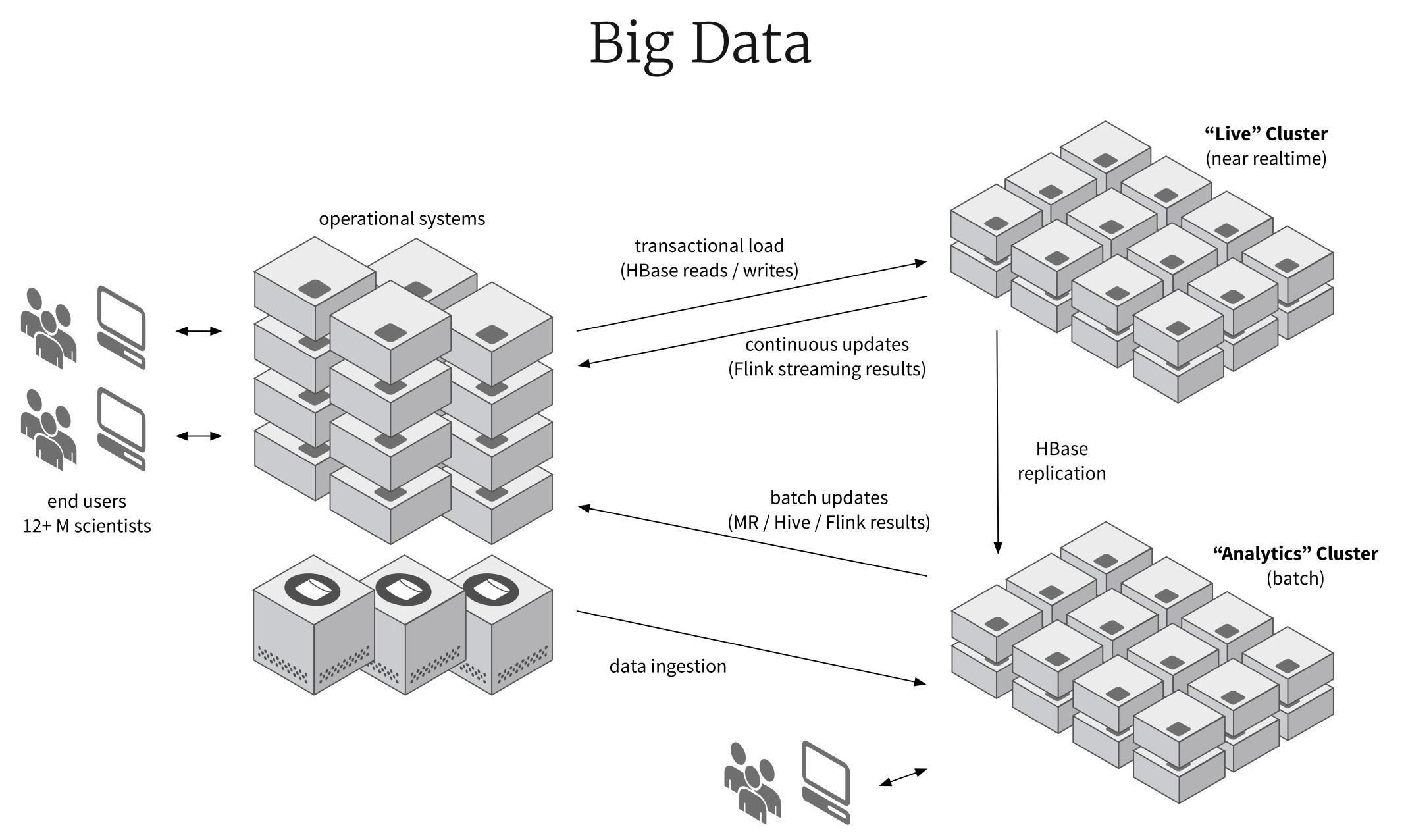
Members

Publications



100+ million

1,500+ million Citations



internal users

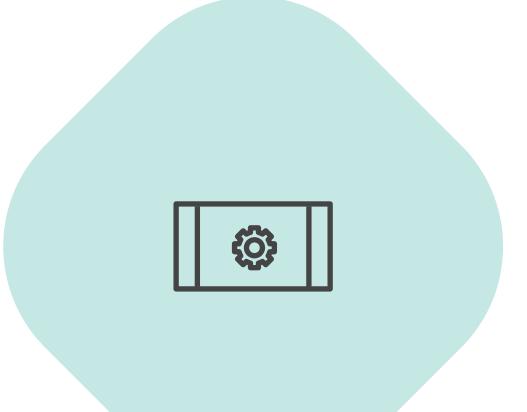




65+

Engineers

Data Ingestion Jobs per Day



370+

3,000+

Yarn Applications per Day



000

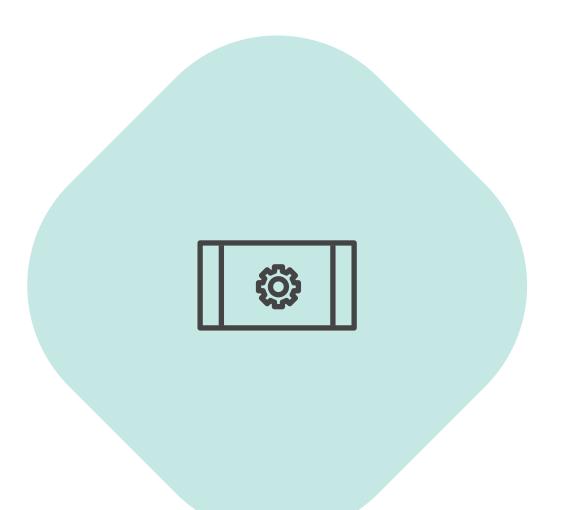
Ease of Maintenance

Ease of Operations

65+

Engineers

Developer Productivity



3,000+

Yarn Applications per Day

Big Data Architecture Integration Patterns & Pricinicples

Integration patterns should be strategic, but also ...

should be driven by use cases

should tackle real world pain points

should **not** be dictated by a **single technology**

Patterns & Principles

Big data is still a fast moving space

Big data **batch processing** today is quite different compared to 5 years ago

Big data stream processing is evolving heavily right now

Big data architecture

must evolve over time

Patterns & Principles

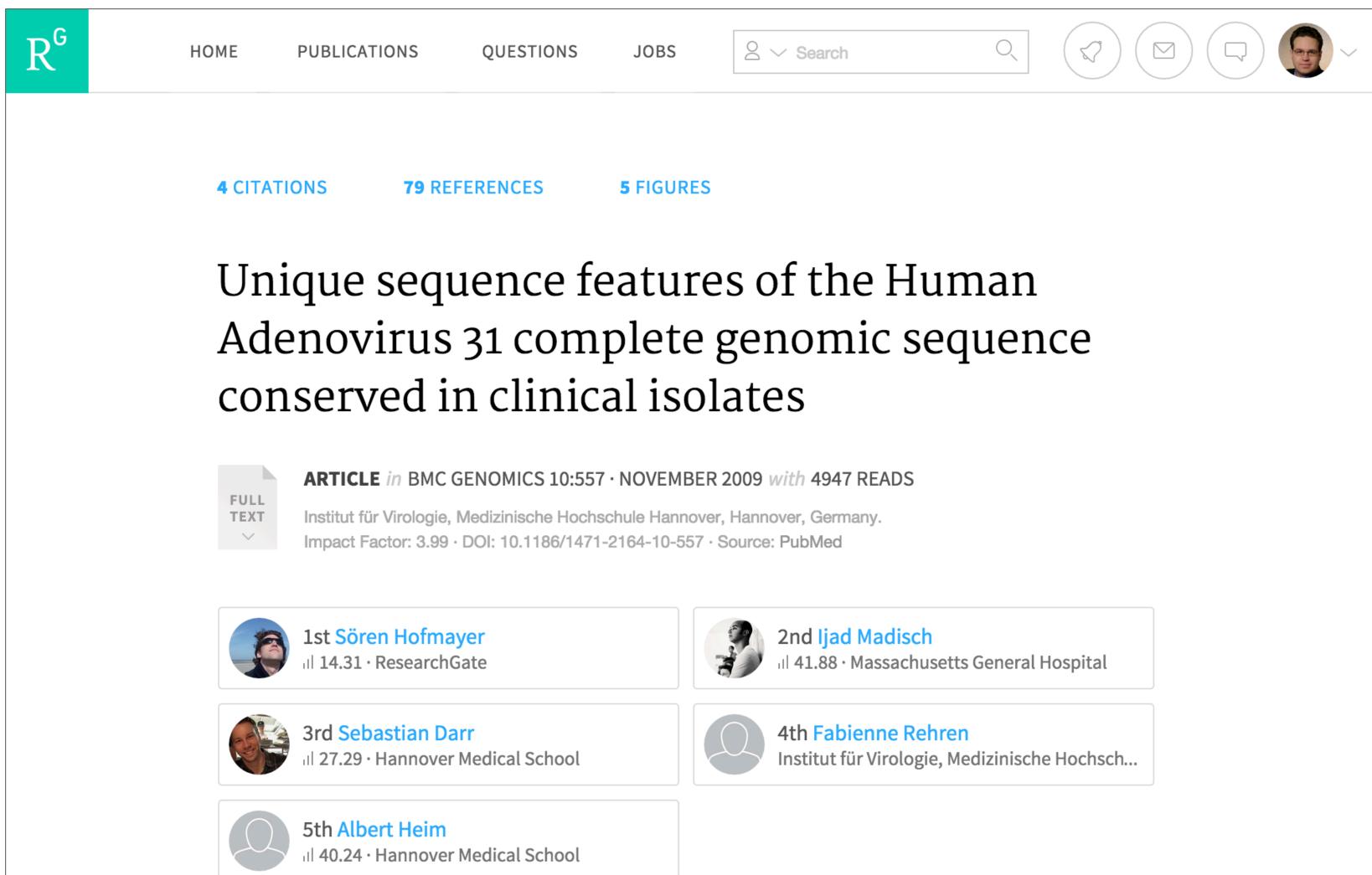
First Big Data Use Case Early 2011, Author Analysis



Author Analysis – Clustering and Disambiguation

<text><text><text><text><text><text><text><text><text><text><text><text><text><text></text></text></text></text></text></text></text></text></text></text></text></text></text></text>	Article: Evidence of Molecular Evolution Driven by Recombination Events Influencing Tropism in a Novel Human Adenovirus That Causes Epidemic Keratoconjunctivitis			
	Robinsor David Sc	Michael P Waleh · Ashish Chintakuntlawar · Christopher M Robinson (Ijad Madisch) Balázs Harrach · Nolan R Hudson · David Schnun · Albert Heim · James Chodosh · Donald Seto · Morris S Jones		
	1k Reads	95 Citations		Article Adeno conse
				Soeren

	Article: Unique sequence features of the Human Adenovirus 31 complete genomic sequence conserved in clinical isolates			
		Soeren Hofmayer · Ijad Madisch · Sebastian Darr · Fabienne Rehren · Albert Heim		
ource	5k Reads	4 Citations		

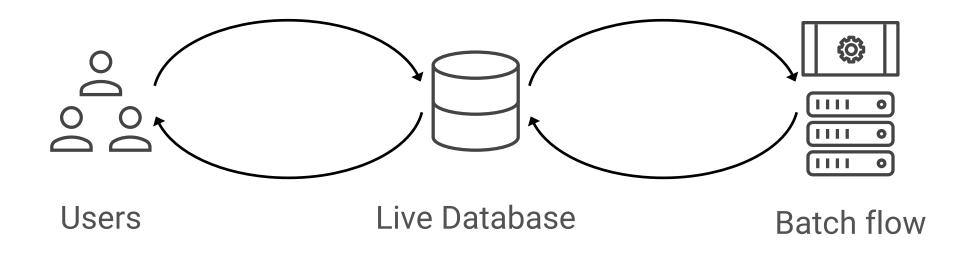


Author Analysis – High Product Impact

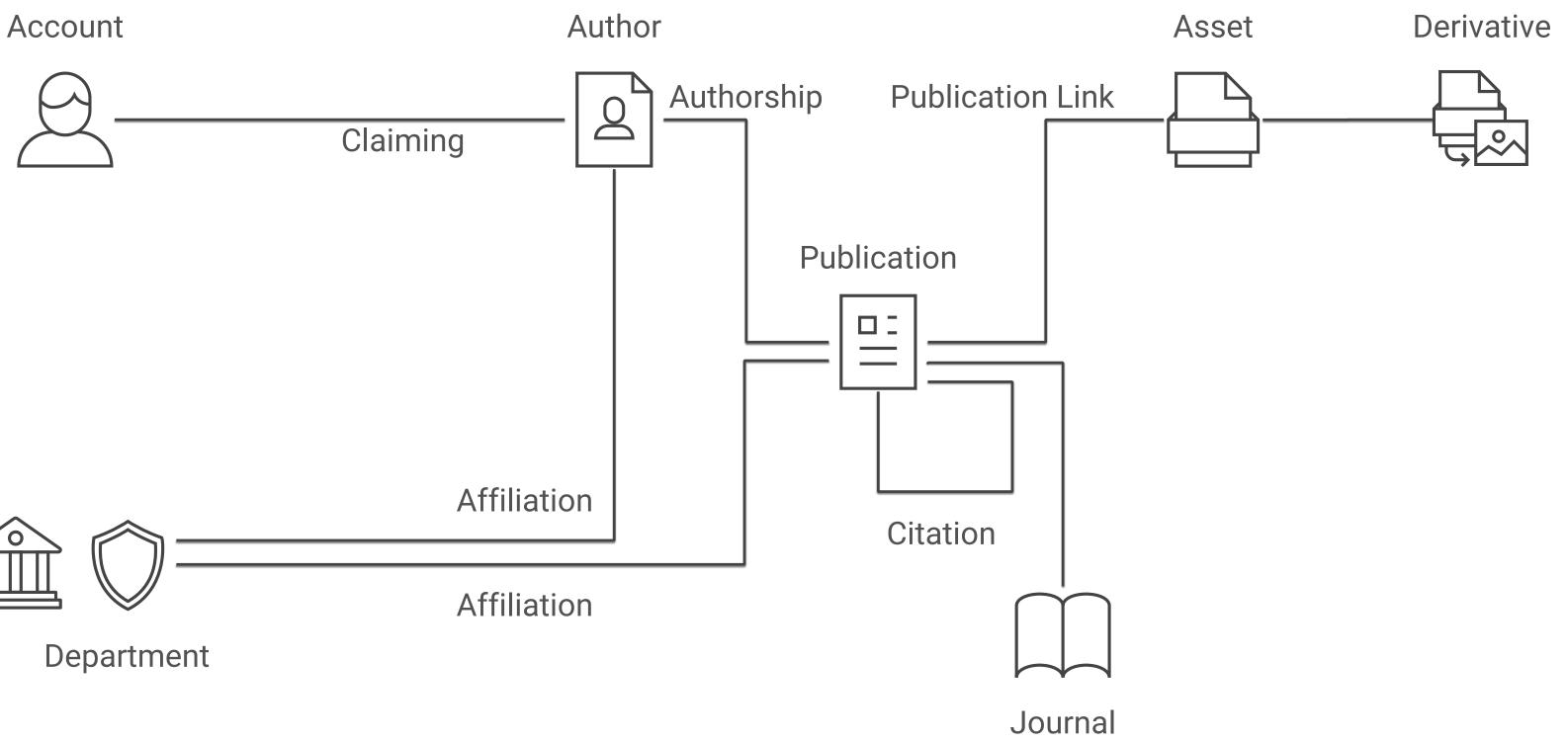
Enriching User Generated Content

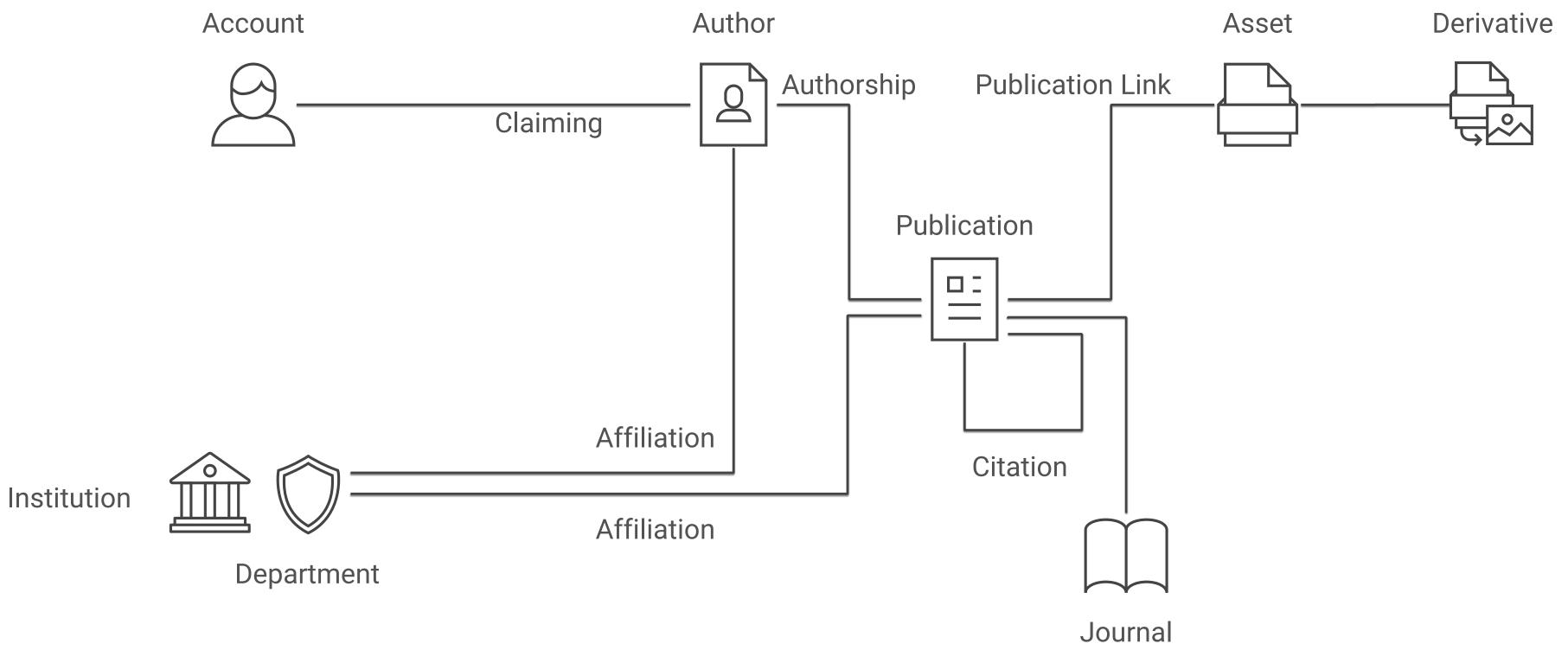
Users and batch flows continuously enrich an evolving dataset

Both user actions and batch flow results ultimately affect the same live database

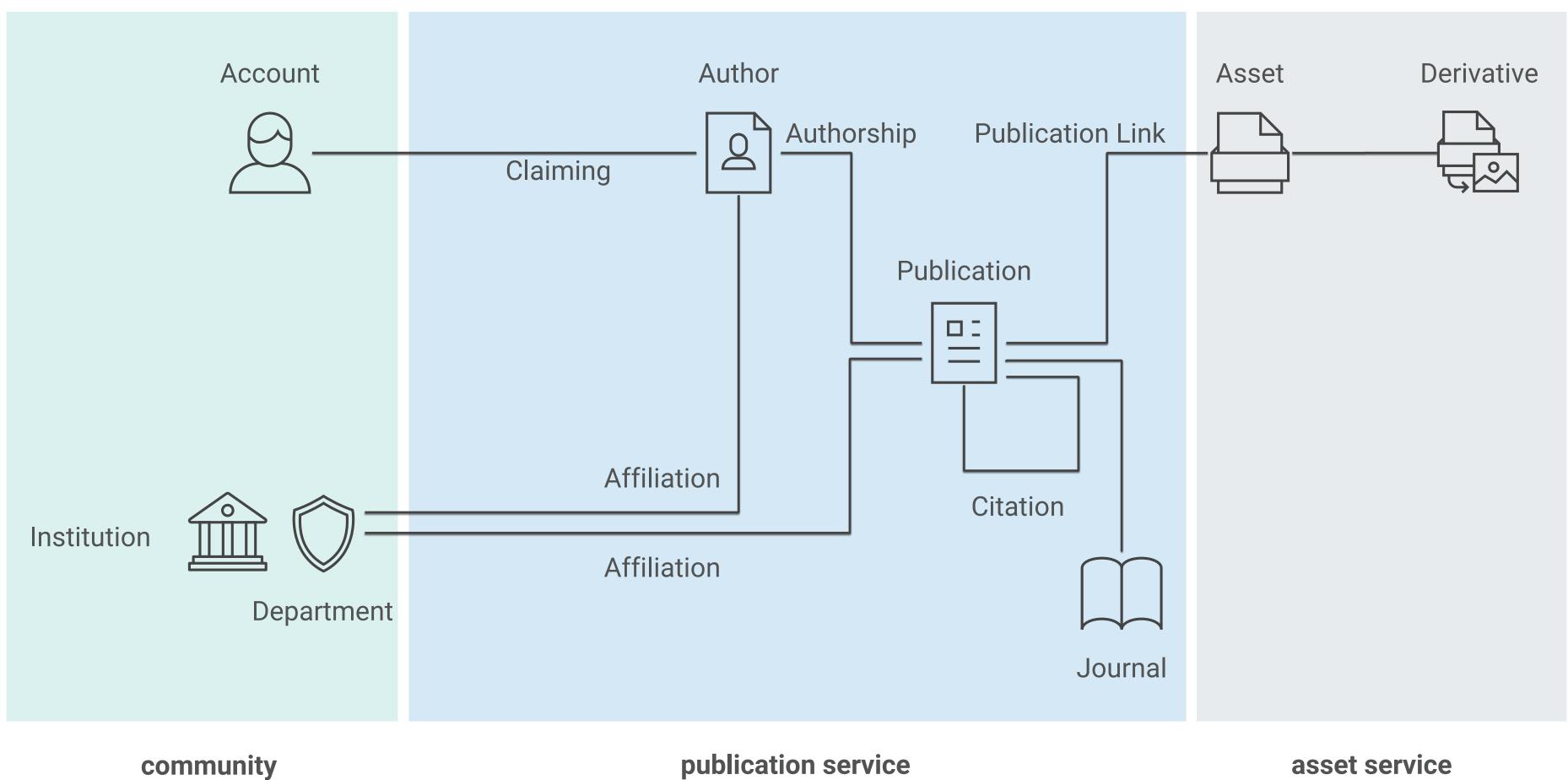


Bibliographic Metadata – Data Model





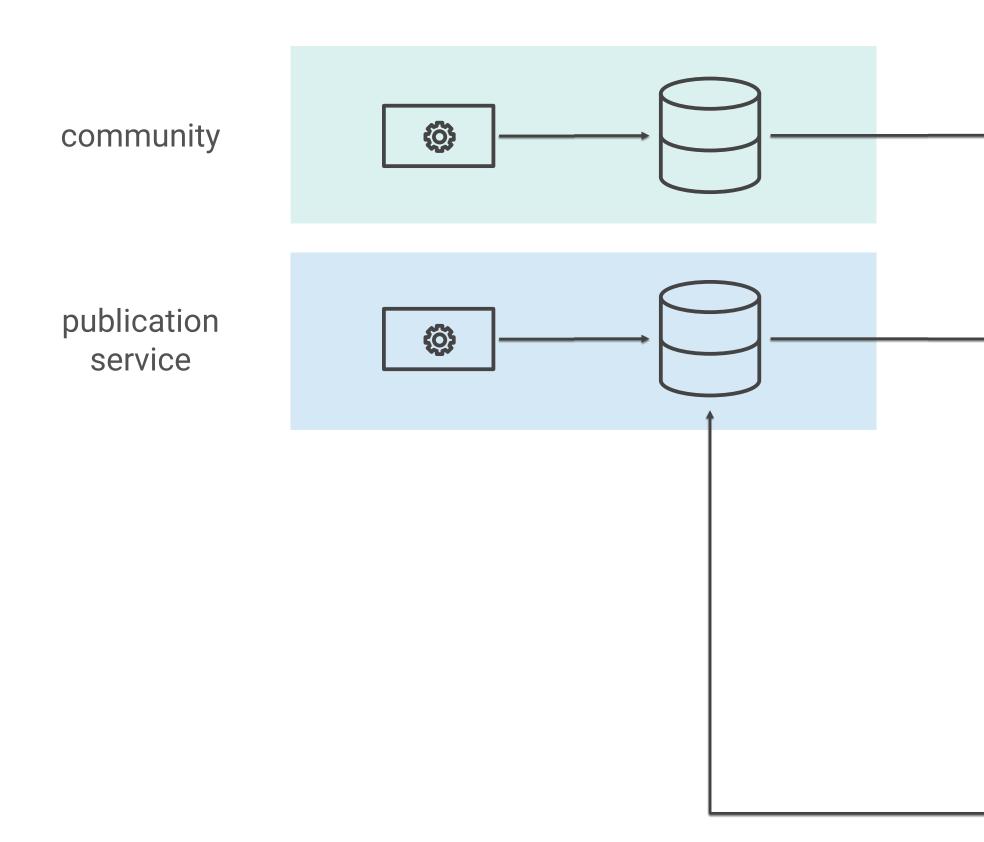
Bibliographic Metadata – Services

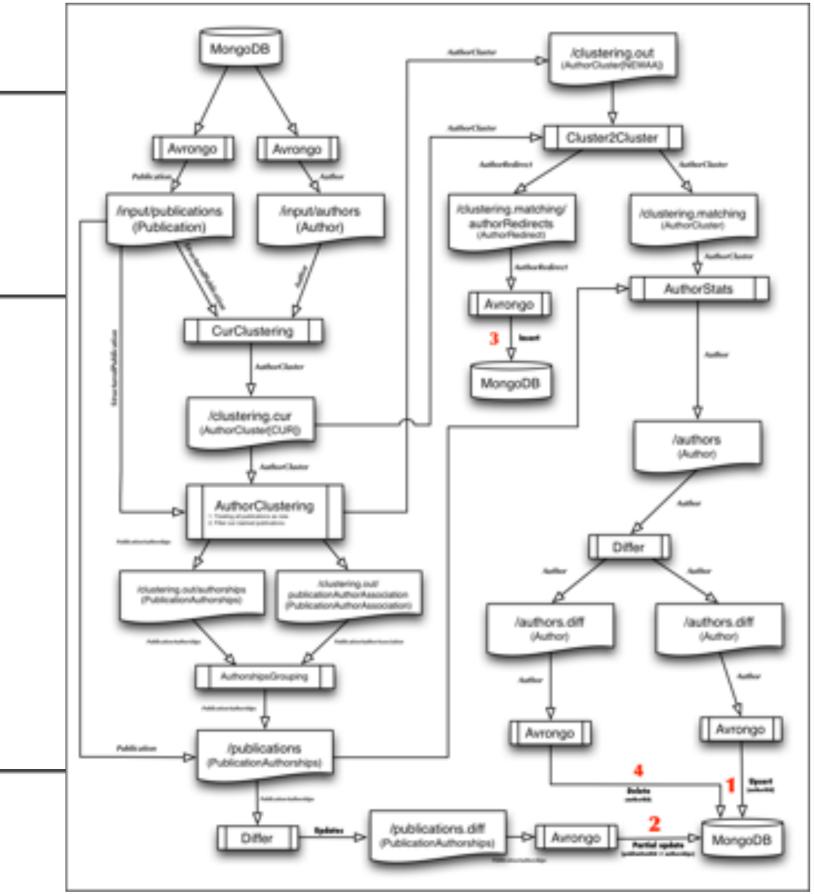


publication service

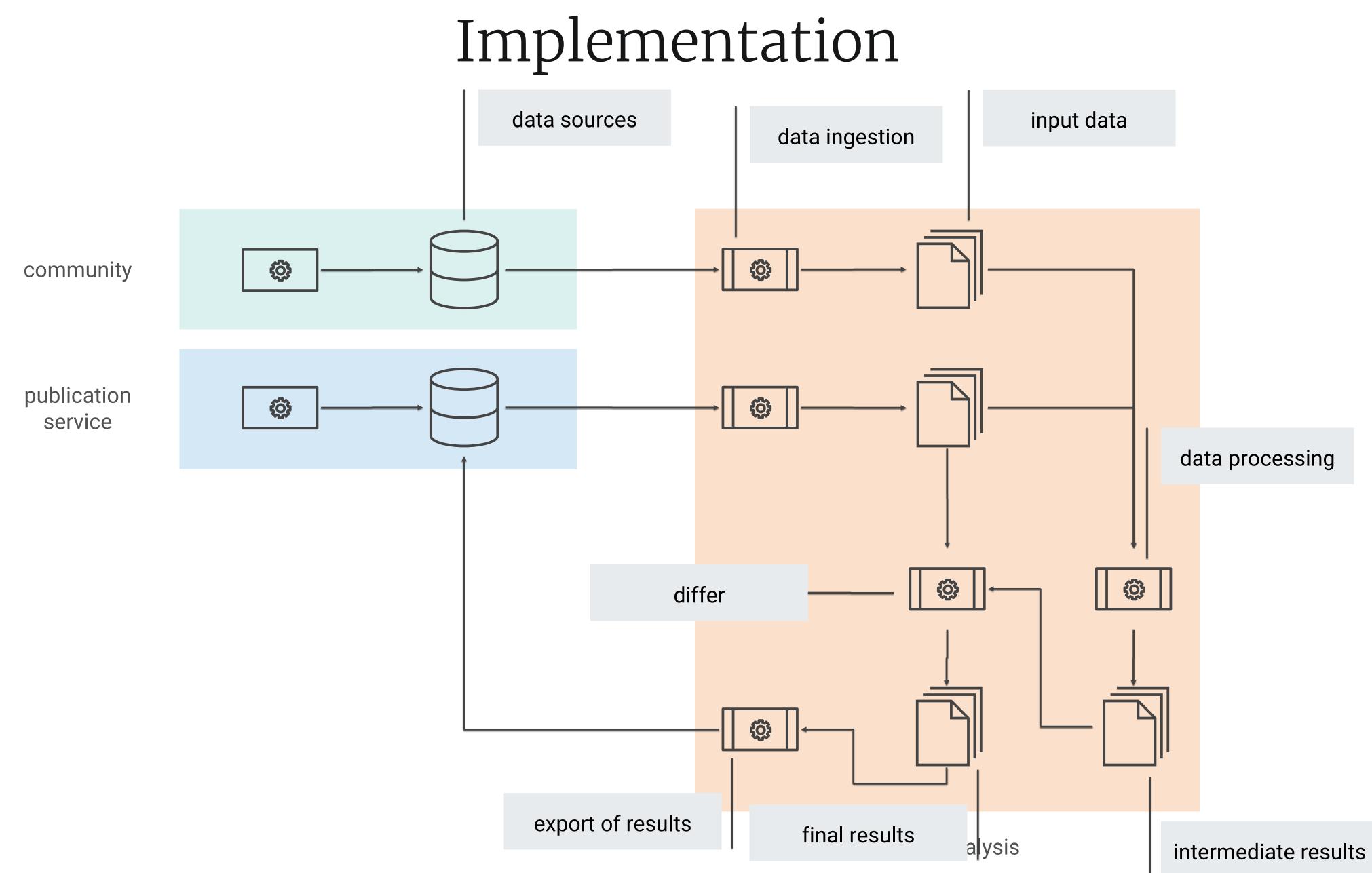
asset service

Implementation

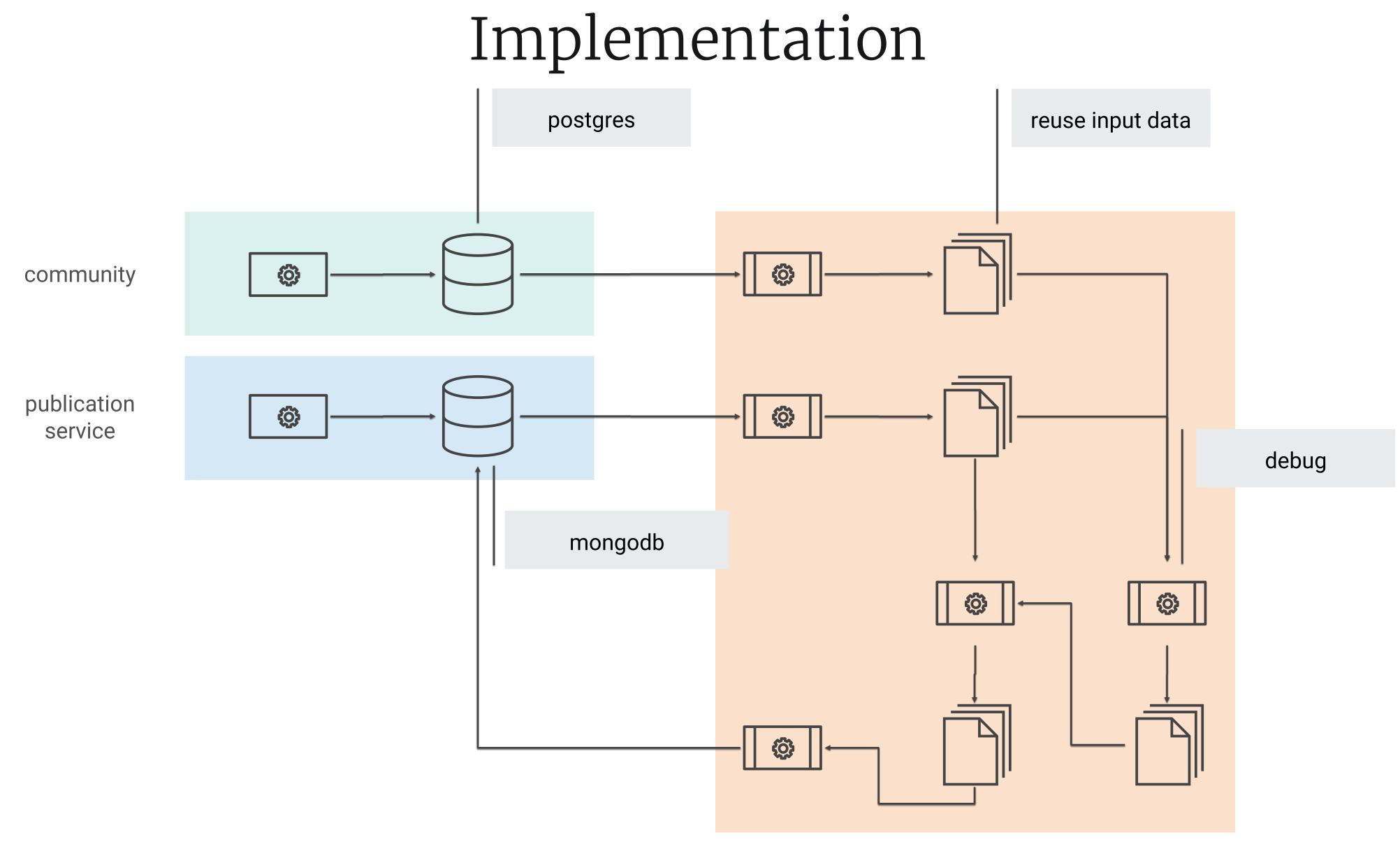




author analysis



#1 Decouple Data Ingestion



author analysis

Debugging an Error on Production

Your flow

has unit and integrations tests but still breaks unexpectedly in production

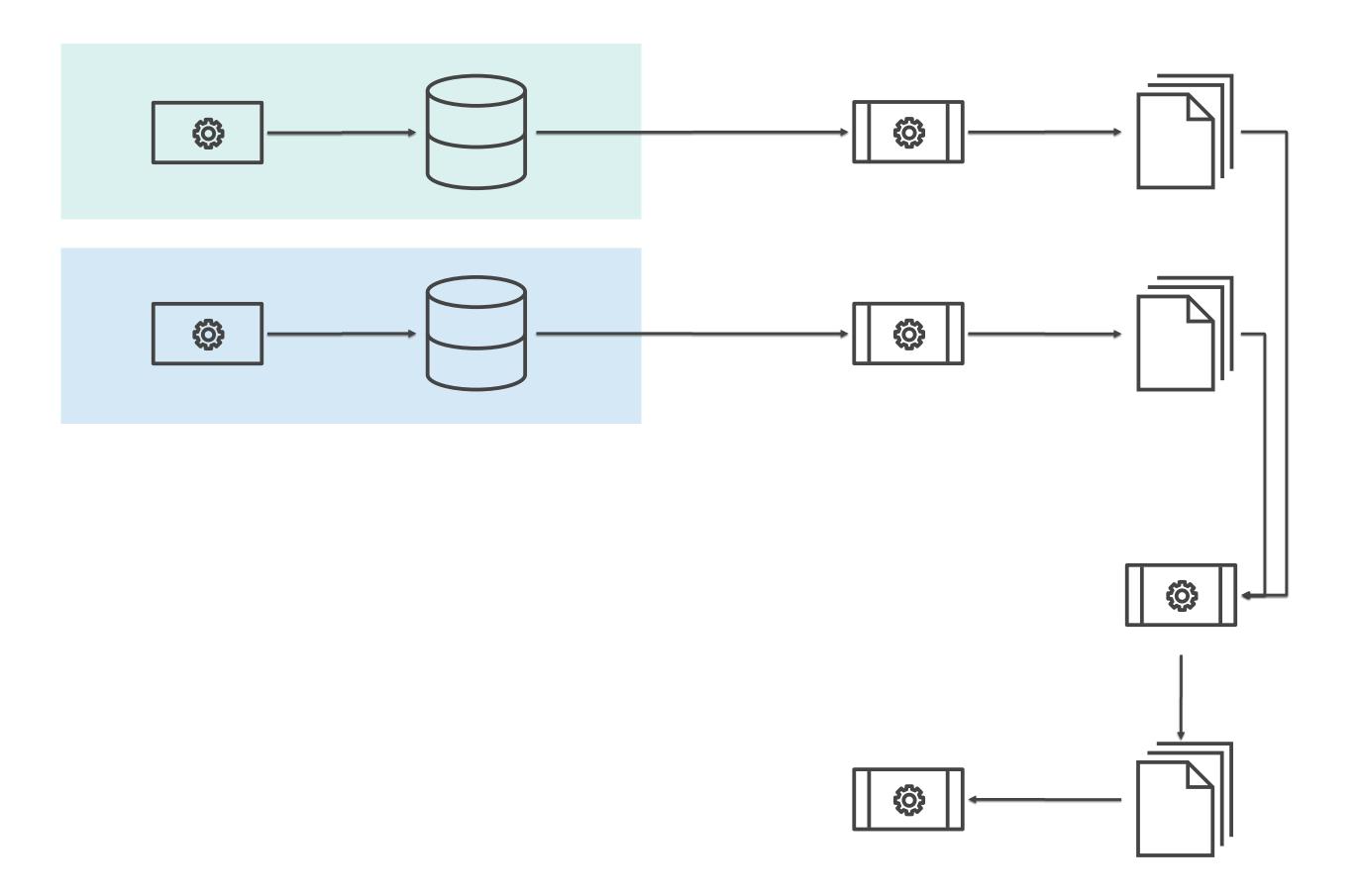
You need to find the root cause

- Is it a change in input data?
- Is it a change on the cluster?
- Is it a race condidition?

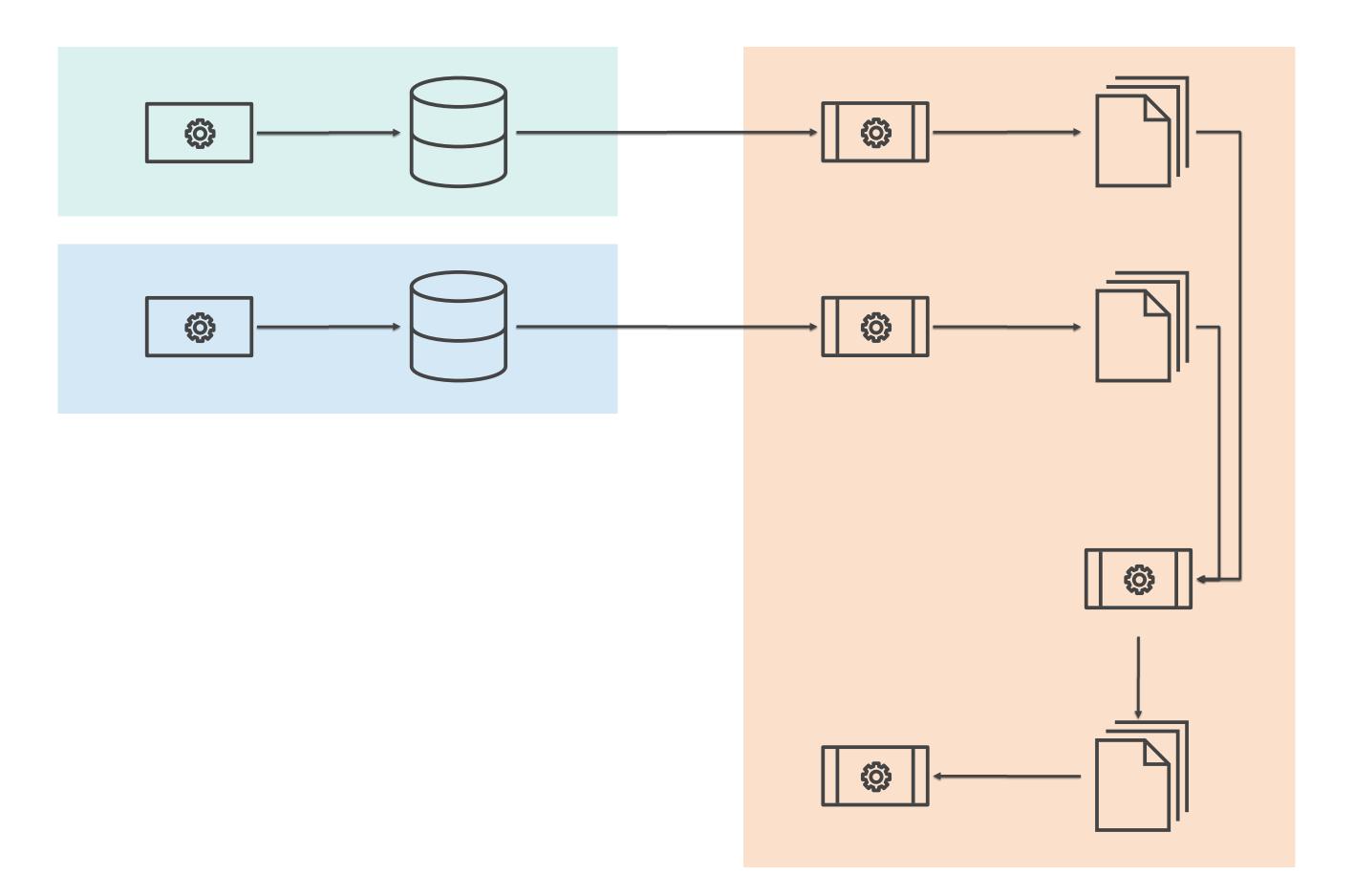
Crucial capabilities

Easy adhoc analysis of all involved data (input, intermediate, result) Rerun current flow with current cluster configuration on yesterday's data Confirm hotfix by re-running on today's data (exactly the same data that triggered the bug)

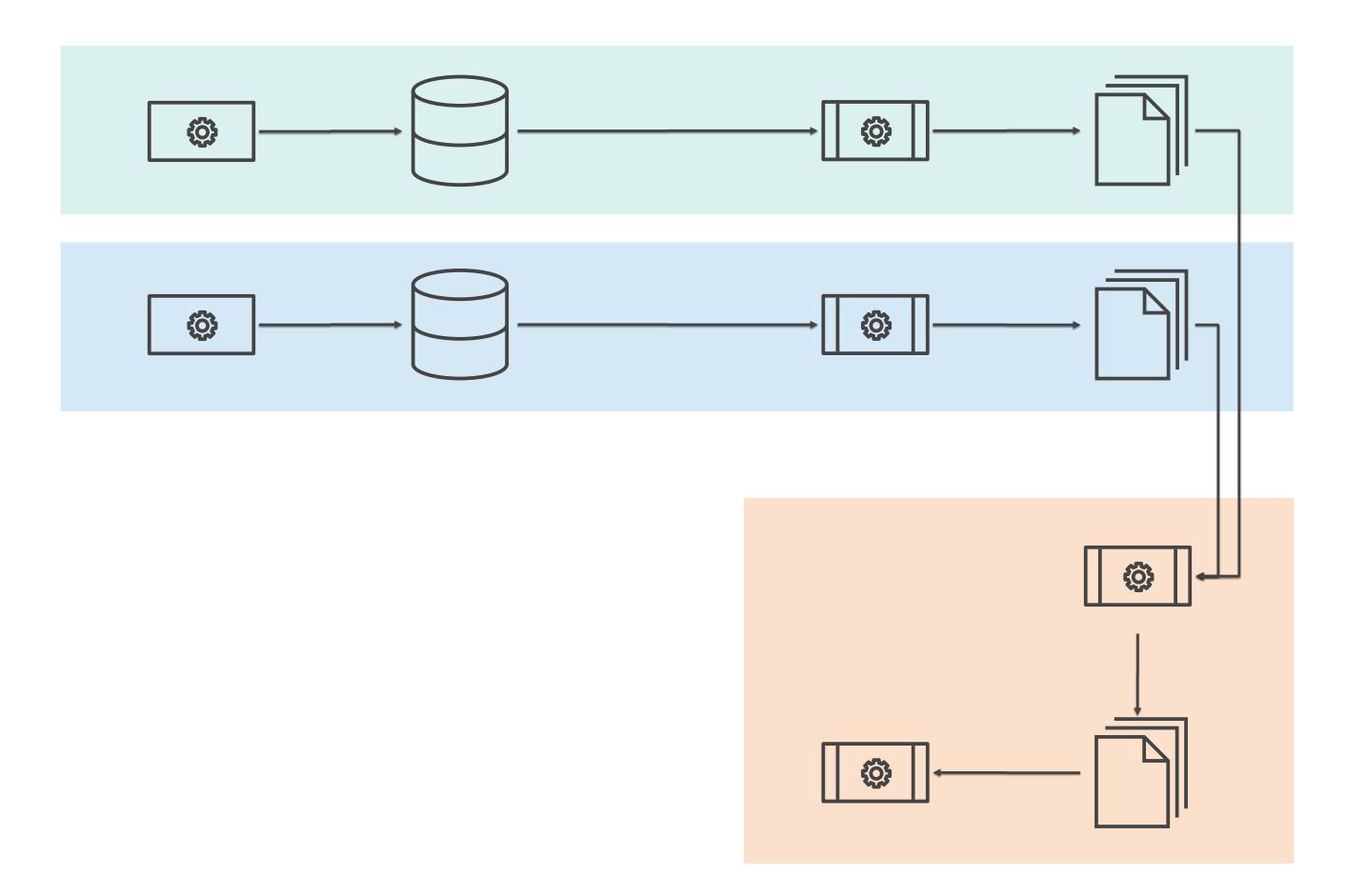
How to decouple?



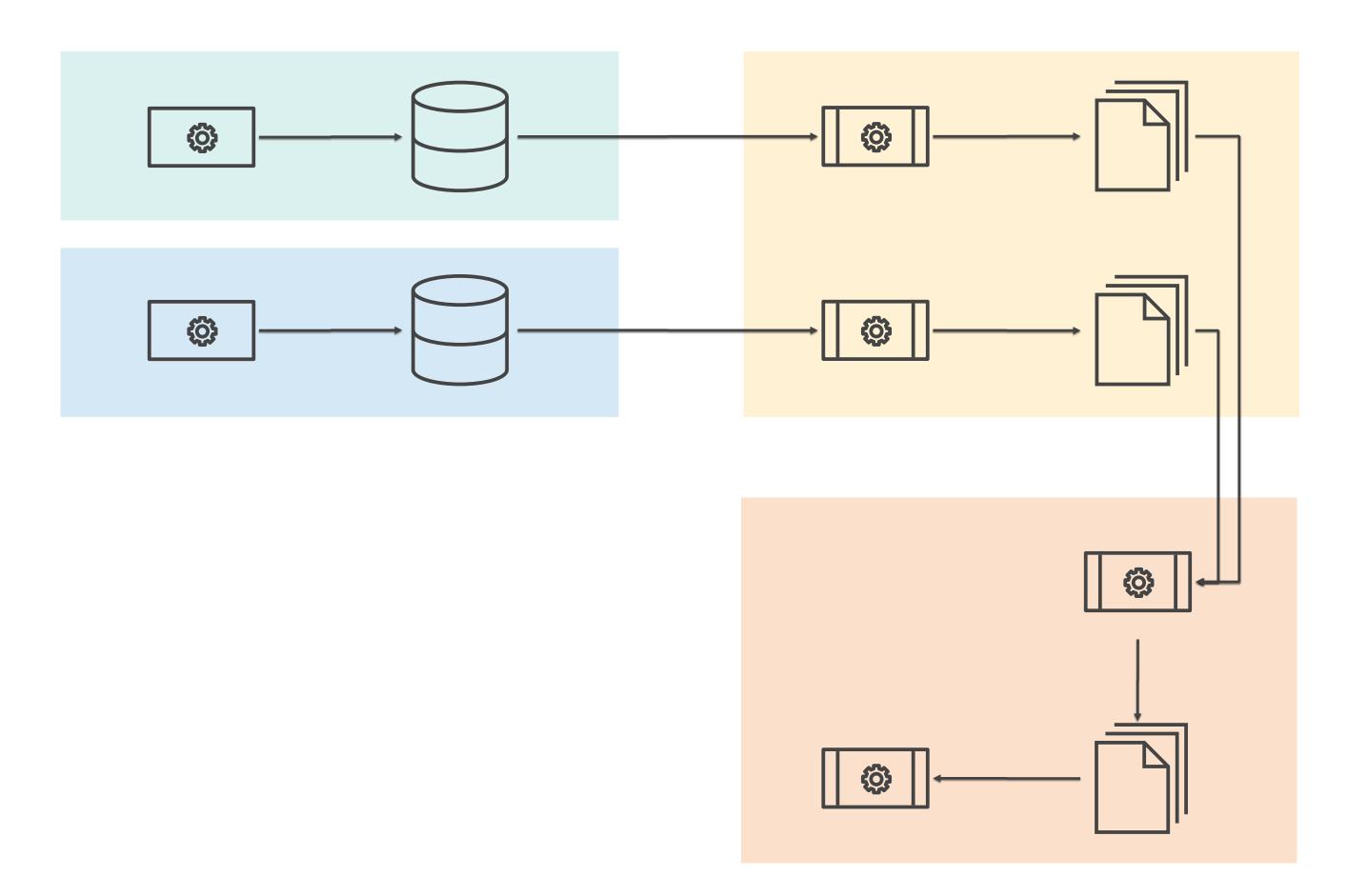
Ingesting Data as Needed?

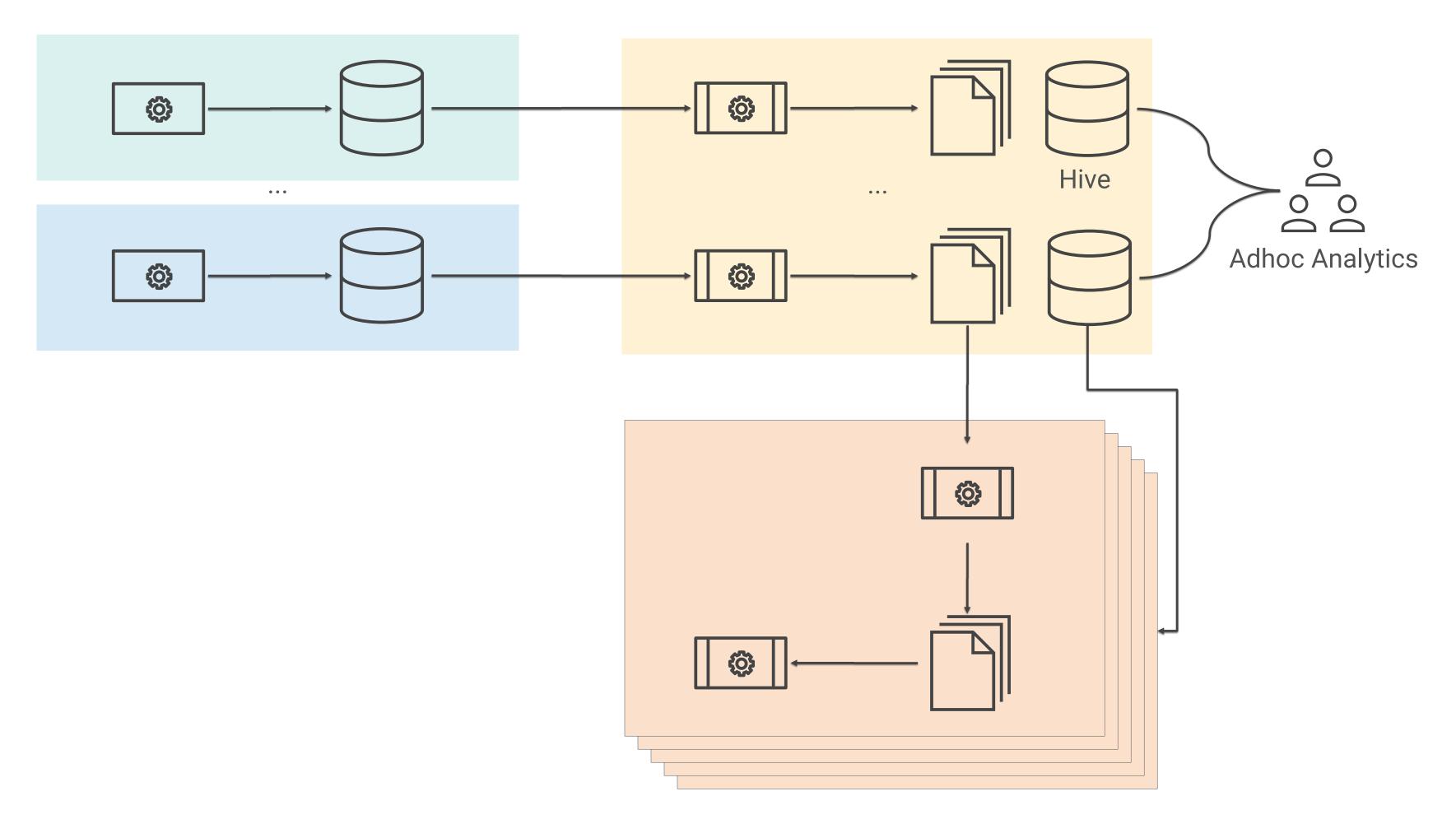


Publishing Data as Needed?



Dedicated Data Ingestion!





Platform Data Import

Dedicated component, but generic

Every team can onboard new data sources, as required by use cases

Every ingested source is immediately available for all consumers (incl. analytics)

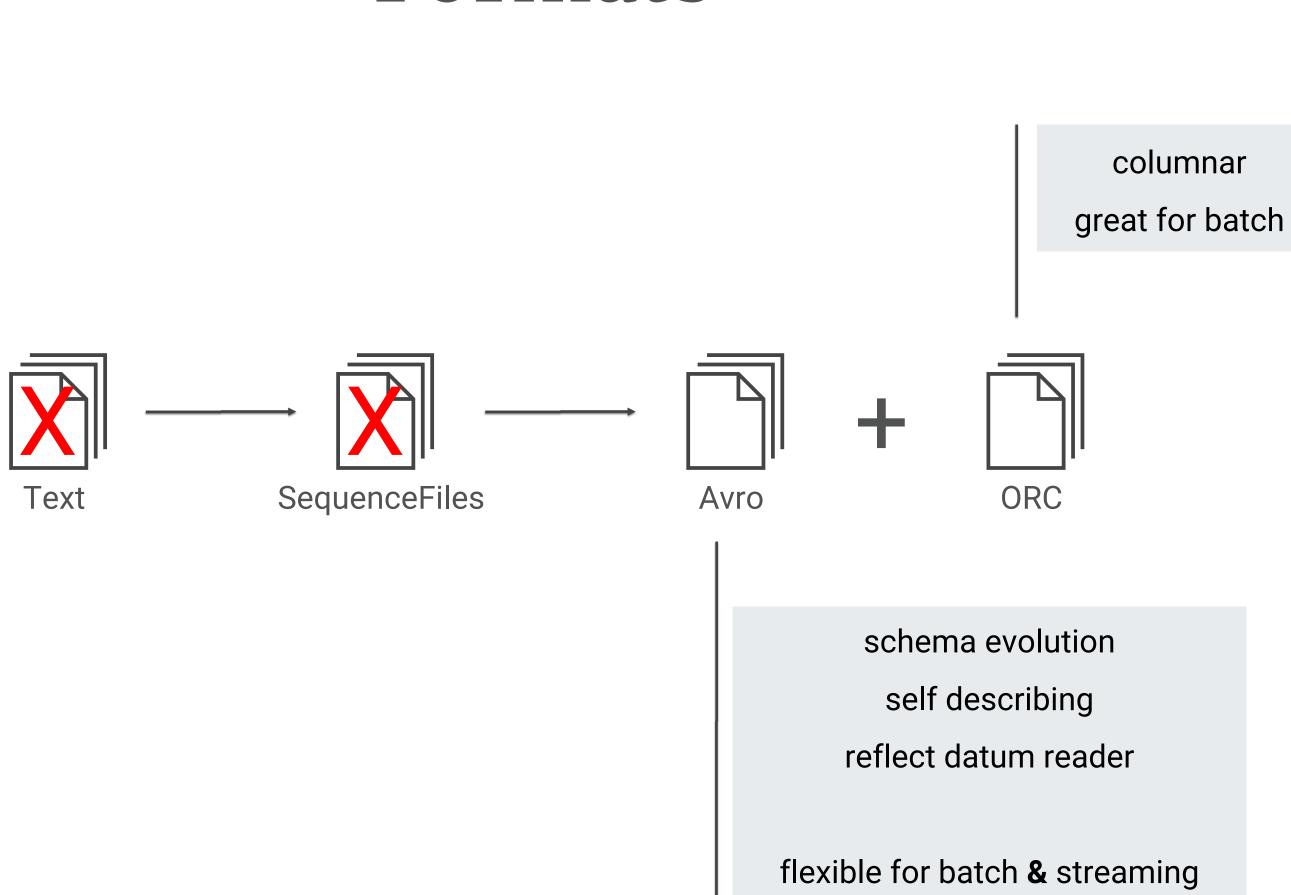
Feature parity for all data sources (e.g., mounting everything in Hive)

Platform Data Import

#2 Speak a common format*

* have at least one copy of all data in a common format (e.g., avro)

Formats



Speak a common format

Have at least one copy of all data in a common format

Your choice of processing framework should not be limited by format of existing data

Every ingested source should be available for all consumers

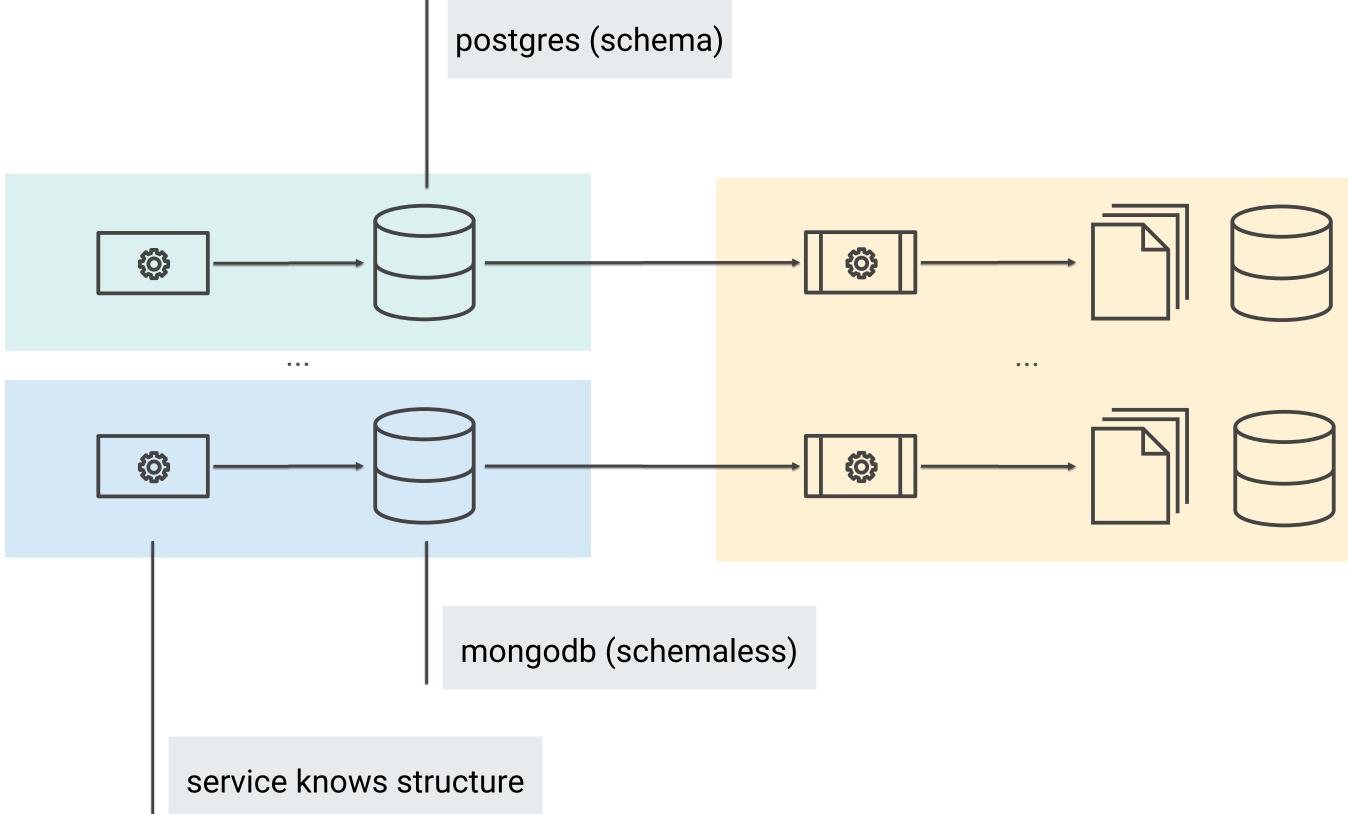
When optimizing for a framework (e.g., ORC for Hive) consider a copy

#3 Speak a common language*

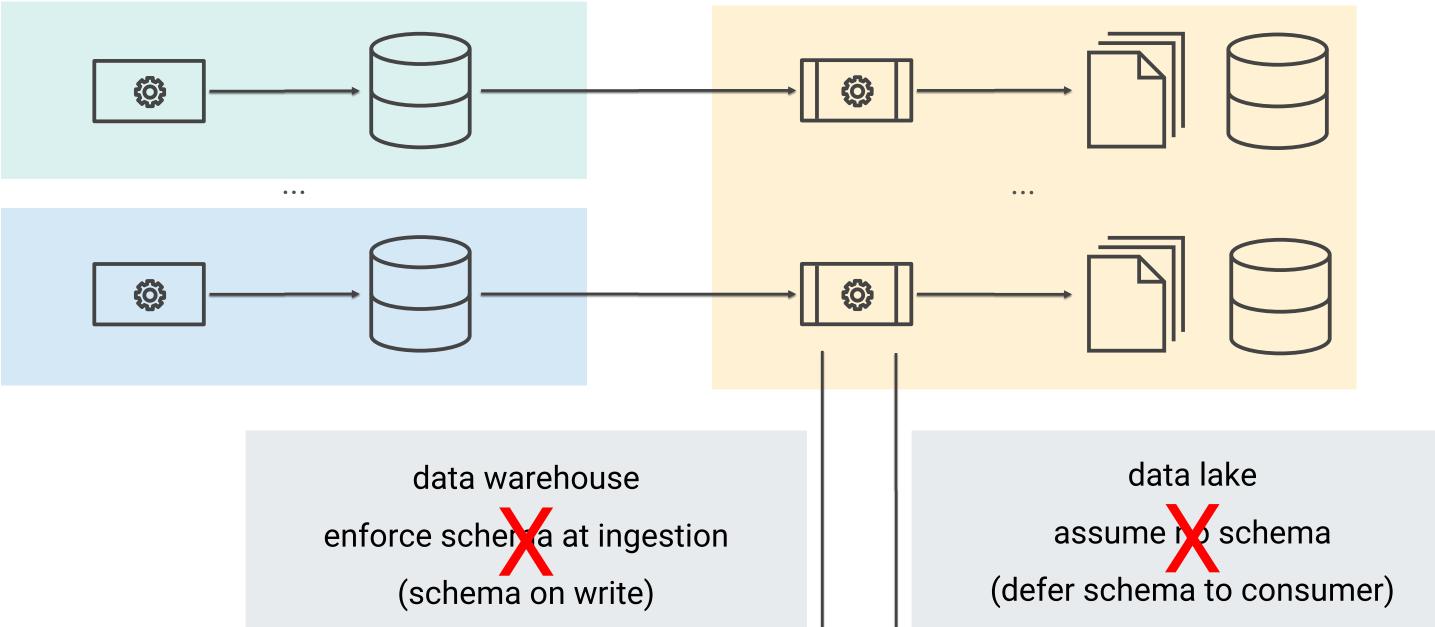
* continuously propagate schema changes



Structured or unstructured data?



Data Warehouse vs. Data Lake



Preserve schema information that is already present some times at database level many times at application level

Preserve full data – be truthful to our data source continuously propagate schema changes

Can we have something like a **Data Lakehouse**?

Can we have both?

Entities Define Schema

Code first

entities within owning service define schema

Auto conversion preferred

conversion to other representations via annotations (JSON, BSON, Avro, ...)



Continuously propagate schema changes

Data ingestion process is generic and driven by avro schema

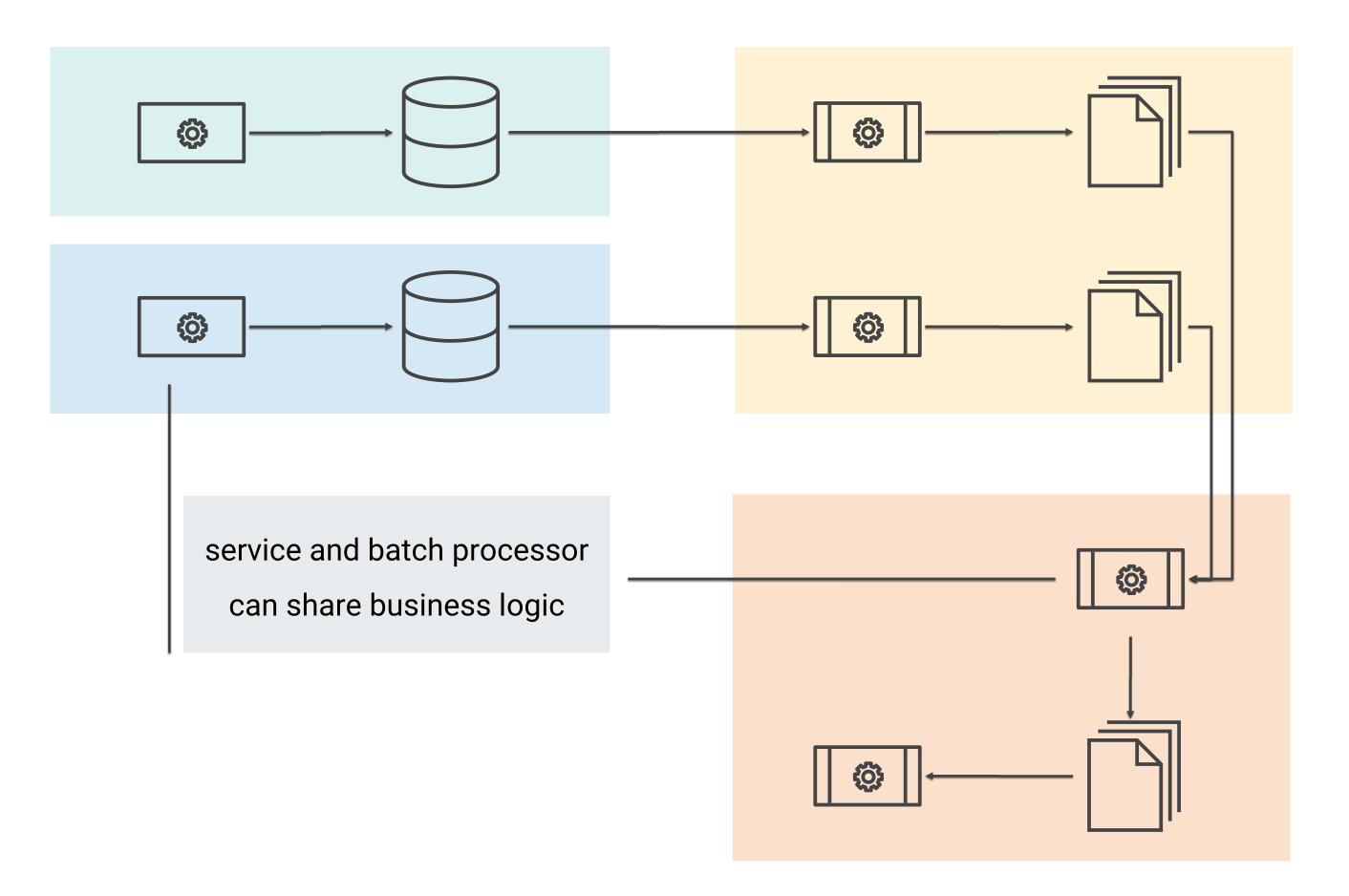
Changes in avro schema are continuously propagated to data ingestion process

Consumers with old schema can still read data due to avro schema evolution

Caveat: breaking changes still have to be dealt with by a change process

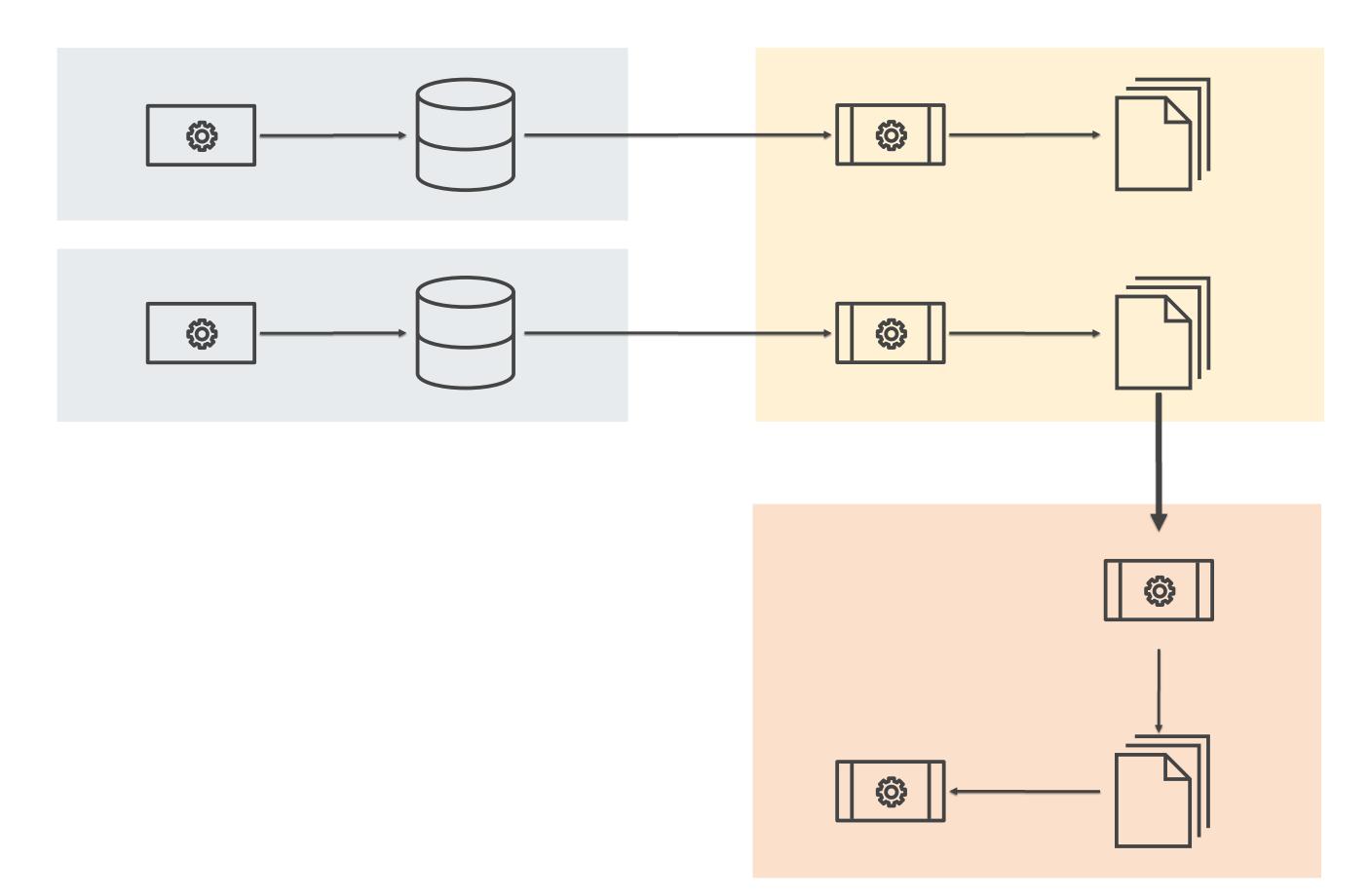
Everyone speaks the same language

Extra Benefit

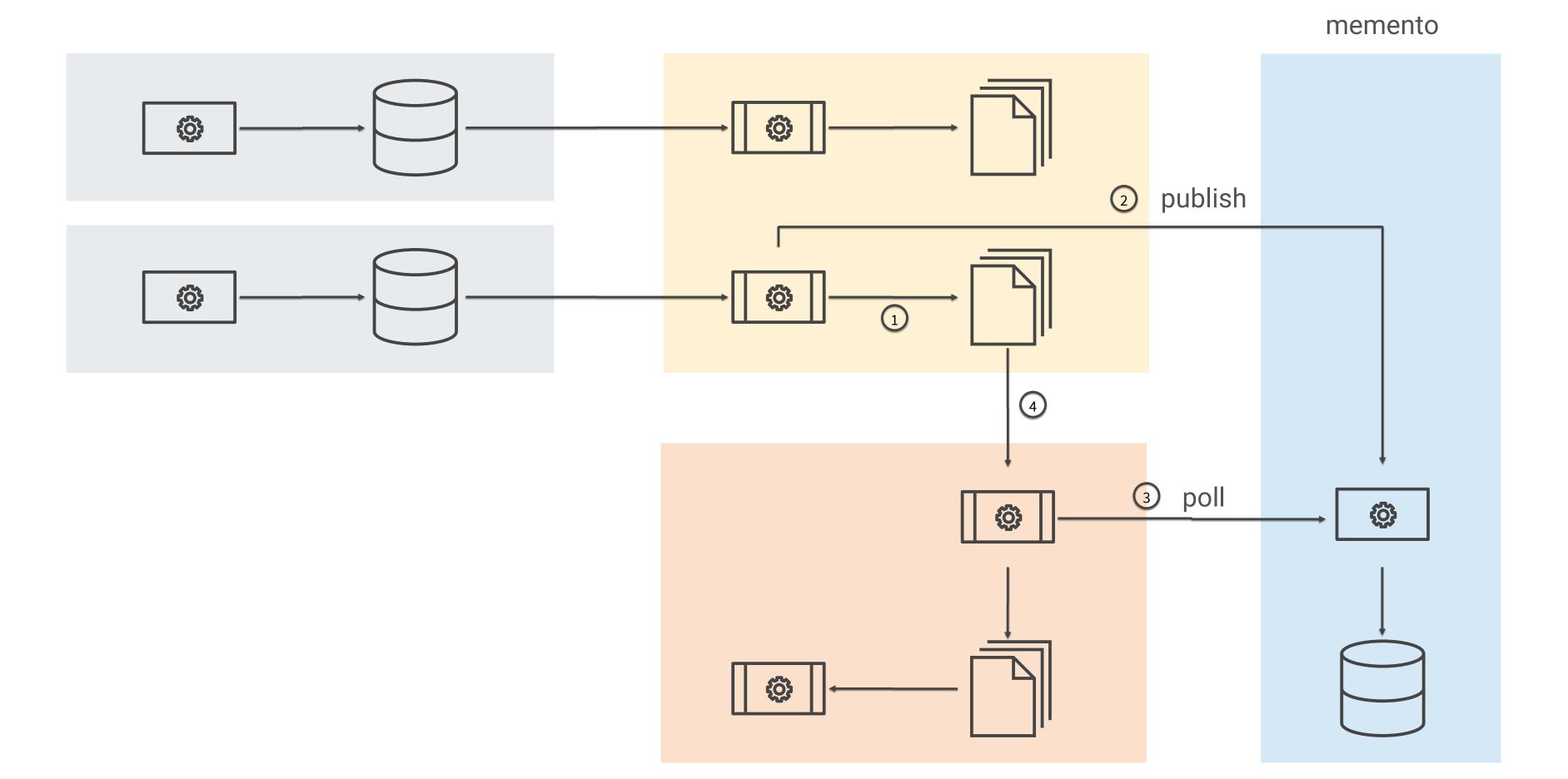


#4 Model Data Dependencies Explicitly

Model Data Dependencies Explicitly



Model Data Dependencies Explicitly



Memento v2

memento publish

```
{
    "namespace": "platform-data",
    "name": "refind.authors",
    "businessDate": "2017-06-11",
    "type": "hdfs",
    "path": "hdfs://platform-data/refind/authors/2017-06-12/attempt-0/data",
    "format": "avro",
    "attempt": 0,
    "timestamp": "2015-06-12T00:42:23.125+0000"
}
```

memento poll <waiting-time>

```
{
    "namespace": "platform-data",
    "name": "refind.authors",
    "businessDate": "2017-06-11",
    "type": "hdfs"
}
```

```
unique artifactId
```

```
{
    "artifactId": "dll4682f02fd",
    "namespace": "platform-data",
    "name": "refind.authors",
    "businessDate": "2017-06-11",
    "type": "hdfs",
    "path": "hdfs://platform-data/refind/authors/2017-06-12/attempt-0/data",
    "format": "avro",
    "attempt": 0,
    "timestamp": "2015-06-12T00:42:23.125+0000"
}
```

Model Data Dependencies Explicitly

More flexible scheduling – run flows as early as possible

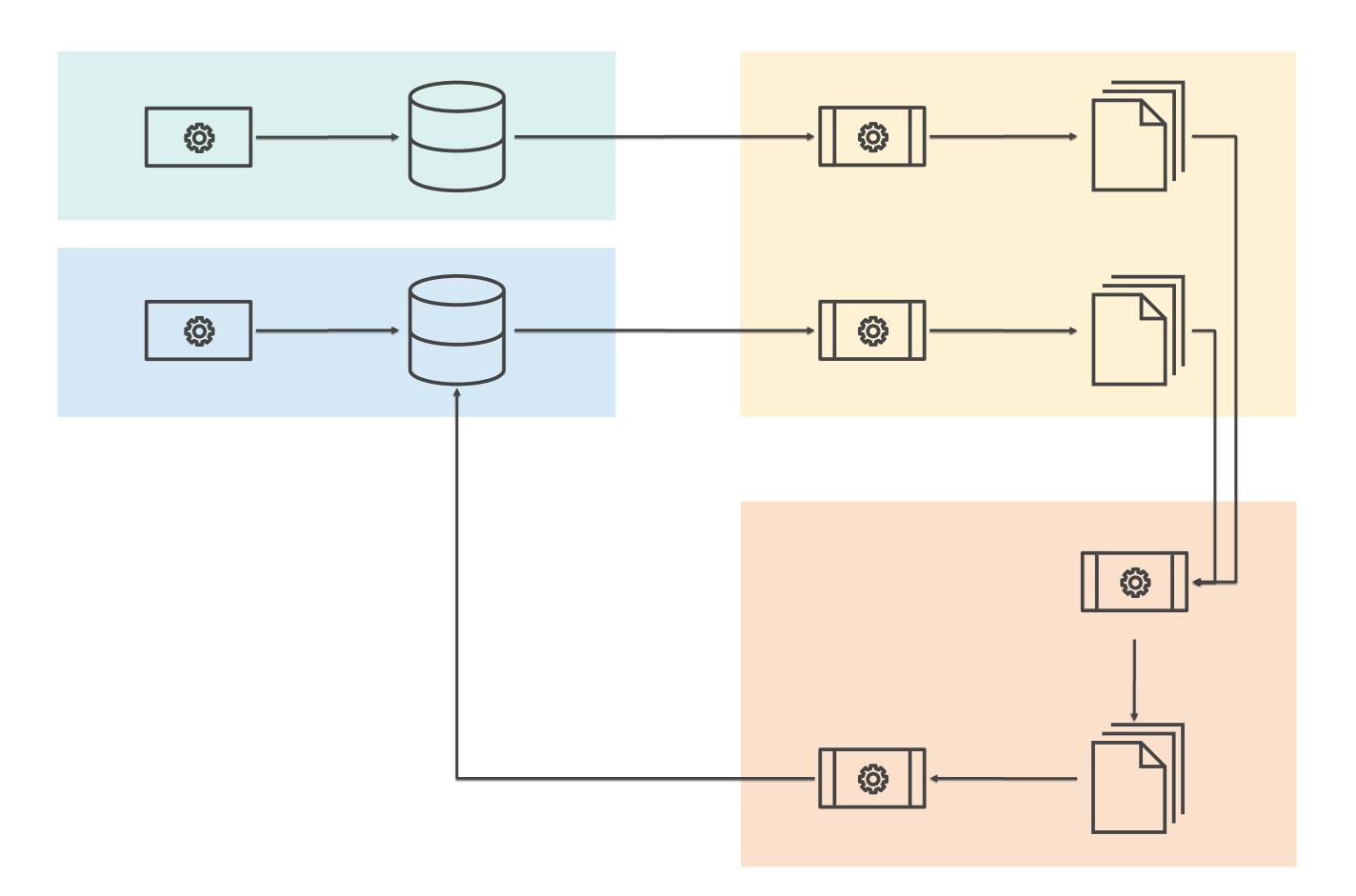
Allows multiple ingestion or processing attempts

Allows immutable data (repeatable read)

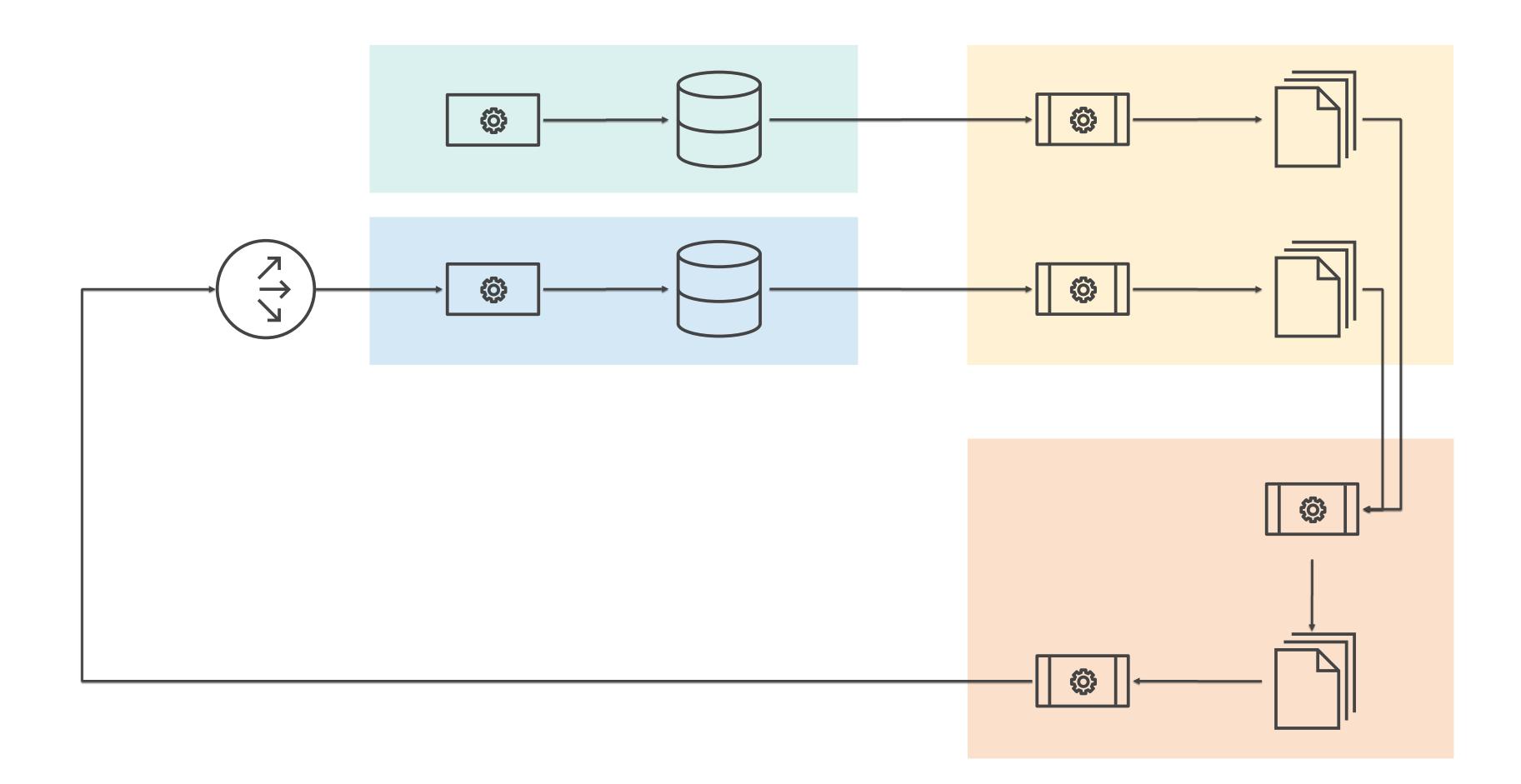
Allows analysis of dependency graph which datasets are used by what flow

#5 Decouple export of results

Decouple export of results



Decouple export of results



Push results via HTTP to service

Export of results just becomes a client of the service service does not have to be aware of big data technologies

Service can validate results, e.g., plausibility checks optimistic locking

Makes testing much easier

Avro → Http

Part of the flow, but standardized component

Handles tracking of progress treats input file as a "queue" converts records to http calls can be interrupted and resumed anytime

Sends standardized headers, e.g.,

X-rg-client-id: author-analysis

Handles backpressure signals from services

#6 Model Flow Orchestration Explicitly

Model Flow Orchestration Explicitly

Consider using an execution system like Azkaban, Luigi, or Airflow

Establish coding standards for orchestration, e.g., inject paths from outside – don't construct them in your flow inject calculation dates – **never** call now() inject configuration settings – don't hardcode -D mapreduce.map.java.opts=-Xmx4096m foresee environment specific settings

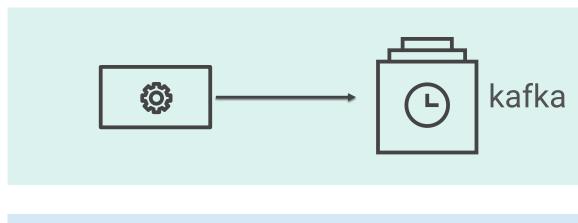
Think about ease of operations tuning of settings upgrades

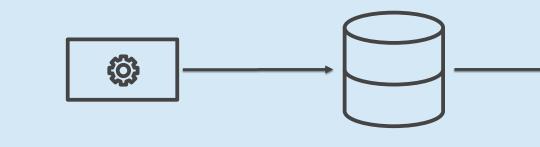
What about Stream Processing?



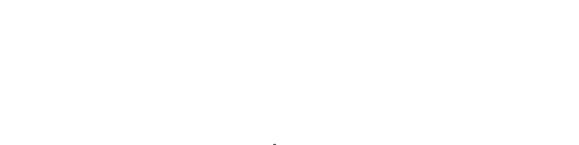
Sources of Streaming Data

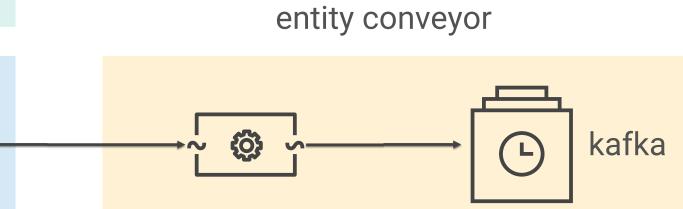
timeseries data

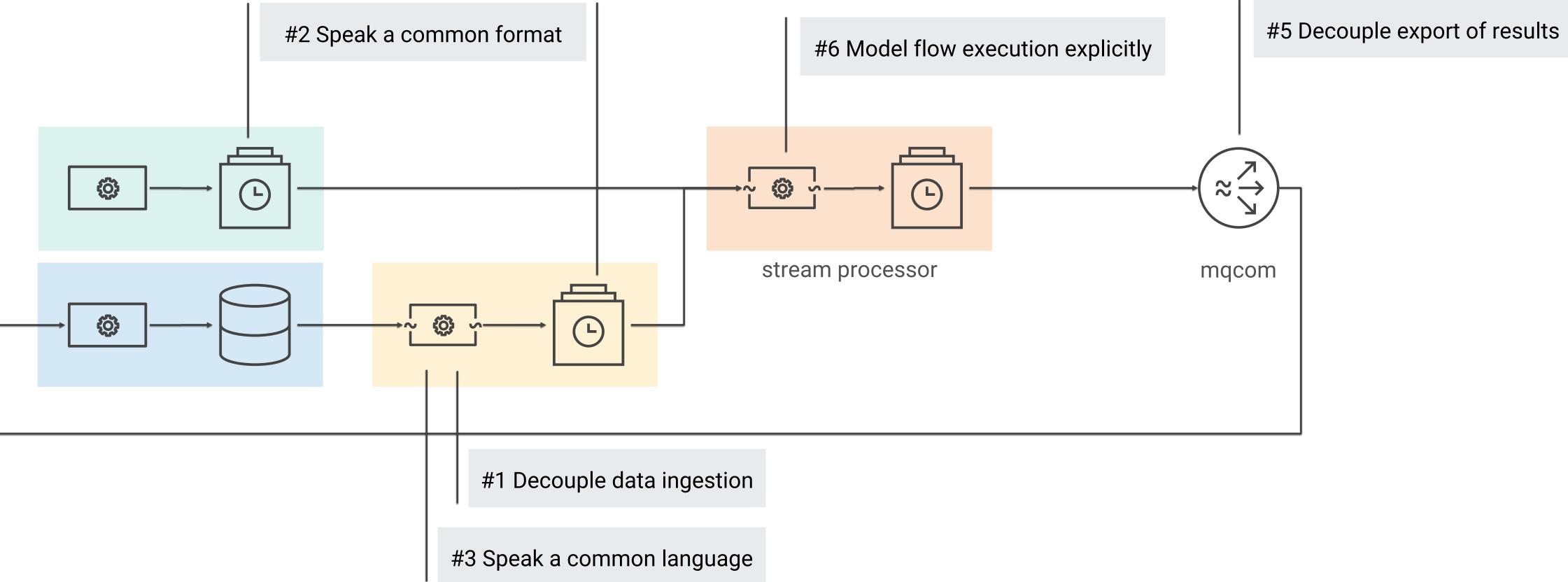




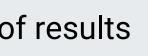
non-timeseries data (e.g., graph data)







Stream Processing



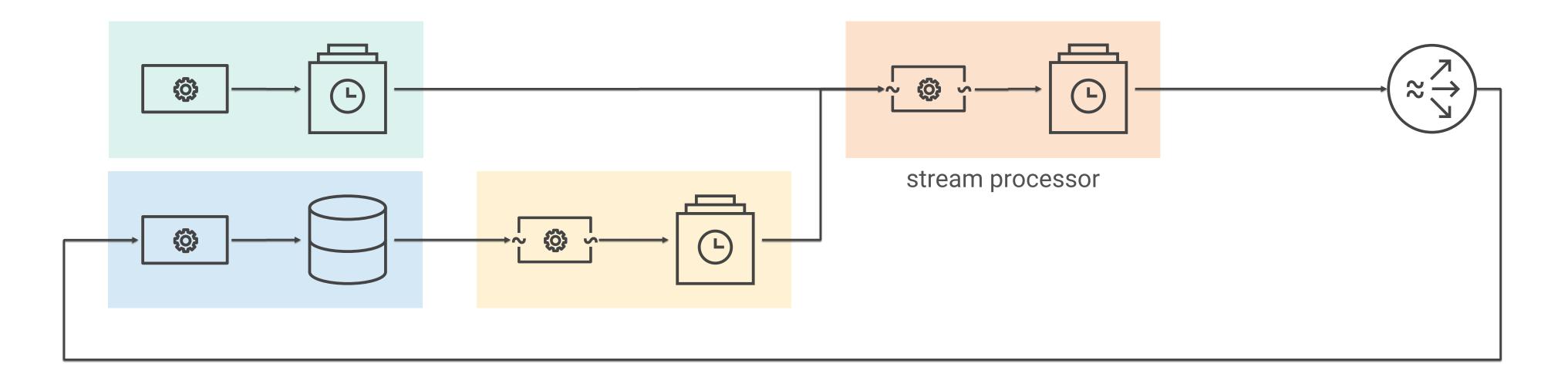
What about #4 ? Model Data Dependencies Explicitly

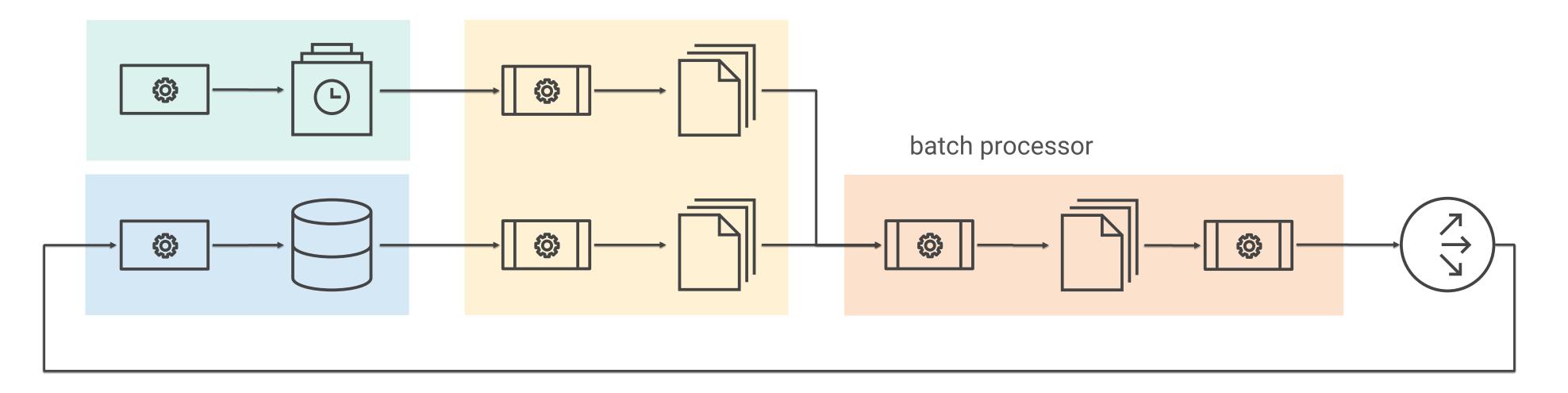
We think about it

Depends on use cases and pain points

Potentially put Kafka topics into Memento storing "offsets of interest" from producers facilitate switching between incompatible versions of stream processors

Evolving Big Data Architecture







Michael Häusler, Head of Engineering https://www.researchgate.net/profile/Michael_Haeusler

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Thank you!

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