



# Adding intelligence to streamprocessing applications

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#### Forecast

Introducing intelligent applications

Introducing microservice architectures

Stream processing with Apache Kafka and Apache Spark

Interactivity time







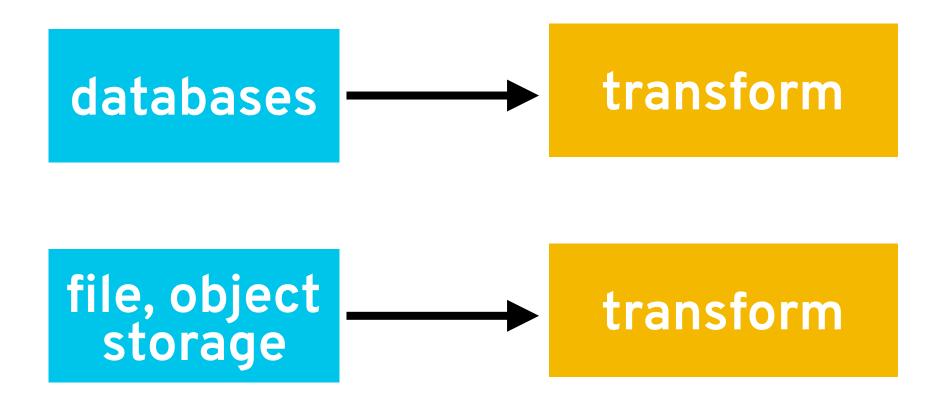


# Intelligent applications

Intelligent applications collect and learn from data in order to provide improved functionality with longevity and popularity.

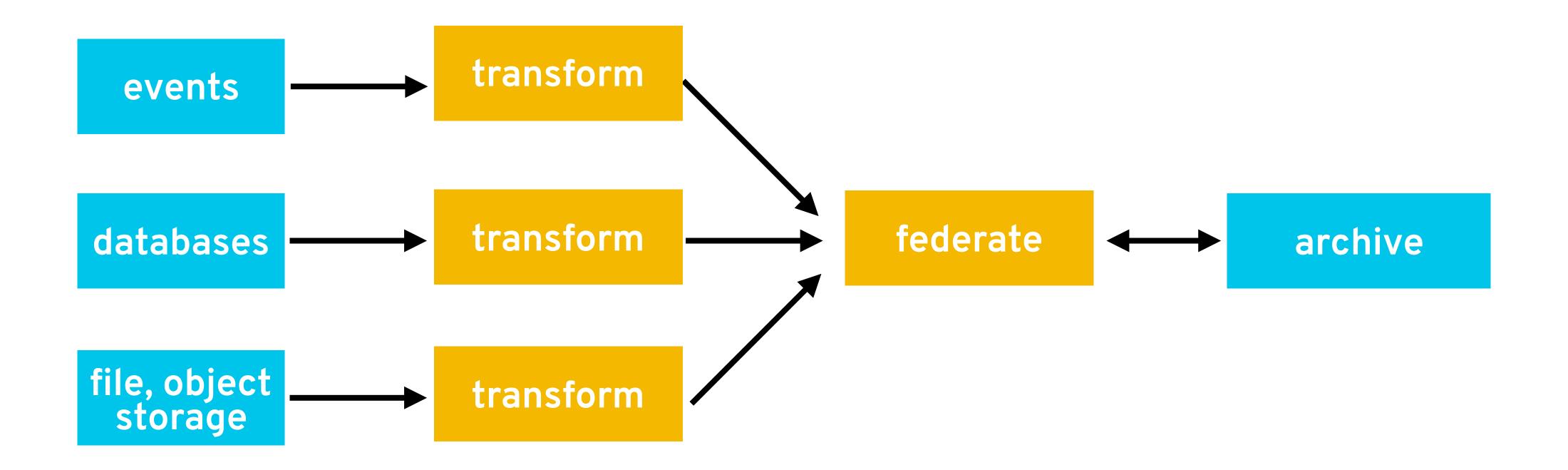






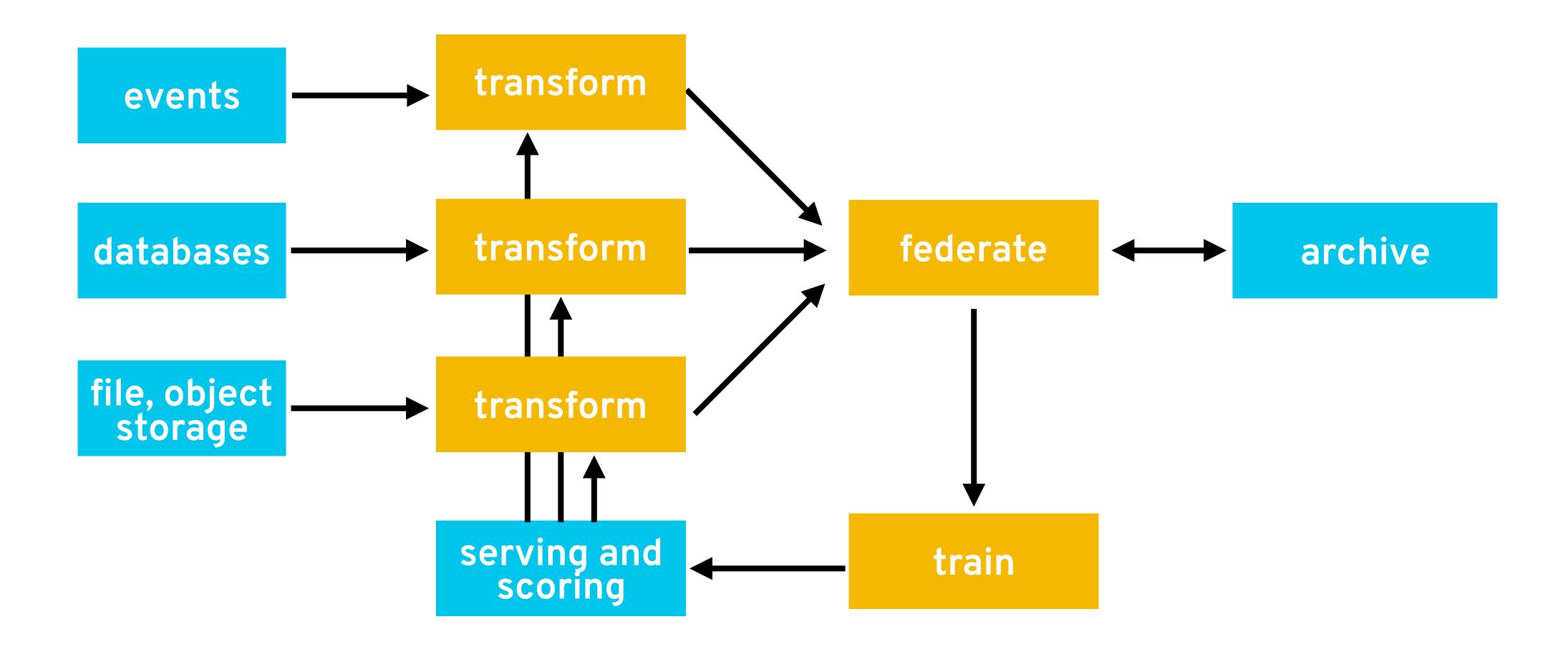






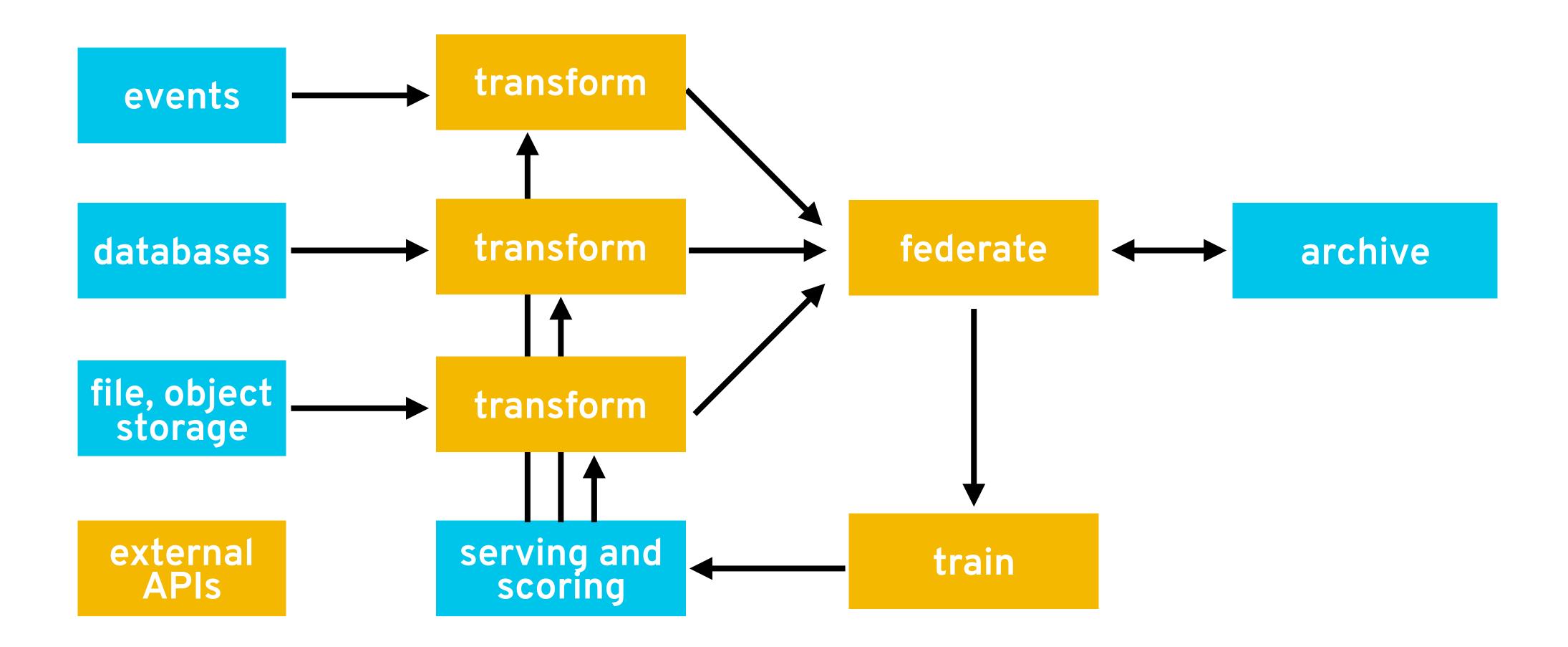






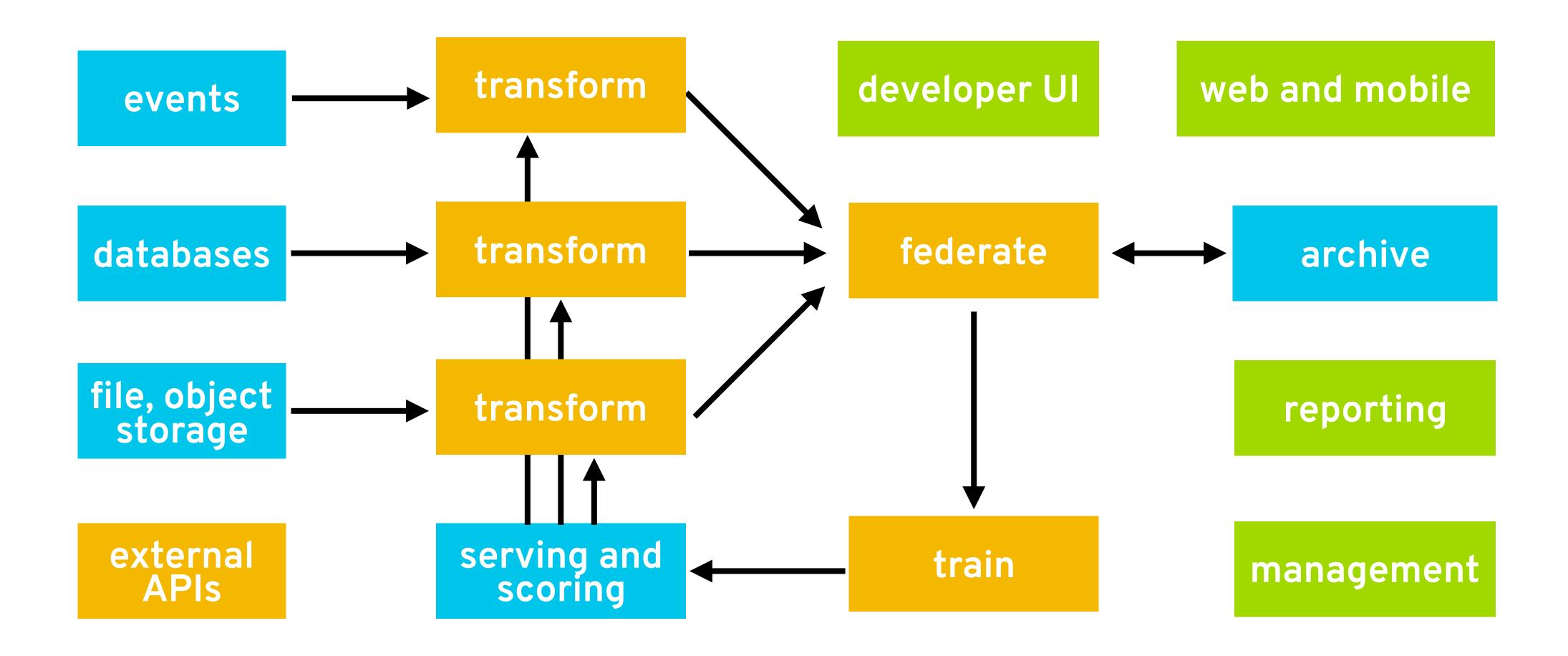






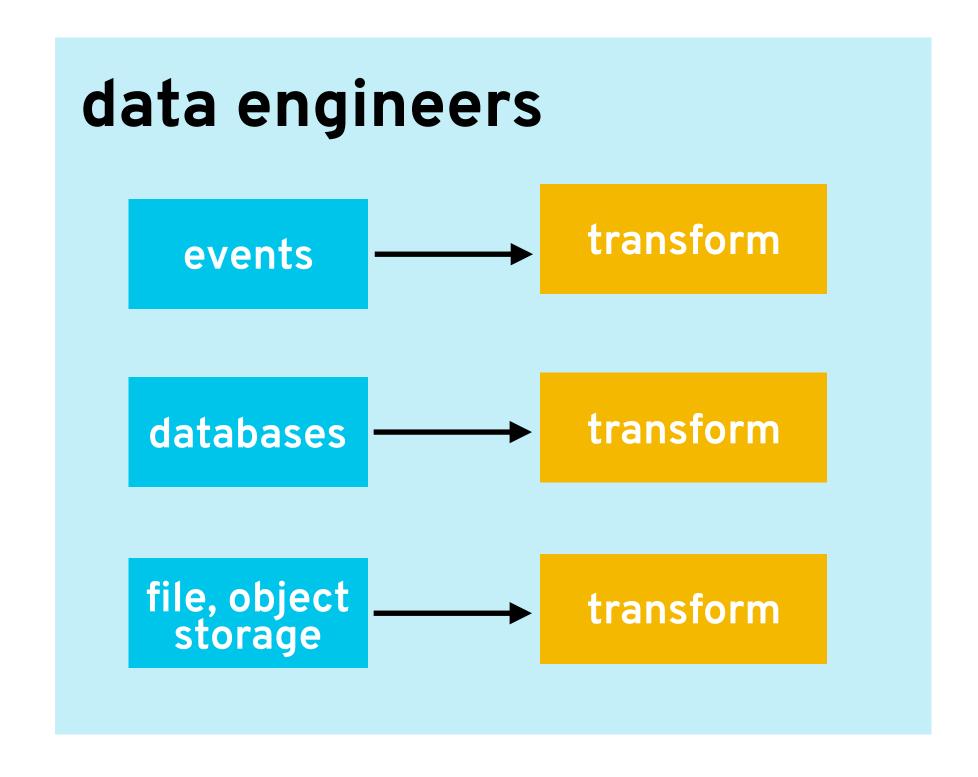








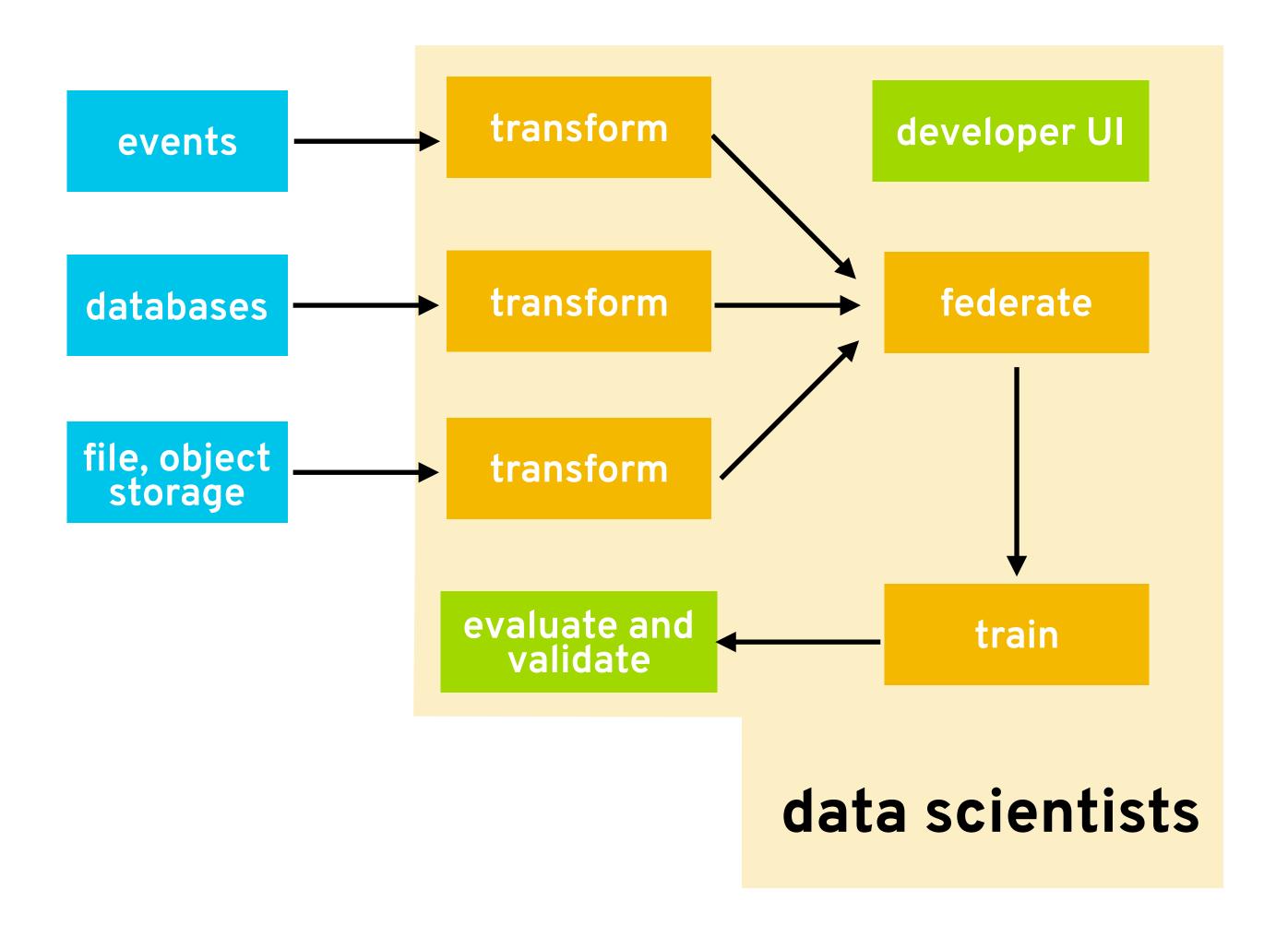






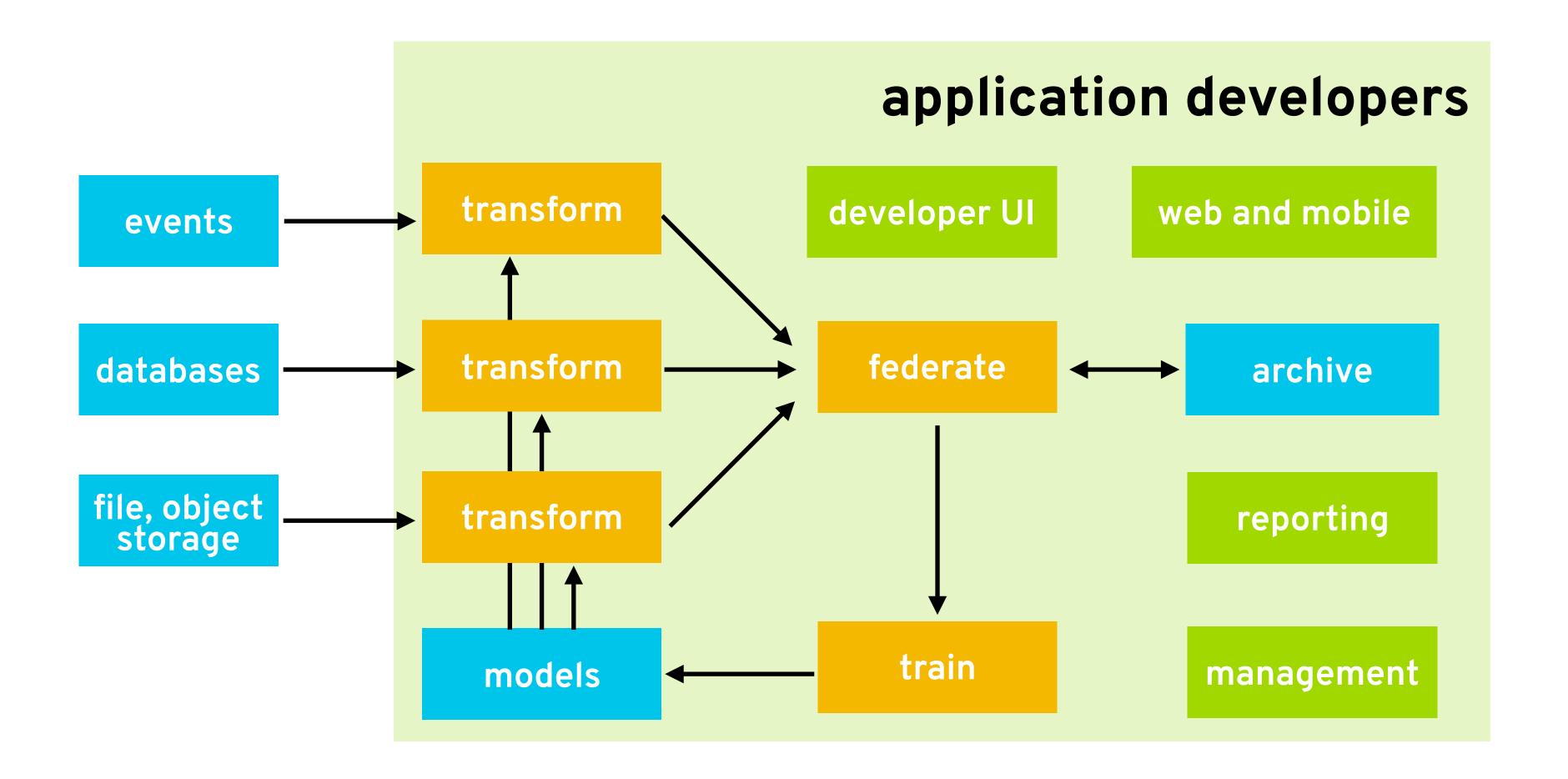






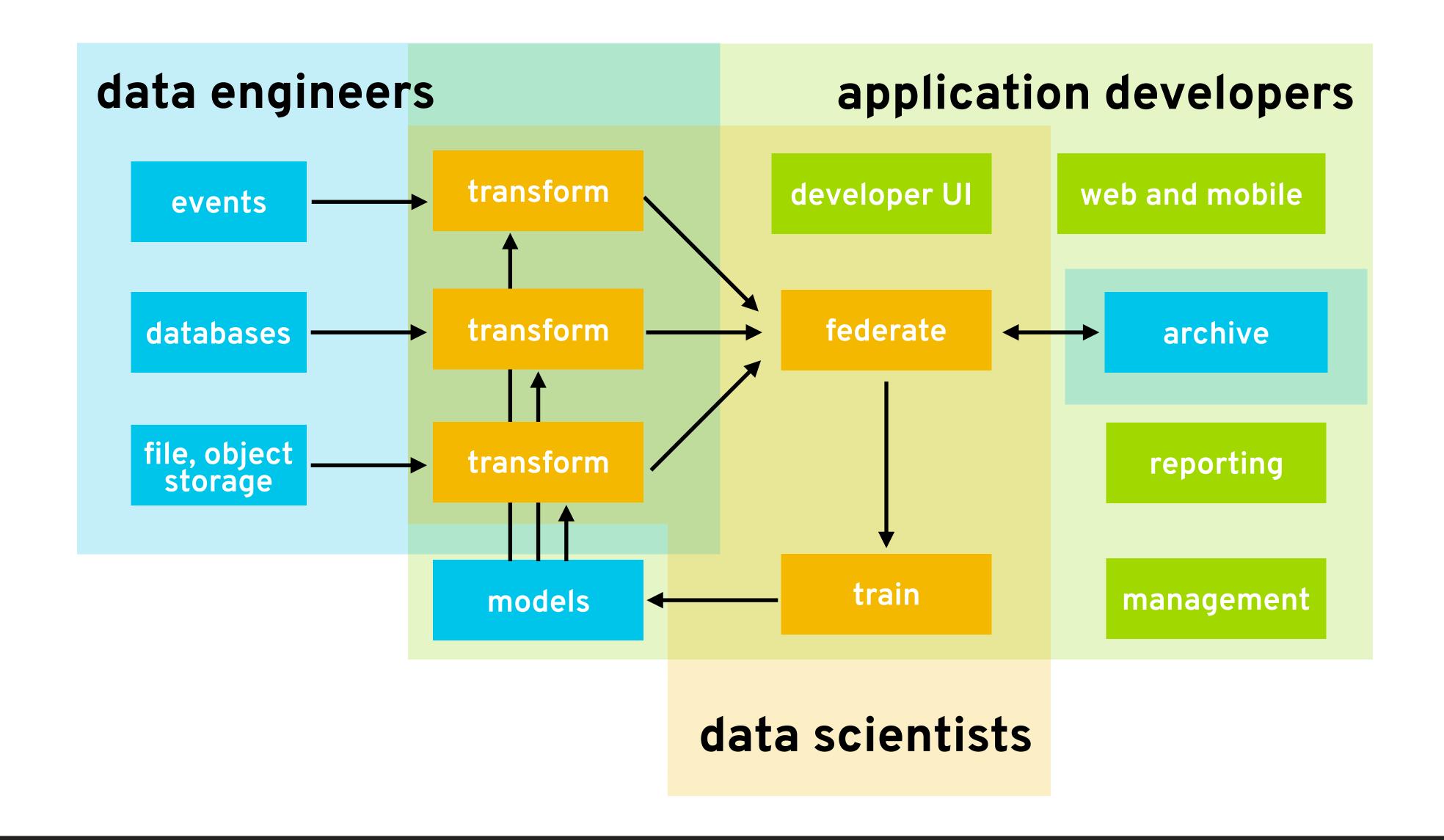




















# Containers: great for application developers and data scientists

# What is a container?







```
% pip install numpy
```





/usr/bin/pip executable arguments pip install numpy environment LANG=en\_US USER=willb ... virtual memory file handles root filesystem process table network routes





executable arguments environment virtual memory file handles root filesystem process table network routes





executable

/usr/bin/pip

arguments

oip install numpy

Software Failure. Press left mouse button to continue.

Guru Meditation #888888884.88888AC8

root filesystem

process table

network routes





executable

/usr/bin/pip

arguments

oip install numpy

Software Failure. Press left mouse button to continue.

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root filesystem

process table

network routes





executable arguments environment virtual memory file handles root filesystem process table network routes





executable arguments environment virtual memory file handles /var/lib/envs/main root filesystem process table network routes





executable

arguments

environment

virtual memory

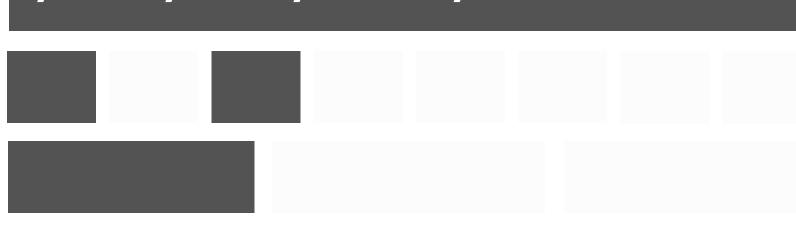
file handles

/var/lib/envs/main

root filesystem

process table

network routes









executable pip install numpy pip install riskylib arguments environment virtual memory file handles /var/lib/envs/main root filesystem /var/lib/envs/risky process table network routes













base image





configuration and installation recipes

base image





user application code

configuration and installation recipes

base image





user application code

configuration and installation recipes

base image

a6afd91e

6b8cad3e

33721112

e8cae4f6

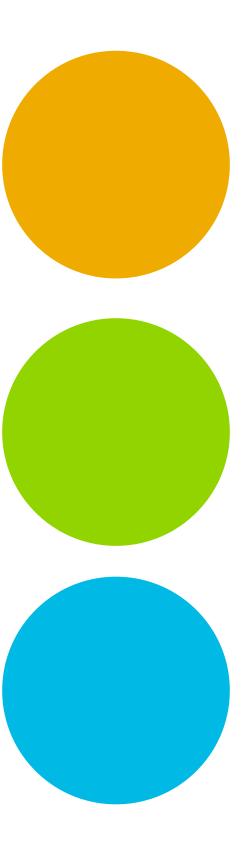
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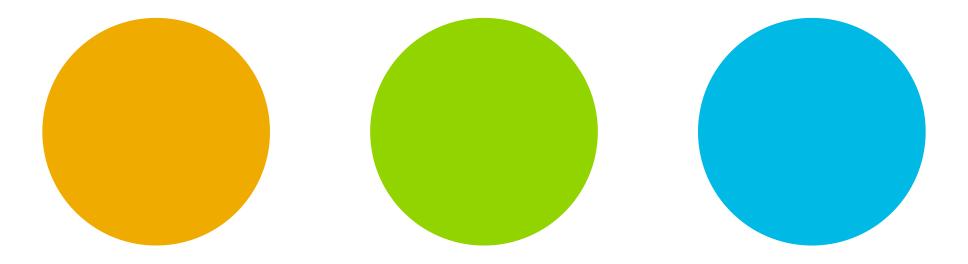






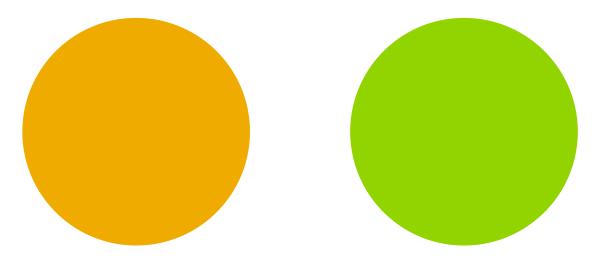






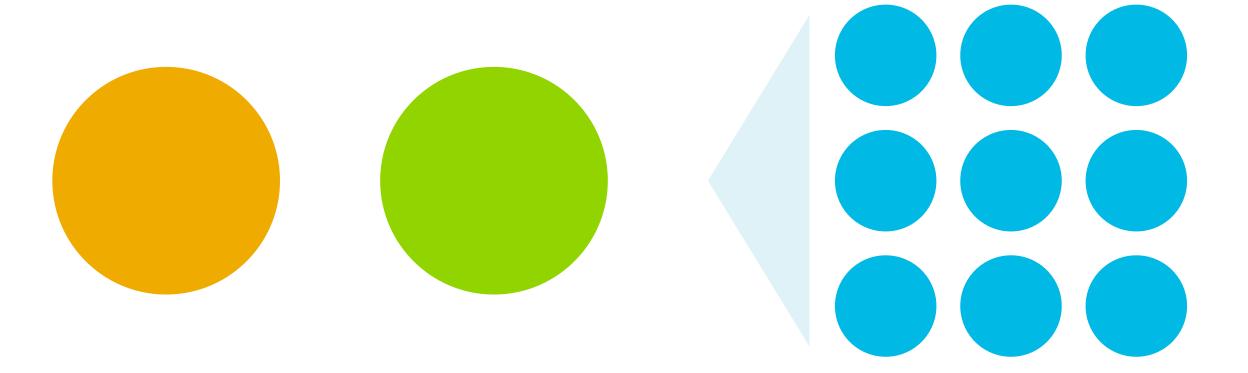








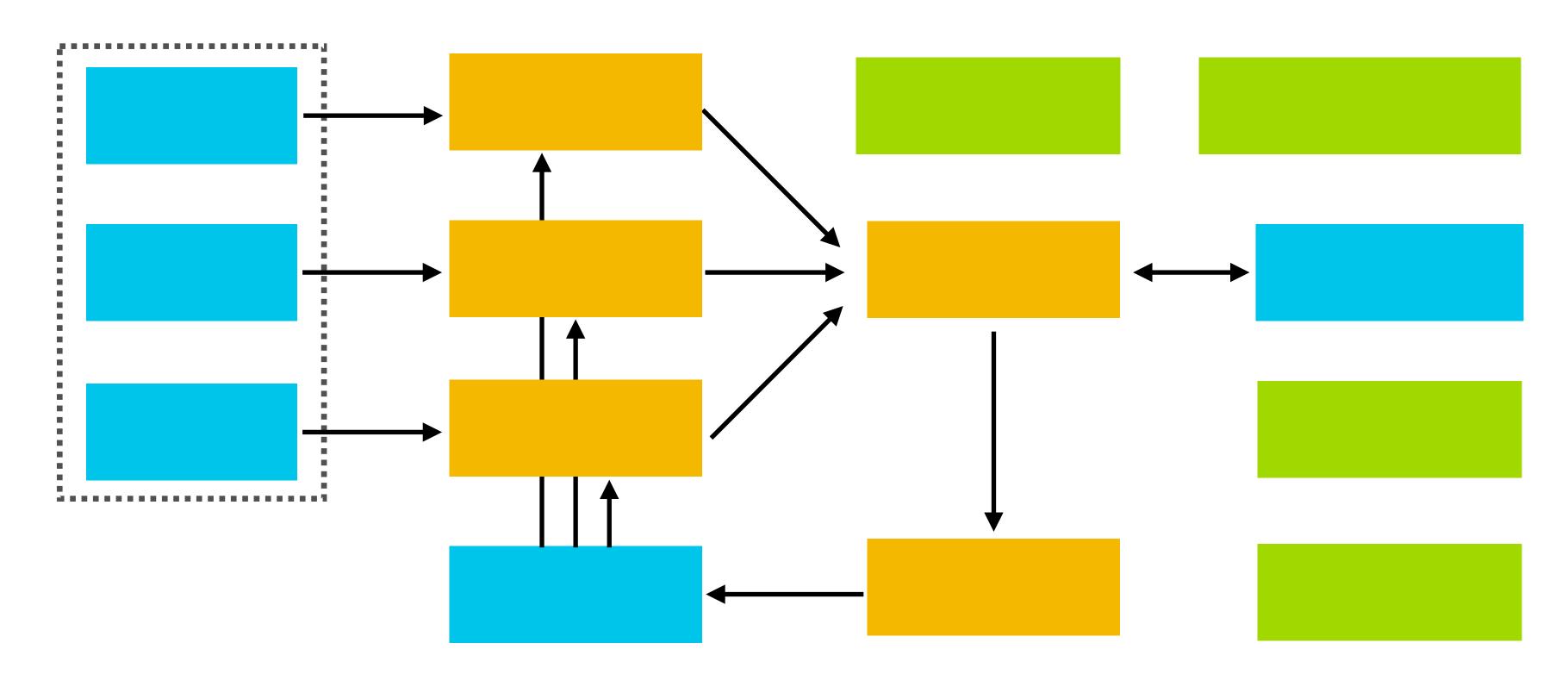








# Declarative app configuration





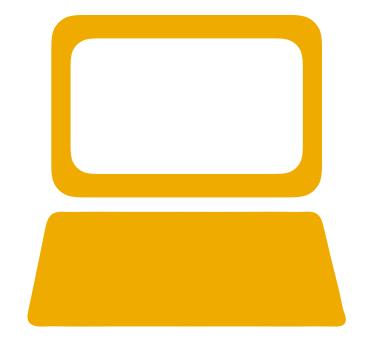


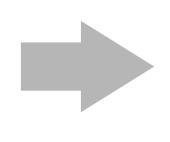
### Integration and deployment

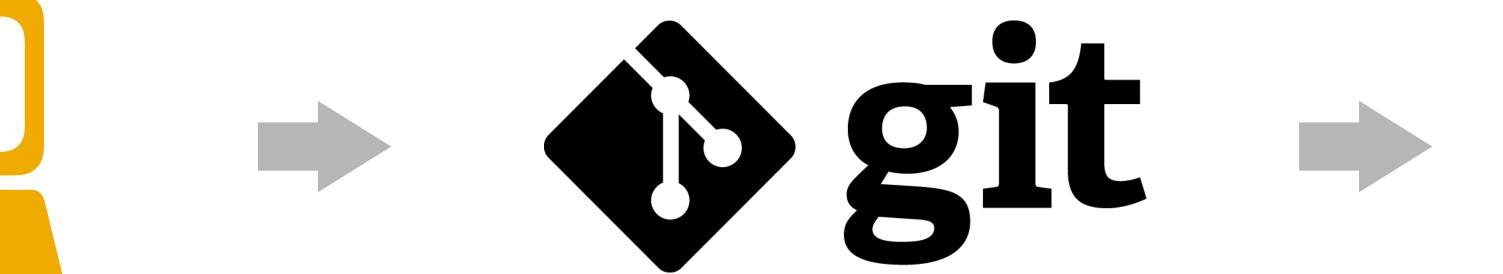


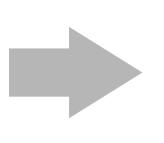








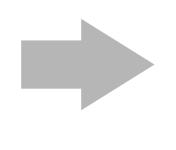


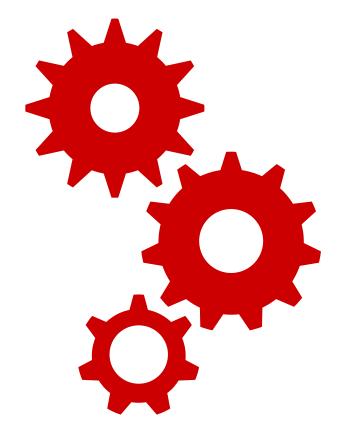


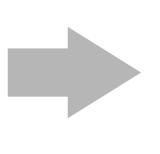












application code

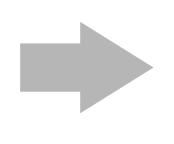
configuration and installation recipes

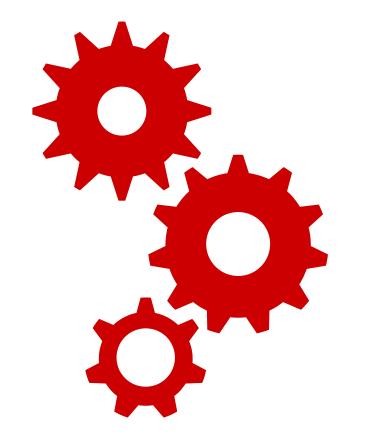
base image

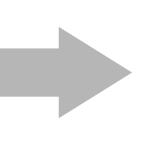












application code

configuration and installation recipes

base image

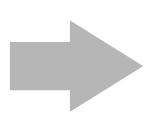


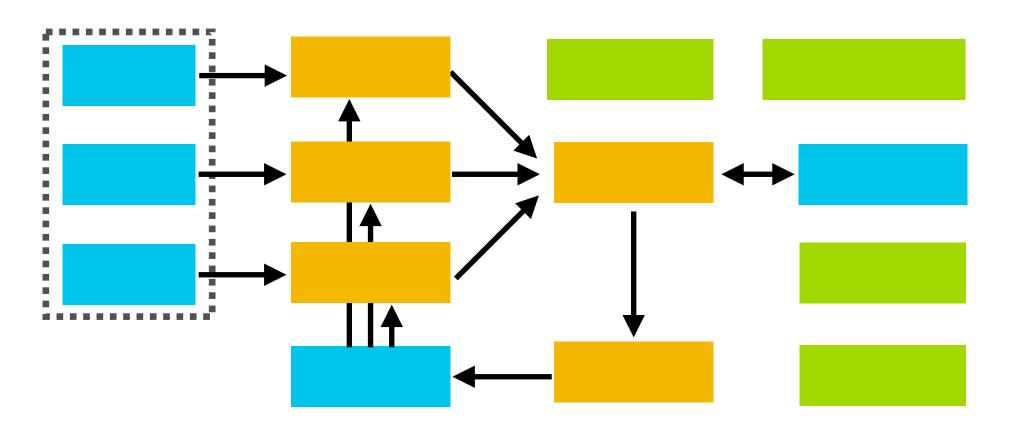


application code

configuration and installation recipes

base image

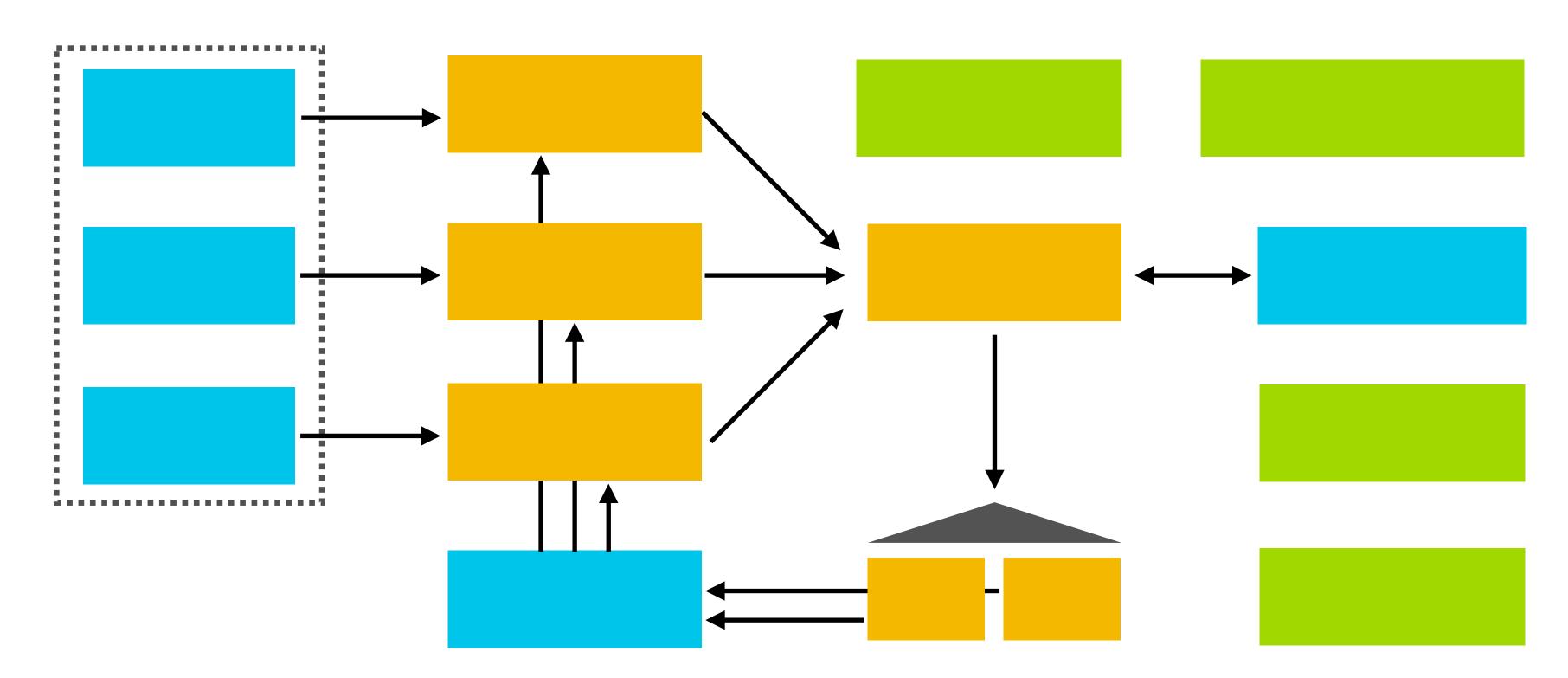








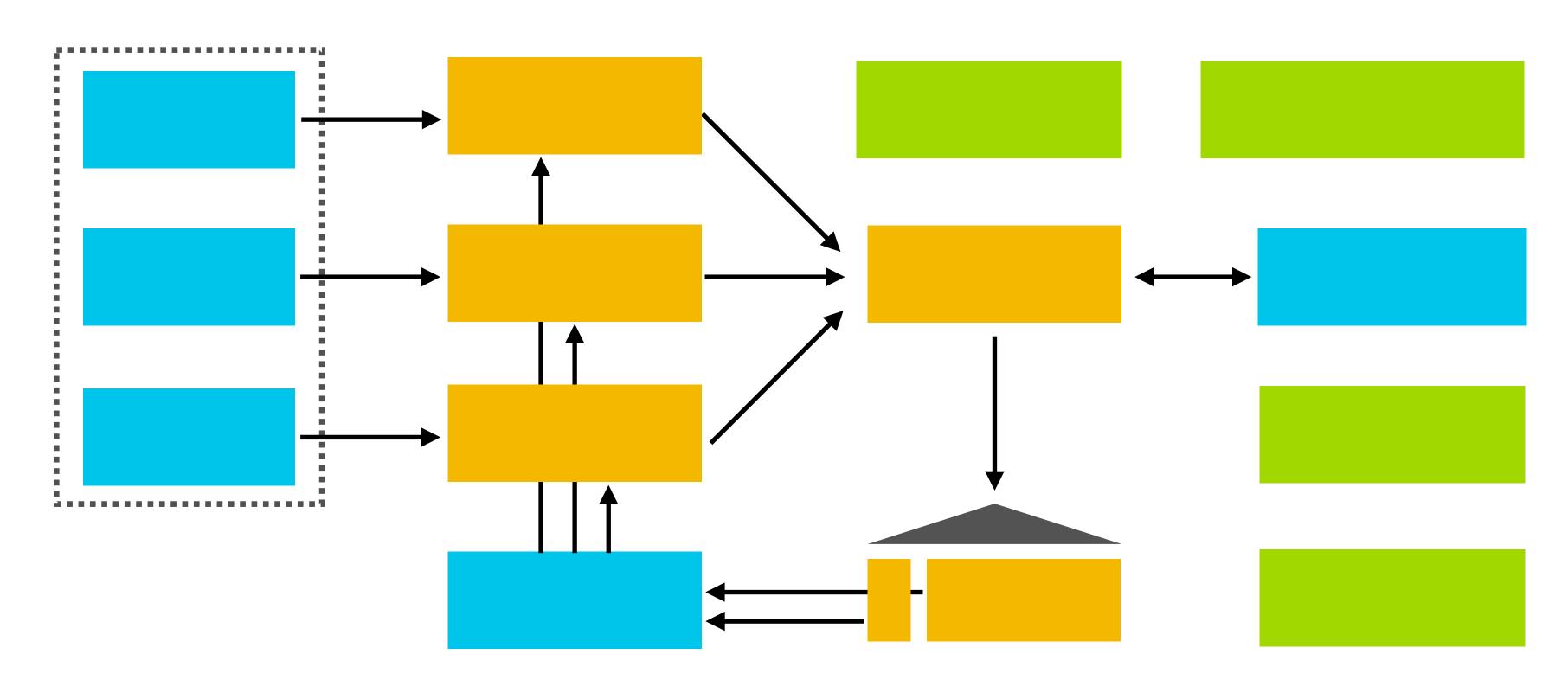
# Flexible service routing







# Flexible service routing







# How can we build intelligent applications in containers?

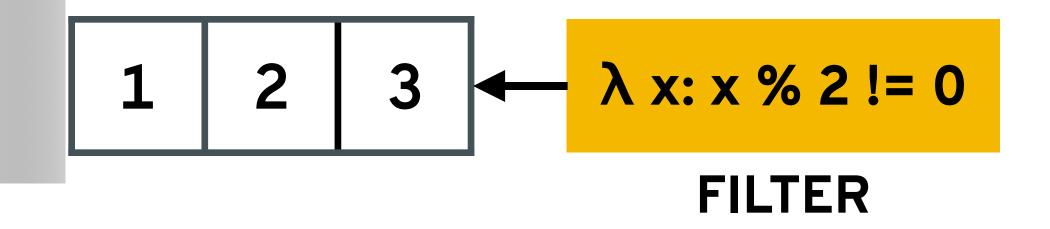




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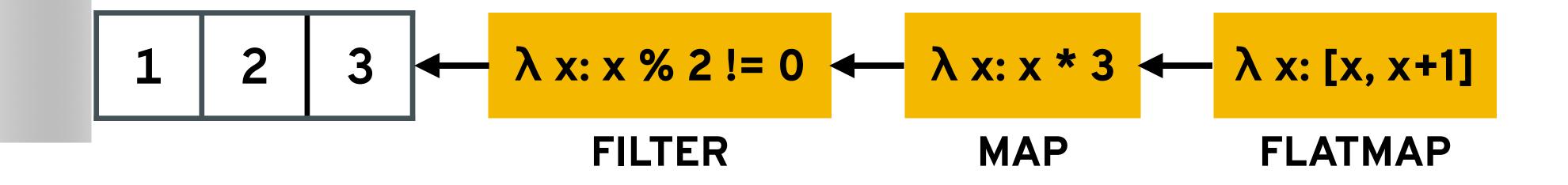






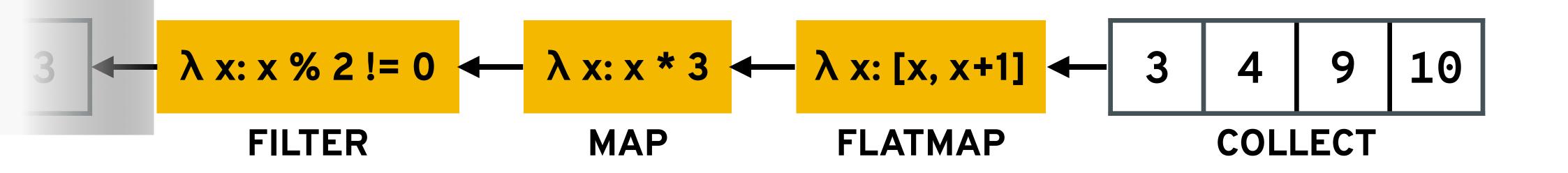












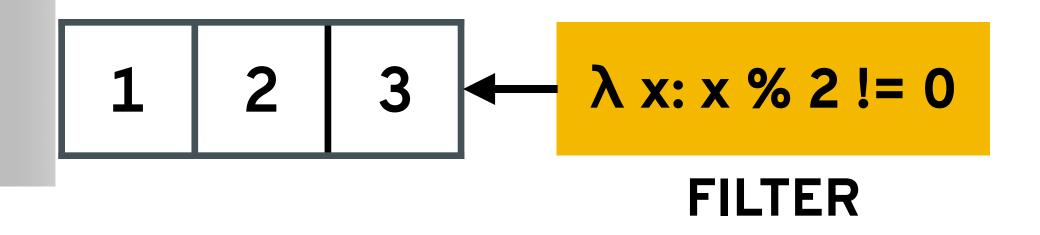




1 2 3

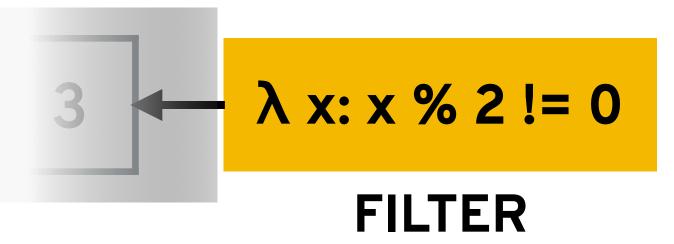






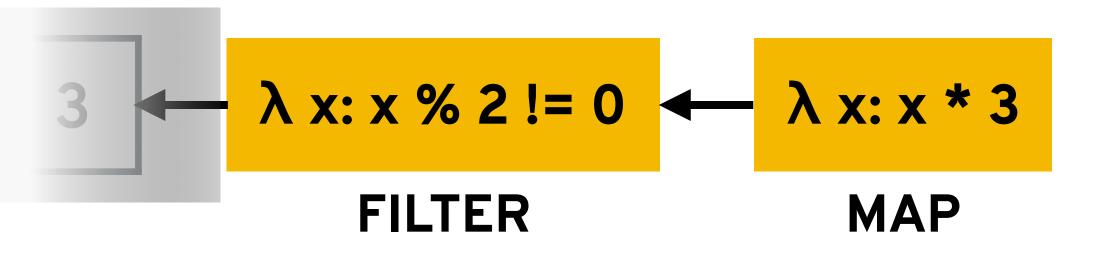






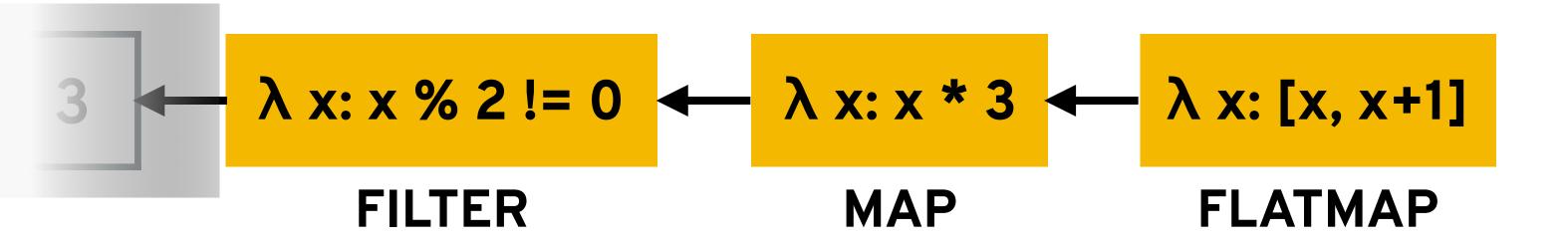






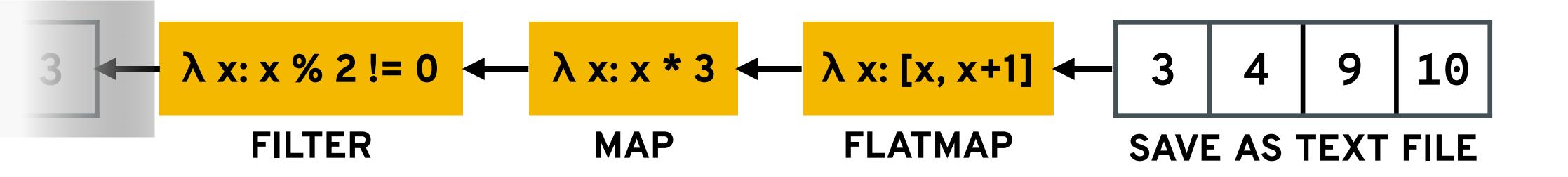






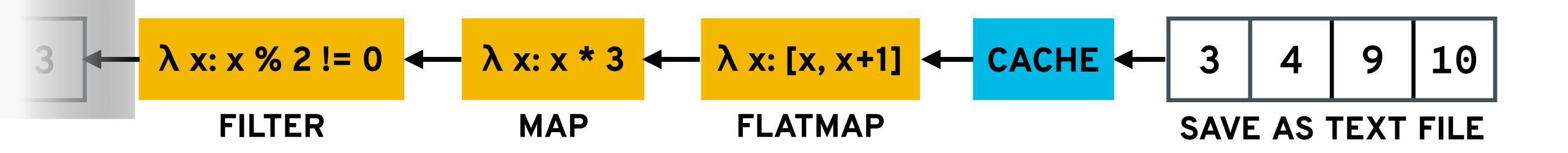
















### Spark core





Graph SQL ML Streaming

Spark core





Graph SQL ML Streaming

Spark core

ad hoc Mesos YARN





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 6
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 9
 10
 11
 12

 executor
 executor
 executor
 executor
 master



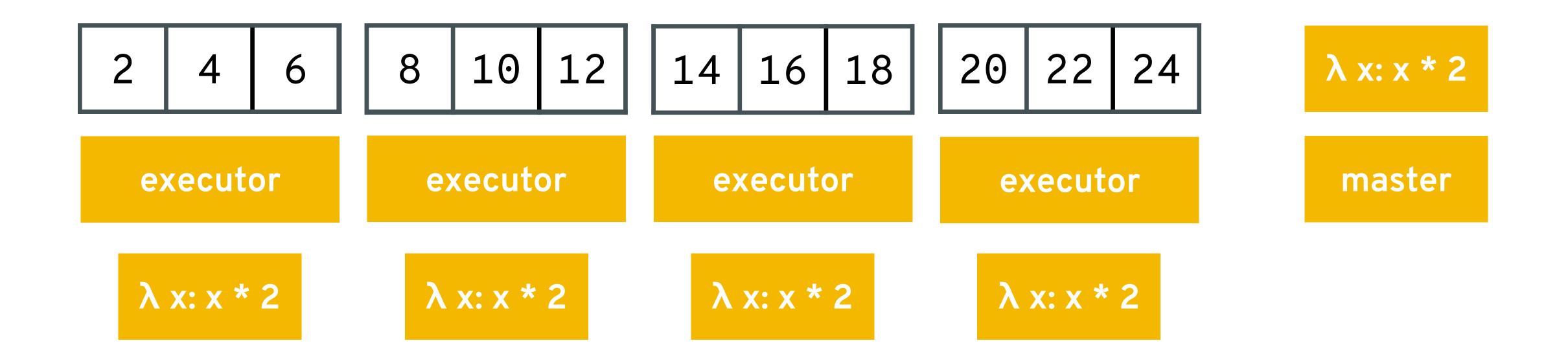


 1
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 11
 12
 λ x: x \* 2

 executor
 executor
 executor
 executor
 master











2 4 6 8 10 12

executor

executor

λ x: x \* 2

λ x: x \* 2

20 22 24

executor

master

λ x: x \* 2

λ x: x \* 2





 2
 4
 6
 8
 10
 12
 14
 16
 18
 20
 22
 24
 λ x: x \* 2

 executor
 executor
 executor
 master

 λ x: x \* 2
 λ x: x \* 2
 λ x: x \* 2
 λ x: x \* 2





Goal: use the **same abstraction** for batch and "streaming" (micro-batch) data by **dividing a stream** into **many small RDDs**.



input stream





Goal: use the **same abstraction** for batch and "streaming" (micro-batch) data by **dividing a stream** into **many small RDDs**.

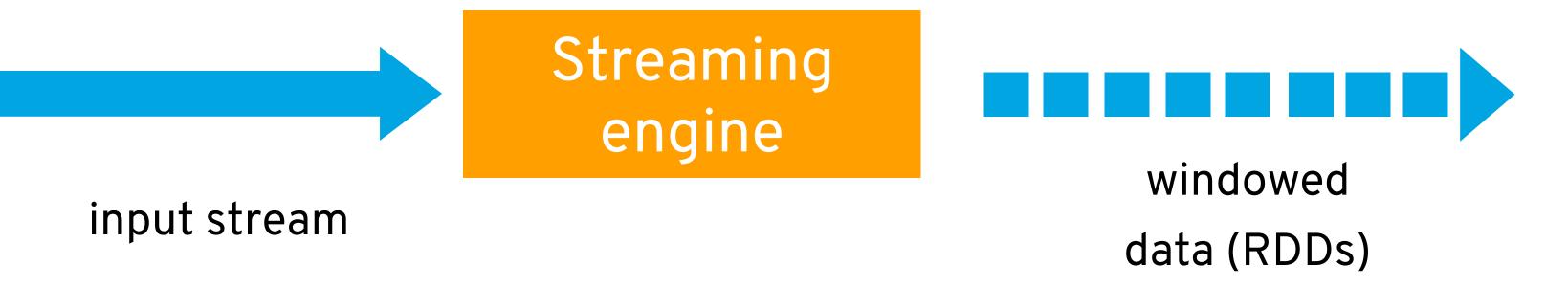


input stream





Goal: use the **same abstraction** for batch and "streaming" (micro-batch) data by **dividing a stream** into **many small RDDs**.







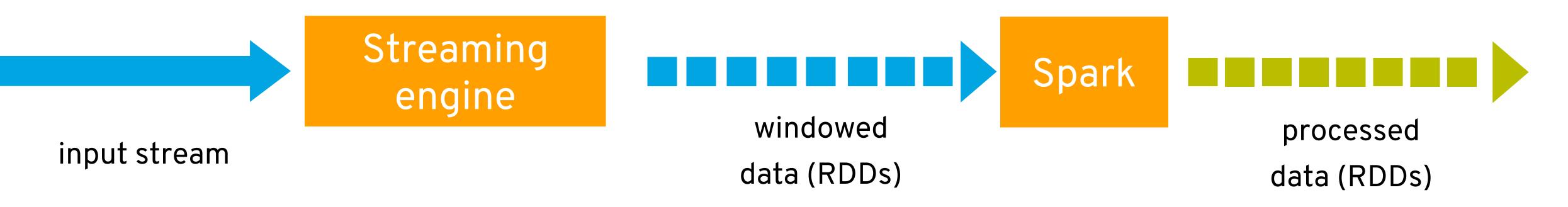
Goal: use the **same abstraction** for batch and "streaming" (micro-batch) data by **dividing a stream** into **many small RDDs**.







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### Structured queries

The capacity to run arbitrary code in RDDs is powerful but comes with **an important tradeoff**: Spark can't rearrange RDD programs to improve their performance.

Writing Spark programs with a query DSL allows Spark to generate optimized execution plans.





# Structured query in Spark

**SQL interface** (unchecked syntax or semantics)
SELECT word, COUNT(\*) FROM words GROUP BY word

Data frame interface (semantics checked at run-time) words.groupBy('word').count()

Dataset interface (mostly checked at compile-time)





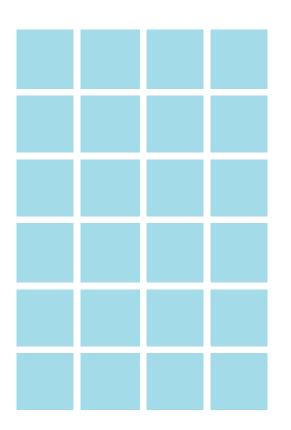
### Query planning

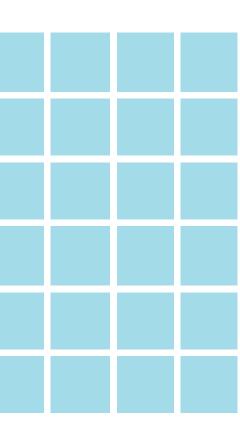
```
SELECT * FROM A, B WHERE
A.ID = B.ID AND
uncommon(A.X) AND
extremelyRare(B.Y)
```





# Anaïve plan

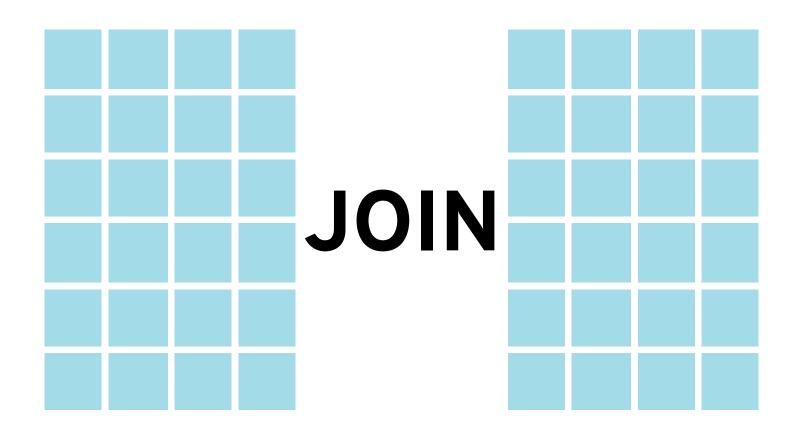








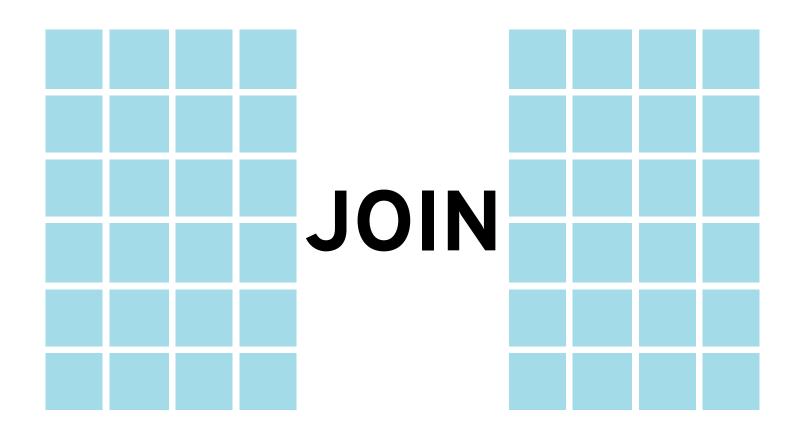
# Anaive plan

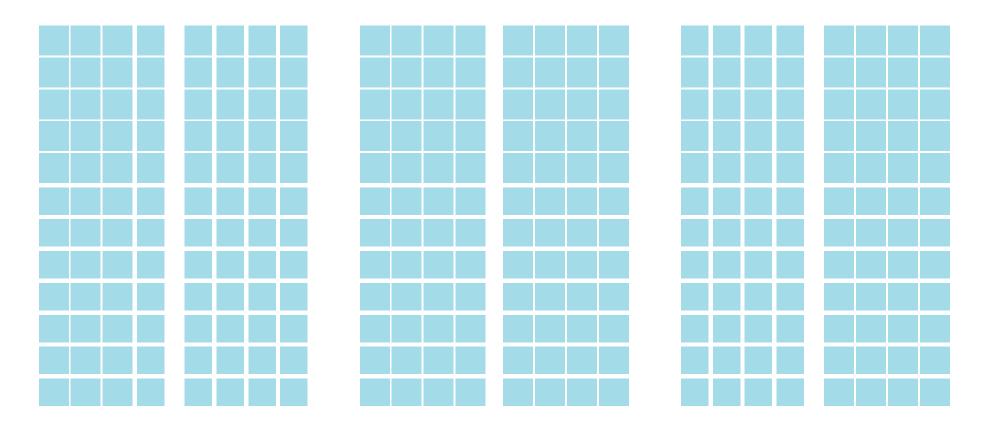






# Anaïve plan

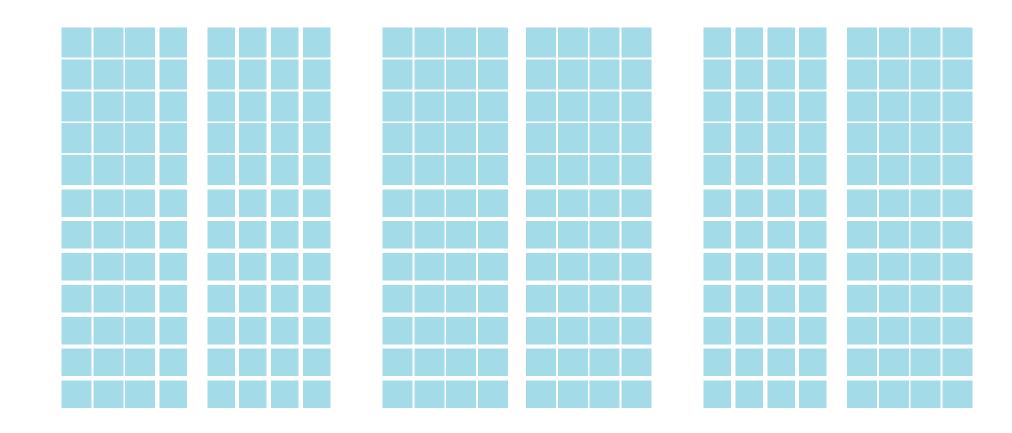








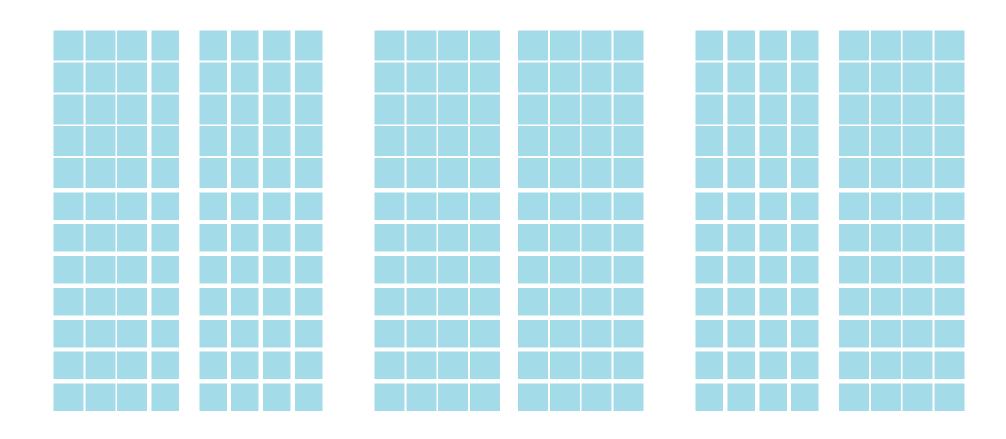
# Anaïve plan







# Anaive plan

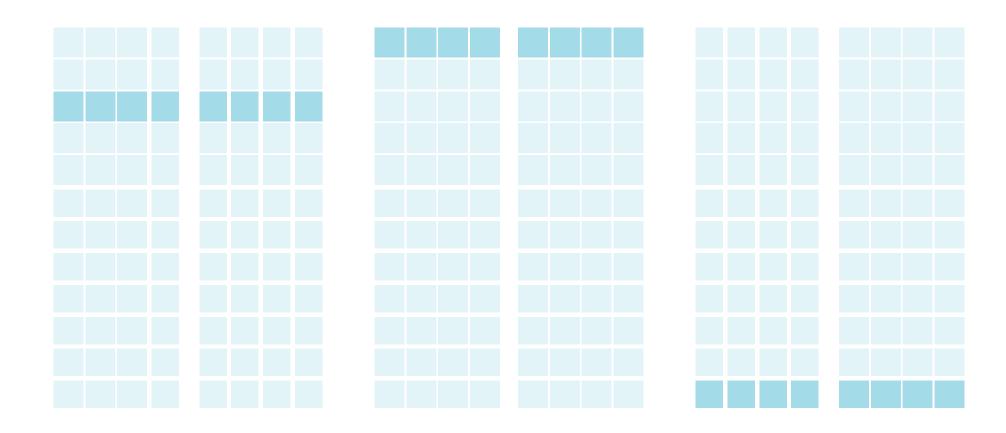


FILTER





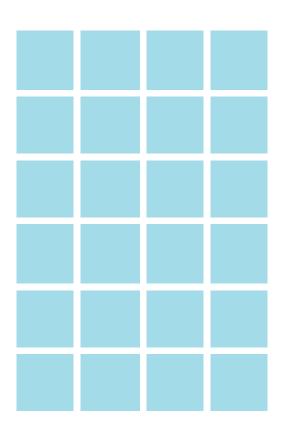
# Anaïve plan

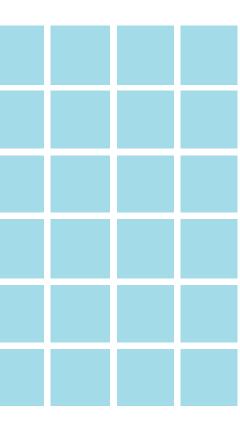


**FILTER** 



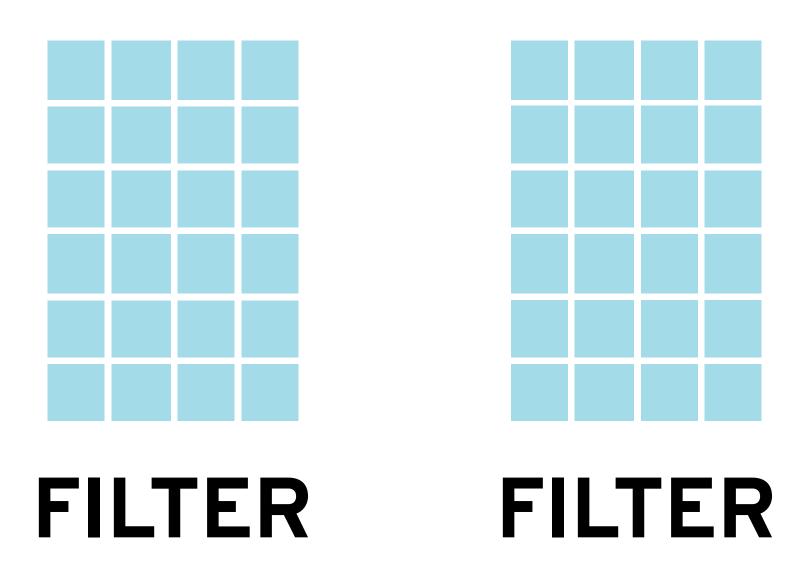






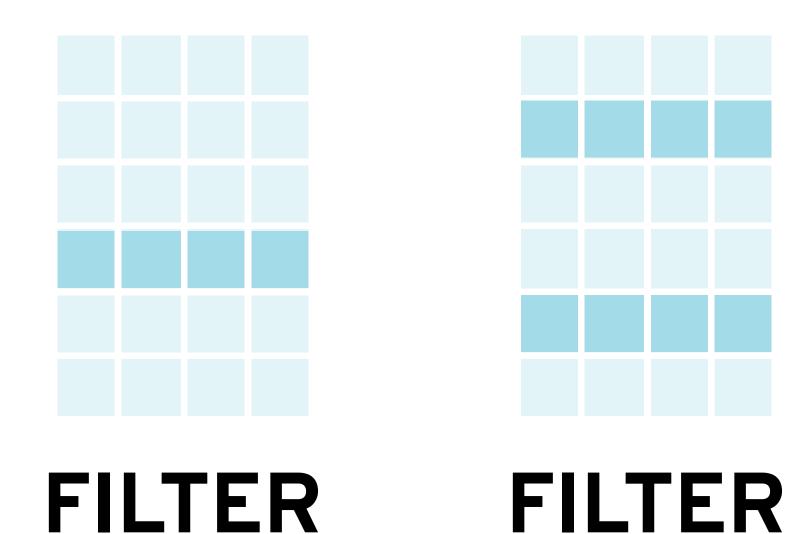






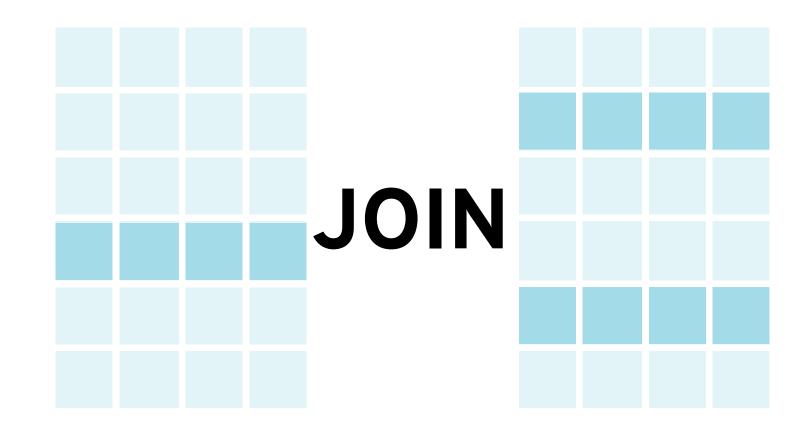






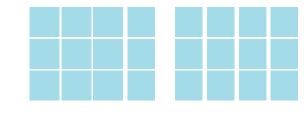
















Structured streaming combines stream processing with query planning for high-performance analytics on events!





events





events message bus

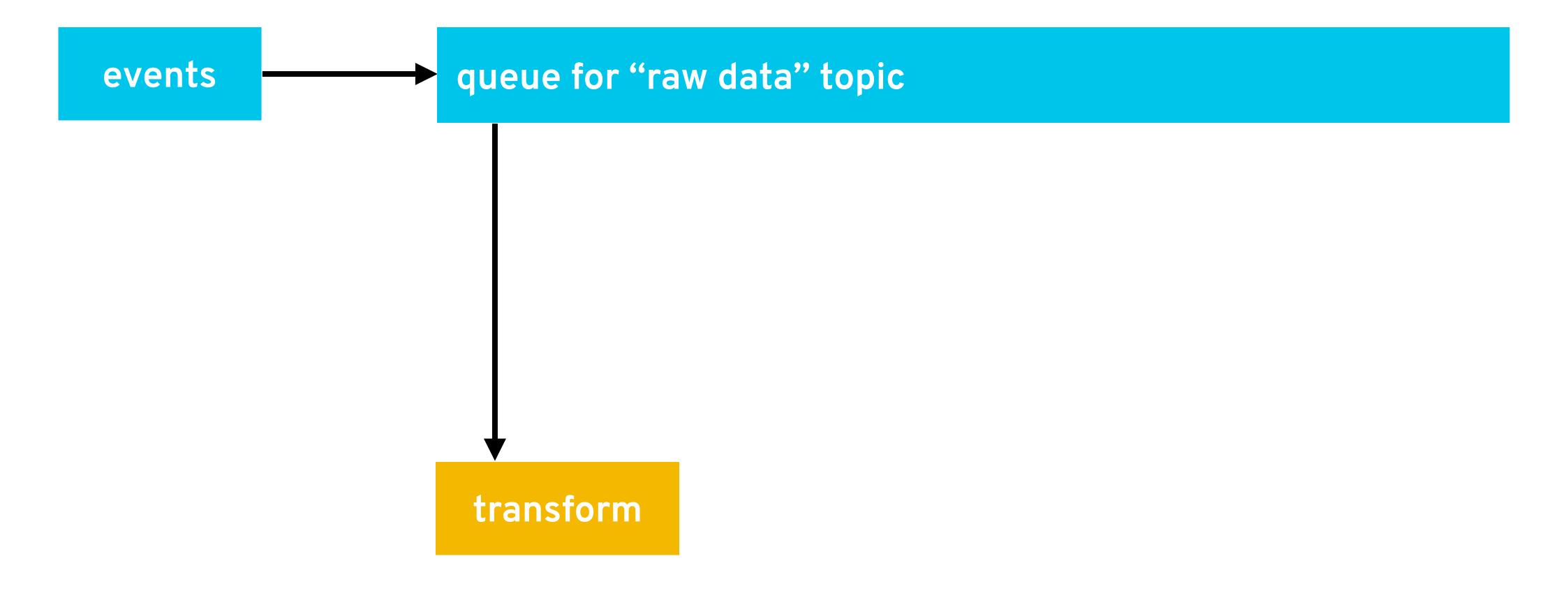




events queue for "raw data" topic

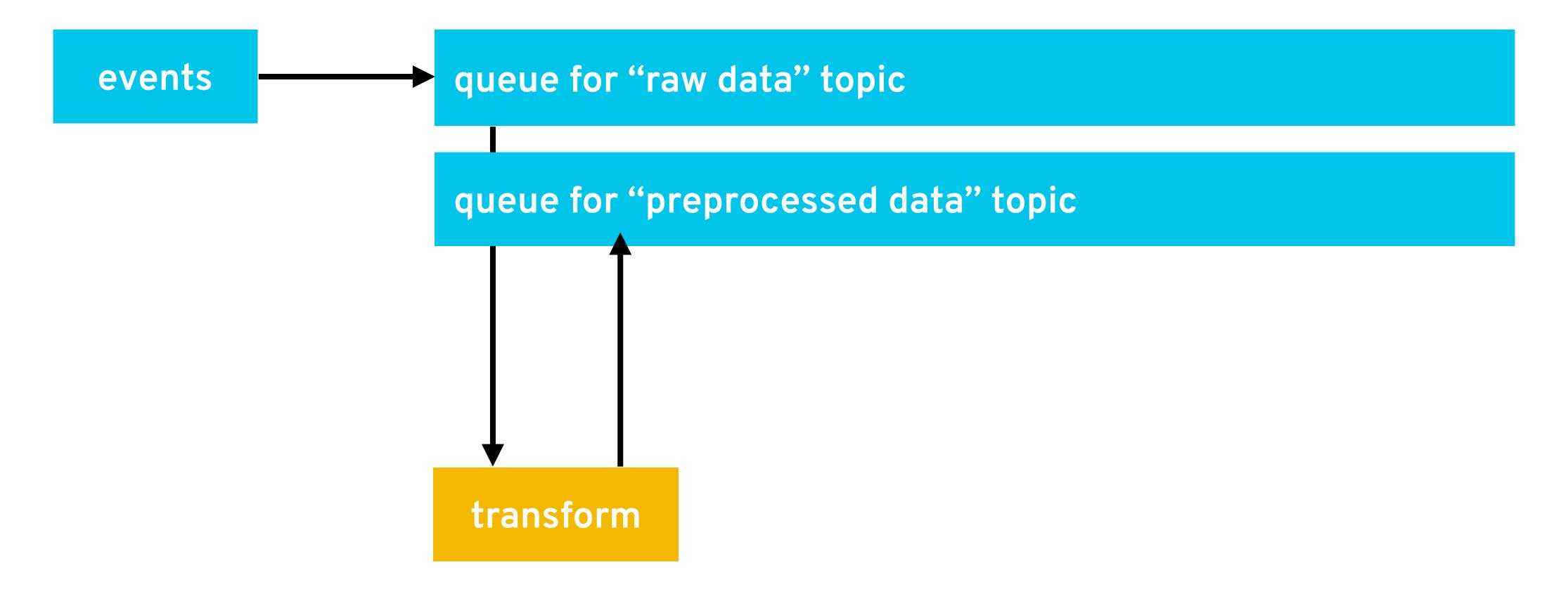






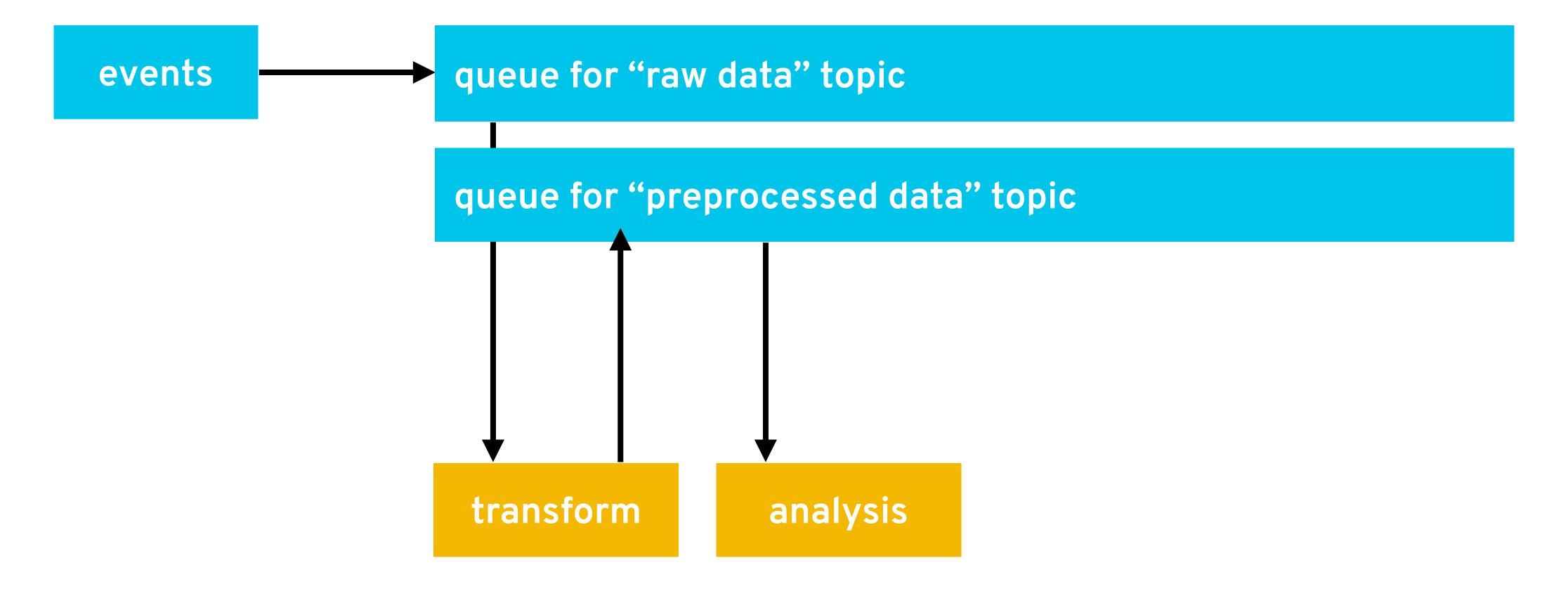






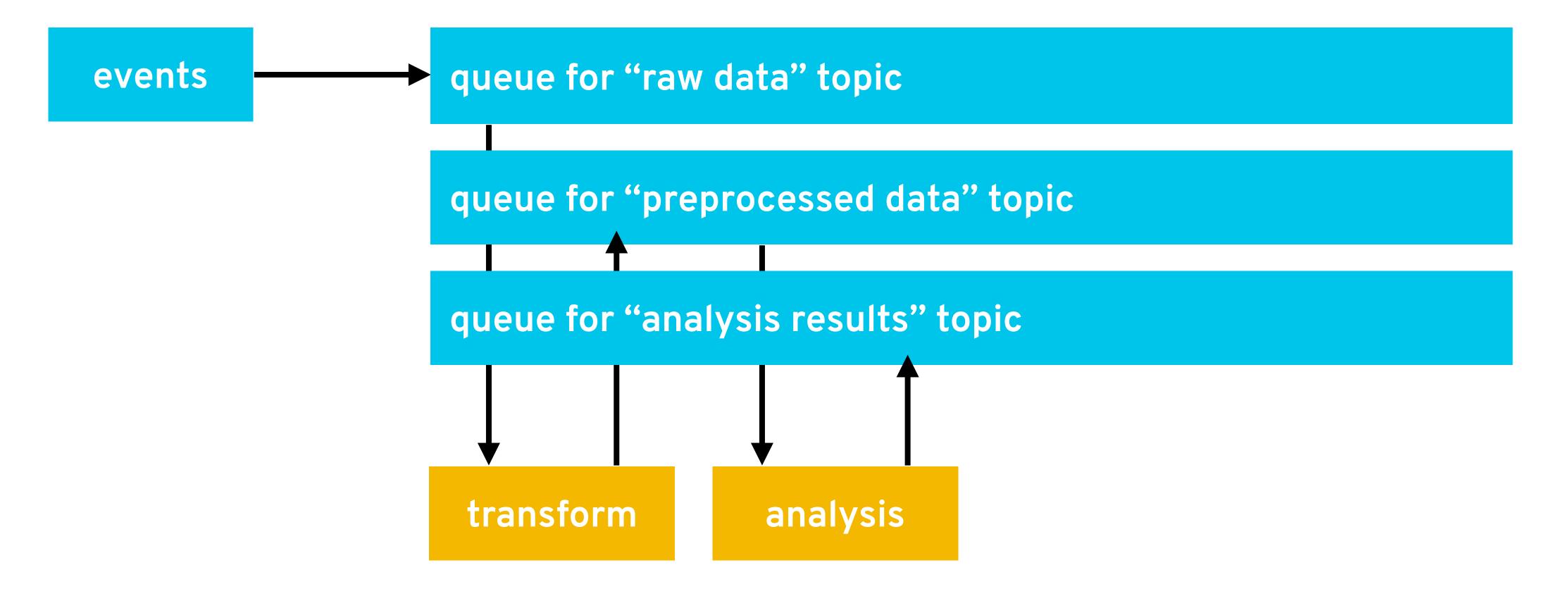






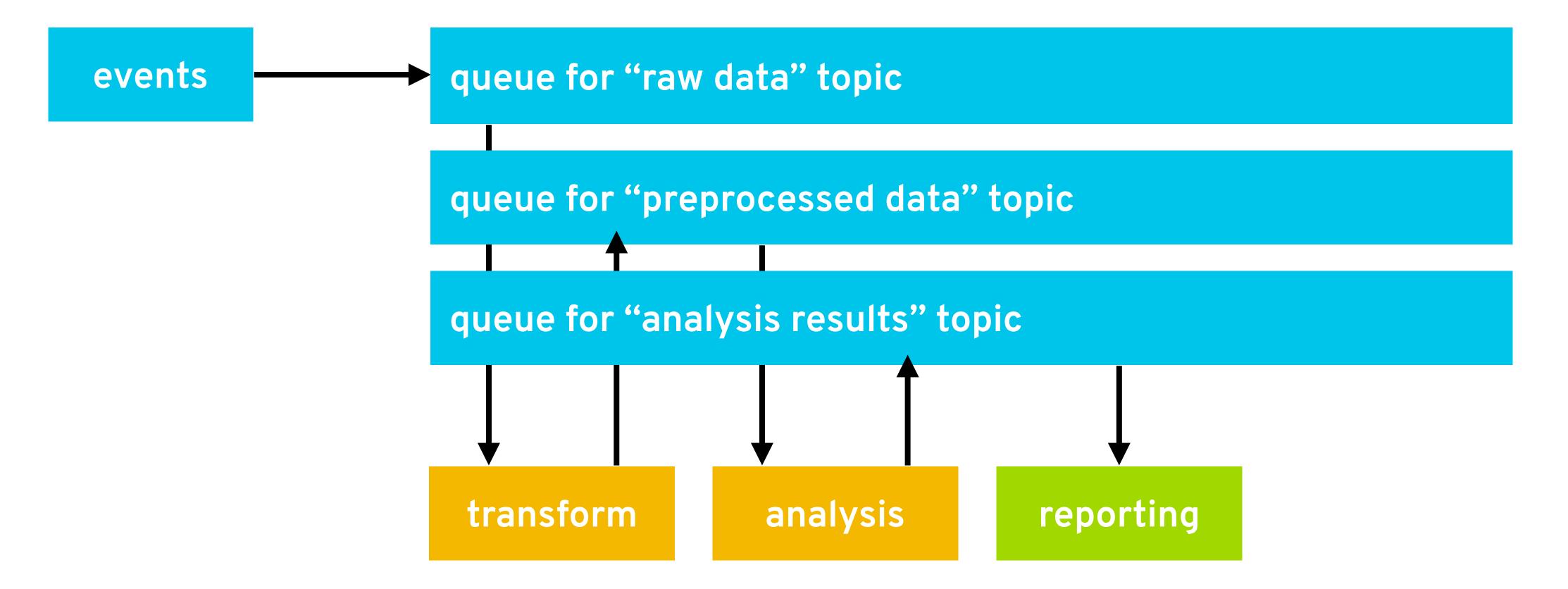






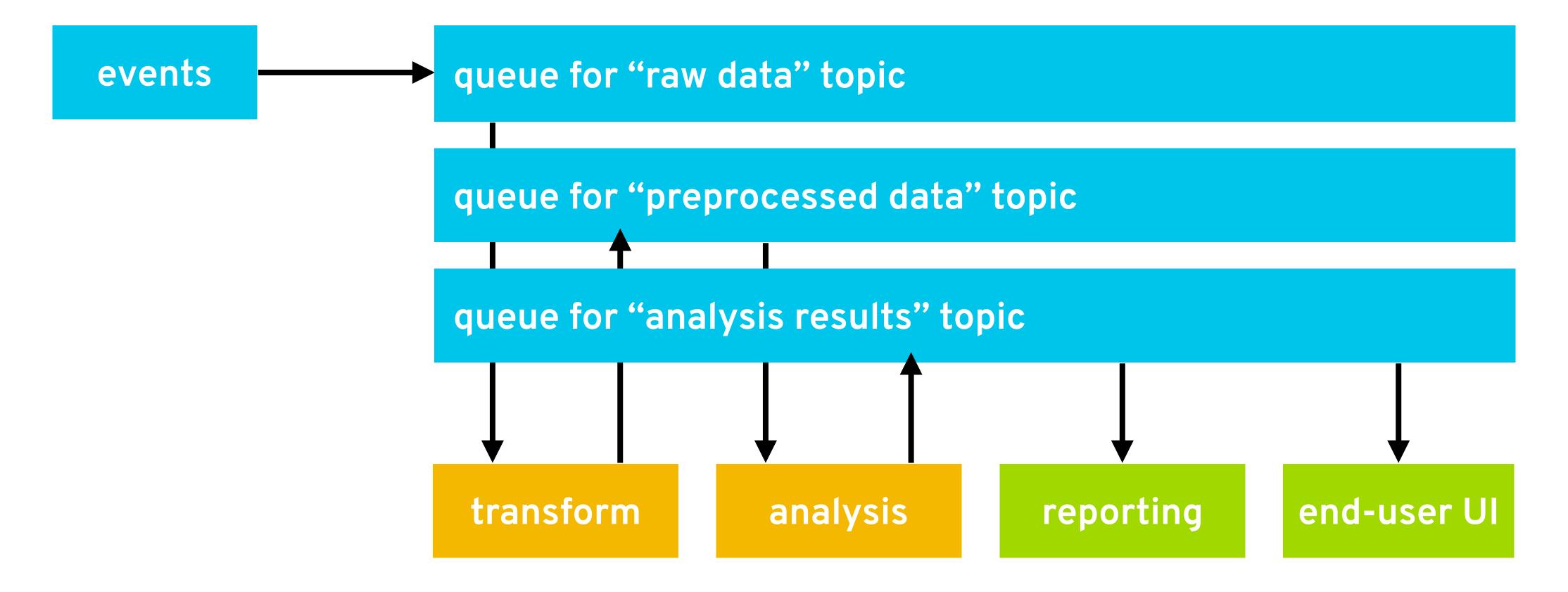












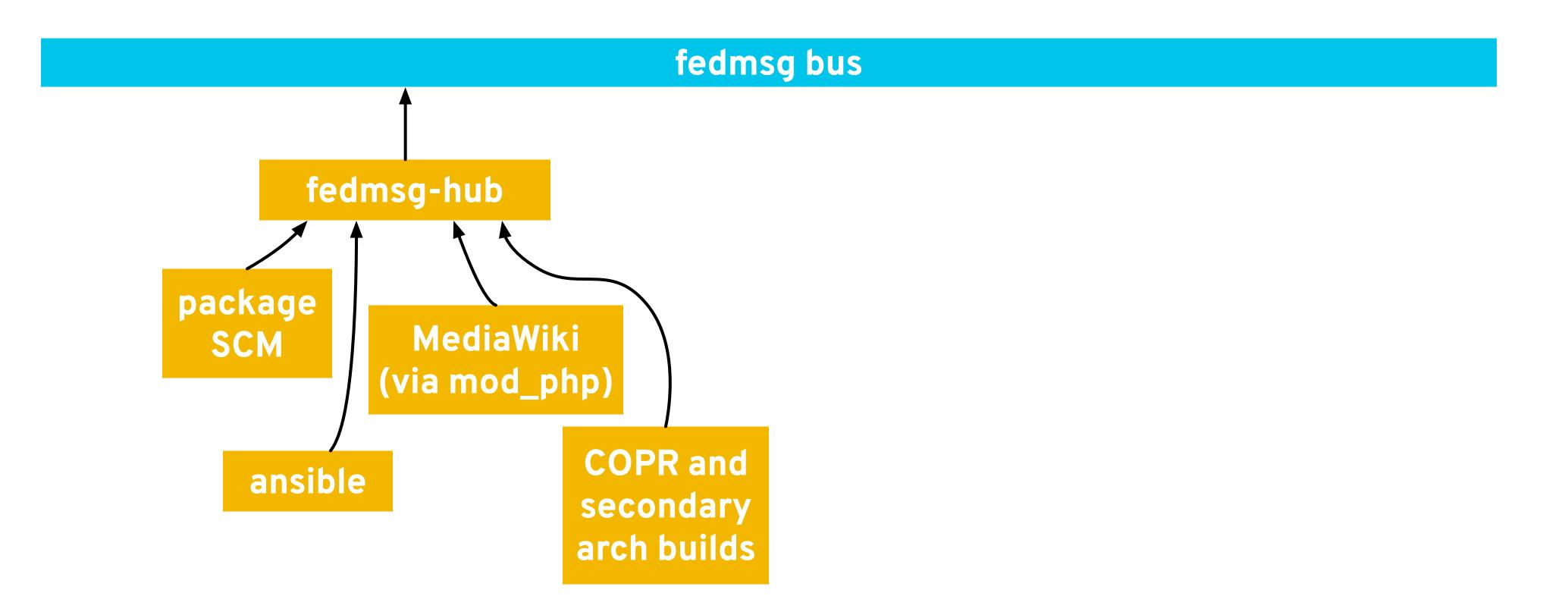




#### fedmsg bus

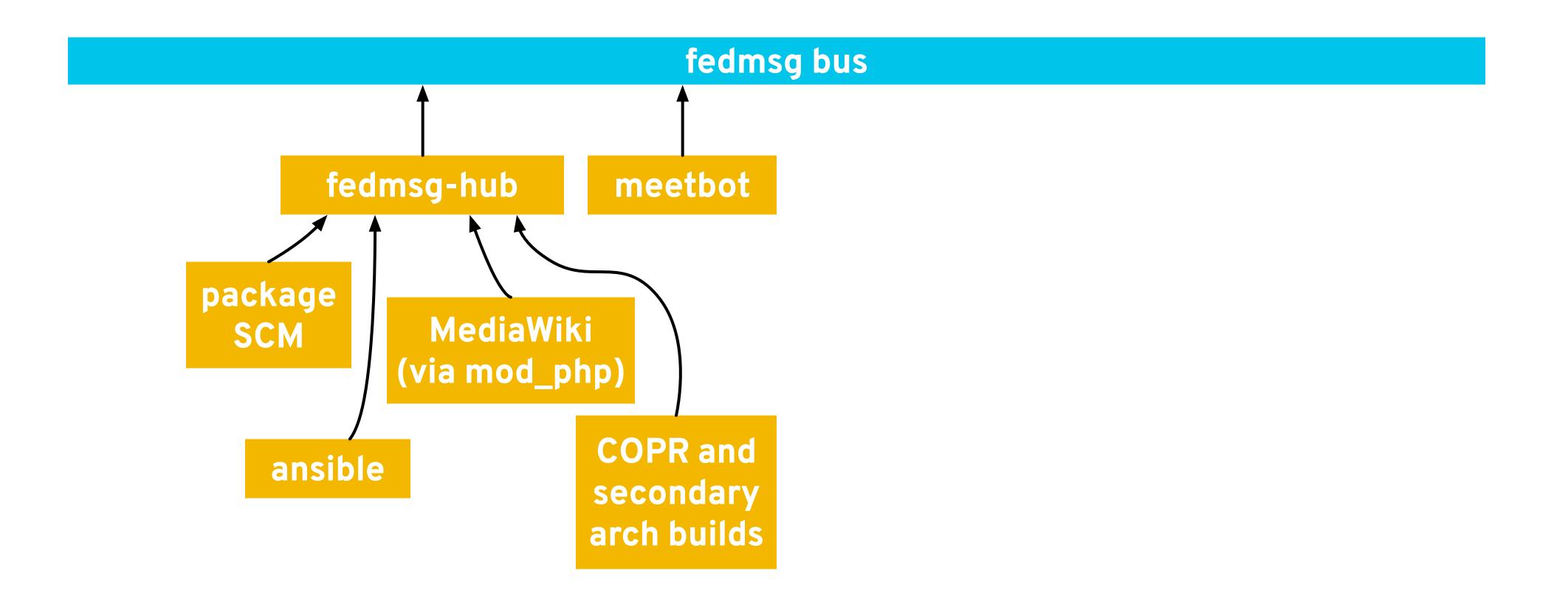






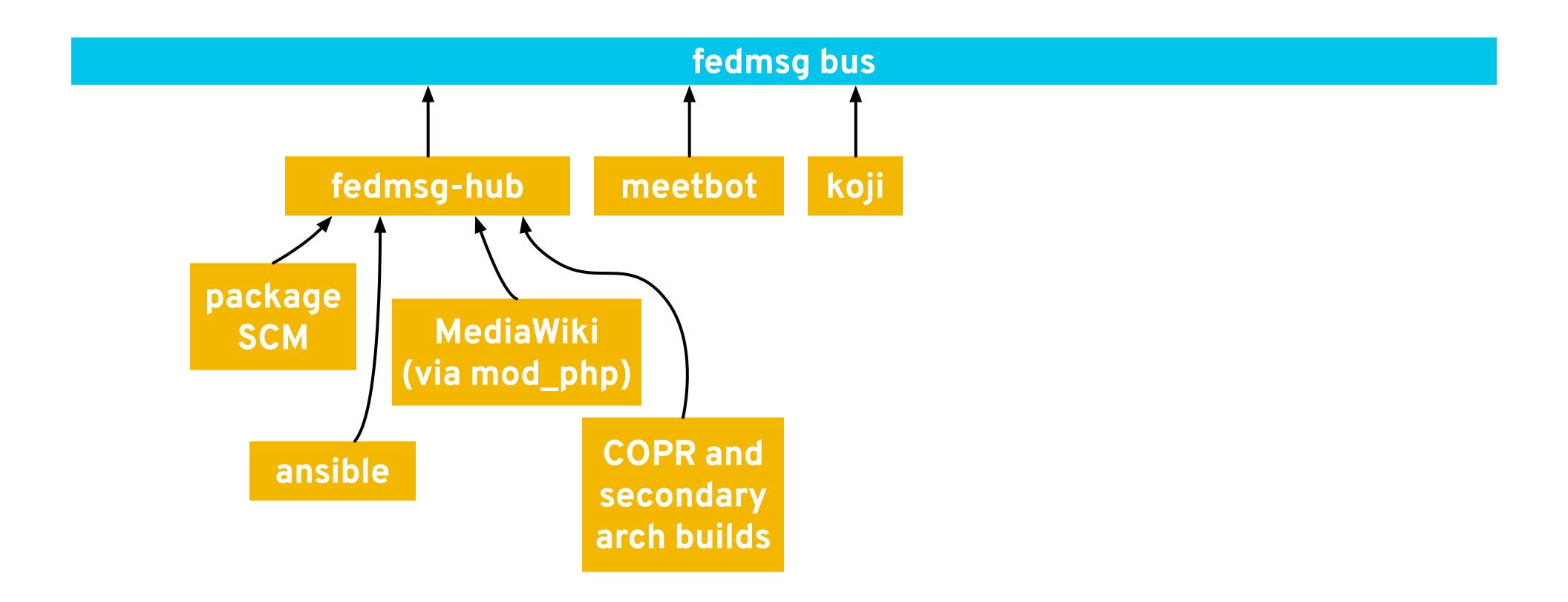






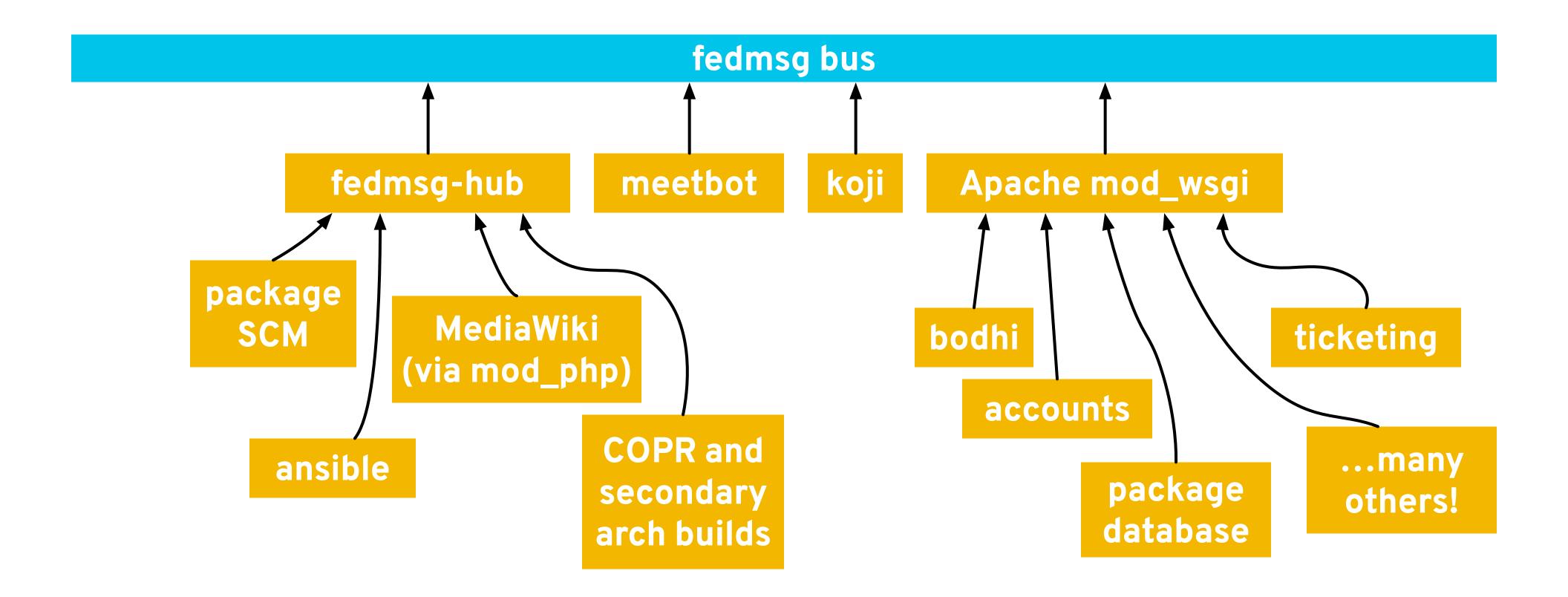






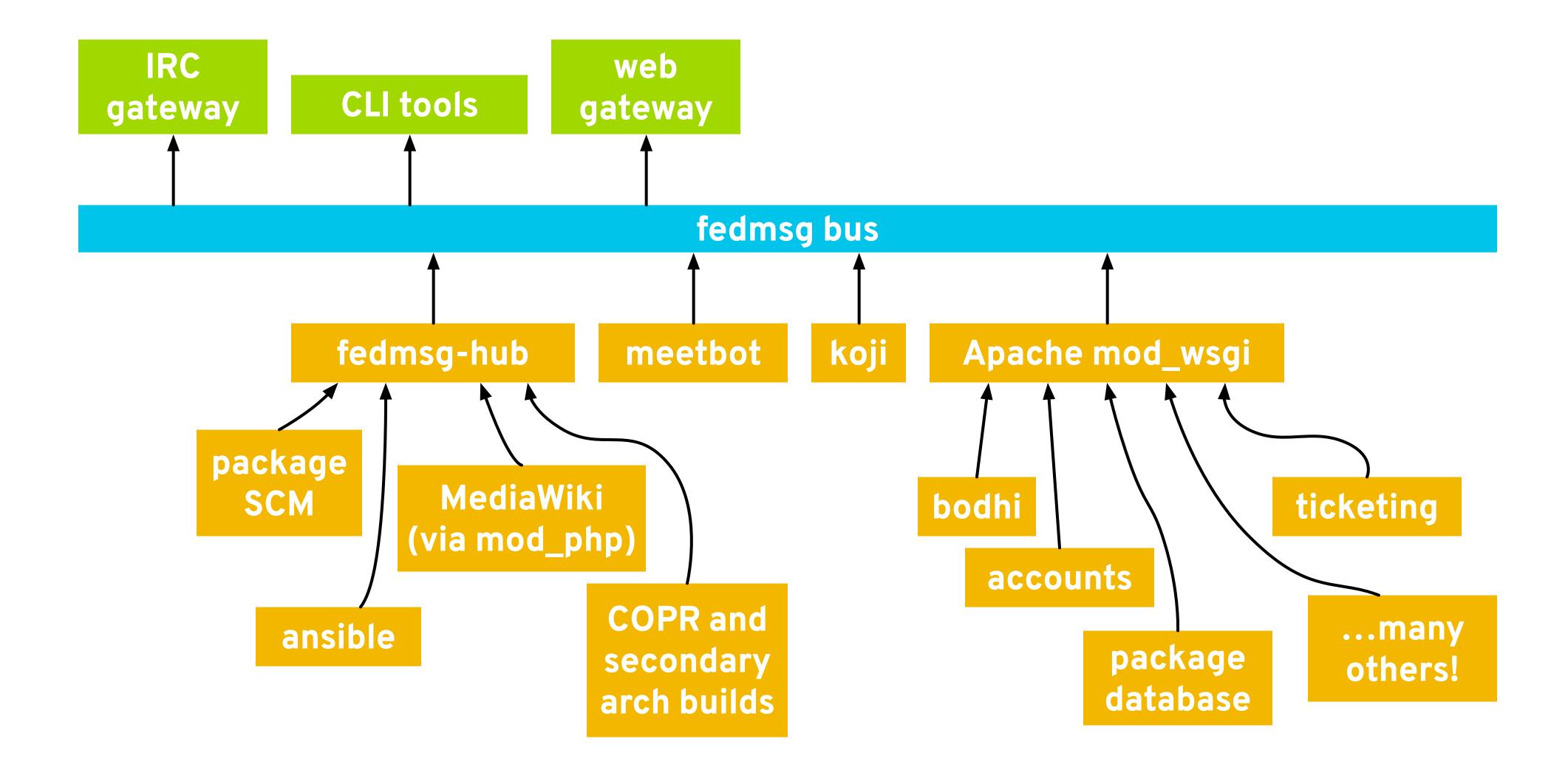






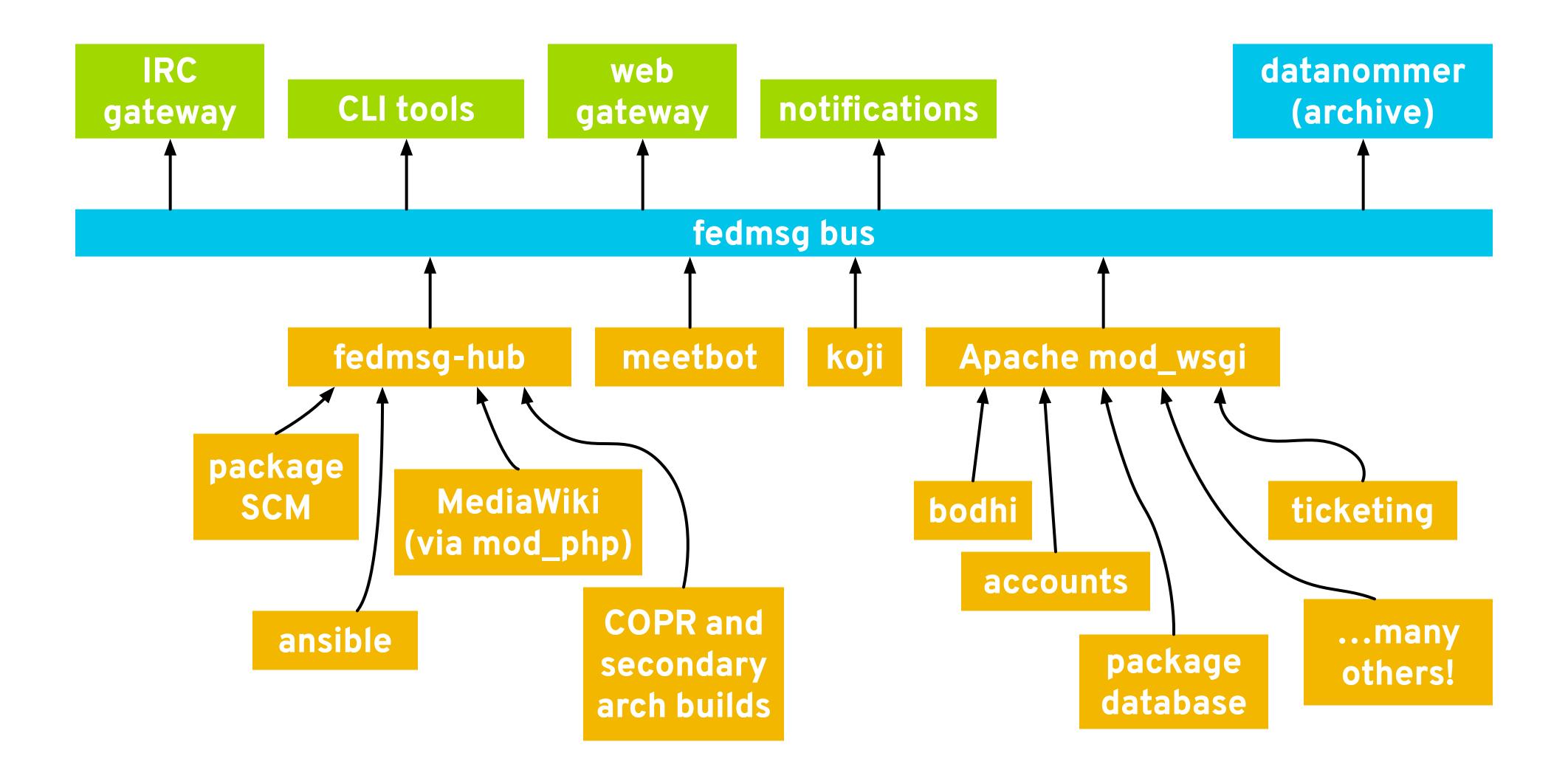






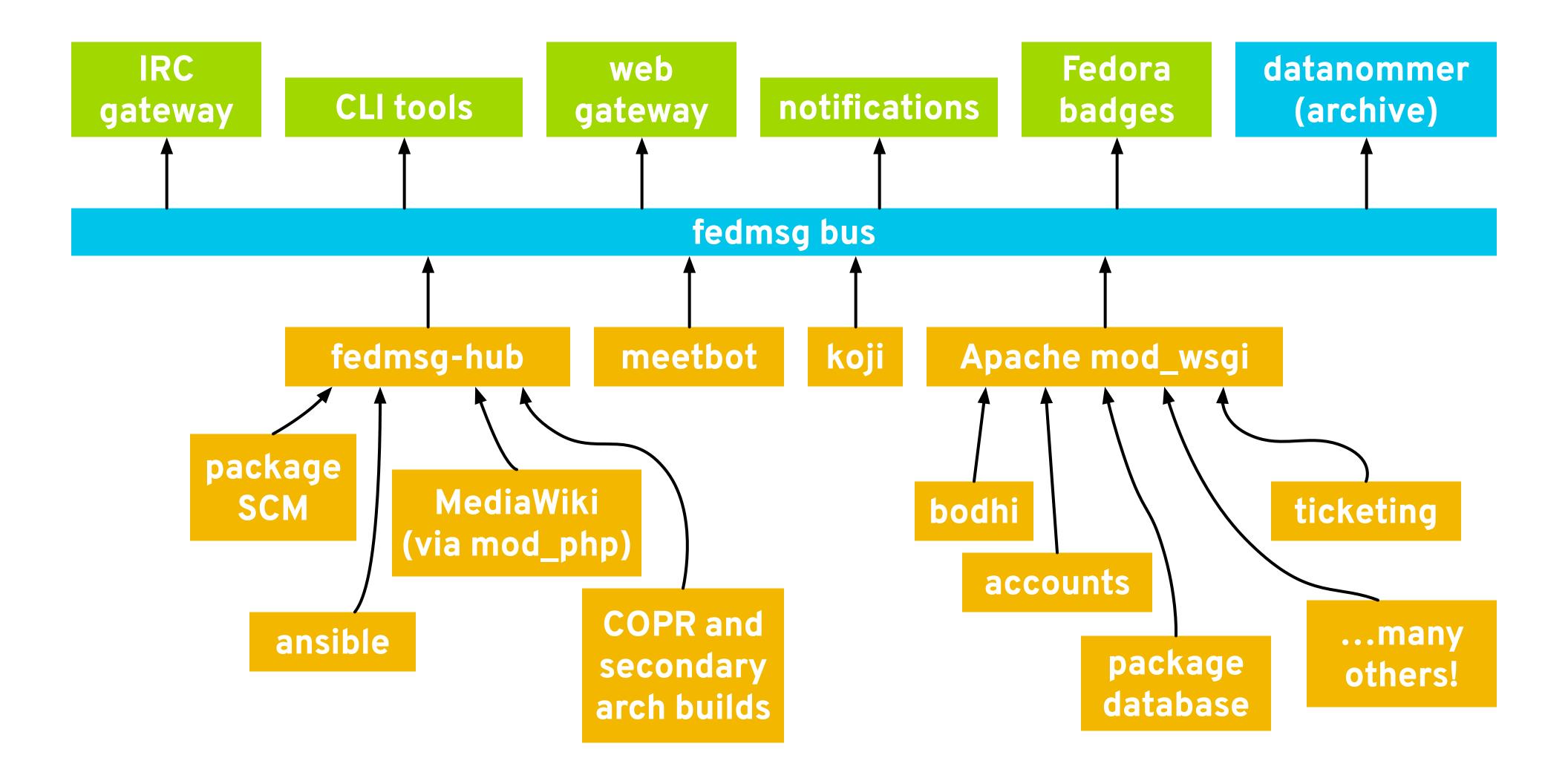








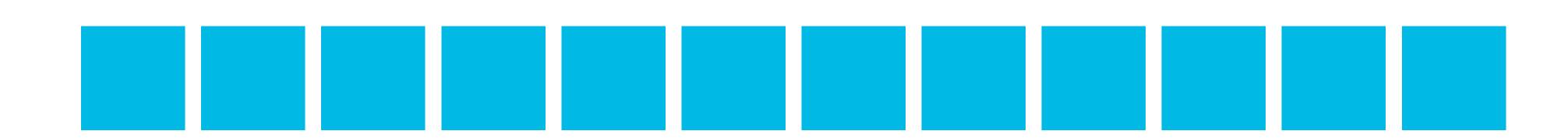








### Apache Kafka







#### Apache Kafka







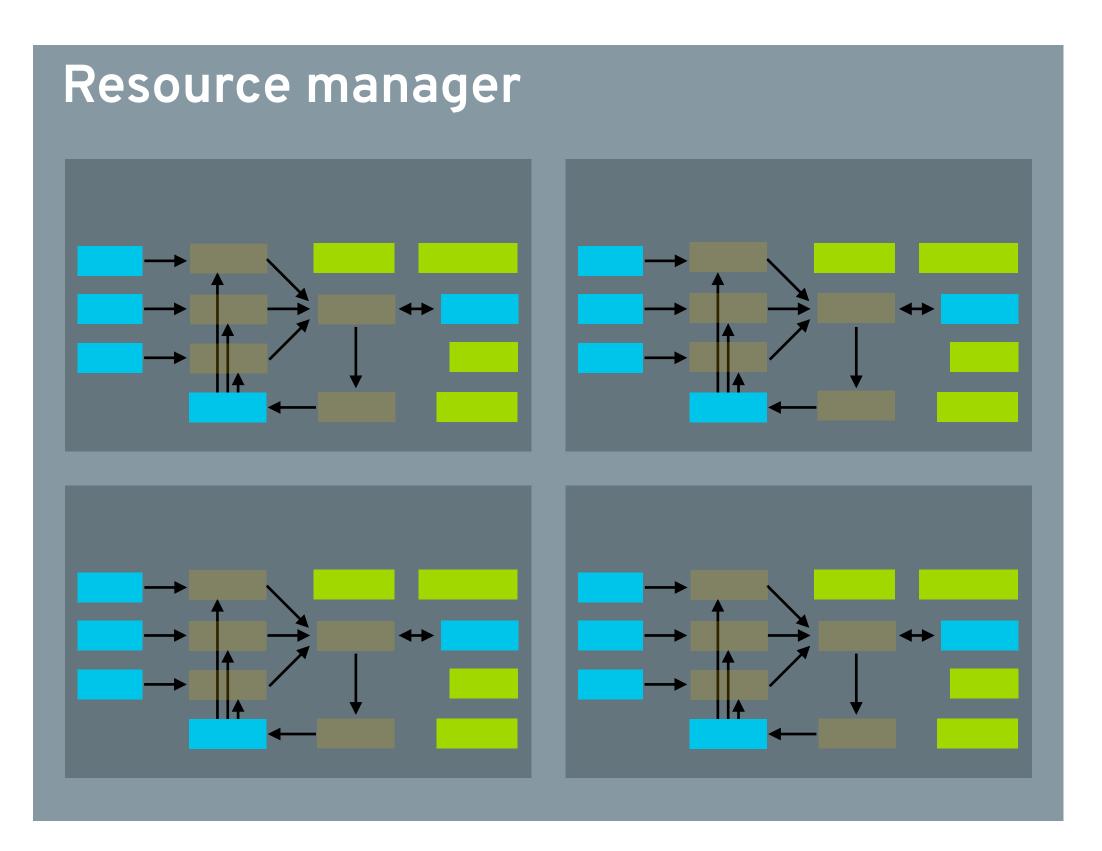
#### Apache Kafka

1 3 7 11 2 4 5 10 6 8 9 12





### Multitenant compute clusters



Spark executor

Spark executor

Spark executor

Spark executor

Spark executor

Spark executor

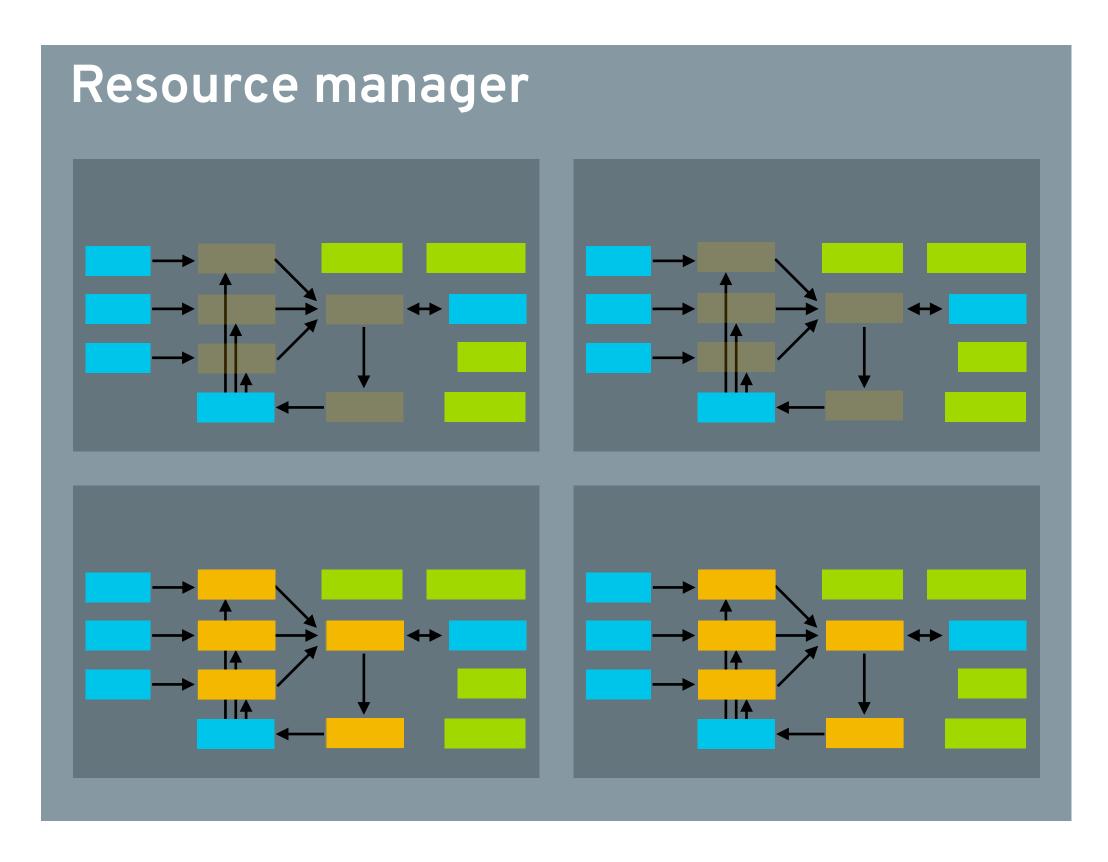
Shared FS / object store

**Databases** 





### Multitenant compute clusters



Spark executor

Spark executor

Spark executor

Spark executor

Spark executor

Spark executor

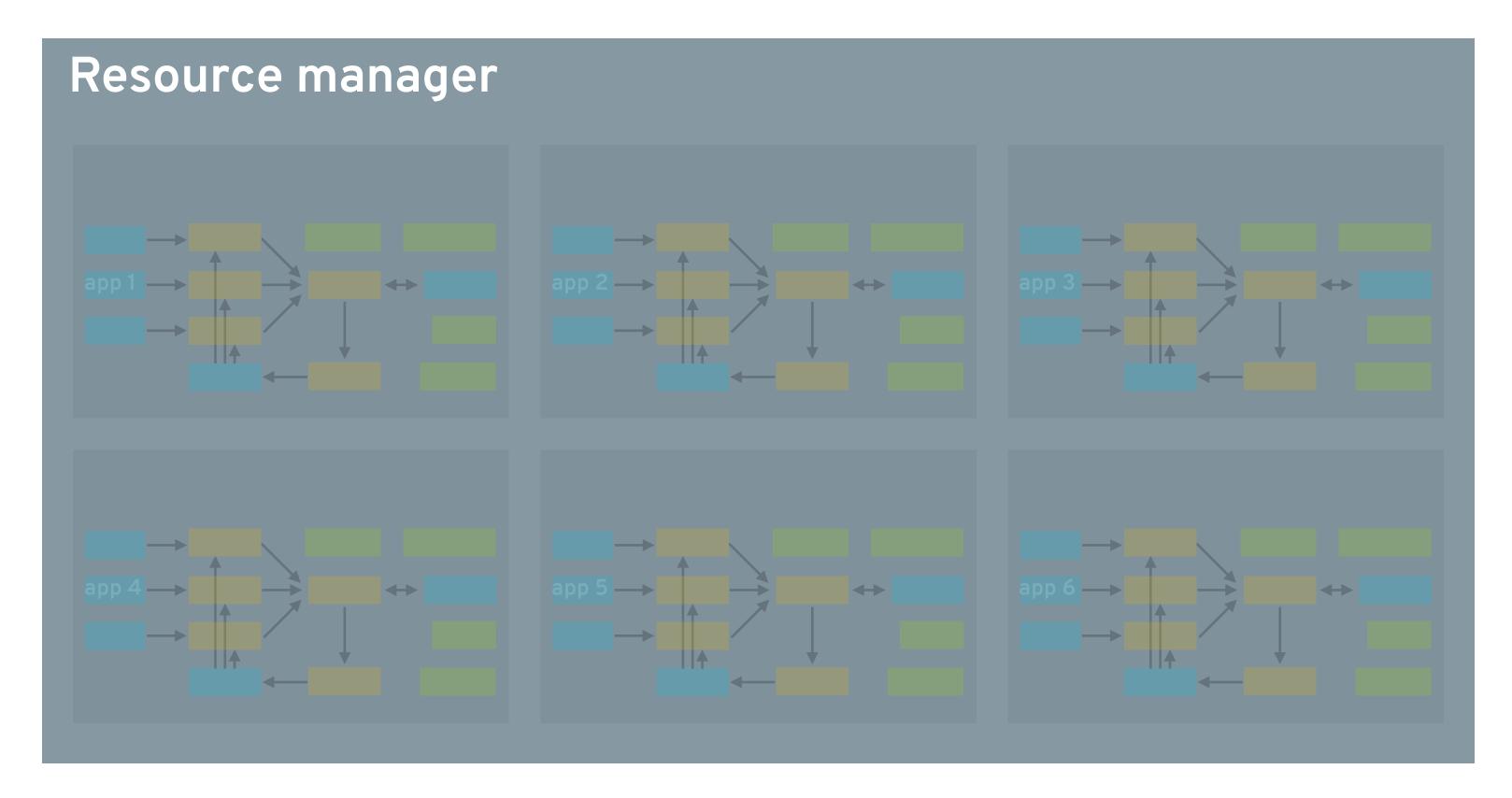
Shared FS / object store

**Databases** 





### One cluster per application



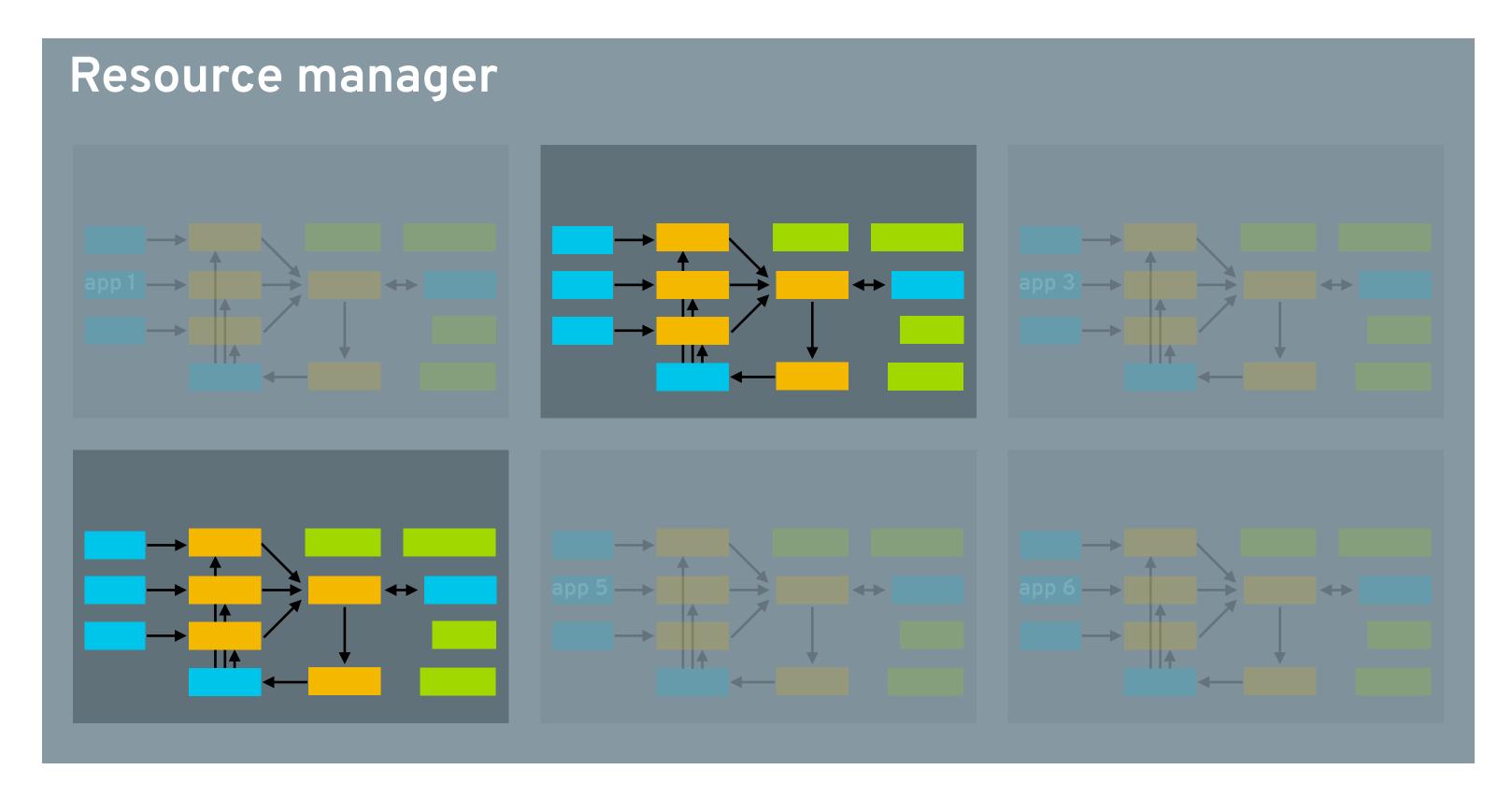
Shared FS / object store







### One cluster per application



Shared FS / object store







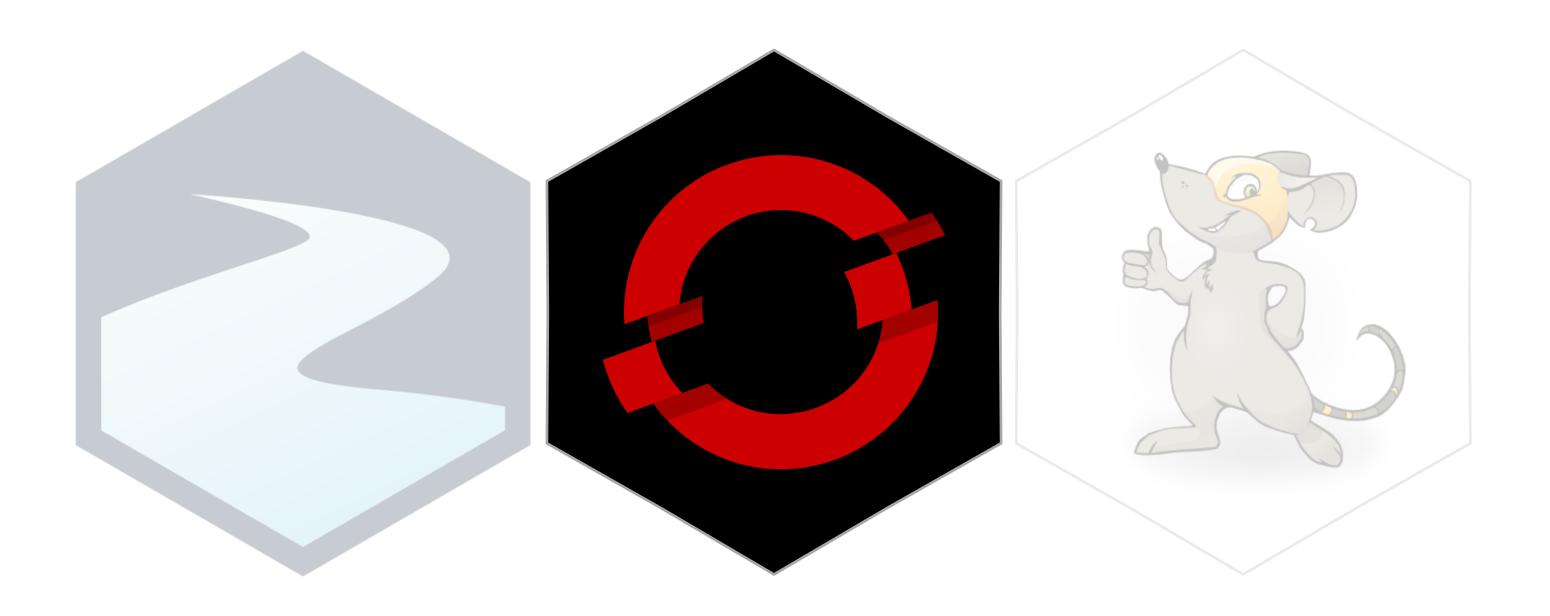
### radanalytics.io

An open-source community enabling intelligent applications on OpenShift Tooling to manage Apache Spark, Jupyter notebooks, and TensorFlow training and model serving



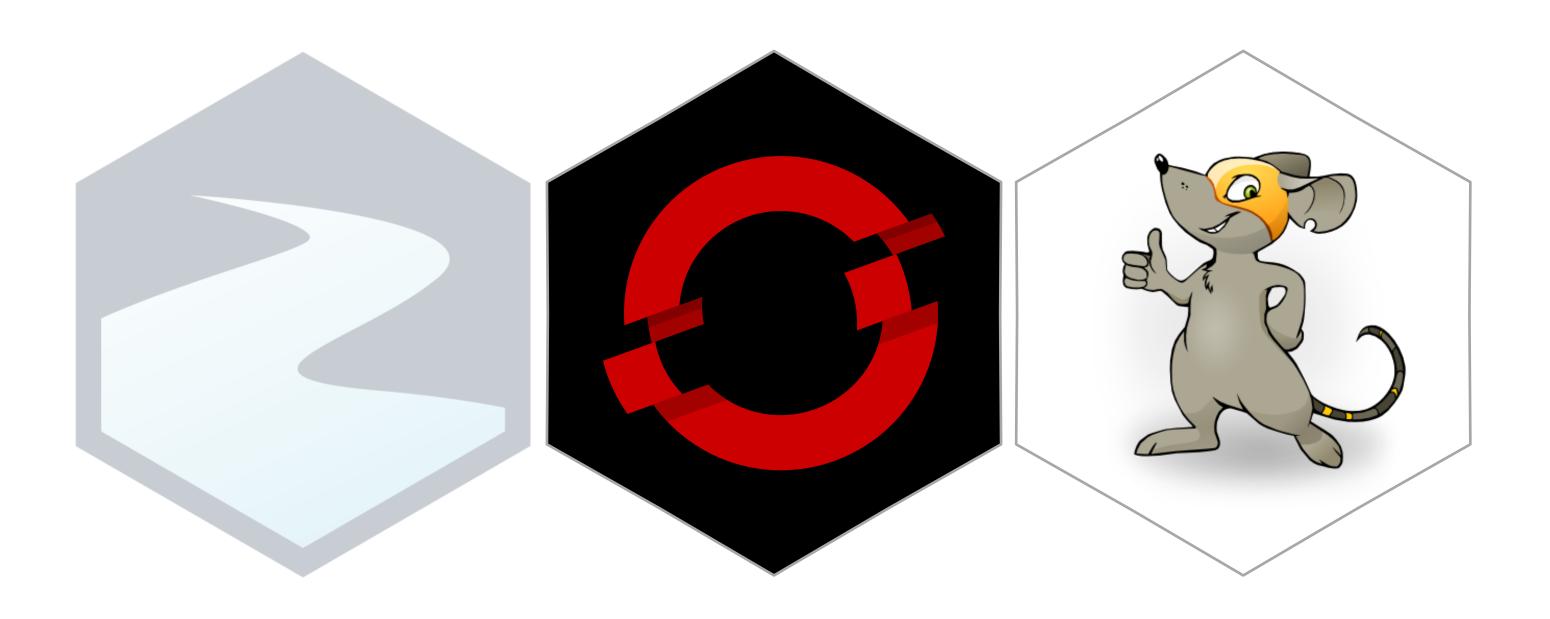






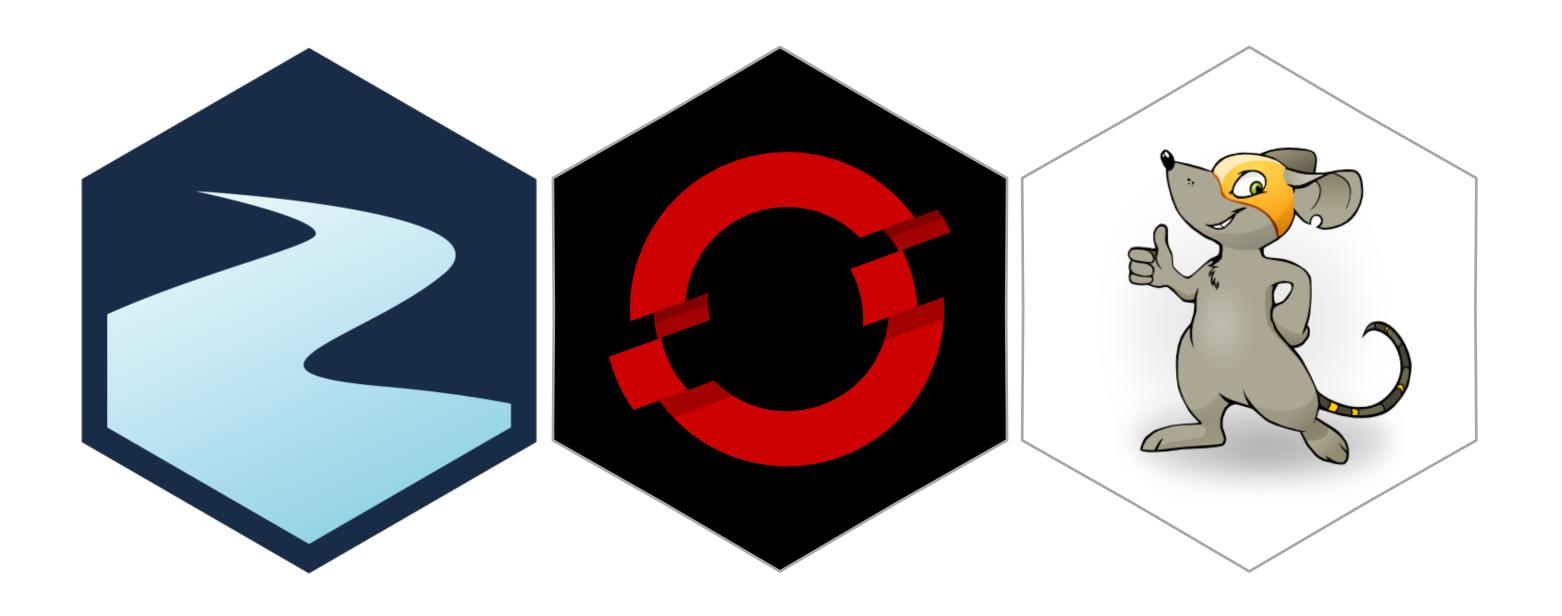


















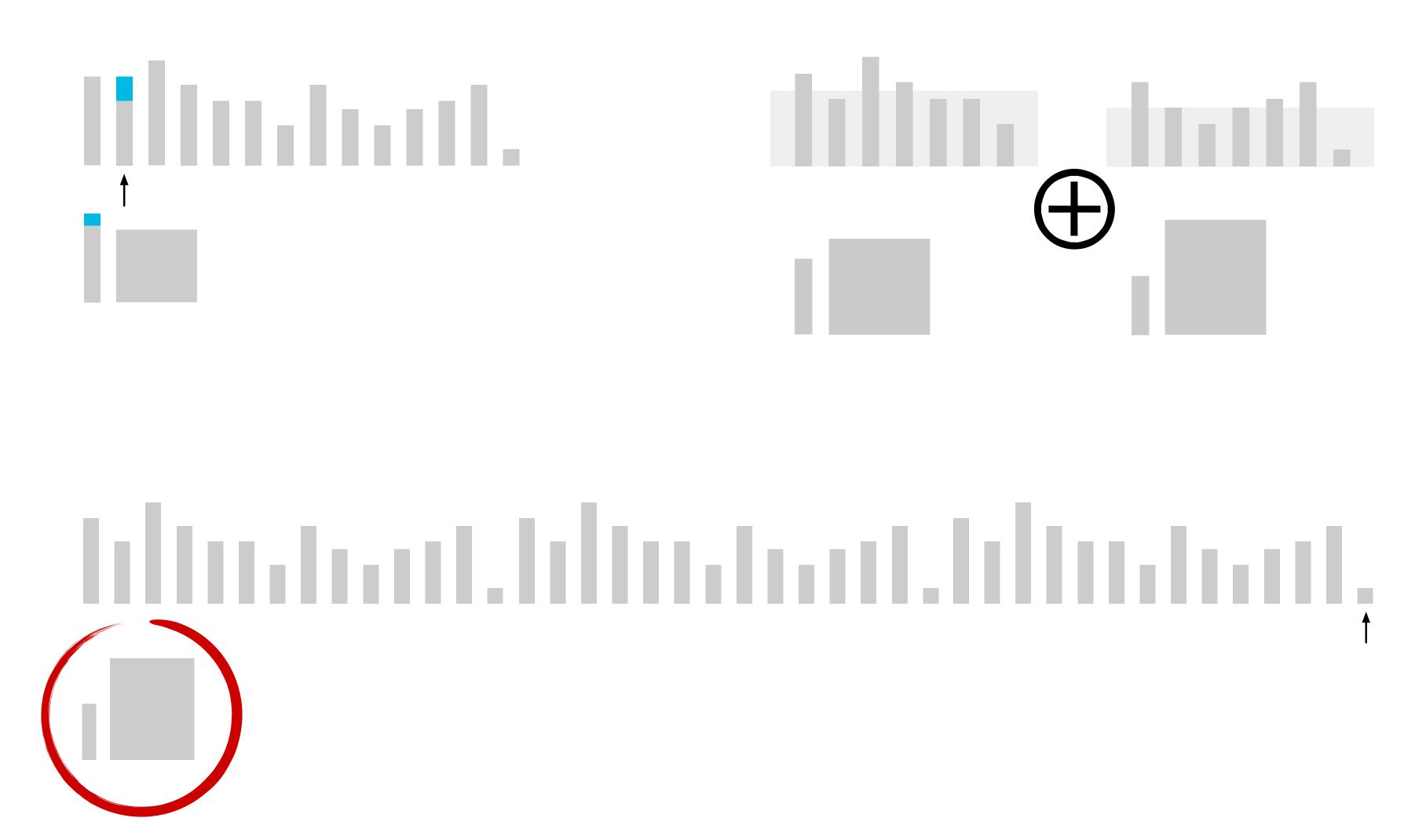


# Properties of useful streaming algorithms























# Interactivity time!

# http://bit.ly/streaming-bds18



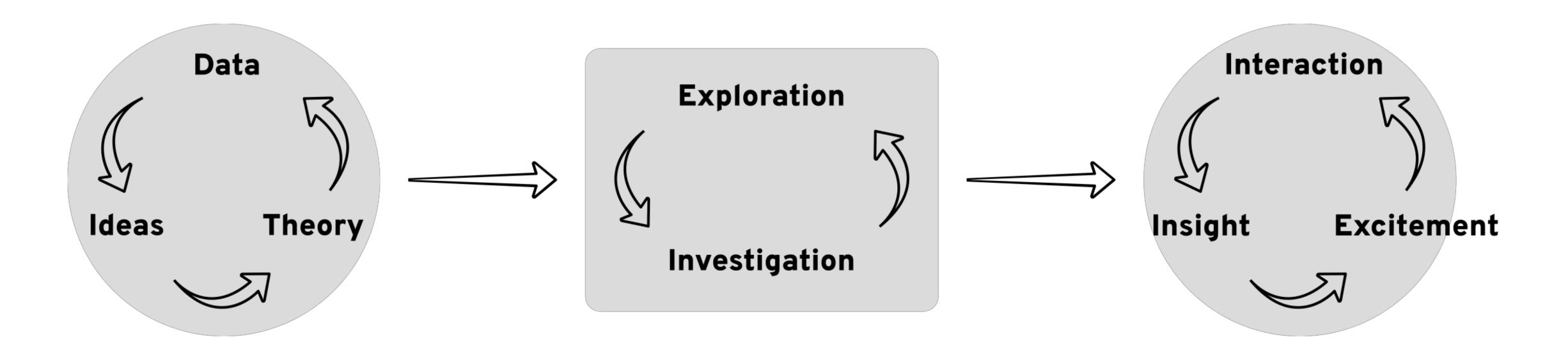






# From notebooks and experiments to intelligent applications

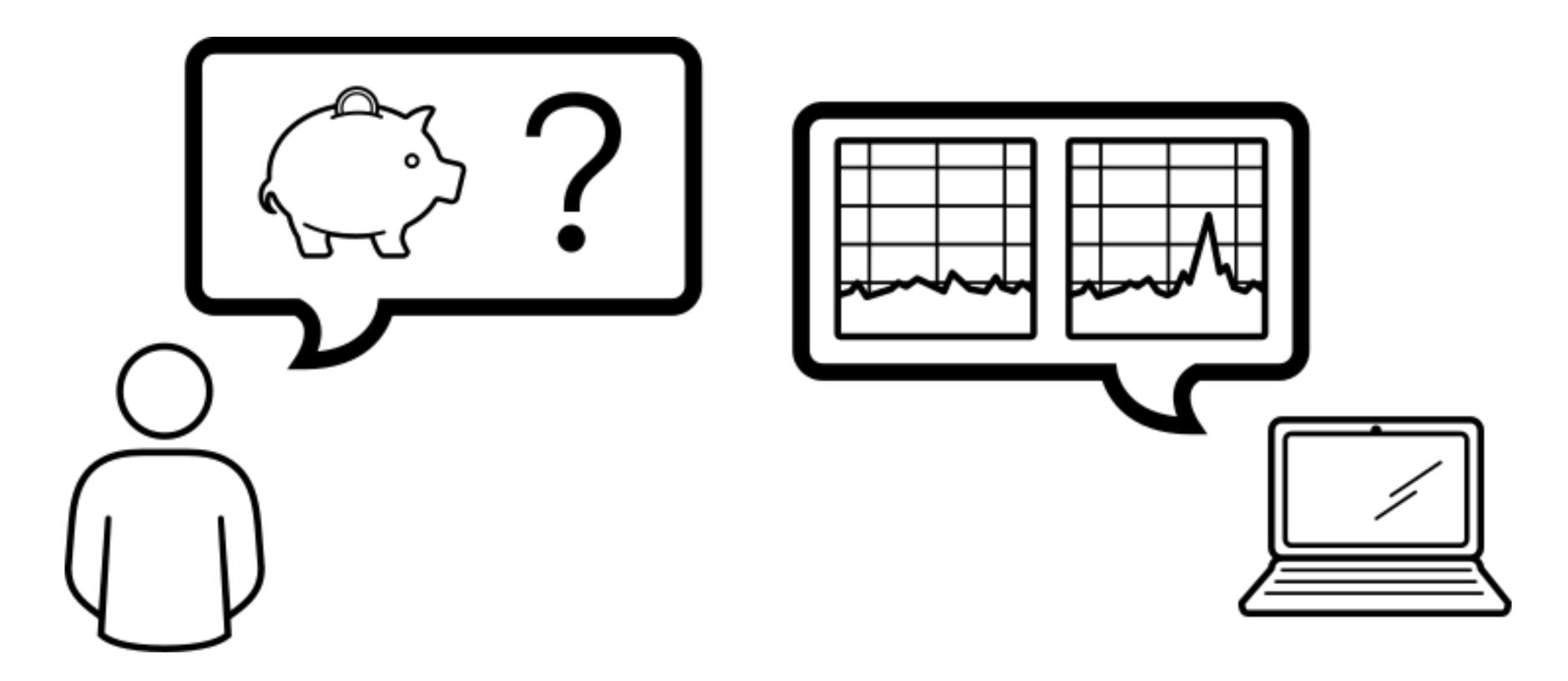
## Searching for solutions







### Consider possible interactions







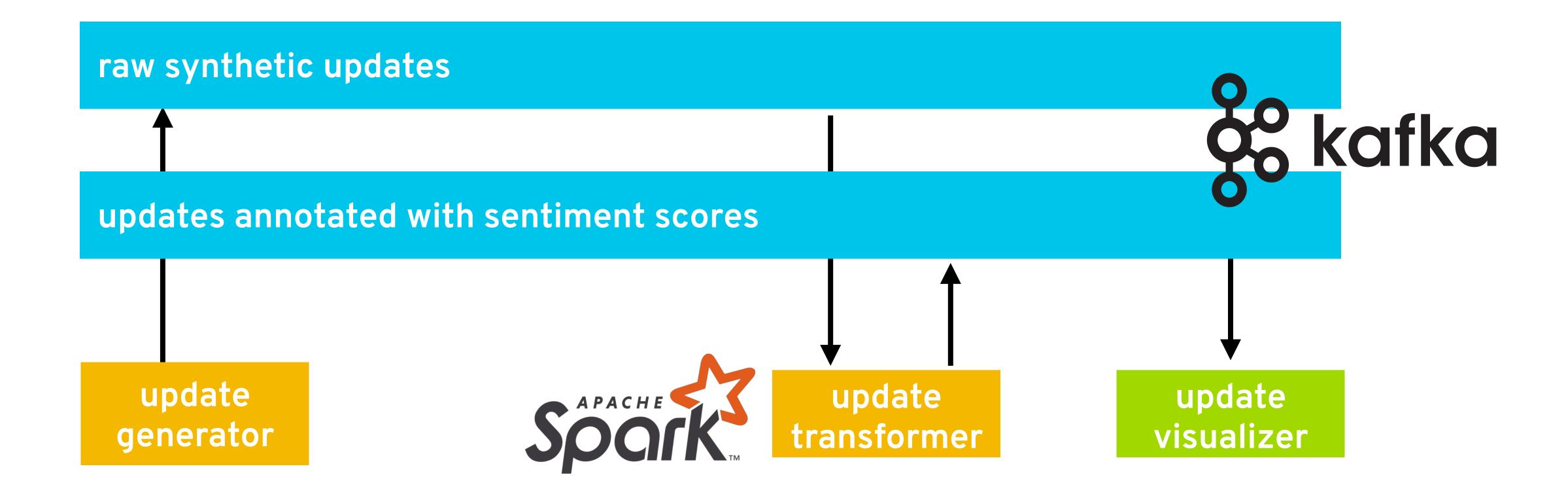
#### Plan for common patterns

Ingest Process Publish



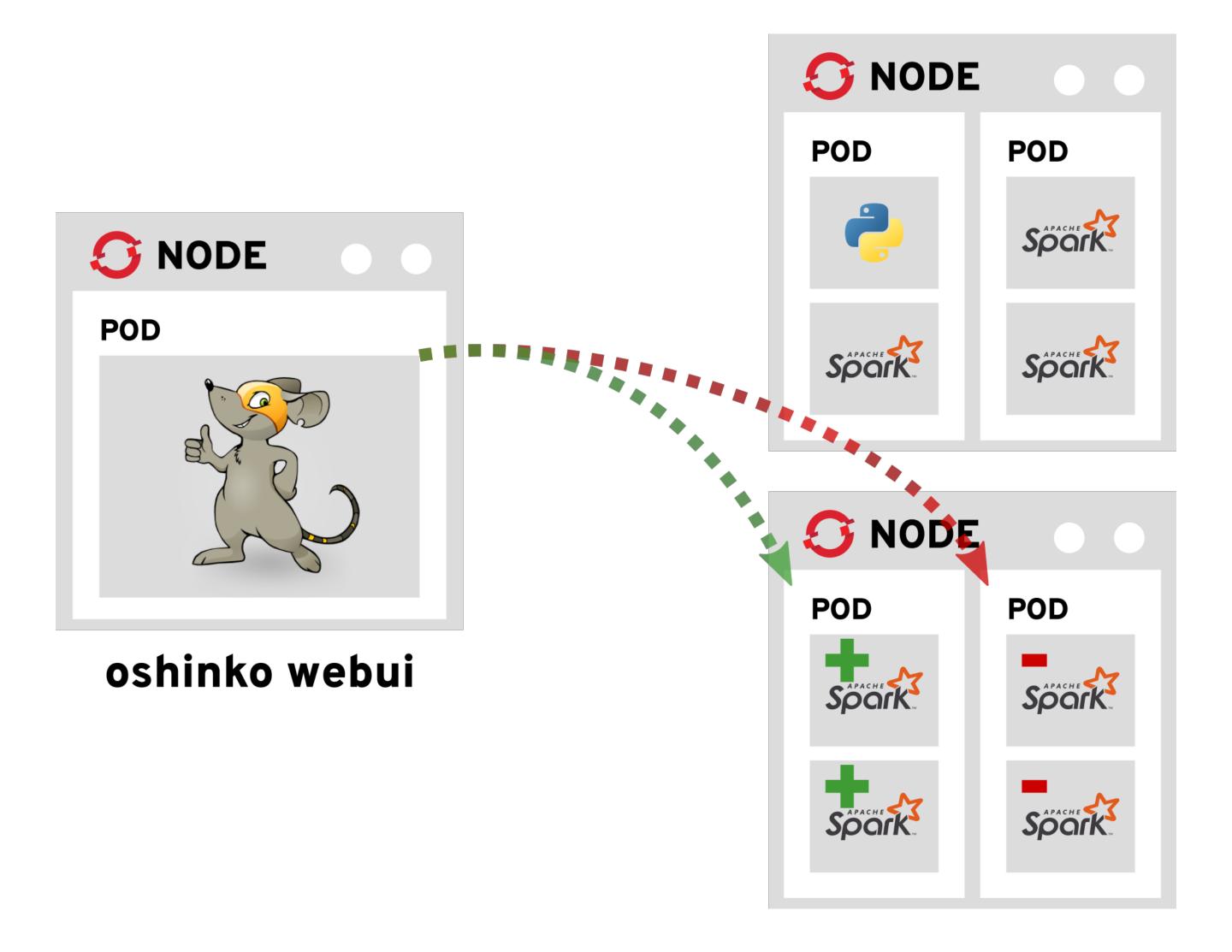


### Kappa in action



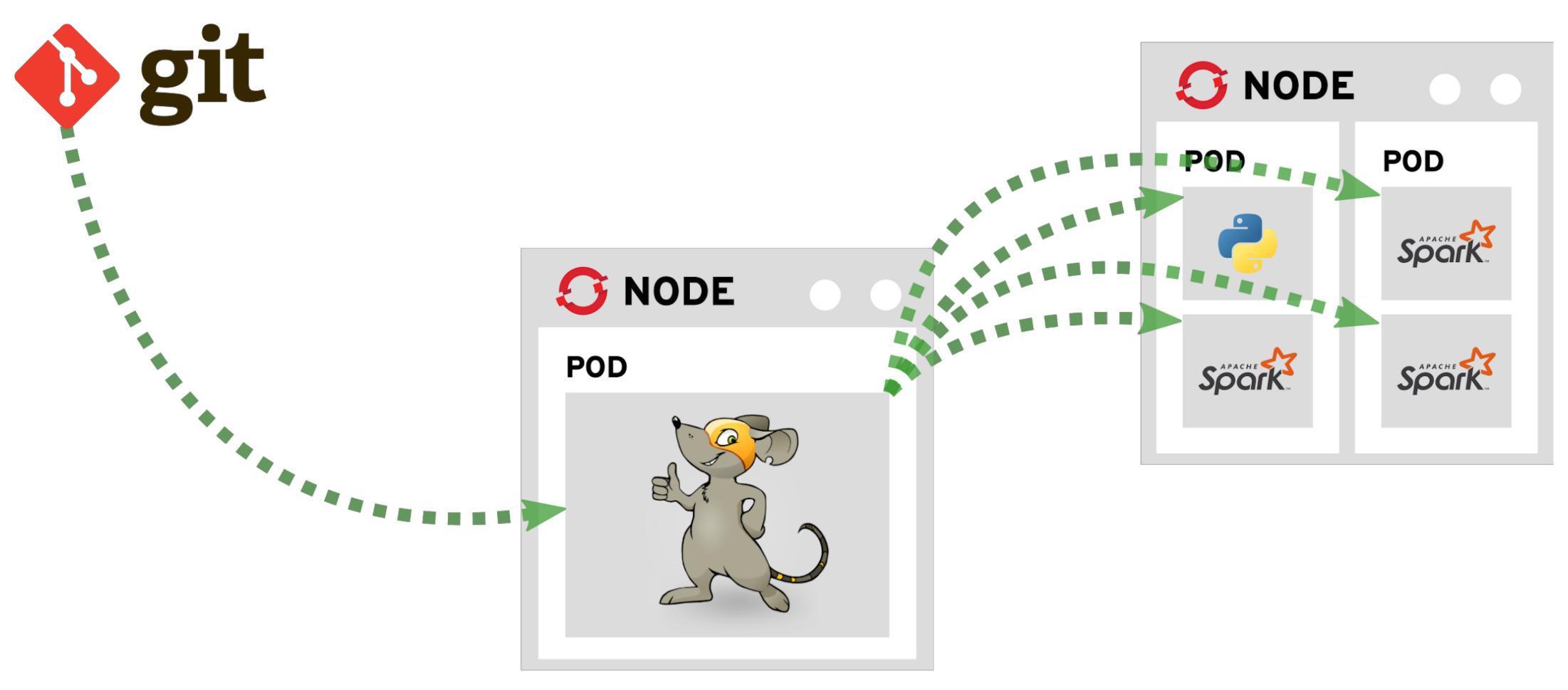












oshinko source-to-image



