

Fast, just-in-time queries on heterogeneous scientific data

Anastasia Ailamaki

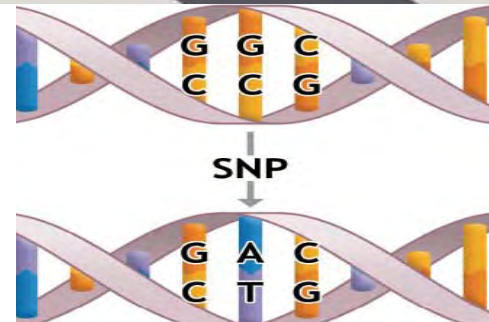
EPFL and RAW Labs SA

work funded by the Swiss NSF and the European Union

Most firms estimate that they are only analyzing 12% of the data that they already have.

Forrester, 2014

- Growing data
- Growing heterogeneity
- Data movement restrictions



Available data impedes business & scientific analytics

needful data

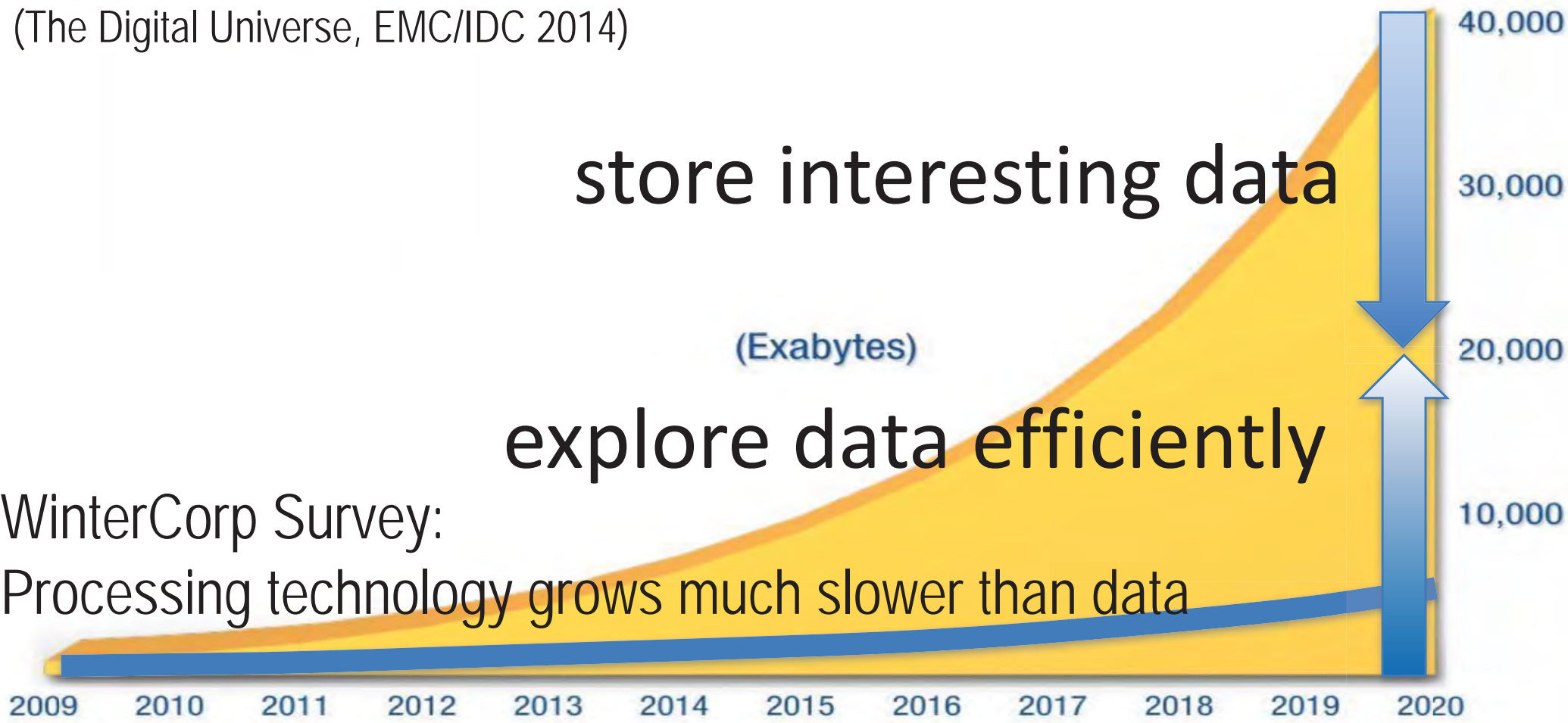
[VLDB12a, IISWC12]

- Cloudera customer jobs use KB-TB data
(from the PBs gathered daily)
- 80% of jobs access 1-8% of bytes
- 90% of FB jobs read <10% of bytes stored
- 80% of reuse is within 3hrs
- 95% of data used is 1 day old (Vahoulis et al)

interesting is  data, re-accessed soon

The Digital Universe: 50-fold Growth from the Beginning of 2010 to the End of 2020

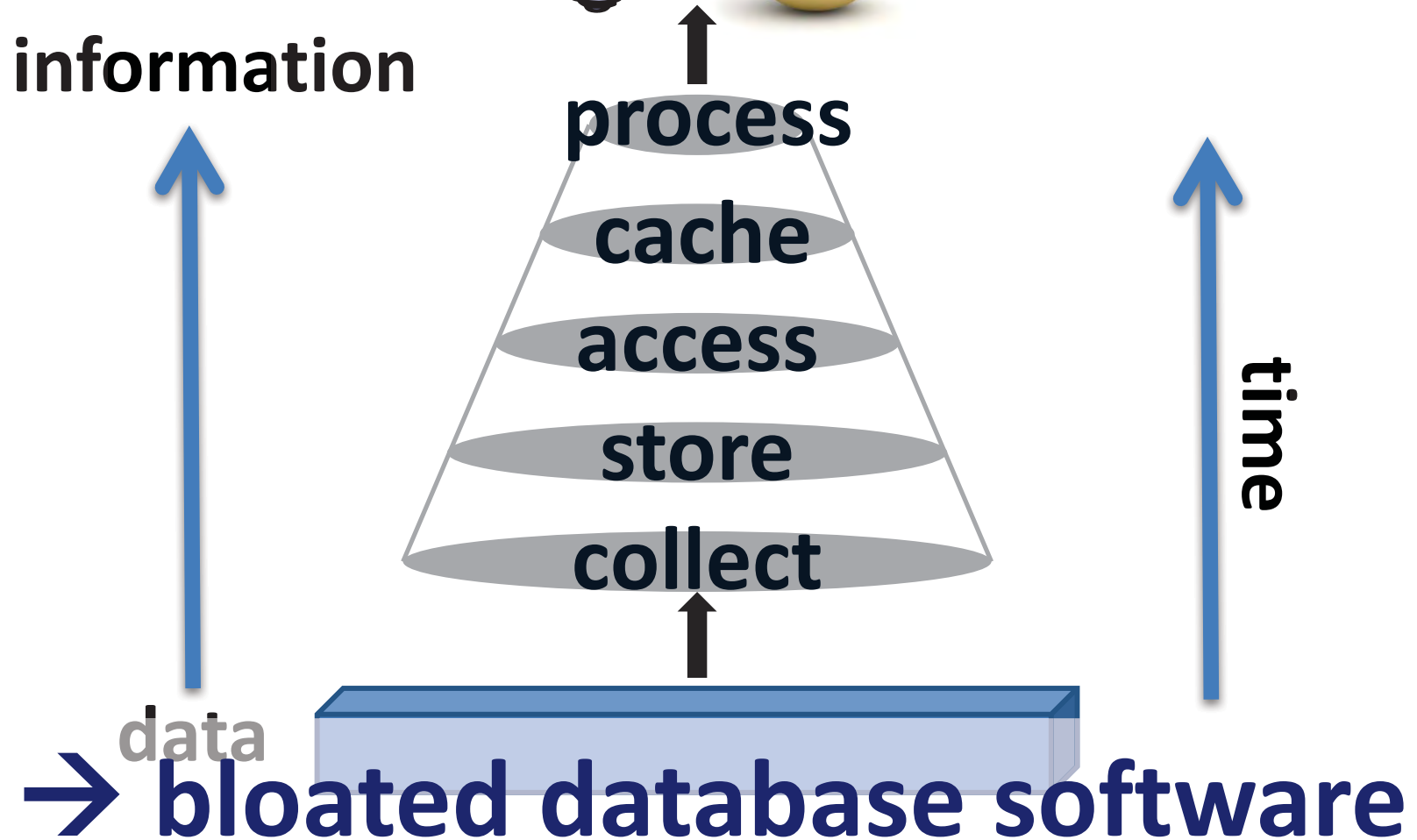
(The Digital Universe, EMC/IDC 2014)



WinterCorp Survey:

Processing technology grows much slower than data

build database to run queries



new: one DB per app/data pair

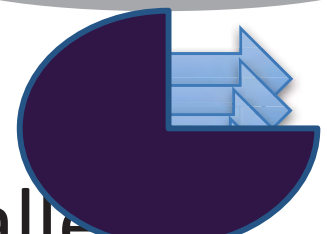
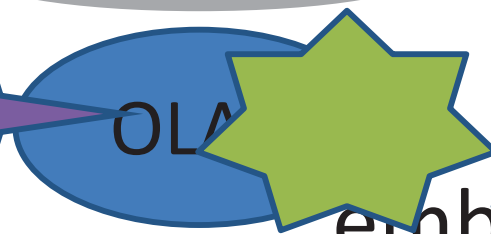
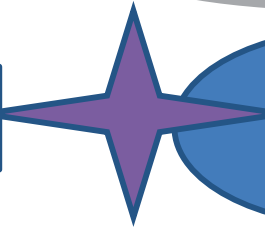
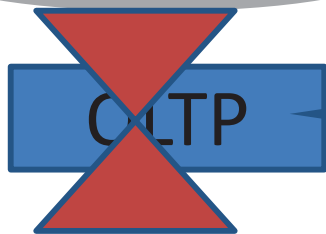
80% of analysts' time goes to data preparation and configuration

Main-memory
DBMS

Column
stores

NoSQL
systems

Stream
DBMS



embarrassingly parallel

Databases will be extinct



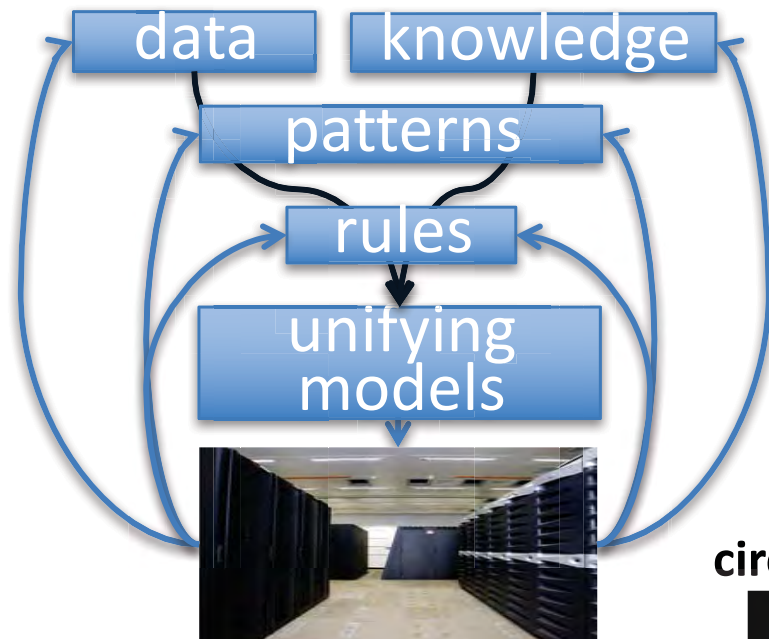
outline

- inspiring use case: the human brain project
 - brain simulation data
 - querying patient records
- a lean and agile database approach
 - adaptive query processing
 - SQL for all data

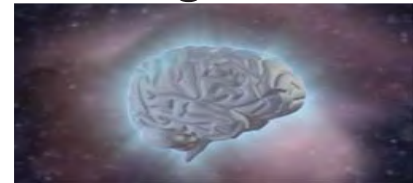
outline

- **inspiring use case: the human brain project**
 - simulation data sets
 - querying patient records
- a lean and agile database approach
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human brain project



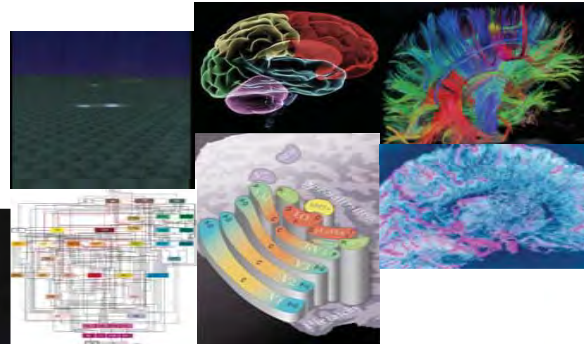
cognition



whole brain



circuits



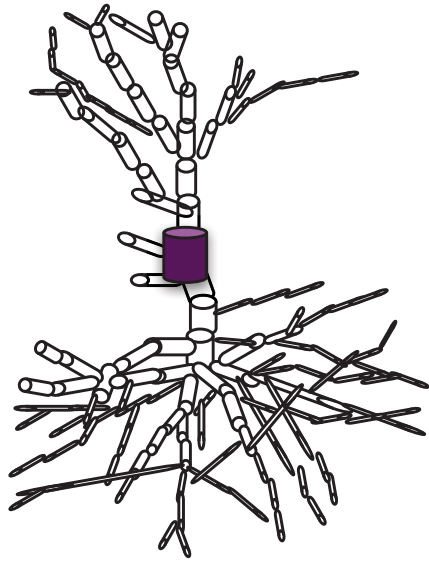
synapses

neurons

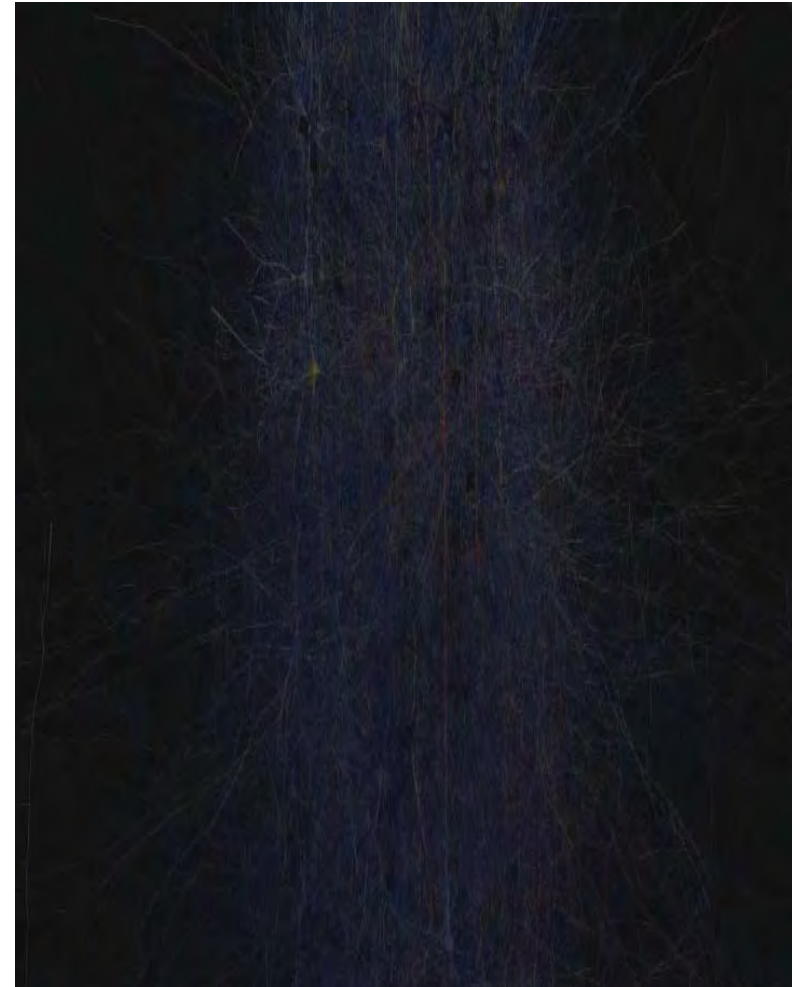


integrate clinical and simulation data

brain simulation



**single neuron,
modeled with 3D cylinders**



database the brain



challenge: navigate in 86B neurons

retrieval (FLAT) [ICDE12]
exploration (SCOUT) [VLDB12]
discovery (TOUCH) [SIGMOD13]
refinement (OCTOPUS) [ICDE14]

1000x simulations

patient data: medical informatics



biological disease signatures

the coupling of
clinical measurements with
validated biomarkers

Example: Alzheimer's disease

Clinical - Phenotype	Proteomic Biomarkers	Genomic Biomarkers	Imaging Biomarkers
Cognition: memory	CSF protein: beta	APP, PSEN1, PSEN2	Volumetric change:
Functional capacity	amyloid		hippocampus, inferior temporal cortex...
General physical health		Common genetic variant: APOE e4e4 ...	Beta amyloid imaging

memory loss/tau level/APOE e4e4/hippocampus atrophy/amyloid

medical informatics platform

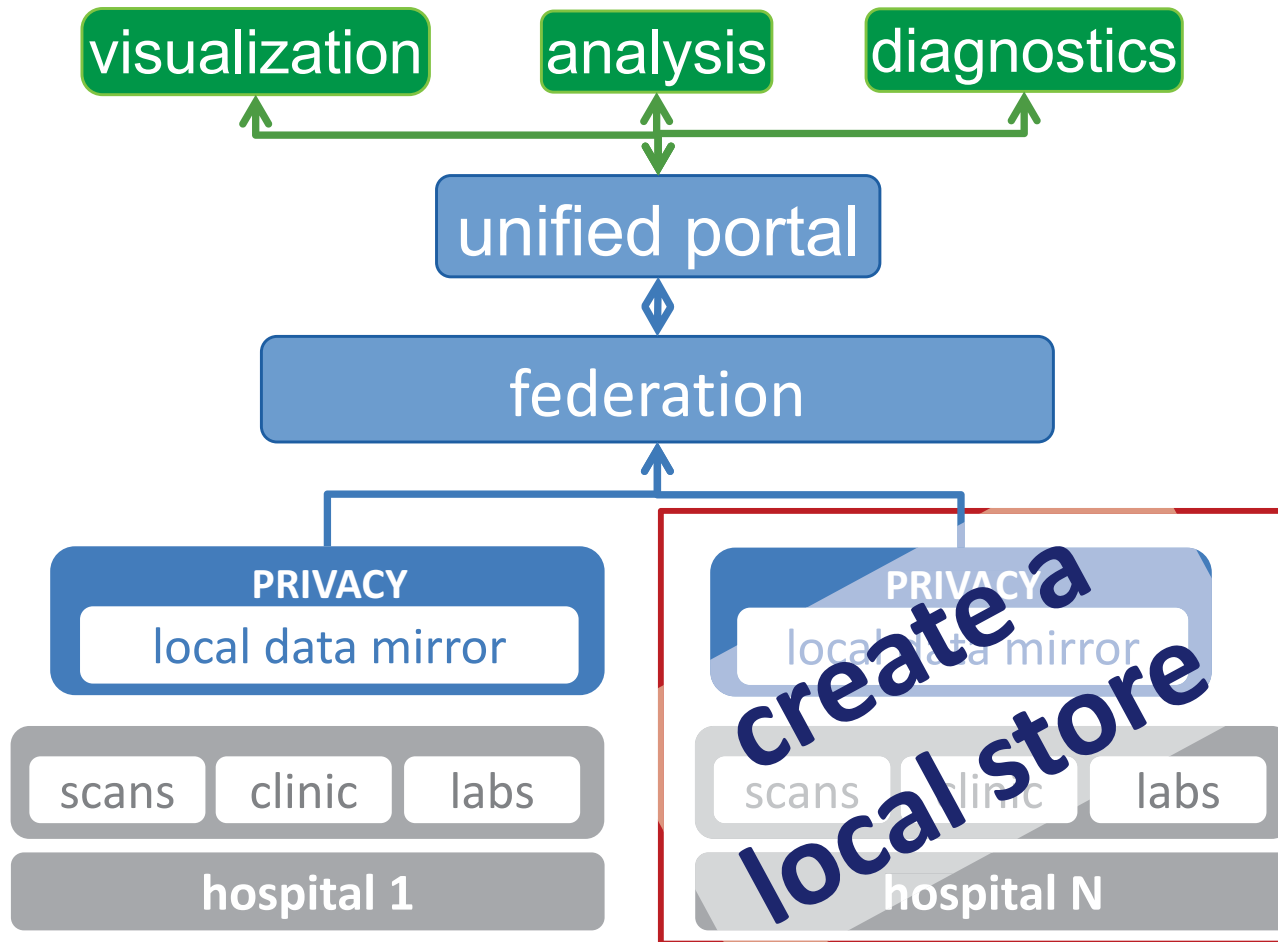
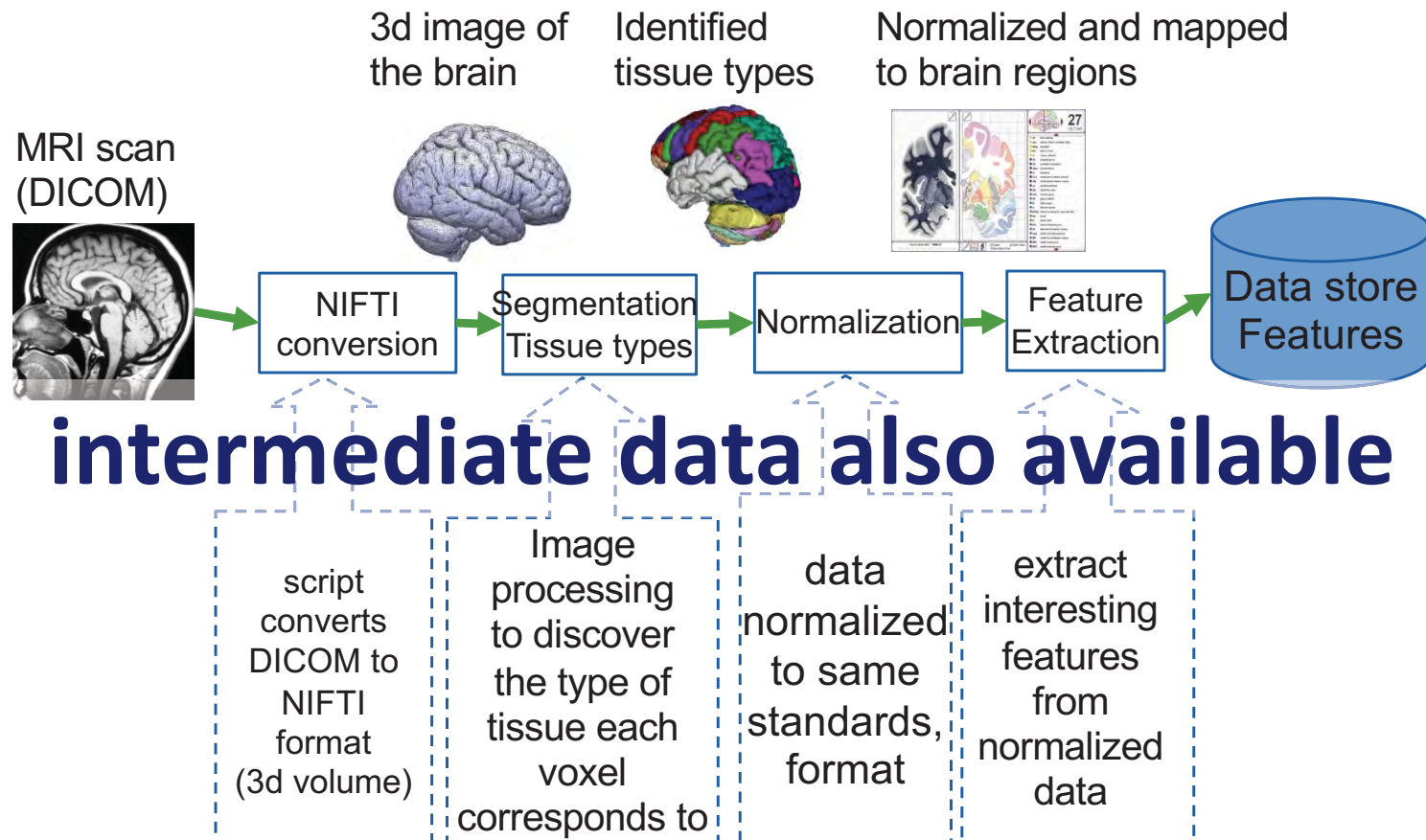
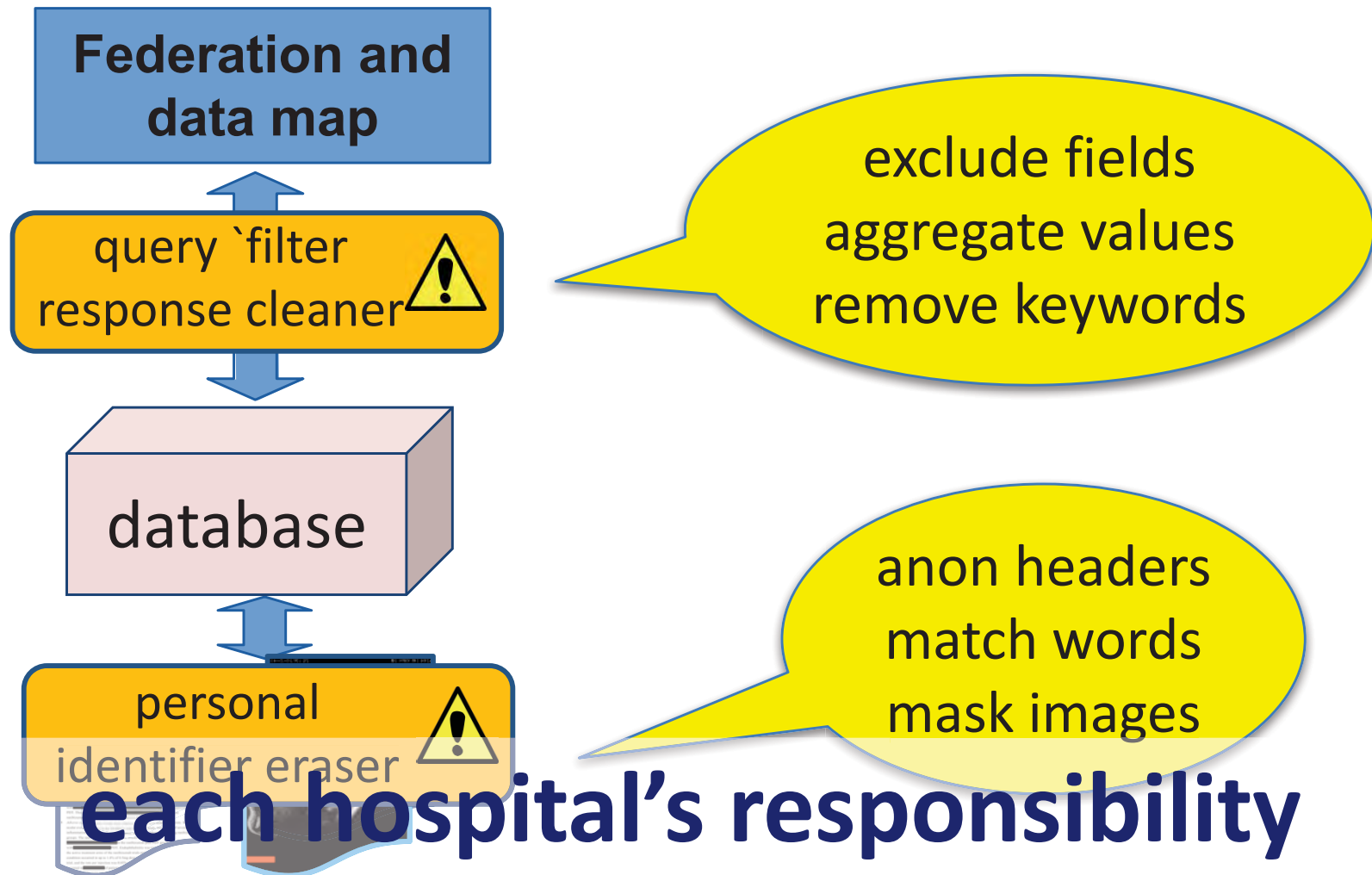


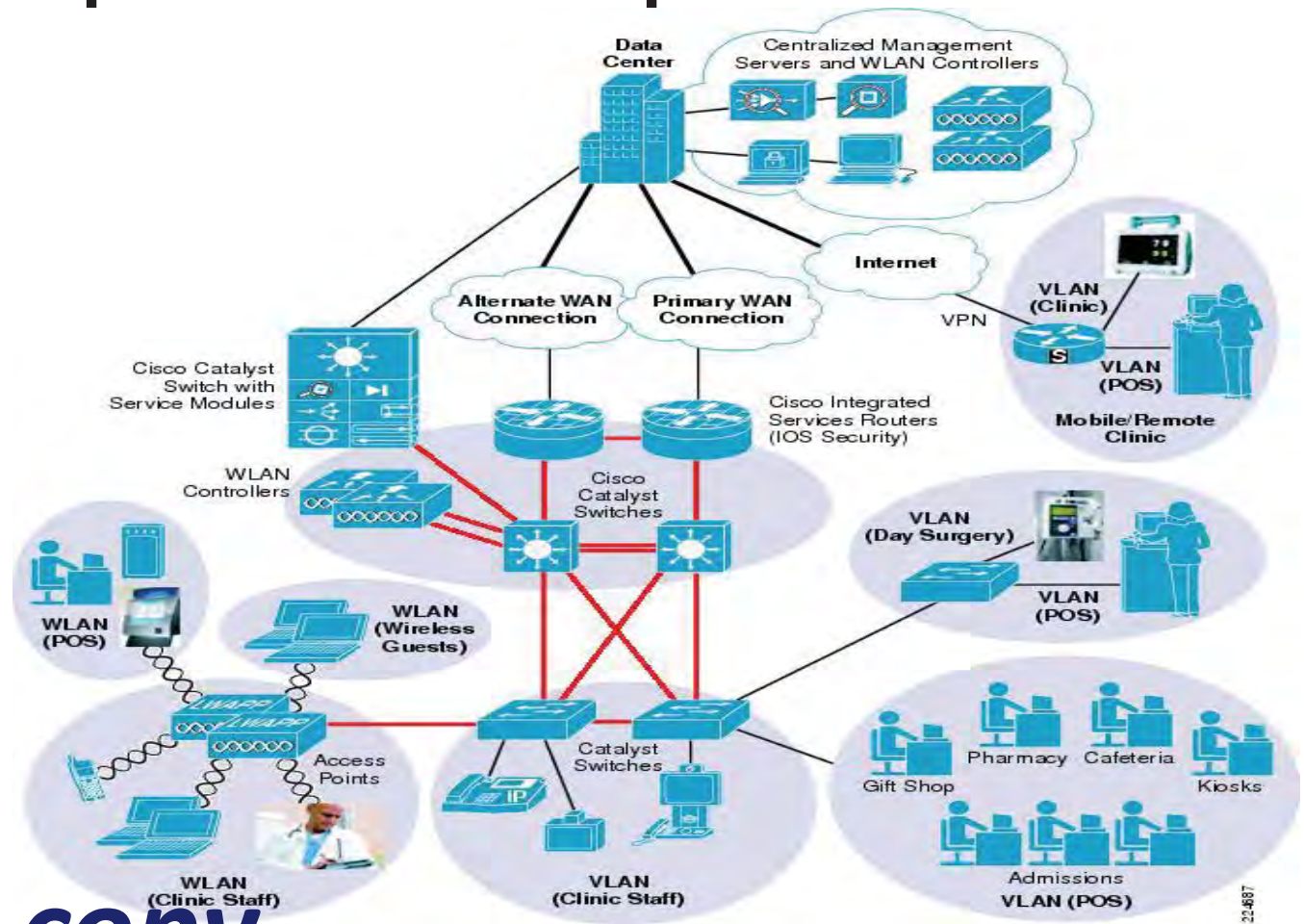
image preprocessing (Lausanne)



two-level anonymization



access to private hospital data



no move, no copy

outline

- inspiring use case: the human brain project
 - simulation data sets
 - querying patient records
- **a lean and agile database approach**
 - **Just-in-time query processing**
 - **SQL for all data**

clinical+genetic+imaging data → signature

Patients (CSV)

id	Protein: AACT	Age	Phenotype	...
1	1.4	45	Trauma	...
2	2	55	Chronic Symptoms	...
3	0.2	56

Brain_GrayMatter (Binary)

	0	1	...	n
0	0.45	0.75	...	0.1
1	0.33	0.3	...	0.38
...
m	0.12	0	...	0.47



BrainRegions (JSON)

```
[{"id": 1,  
  "amygdala": {"X":15,"Y":20, "Vol": 0.5},  
  "hippocampus": {"X":17, "Y":10, "Vol":0.2}},  
{"id": 2, ...},  
{"id": 3, ...}]
```

signature:

age > 50

AND

amygdala.Vol > 0.3

AND

AACT < 1

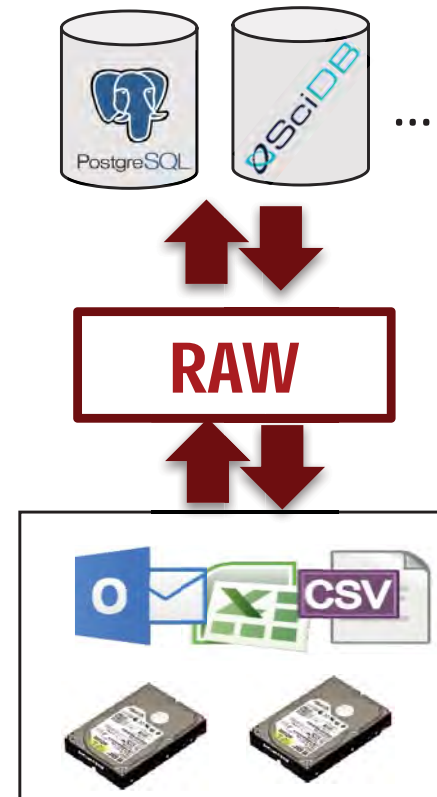
queries on heterogeneous data

cannot load into a Database System!

- diverse formats
- legacy software
- privacy limitations
- data “owned” by one database

RAW: interface to raw data

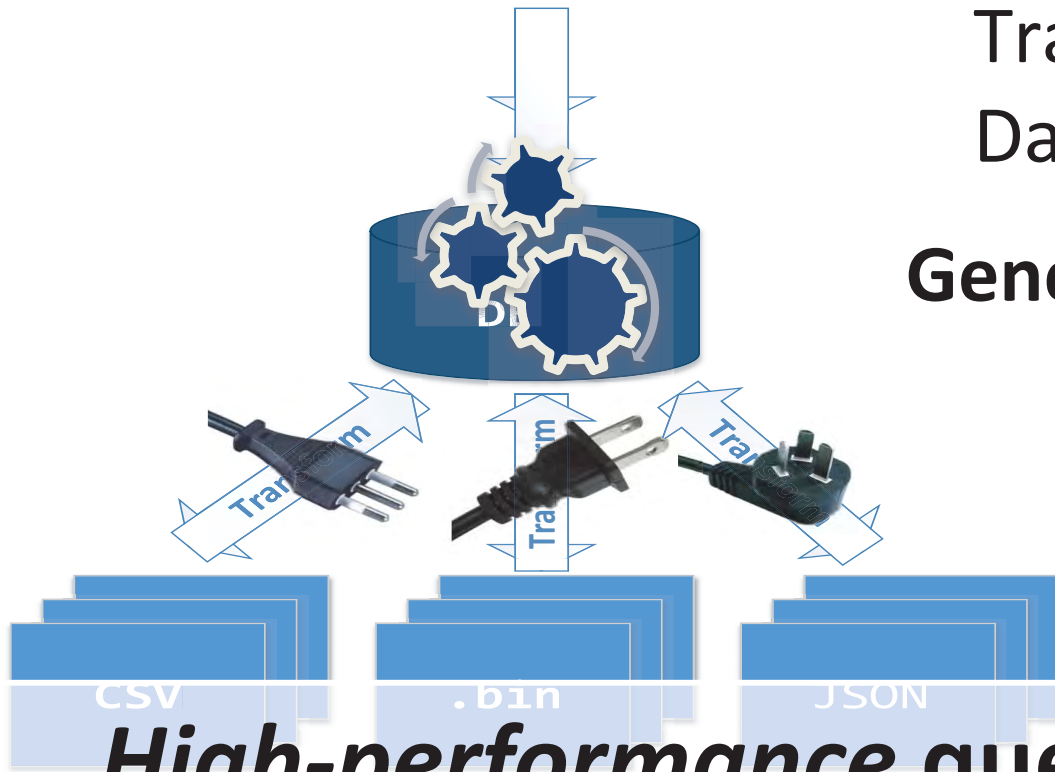
SQL, SCALA, notebooks
code-generated engine



key: data virtualization

Adapting a query engine to data

Query



Traditional:

Data adapts to engine

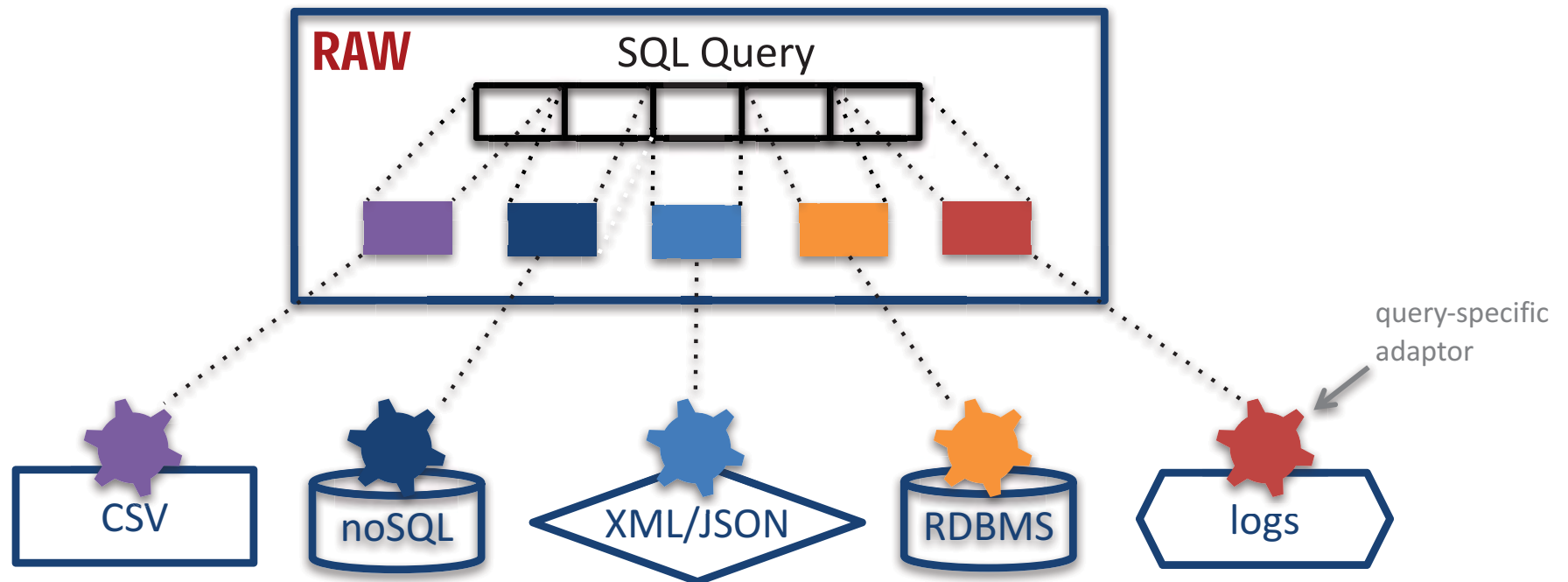
Generate plug-in per data source



Treat each source as
native storage format

***High-performance* querying... while using
original data formats, files, and scripts**

How it works



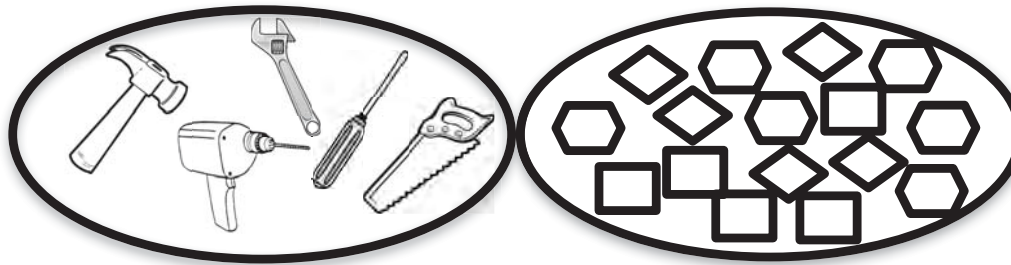
query-specific adaptors auto-generated
Tools and data cached for efficiency

As queries run, RAW gathers information on usage of data and generates software tools.



RAW

just-built
useful
tools



just-gathered
query-specific
data

query-specific
adaptor



CSV



NoSQL



XML/JSON



RDBMS



logs

What you can do with RAW 1/3

- Query raw files (CSV, JSON, XML, ...) directly *without ETL*:
 - without data ingestion
 - without data cleaning
 - without “flattening hierarchies”
 - ... i.e. without “losing information” during ETL
- How?
 - With an extended SQL language
 - No building scripts or using separate ETL tools on the side

What you can do with RAW 2/3

- Query datasets regardless of size or location:
 - Small CSV or Excel file on your Dropbox
 - PBs of Parquet data on Hadoop
 - Tables on a RDBMS
- How?
 - RAW code-generates to your execution environment: local node, Hadoop/Spark cluster, RDBMS backend, ...

What you can do with RAW 3/3

- Build and preview your queries/reporting/analytics, live, step-by-step
- Build dashboards with a few clicks
- How?
 - The user interface is built into RAW's engine.
 - RAW comes with an IDE.

Your Queries 6

New +



patients_overview

patients_gender_city

patients_map

authors_with_ratings

authors

Your Sources 7

Add +



authors

patients

publications

author_ratings

log_example

Editor



```
1 select d.description, d.sub_category, d.category
2 from patients p,
3      p.patientBasedData.diagnosis d
4
```

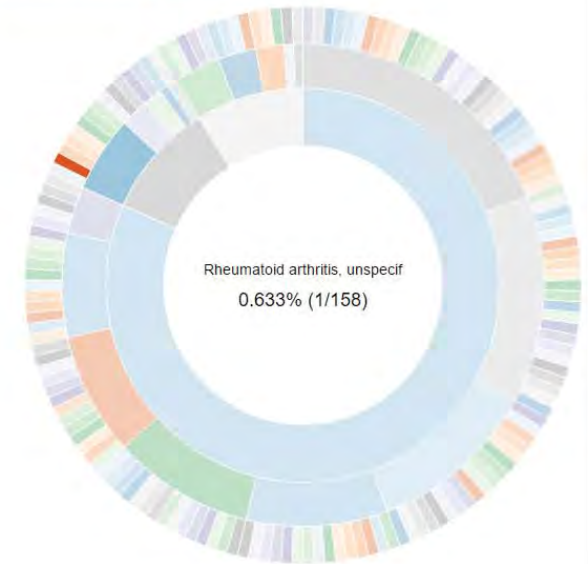
Plot

sunburst ▾

Full results



Diseases of the musculoskeletal system → Inflammatory polyarthropathy → Rheumatoid arthritis, unspecified → 0.633%



Editor



Plot

No plots available

Query failed during execution

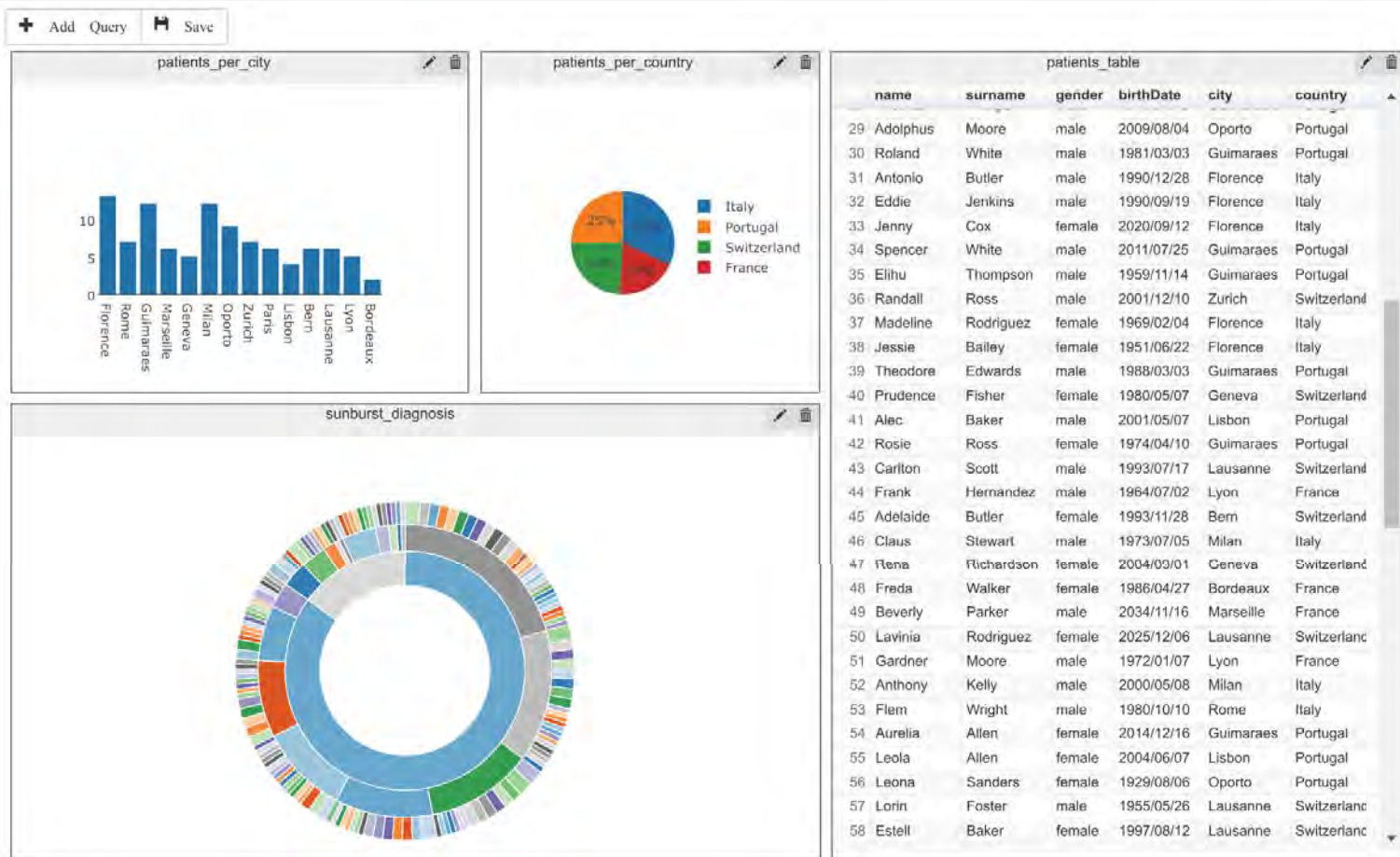
```
❌ select * from log_example parse as r"""\d+\""
```

```
incompatible input: 199.72.81.55 - - [01/Jul/2015:00:00:01 -0400] "GET /history/apollo/ HTTP/1.0" 200 6245
```

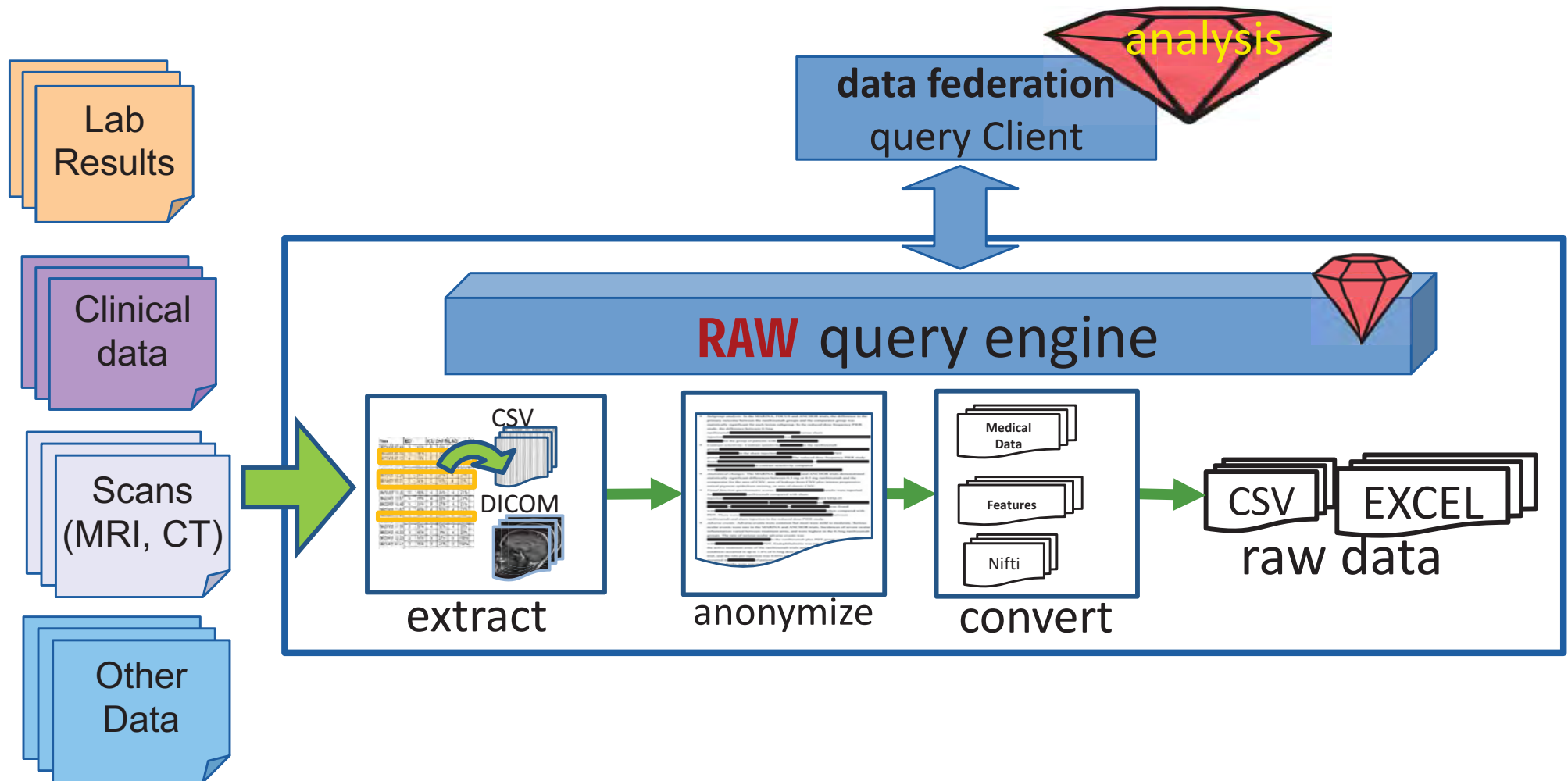
Your Dashboards 2 New +

patients

test



deployment: hospital data mirror



Medical Informatics Platform

visualization analysis diagnostics

unified portal

CHUV (CH)

User Interface

Federation and
Data Map

RAW Query Engine

Hospital Data

UniClinic (DE)

User Interface

Federation and
Data Map

RAW Query Engine

Hospital Data

Niguarda (IT)

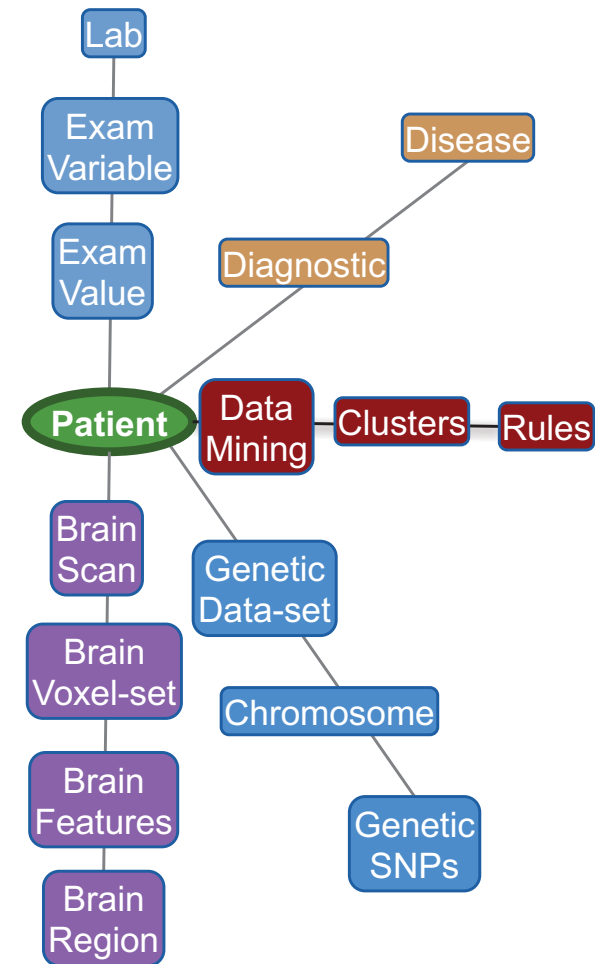
User Interface

Federation and
Data Map

RAW Query Engine

Hospital Data

the user's perspective



Summary

- currently data management cost grows with data owned
- impossible to pre-cook a database system suitable for all data
- from manual ingestion to automatic adaptation: just-in-time query processing

Thank you!

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