Load - II

big data technology

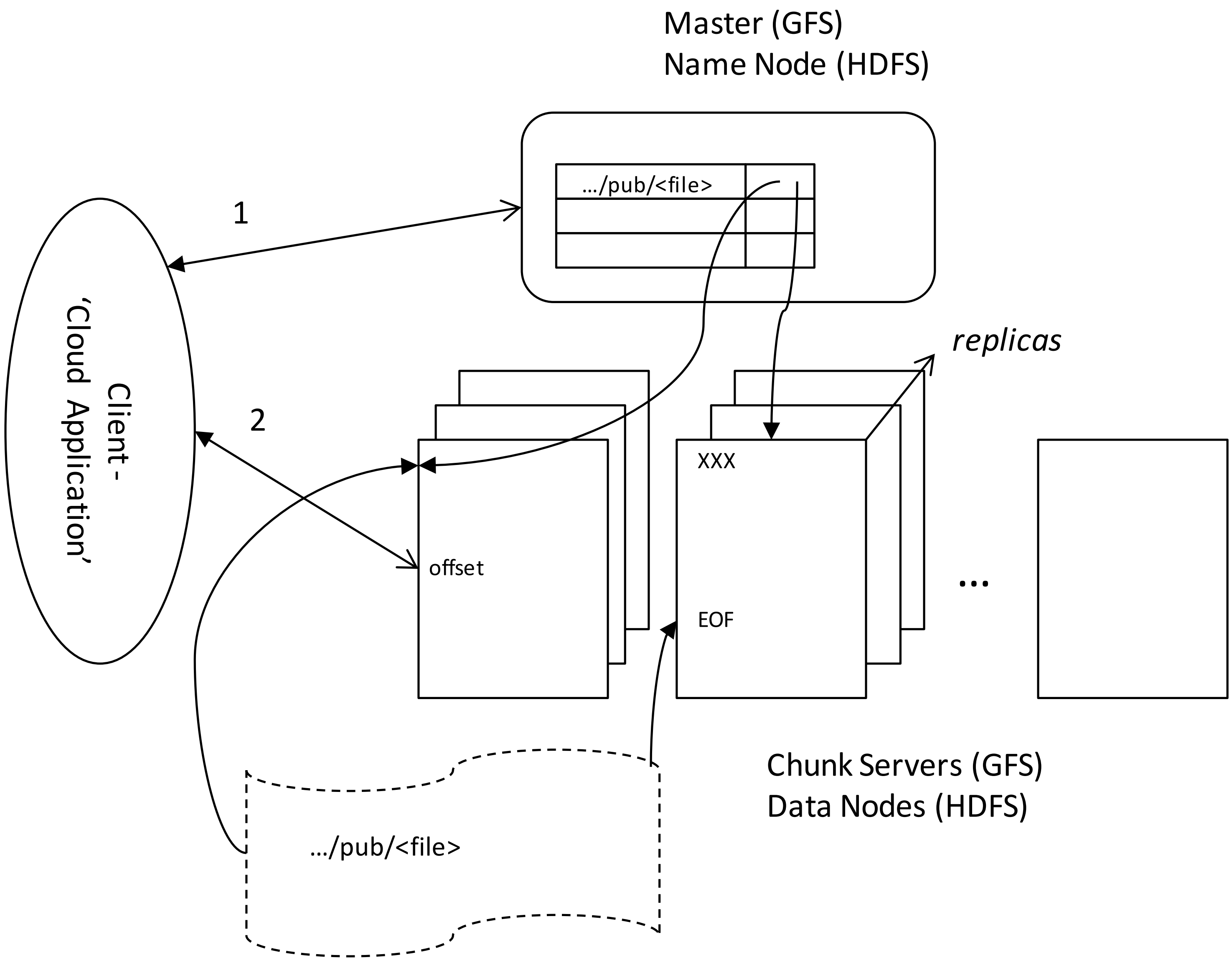
week 3:

map-reduce and programming assignment

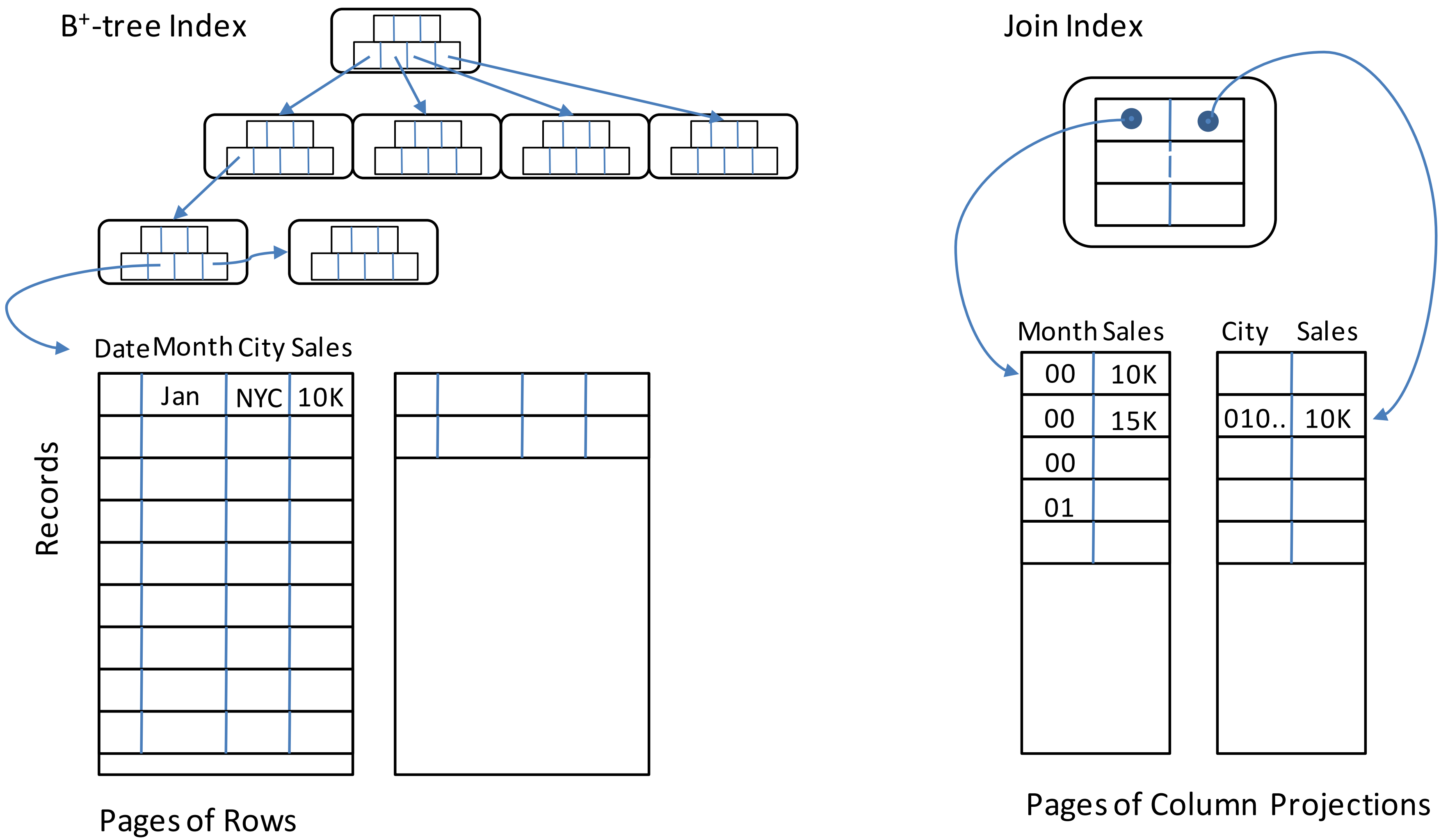
week 4:

distributed file-systems, databases, and trends

distributed file systems (GFS, HDFS)



overview of relational databases

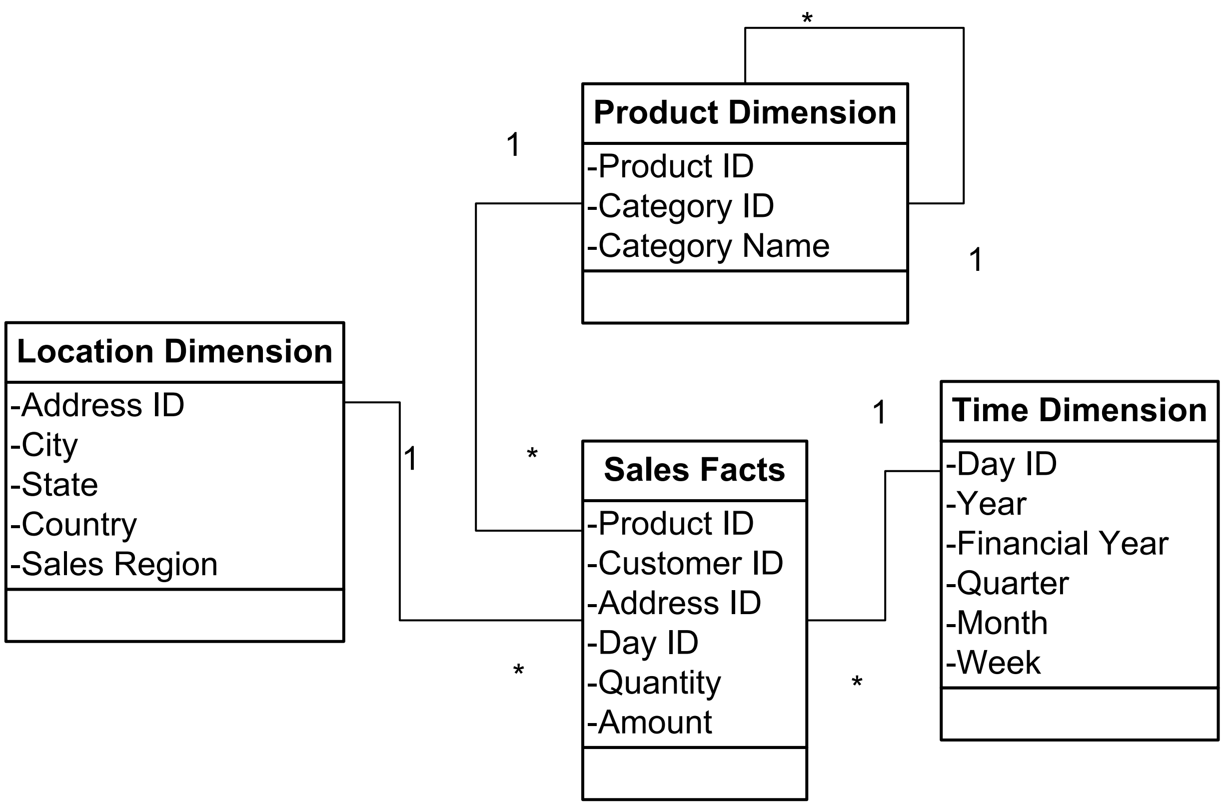


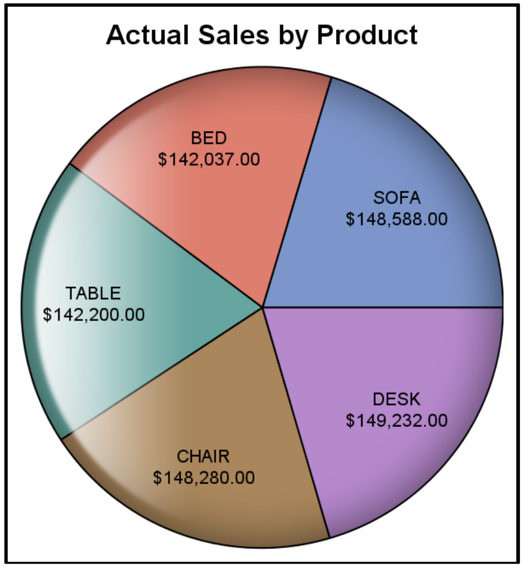
Row Oriented Database

Column Oriented Database

OLAP ("online analytical processing")

e.g.: select SUM(S.amount), S.pid, P.catname from S where S.did=T.did S.pid = P.pid and T.qrtr = 3 group by catname



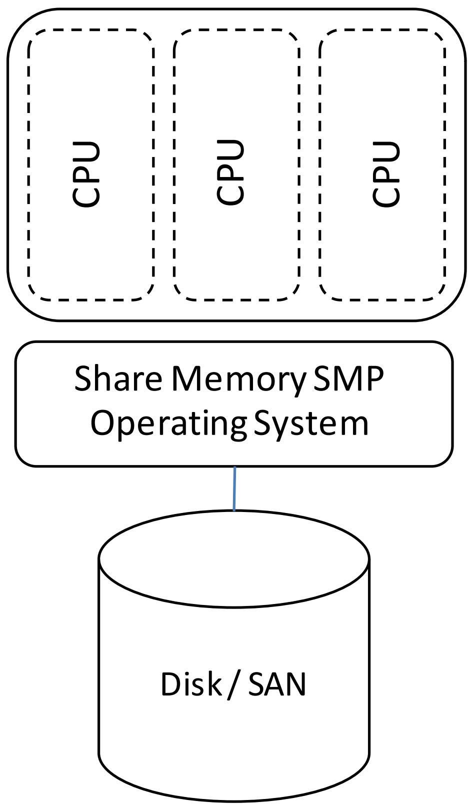


databases: why?

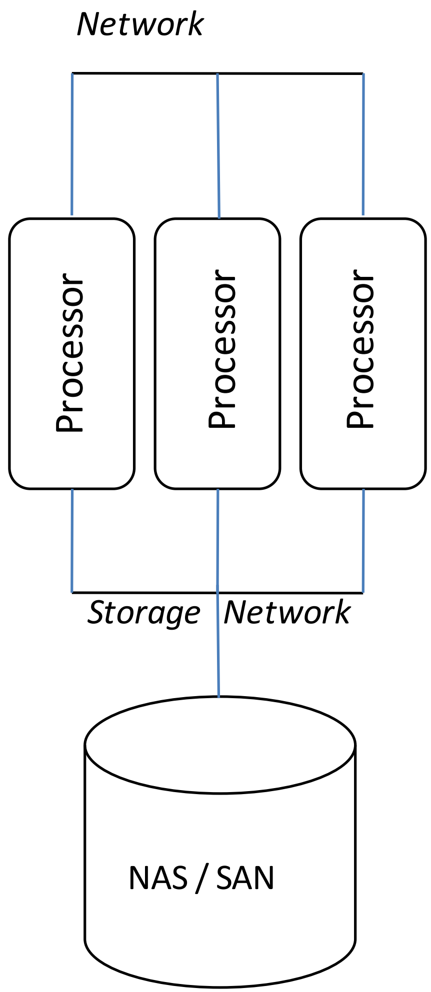
- transaction processing (ACID properties)
- SQL – queries and indexing

- transaction processing *not* need for analytics
  - though there *may* be advantages in not having to move data out of a transaction store if avoidable
- queries – yes, but if large volumes of data are being touched (e.g. joins, large-scale counting, building classifiers, etc.); indexes become *less* relevant
  - resilience to hardware failures, which MR provides, *is* vital.
- *but* OLAP – can be viewed as computing a part of the joint distribution $P(f_1 \ldots f_n)$ – using intuition to select

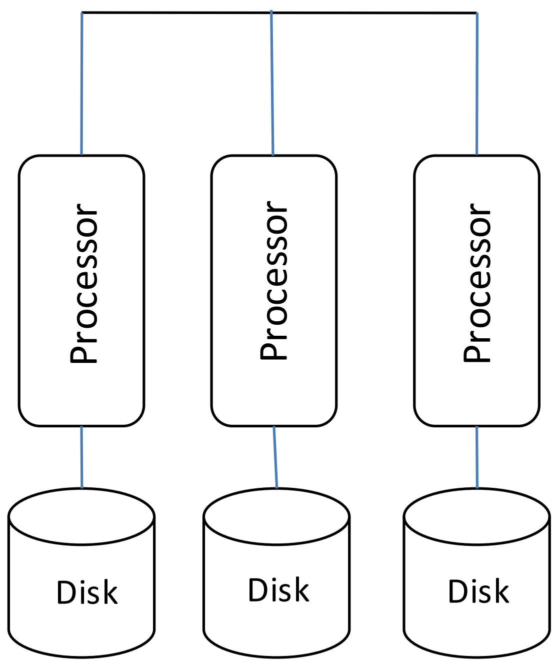
parallel databases



Shared Memory



Shared Disk



Shared Nothing

database evolution

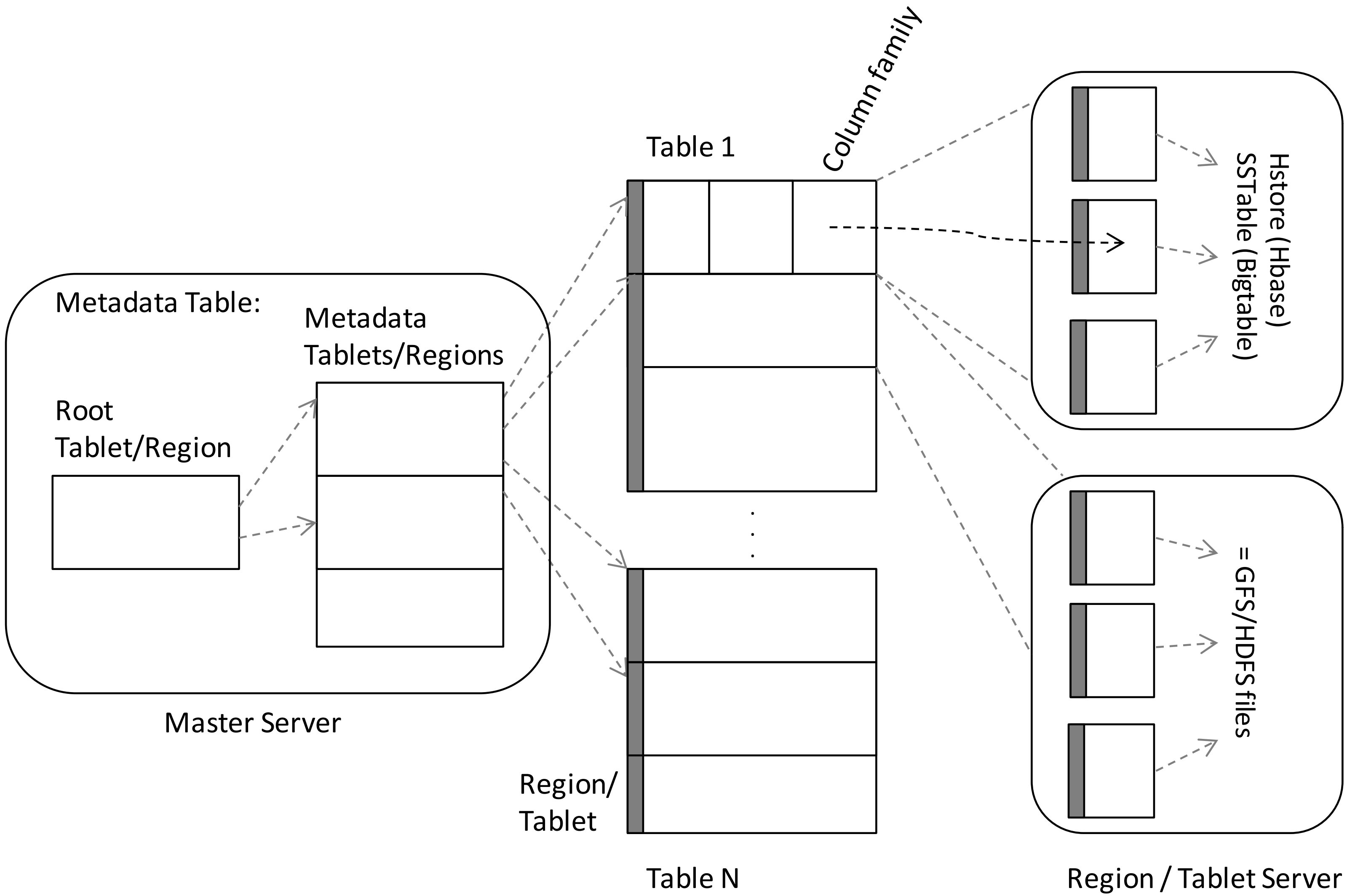
noSQL databases

- no ACID transactions
- *sharded* indexing
- restricted joins
- support columnar storage (if needed)

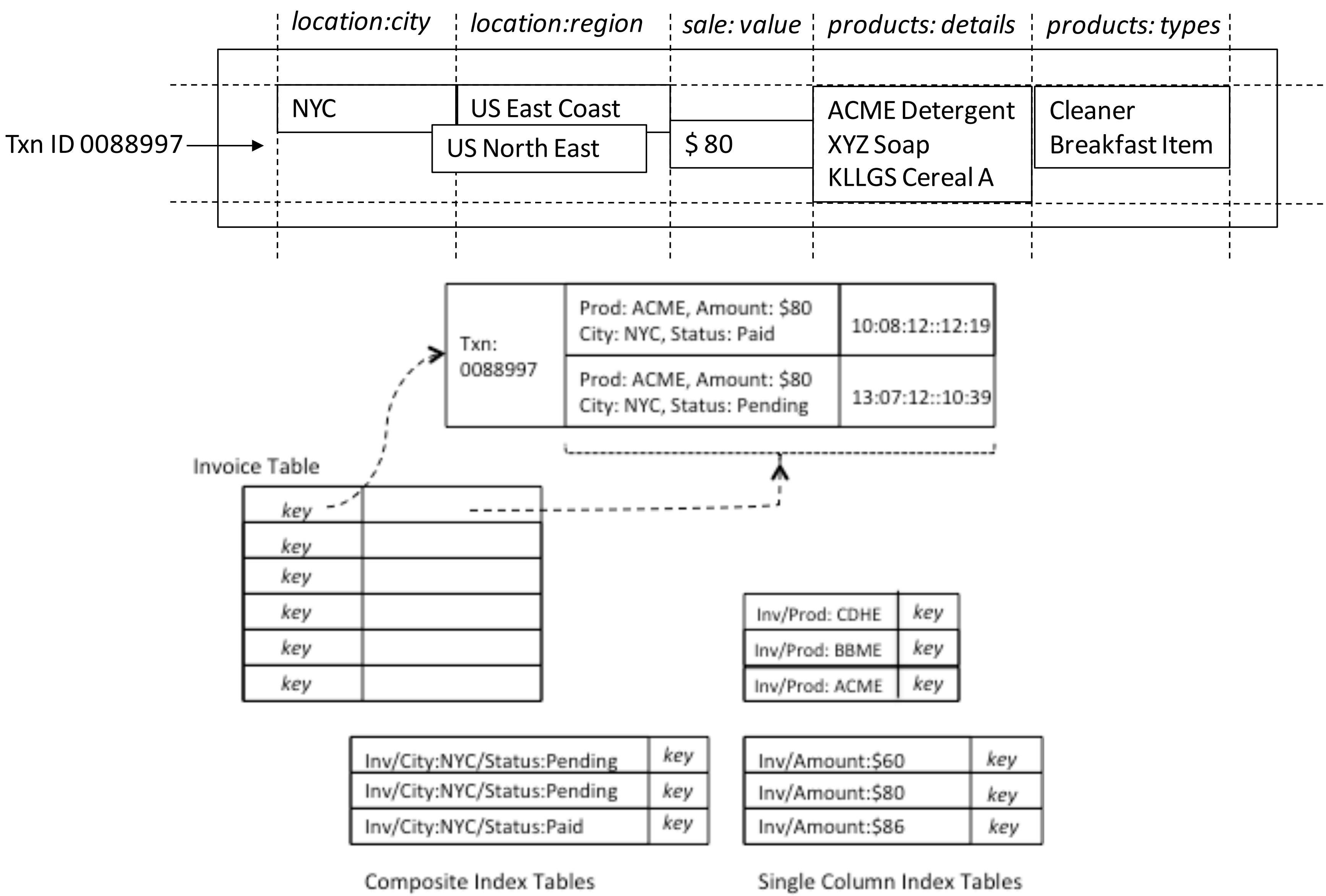
in-memory databases

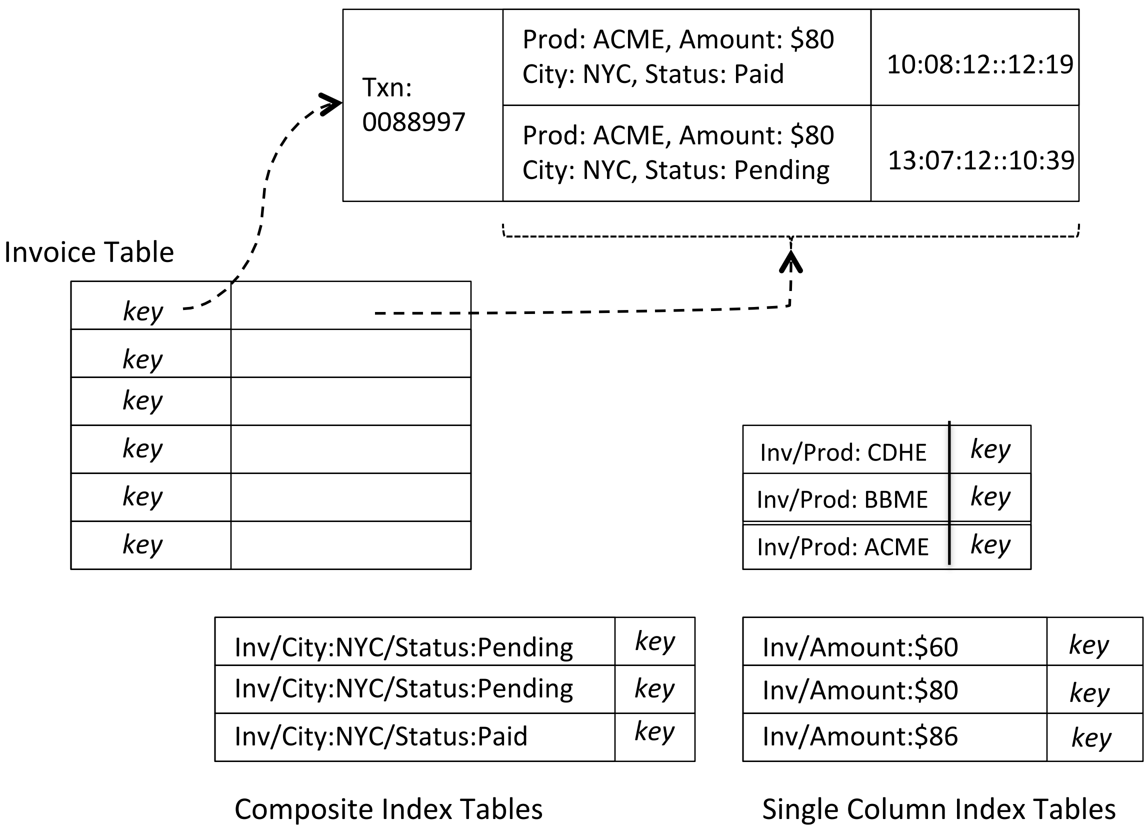
- real-time transactions
- variety of indexes
- complex joins

big-table (HBase)

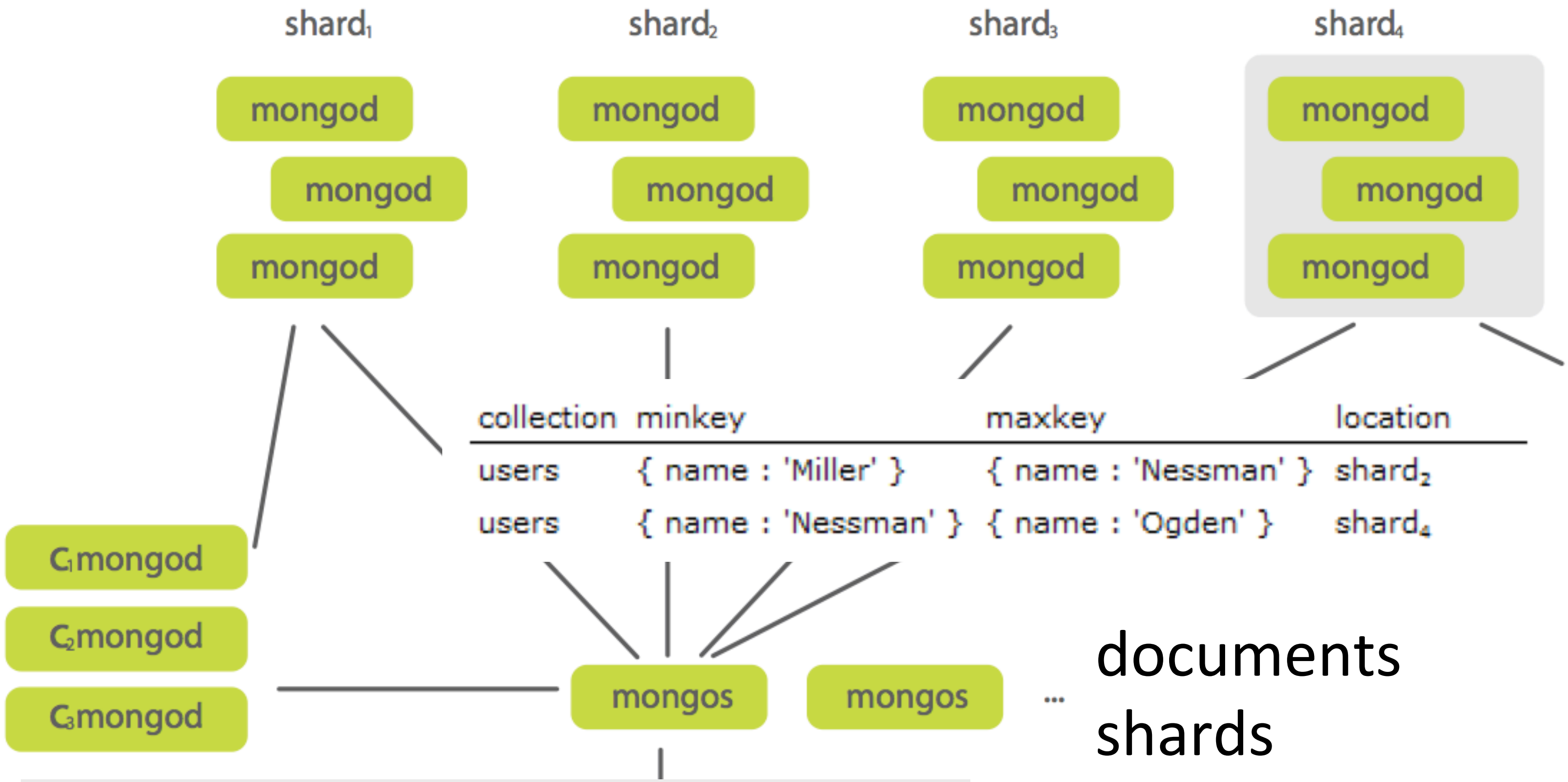


e.g. indexing using big-table




Txn:
0088997
Prod: ACME, Amount: $80
City: NYC, Status: Paid
10:08:12::12:19
Prod: ACME, Amount: $80
City: NYC, Status: Pending
13:07:12::10:39
Invoice Table
key
key
key
key
key
key
Inv/Prod: CDHE
key
Inv/Prod: BBME
key
Inv/Prod: ACME
key
Inv/City:NYC/Status:Pending
key
Inv/City:NYC/Status:Pending
key
Inv/City:NYC/Status:Paid
key
Inv/Amount:$60
key
Inv/Amount:$80
key
Inv/Amount:$86
key
Composite Index Tables
Single Column Index Tables

mongo DB




```
{
  title: "MongoDB",
  last_editor: "172.5.123.91",
  last_modified: new Date("9/23/2010"),
  body: "MongoDB is a...",
  categories: ["Database", "NoSQL", "Document Database"] ,
  reviewed: false
}
```

documents

shards

indexes – incl. *text*

map-reduce

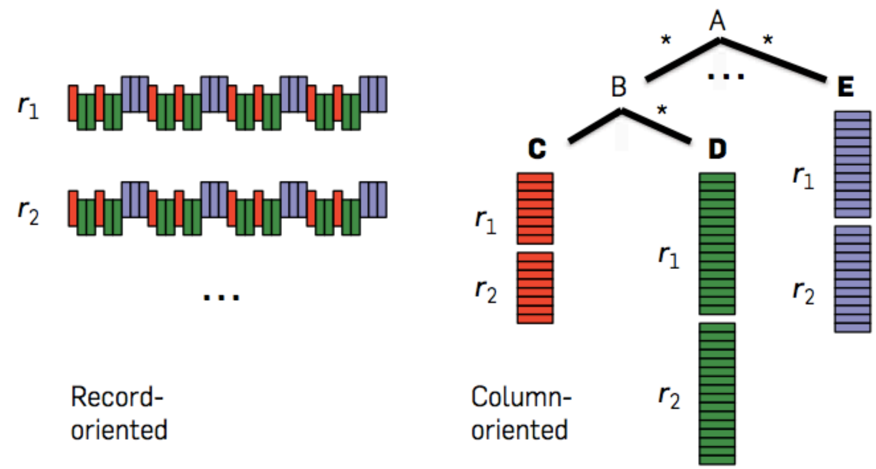
- (JavaScript)

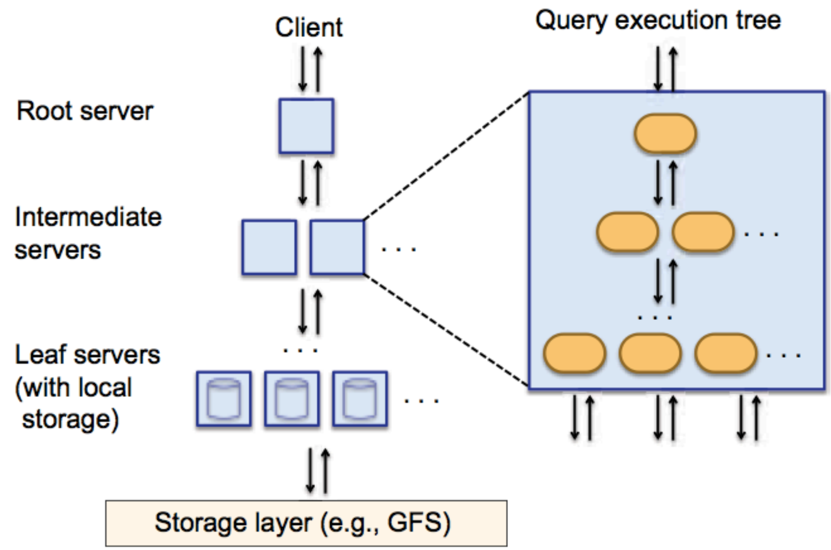
Dremel – new 'kid' on the block?

powers Google's "BigQuery"

two important innovations:

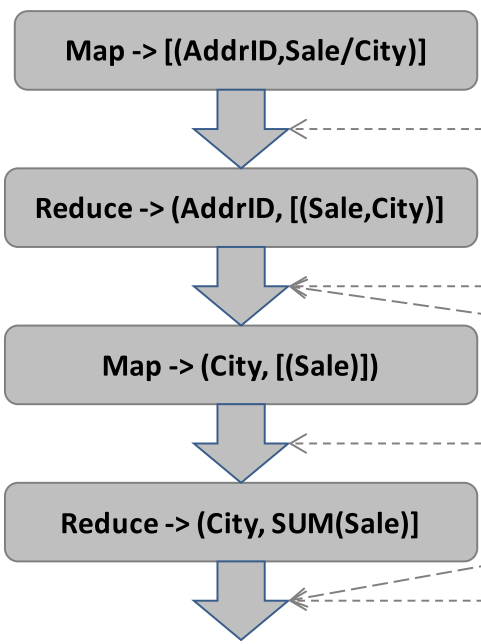
- columnar storage for *nested*, possibly non-unique fields – *leaf servers*
- tree of *query servers* pass intermediate results from root to leaves and back





➢ *orders* of magnitude better than MR on petabytes of data – speed and storage

SQL evolution: SQL-like MR coding



Pig Latin:

```
tmp  = COGROUP Sales  BY AddrID, Cities by AddrID
ioin = FOREACH tmp GENERATE FLATTEN(Sales), FLATTEN(Cities)
grp = GROUP join BY City
res = FOREACH grp GENERATE SUM(Sale)
```

HiveQL:

```
INSERT OVERWRITE TABLE join
SELECT s.Sale, c.City FROM Sales s
        JOIN  Cities c ON s.AddrID=c.AddrID;

INSERT OVERWRITE TABLE res
SELECT  SUM(join.Sale) FROM join GROUP BY join.City
```

SQL: SELECT SUM(Sale), City from Sales, Cities WHERE Sales.AddrID=Cities.AddrID GROUP BY City

SQL evolution: in-DB statistics, in parallel

```
sql> SELECT * FROM training;
 id | class | attributes
----+-------+------------
  1 |     1 | {1,2,3}
  2 |     1 | {1,2,1}
  3 |     1 | {1,4,3}
  4 |     2 | {1,2,2}
  5 |     2 | {0,2,2}
  6 |     2 | {0,1,3}
(6 rows)
```

```
sql> select * from toclassify;
 id | attributes
----+------------
  1 | {0,2,1}
  2 | {1,2,3}
(2 rows)
```

```
sql> SELECT madlib.create_nb_prepared_data_tables(
'training', 'class', 'attributes', 3, 'nb_feature_probs', 'nb_class_priors');
```

```
sql> SELECT madlib.create_nb_probs_view (
'nb_feature_probs', 'nb_class_priors', 'toclassify', 'id', 'attributes', 3, 'nb_probs_view_fast');

sql> SELECT * FROM nb_probs_view_fast;
 key | class | nb_prob
-----+-------+---------
   1 |     1 |     0.4
   1 |     2 |     0.6
   2 |     1 |    0.75
   2 |     2 |    0.25
(4 rows)
```

map-reduce evolution: iteration

many applications require repeated MR:

e.g. page-rank, continuous machine-learning …

1. iterate MR

   but make it more efficient: avoid data copy (HaLoop, Twister)

2. generalized data-flow graph of map->reduce tasks

   tasks are 'blocking' for fault-tolerance (Dryad/LINQ, Hyracks …)

3. direct implementation of recursion in MR

   how to recover from non-blocking tasks failing?

   graph model: (Pregel, Giraph)

   stream model: (S4)

hidden-agenda again…

is the brain's processing highly parallel – yes

does the brain do map-reduce – probably not
does the brain do indexing / databases – no

does the brain classify – appears to do so, yes
so how, i.e. what is its *architecture*?
we'll return to this question in 'predict'

summary

- distributed files – 2^{nd} basic element of big-data
- what databases are good for
  - and why traditional DBs were a happy compromise
- evolution of databases
- evolution of SQL
- evolution of map-reduce

Next week (5)

- ➢ no lecture; only ‘office hours’ based on forum
- ➢ following week (6): Learn: ‘facts’ from data