## **CS213**Principles of Database Systems(H)

### Chapter 13

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## 13.1 Query Optimizer

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**1978 Version 1** 

1985 Version 5

1988 Version 6







1994 First release

**ALL** major DBMS products were designed in the 1980s (MySQL is more recent but was based on SQL Server). They CANNOT change their architecture because it would be, business-wise, suicidal for them to require from customers dumping and reloading today's massive databases for a migration. Only incremental improvements are possible.

#### This is what a \$M 1 machine was

Mid 1980s looking like in the 1980s.

VAX 8600 (high-end)

32-bit architecture

Processor, ~10 to 20MHz clock

4 to 256 Mb of memory

I/Os ~10 to 30 Mb/s



Beginning of multi-processor computers

Keep in mind that the big DBMS products were designed for this.

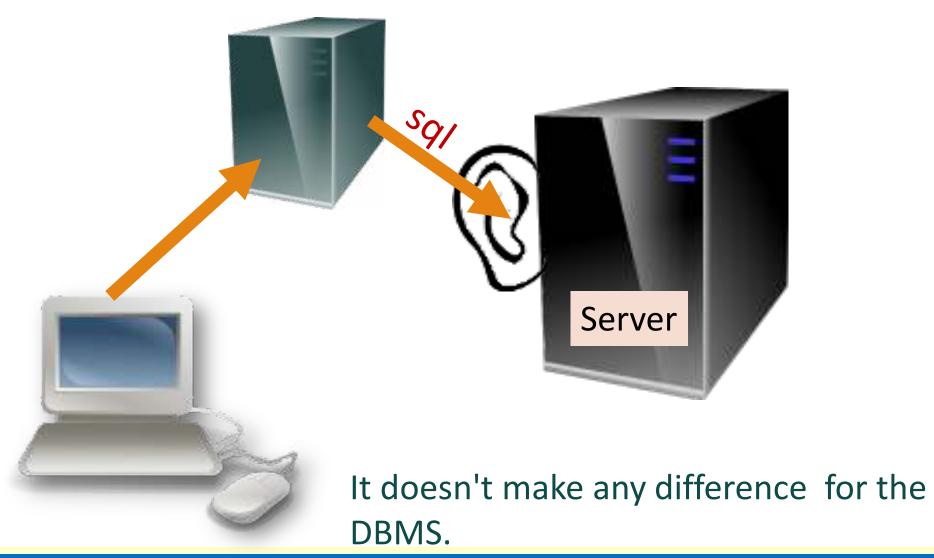


A 4 MB memory board for the VAX 8600

SQL queries will be directly sent to this server. The listener is only here



In many cases (HTTP server, application server) the end user isn't directly talking to the DBMS server.



```
select m.title, m.year_released from movies m inner join credits c on c.movieid = m.movieid inner join people p on p.peopleid = c.peopleid where p.first_name = 'Tim' and p.surname = 'Burton' and c.credited_as = 'D'
```



One way to improve efficiency is to keep data dictionary information (meta-data) in a shared cache to avoid additional queries.

meta-data

### Kept in memory

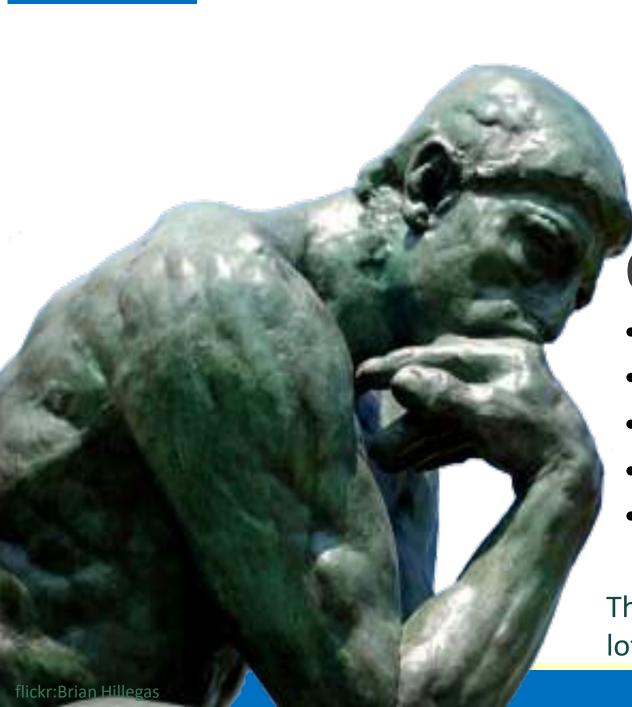
```
select m.title, m.year_released
from movies m
  inner join credits c
  on c.movieid = m.movieid
  inner join people p
  on p.peopleid = c.peopleid
  where p.first_name = 'Tim'
  and p.surname = 'Burton'
  and c.credited_as = 'D'
```



Another crucial phase is the one when the optimizer tries to determine the most efficient way to access data.

### Kept in memory

meta-data



QUERY OPTIMIZER

Logical transforms

Indexes Volumes Storage

Hardware performance

System load

Settings

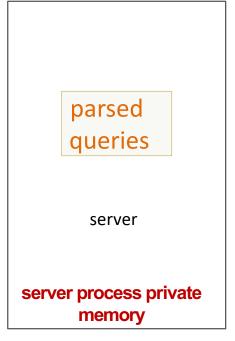
The optimizer has to (or can) take into account a lot of factors.

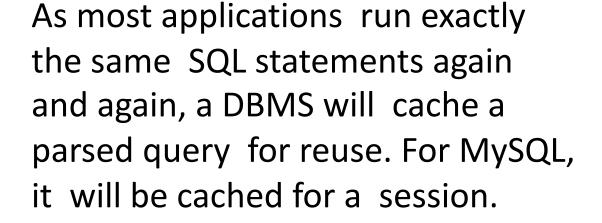
As a result

# PARSING takes time

Let's put it another way: we'd rather not parse exactly the same query many times.

#### Keep parsed queries in memory







## Query cache management LRU

#### Least Recently Used

Of course we cannot hold in cache zillions of parsed queries. We need to manage the cache, and replace queries that haven't been executed in a while with new ones.

select m.title, m.year released from movies m inner join credits c on c.movieid = m.movieid inner join people p on p.peopleid = c.peopleid where p.first name = 'Tim' and p.surname = 'Burton' and c.credited as = 'D'

### Checksum

We primarily recognize identical queries by computing a text checksum.

+ check tables are same and context identical

There are a lot of things to do for optimization because the resource is limited even we have a powerful server.

## 13.2 SCALING

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## SCALING UP

For many years, the answer to a database outgrowing the processing power of its server has been to replace the server by a bigger server.

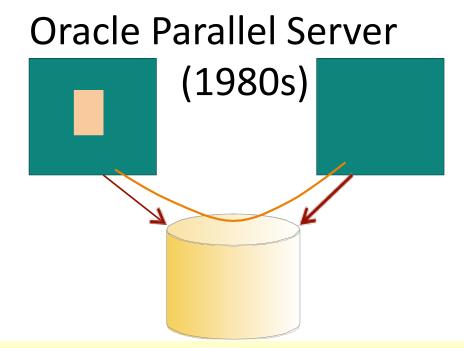


Having a single server is taking a risk if it breaks, and additionally migrating to another server is disruptive.



This is why people quickly thought of an alternative, adding more servers and making them share the load.

In the 1980s, Oracle tried connecting multiple servers to a single database. Complete disaster when the two servers wanted to modify the same data (or simply when one wanted to see what had just been modified by the other), data had to transit via the files.



# Oracle RAC (2001)

Real Application Clusters

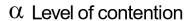


#### Neil Gunther's **Universal Law of Computational Scalability**

Relative capacity C(N) of a computational platform:

$$\mathbf{C}(N) = \frac{N}{1 + \alpha (N-1) + \beta N (N-1)}$$

$$0 \le \alpha, \beta < 1$$



 $\beta$  Coherency delay (latency for data to become consistent)



www.perfdynamics.com

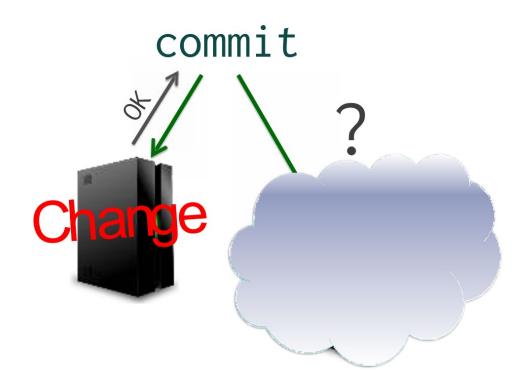
The problem with clustering is that it follows a law of diminishing returns: adding a second server will less than double your capacity, and in practice people have clusters of 2, 3 or 4 machines at most (Neil Gunther is a famous Australian consultant/academic specializing on performance)

One big problem is with transactions that involve several servers. Remember that transactions are meant to be atomic operations.

#### Distributed Transactions



It may happen that when we commit we know for sure it worked on one server, and we get no acknowledgment from the other.

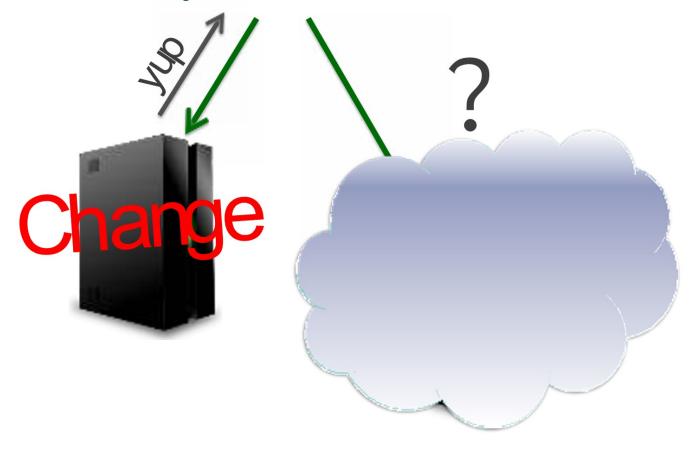


No way to rollback where it's committed, and we don't know if the other node failed before or after having committed the change.

### TWO-PHASE COMMIT

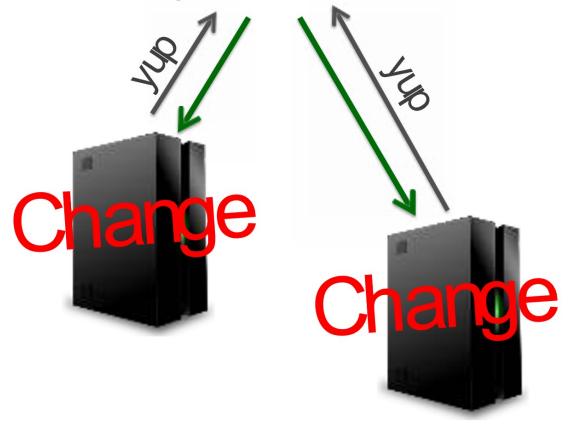
One algorithm was devised (a long time ago), called a "two-phase commit".

#### Ready to COMMit?

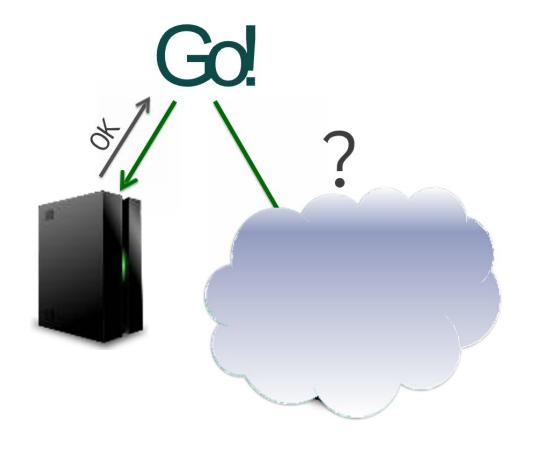


JUST before committing, you ask to all servers involved whether they are ready to commit. If you don't get an all clear, you can rollback.

#### Ready to Commit?



Otherwise, you can then (2<sup>nd</sup> phase) send the official "commit" signal



Odds that something will fail are far lower than with a single-phase commit, but not zero.

## Latency

Additionally, you have latency issues. All machines in a cluster may not be sitting next to each other, they may be a few miles apart in different data centers for security reasons (fire, flood ...)

# $1 \, \text{KM} = 0.0033 \, \text{ms}$ $2000 \, \text{KM} = 6.6 \, \text{ms}$

Distance from Shenzhen to Beijing

Even if information travels fast, multiplying exchanges (two-phase commit) may become a sensitive issue.