

Week 01.2: Storage and computation

DS-GA 1004: Big Data

Instructor: Brian McFee

What are our resources?

Storage

- Where and how is data kept?
- How much data can we keep?

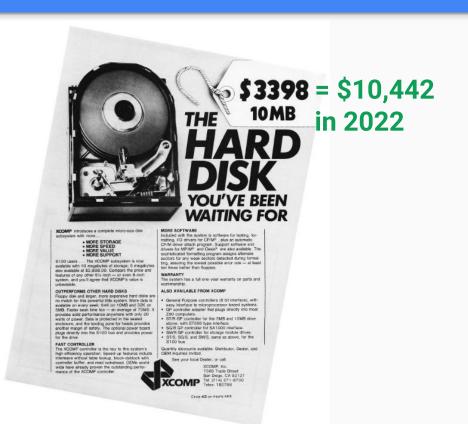
Computation

How quickly can we process data?

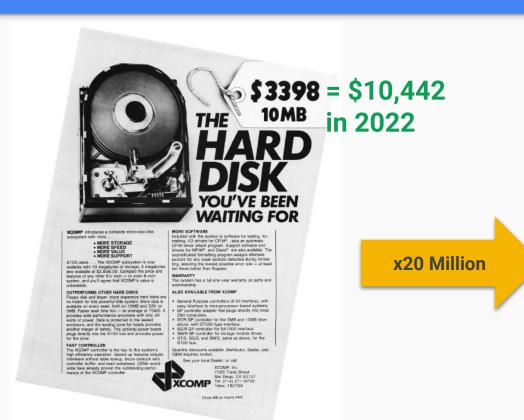
Communication

How quickly can we move data between locations?

The cost of storage over time...



The cost of storage over time...





x20

EGGE

HGST Ultrastar He10 HDD 10000GB Serial ATA III internal hard drive

\$8,591.00

Free Shipping

Volume and velocity

- (2019) Facebook generates 4 Petabytes (4e15 bytes) of new data per day https://research.fb.com/facebook-s-top-open-data-problems/
- (2019) YouTube collects 300 hours of HD video every minute http://www.businessofapps.com/data/youtube-statistics/
- (2017) Twitter stores >500PB of data

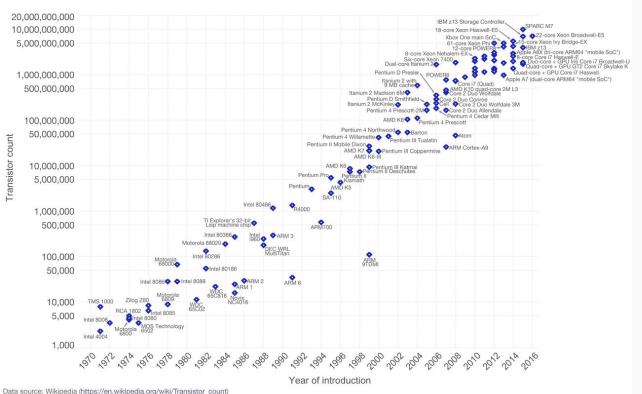
 https://blog.twitter.com/engineering/en_us/topics/infrastructure/2017/the-infrastructure-behind-twitter-scale.html
- (2017) CERN data center: > 200PB

 https://home.cern/news/news/computing/cern-data-centre-passes-200-petabyte-milestone

Moore's Law – The number of transistors on integrated circuit chips (1971-2016)

Moore's law describes the empirical regularity that the number of transistors on integrated circuits doubles approximately every two years.

This advancement is important as other aspects of technological progress – such as processing speed or the price of electronic products – are strongly linked to Moore's law.



Note: transistor count ≠ speed!

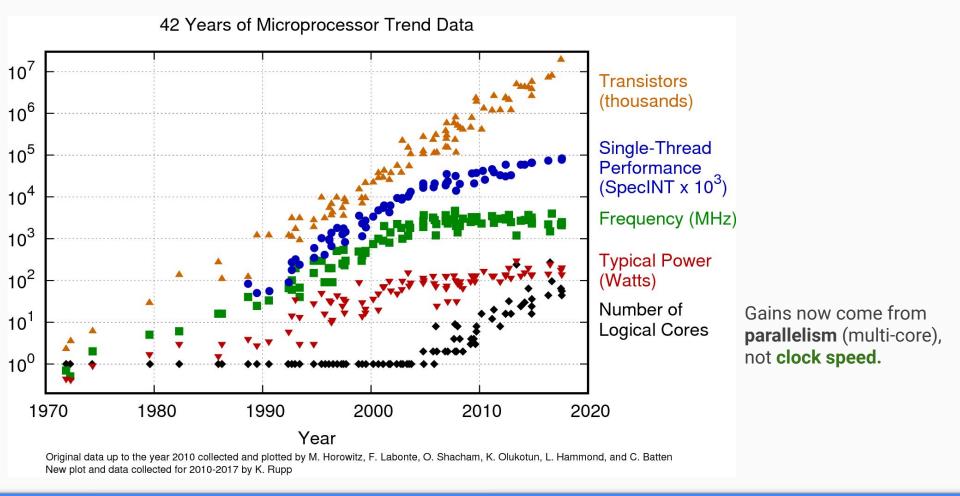
Moore in 2015:

Licensed under CC-BY-SA by the author Max Roser

"I see Moore's law dying here in the next decade or so."

Gordon Moore (1965, 1975)

The data visualization is available at OurWorldinData.org. There you find more visualizations and research on this topic.



What does this tell us?

- Storage capacity (per \$)
 continues to increase
- Data velocity is increasing
- CPU speed is not keeping up

In pioneer days, they used oxen for heavy pulling, and when one ox couldn't budge a log, they didn't try to grow a bigger ox.

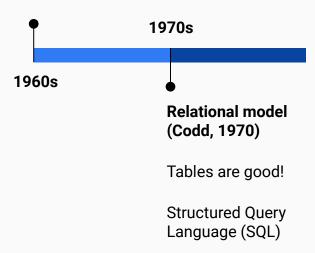
We shouldn't be trying for bigger computers, but for more **systems of computers**.



Custom software for each application / query



Custom software for each application / query

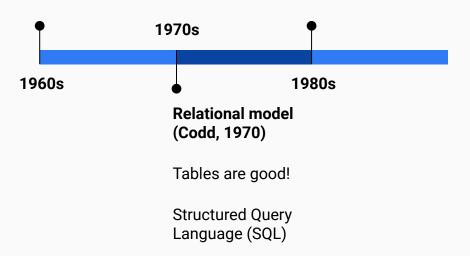


Custom software for each application / query

RDBMS takes off

Databases for commodity computers

SQL standardizes

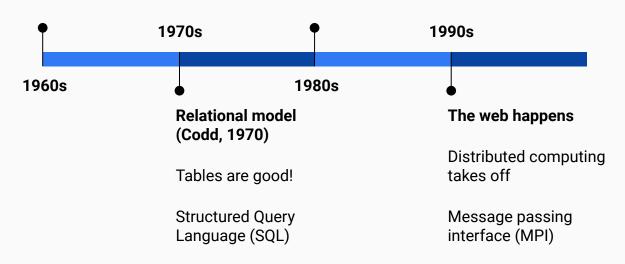


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Custom software for each application / query

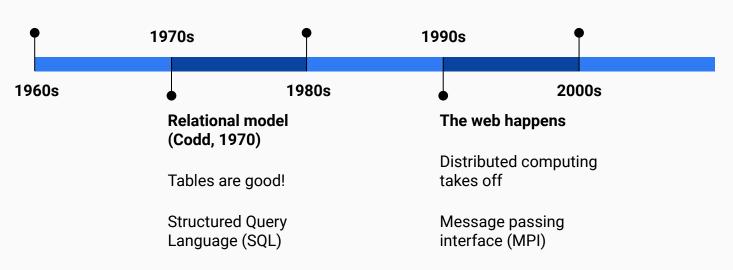
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Databases for commodity computers

SQL standardizes

MapReduce, NoSQL

Simpler abstractions for distributed computation and storage



Custom software for each application / query

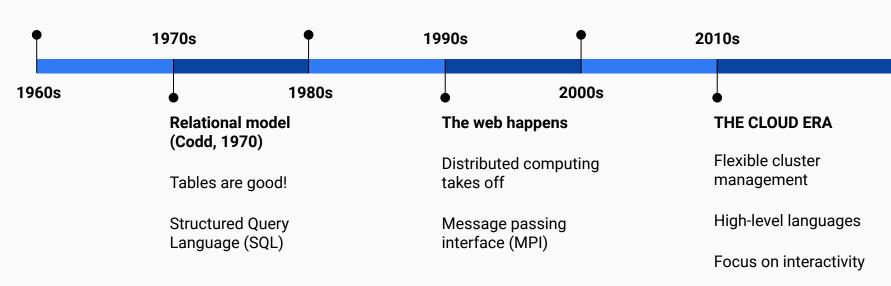
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Databases for commodity computers

SQL standardizes

MapReduce, NoSQL

Simpler abstractions for distributed computation and storage



Starting at the beginning

File systems!

- To understand modern tools, it helps to understand the previous generation.
- What problems did people face in designing:
 - o Relational databases?
 - o Map-reduce?
 - o Spark?
 - o Dask?

```
bmcfee@mariana.cims.nyu.edu /scratch/bmcfee/data/MSD
→ find data |head -30 |grep h5
data/A/A/A/TRAAAZF12903CCCF6B.
data/A/A/A/TRAAAAK128F9318786.h
data/A/A/A/TRAAAYX128F4263BC0.h
data/A/A/A/TRAAAAV128F421A322.h5
data/A/A/A/TRAAAAW128F429D538.h5
data/A/A/A/TRAAAAY128F42A73F0.h5
data/A/A/A/TRAAABD128F429CF47.h
data/A/A/A/TRAAACN128F9355673.hs
data/A/A/A/TRAAACV128F423E09E.h
data/A/A/A/TRAAADJ128F4287B47.hs
data/A/A/A/TRAAADT12903CCC339.h!
data/A/A/A/TRAAADZ128F9348C2E.h
data/A/A/A/TRAAAEA128F935A30D.h5
data/A/A/A/TRAAAED128E0783FAB.h5
data/A/A/A/TRAAAEF128F4273421.h5
data/A/A/A/TRAAAEM128F93347B9.h5
data/A/A/A/TRAAAEW128F42930C0.h5
data/A/A/A/TRAAAFD128F92F423A.h!
data/A/A/A/TRAAAFI12903CE4F0E.h5
data/A/A/A/TRAAAFP128F931B4E3.h5
data/A/A/A/TRAAAFW128F42A4CFD.h5
data/A/A/A/TRAAAGF12903CEC202.h5
data/A/A/A/TRAAAGR128F425B14B.hs
data/A/A/A/TRAAAGW12903CC1049.h5
data/A/A/A/TRAAAHD128F42635A5.h5
data/A/A/A/TRAAAHE12903C9669C.h
```

- Use directories to organize your data
- Structured data can be stored as files
 - ⇒ data persists across application runs

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data/A/A/A/TRAAADZ128F9348C2E.h
data/A/A/A/TRAAAEA128F935A30D.h
data/A/A/A/TRAAAED128E0783FAB.hs
data/A/A/A/TRAAAEF128F4273421.h5
data/A/A/A/TRAAAEM128F93347B9.hs
data/A/A/A/TRAAAEW128F42930C0.h
data/A/A/A/TRAAAFD128F92F423A.h
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```

- Use directories to organize your data
- Structured data can be stored as files
 - ⇒ data persists across application runs
- Some great properties:
 - Easy to implement
 - Portable across systems
 - Network file systems (NFS) enable limited distributed processing

File systems can be awesome!

- If the data doesn't have obvious structure to index
 - Or if your data is naturally organized by the file system (name and time)
 - o E.g., system logs
- **Or** if indexing isn't worth the cost
 - E.g., one-off jobs
- Or if you want maximum portability across platforms

Reasons not to rely (only) on file systems

- Does not expose or exploit the structure of data
 - What if I want to search by file contents?
 Better options than brute force?
- Each query / analysis required writing a new program
 - Little re-usability between similar analyses
 - Or the same analysis with slightly different data structures
- Directory hierarchies may be too restrictive
- Practical limitations of filesystems (inode count, etc)

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data/A/A/A/TRAAAGF12903CEC202.
data/A/A/A/TRAAAGR128F425B14B.
data/A/A/A/TRAAAHD128F42635A5.
data/A/A/A/TRAAAHE12903C9669C.
```

When are file systems not awesome?

- When your data is structured along multiple axes
- When data have complex interactions
- When your analyses are complex
- When the benefits of indexing outweigh the costs
 - E.g., finding items in sub-linear time
- Relational databases to the rescue!

Databases did not replace file systems

- File archives are still the most common way to share large datasets
 - But we usually include some metadata/indexing structure as well
- As we'll see soon, Hadoop relies on a distributed file system
 - DB abstractions can be built on top
 - But this comes with restrictions on file contents
- Key differences now:
 - Standardized (restricted) file formats ("CSV", Parquet, HDF5, NPY, etc.)

Next week

Relational databases

- Structure can be a very good thing!
- Relational model and SQL
- Databases as a service