

Slides: Tatyana Mozgacheva

tmozgach@sfu.ca

Challenges and Opportunities with Big Data



Agenda

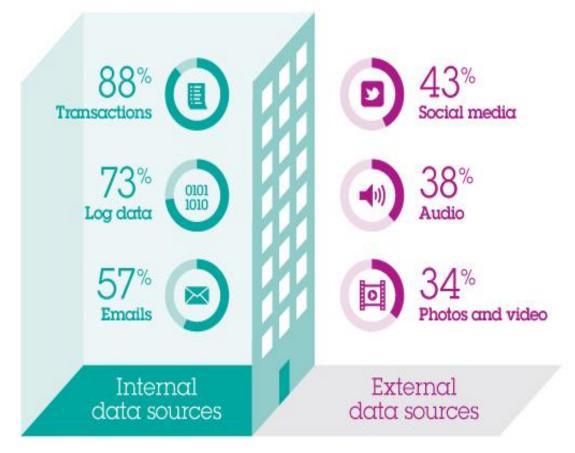
- ✓ What is the Big Data?
- ✓ Application of the Big Data
- √ 3 Vs of the Big Data
- ✓ The Big Data analysis pipeline
- √ Challenges





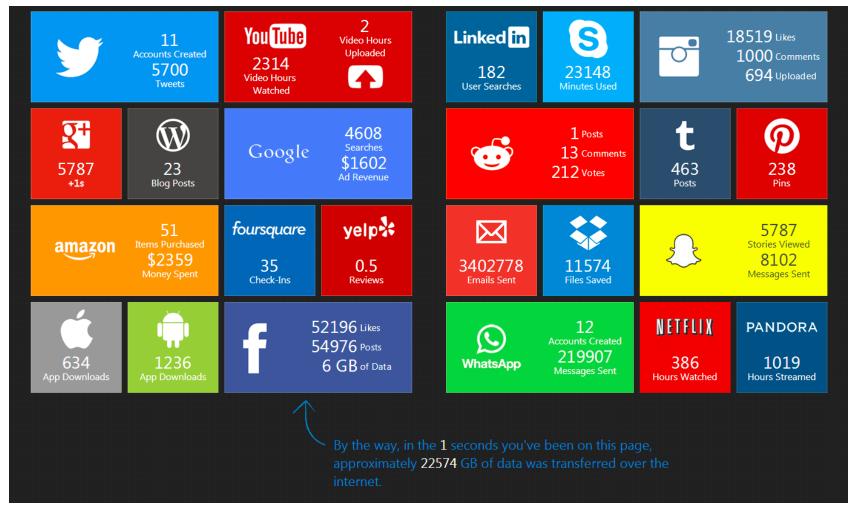
Big Data

Every day, we create **2.5 quintillion bytes of data** — so much that 90% of the data in the world today has been created in the **last two years alone**.





How much data is generated in social media in 1 second?





Application Of Big Data

Smarter Healthcare



Multi-channel sales



Homeland Security



Telecom



Traffic Control



Trading Analytics



Manufacturing



Search Quality

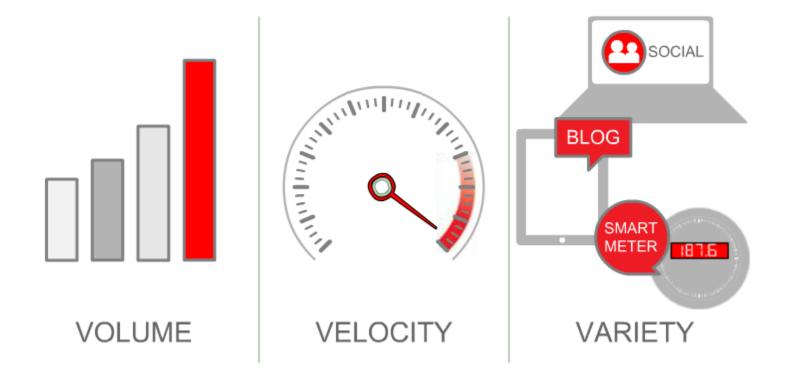




Our future?

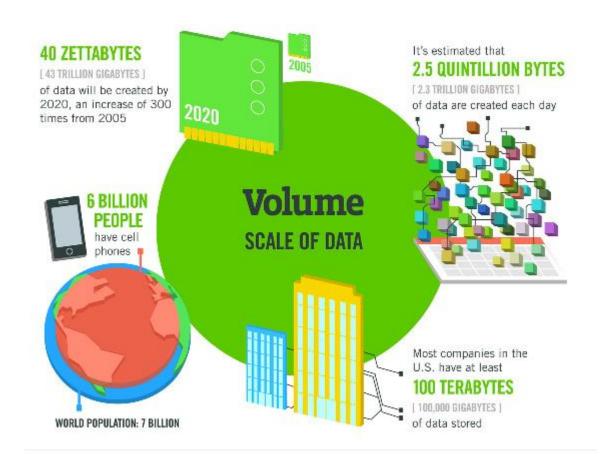


Big data spans three dimensions:



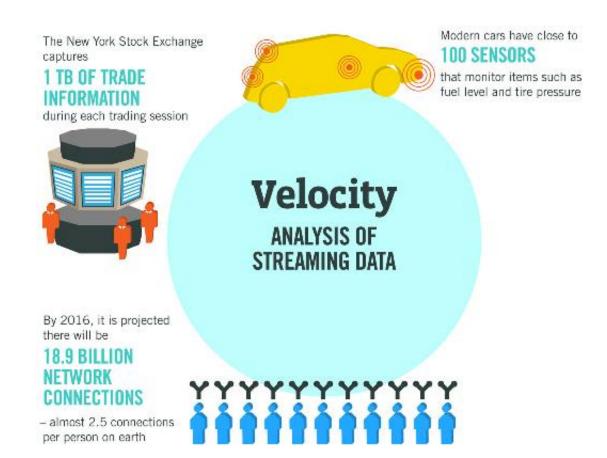


Volume



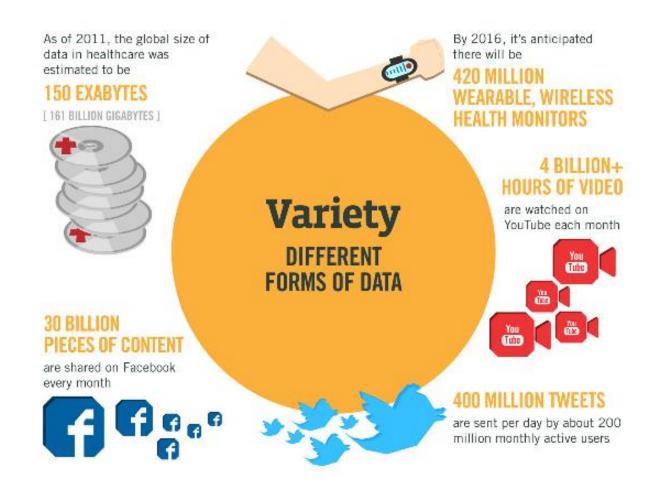


Velocity



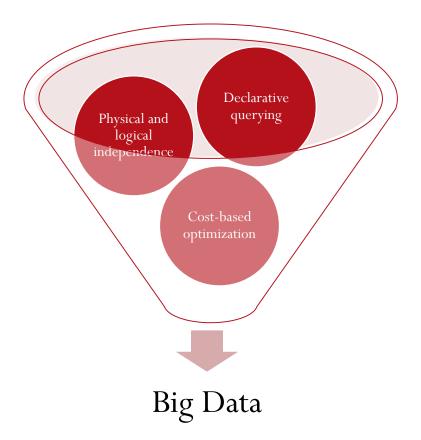


Variety



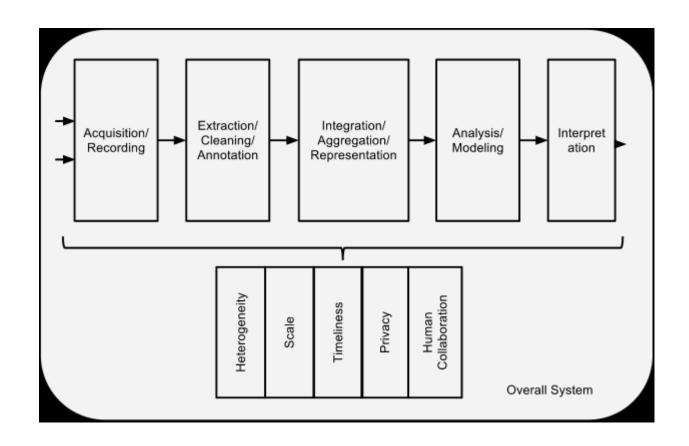


What has been Achieved...



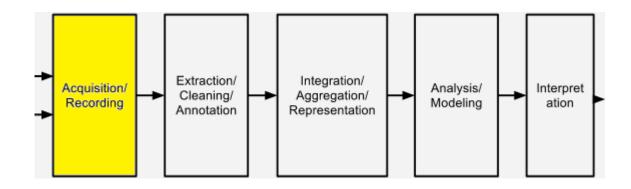


Phases and challenges in the Big Data Analysis Pipeline





The Big Data Analysis Pipeline





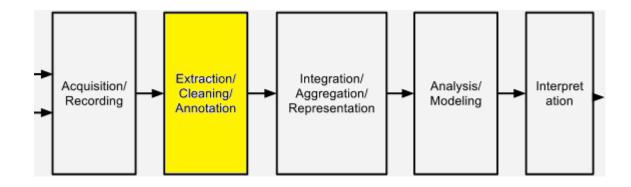
Data Acquisition and Recording

- ☐ What data to **keep**?
- What to discard?
- ☐ How to filter out the data on the fly?
- ☐ What is right **metadata**?



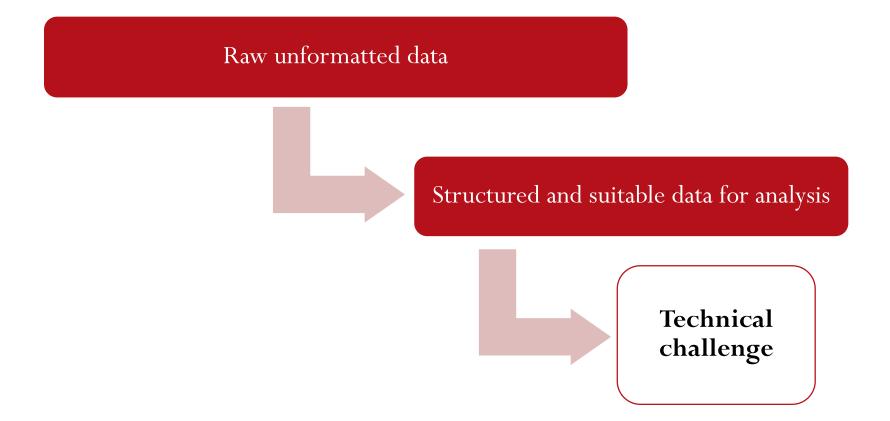


The Big Data Analysis Pipeline



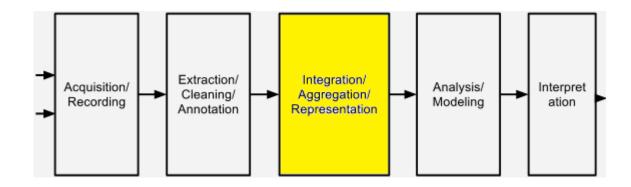


Information Extraction and Cleaning





The Big Data Analysis Pipeline





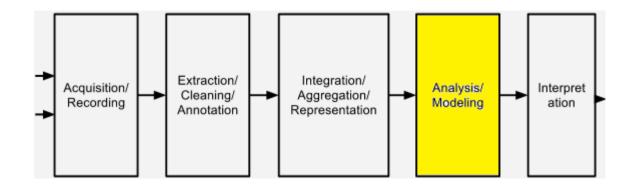
Data Integration, Aggregation, and Representation

- > How to combine **heterogeneous** data?
- How to select a suitable database design?





The Big Data Analysis Pipeline



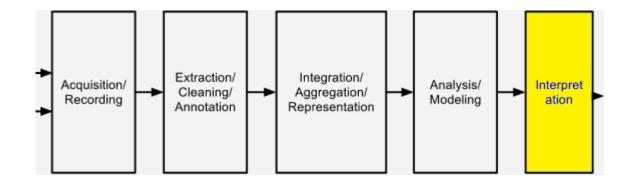


Query Processing, Data Modeling, and Analysis

- ➤ Querying and mining Big Data are fundamentally **different** from traditional statistical analysis.
- > Information redundancy can be explored for:
- missing data,
- to crosscheck conflicting cases,
- to validate trustworthy relationships,
- to uncover hidden relationships and models.
- ➤ Lack of coordination between database systems with analytics packages (e.g. statistical analyses).



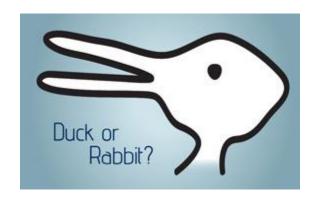
The Big Data Analysis Pipeline





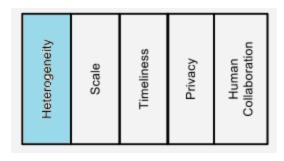
Interpretation

- > Simplify life of analyst.
- ➤ Necessity of **provenance** data to repeat the analysis with different assumptions, parameters, or data sets.





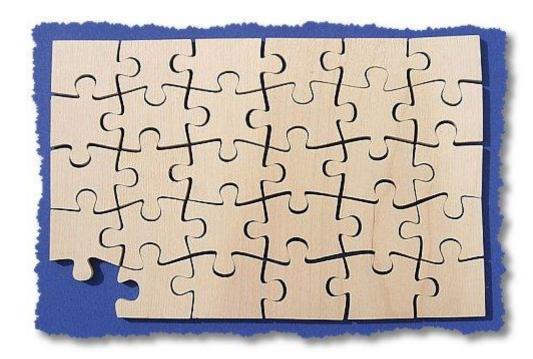
Challenges in the Big Data Analysis Pipeline





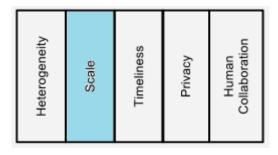
Heterogeneity and Incompleteness

- > Data from **different sources**/platforms.
- > Data formats are different.
- > Data missing due to security, privacy, or other reasons.





Challenges in the Big Data Analysis Pipeline





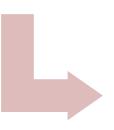
Scale

Data volume is **scaling faster** than compute resources.

Dramatic shift:

Increasing numbers of cores

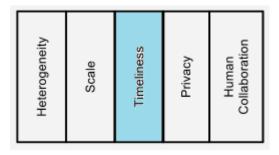
Parallelism within a single node



Parallel data processing techniques that were applied in the past for processing data across nodes **don't directly apply** for intra-node parallelism



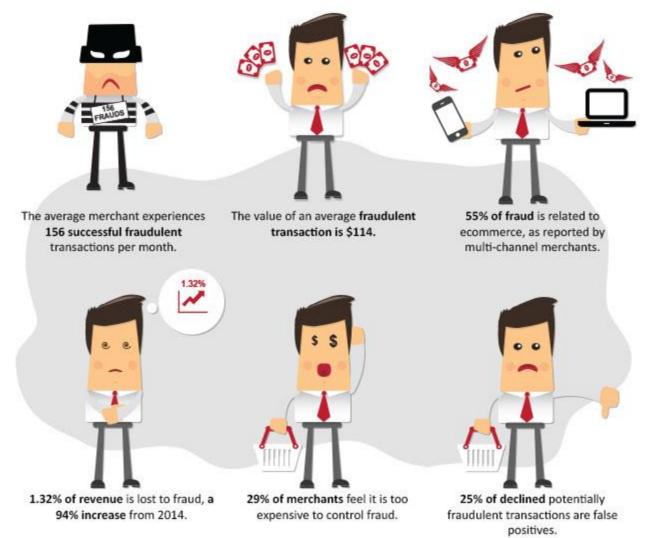
Challenges in the Big Data Analysis Pipeline





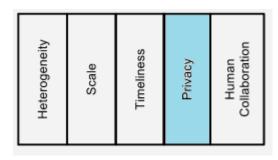
Timeliness

A full analysis of data is not feasible in real-time.



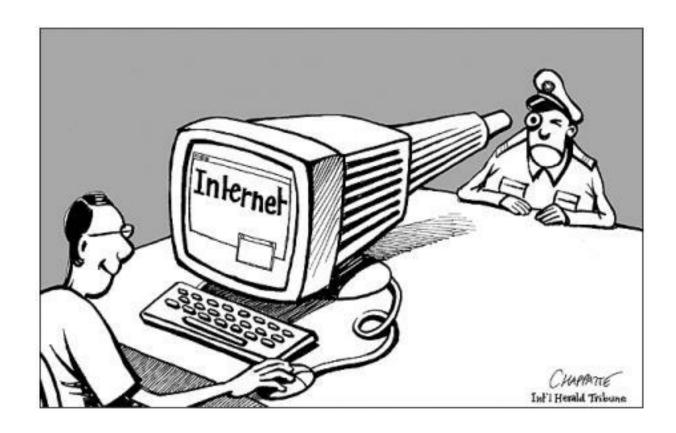


Challenges in the Big Data Analysis Pipeline



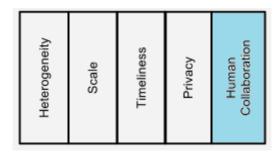


Privacy





Challenges in the Big Data Analysis Pipeline

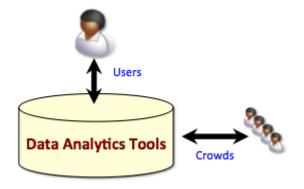




Human collaboration



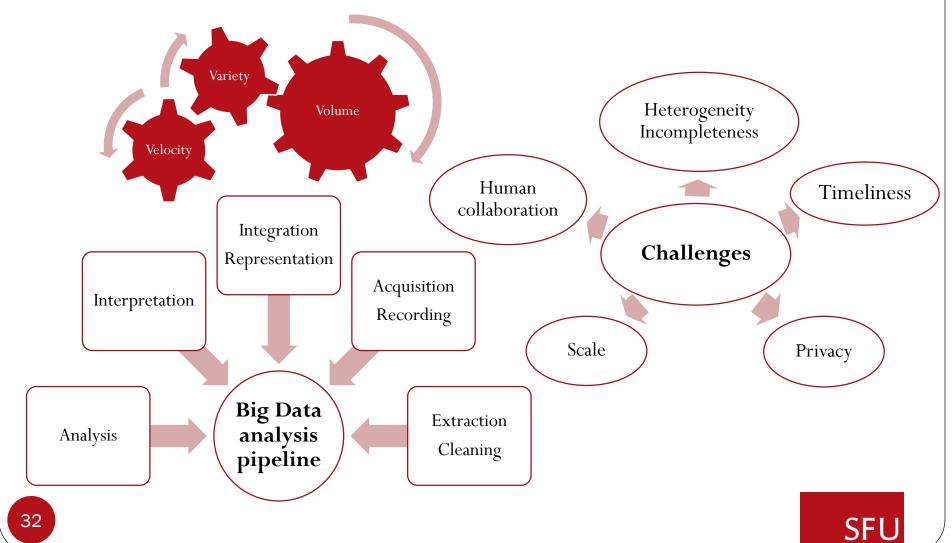






Conclusion

Objective: many technical and ethnical **challenges** must be **addressed before** this potential can be realized fully.



THANK YOU FOR YOUR ATTENTION ANY QUESTIONS