



Chapter 1:

Introduction to Big Data — the four V's

This chapter is mainly based on the Big Data script by Donald Kossmann and Nesime Tatbul (ETH Zürich)



Goal of Today



- What is Big Data?
 - introduce some major buzz words
- What is not Big Data?
 - get a feeling for opportunities & limitations





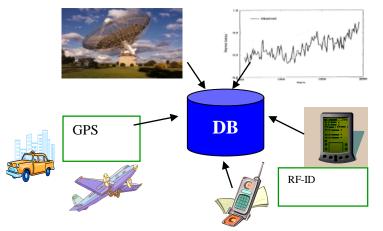
- Problem:
 - sales for lollipops are going down
- Data:
 - all sales data by customer, region, time, ...
- Information:
 - Iollipops bought by people older than 25 (but eaten by people younger than 10)
- Knowledge:
 - moms believe: lollipops = bad teeth
- Value:
 - dentists advertise your lollipops



Why is this difficult?



- You need more data than your data warehouse.
 - you need more data that you have
 - logs, Twitter feeds, blogs, customer surveys, ...



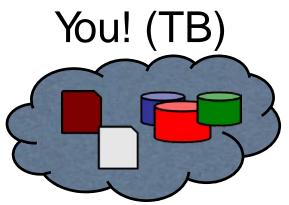
- You need to ask the right questions.
 - data alone is silent
- You need technology and organization that help you concentrate on asking the right questions.



From "Small Data" to "Big Data"



• Step 1:

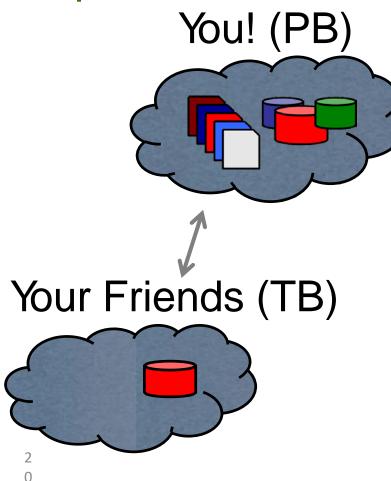




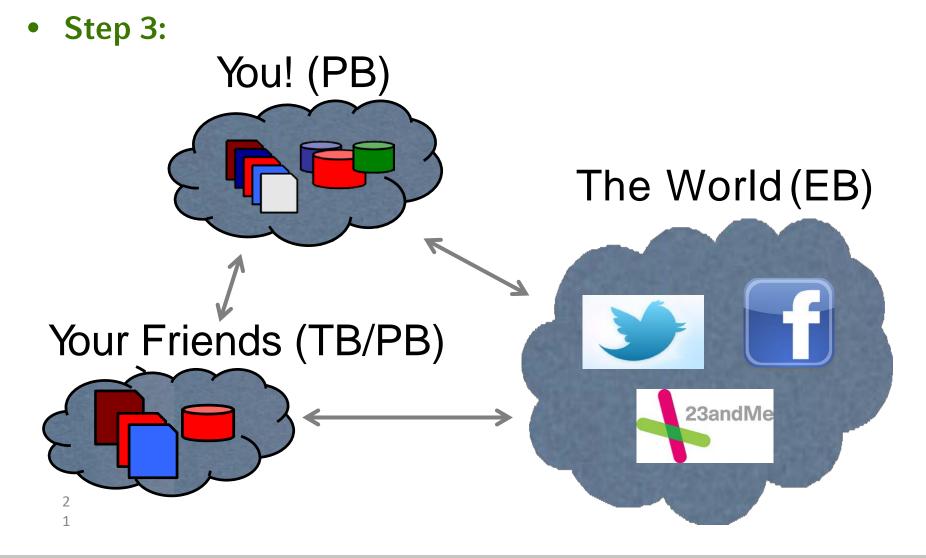
From "Small Data" to "Big Data"



• Step 2:



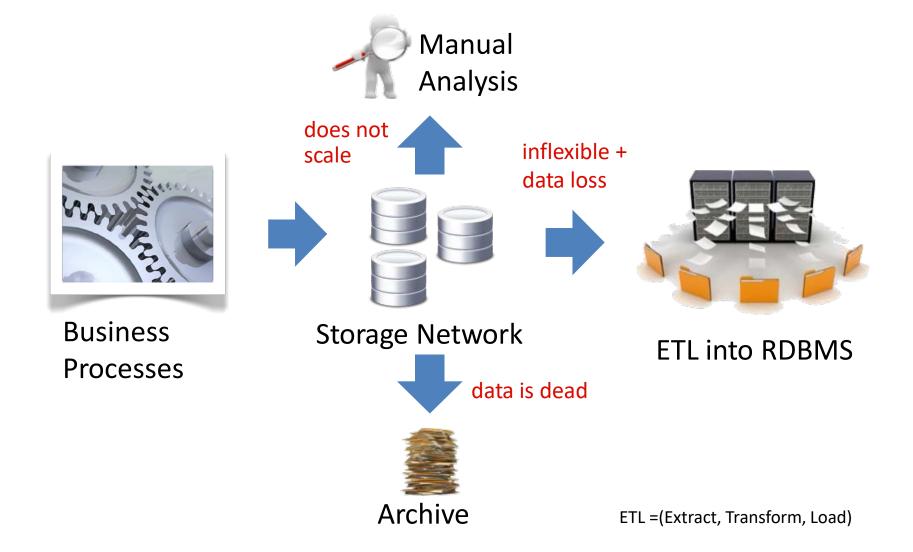






Limitations of State of the Art

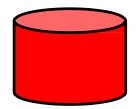








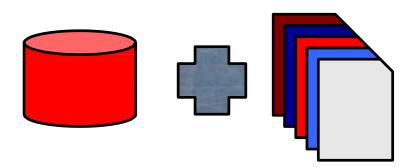
- Take Steps 0 to 3
 - Step 0: Data Warehouses (relational Databases)
 - Step 1: Data Warehouses + Hadoop (HDFS)
 - Step 2: Business Processes + Analytics + Exchange
 - Step 3: BP + Analytics + Exchange + Real-Time







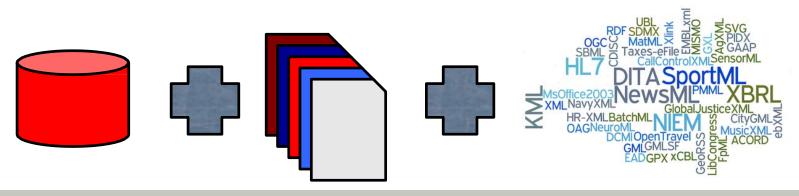
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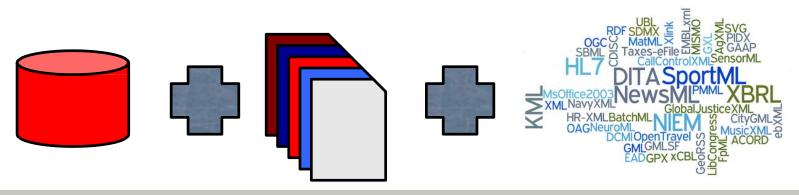
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- Take Steps 0 to 3
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 - Step 1: Data Warehouses + Hadoop (HDFS)
 - Step 2: Data Warehouses + Hadoop + XML (Standards)
 - Step 3: Data Warehouses + Hadoop + XML + ?





What needs to be done? (Organisation)



- Static Business Model -> Agile Business Model
 - You and your customers adapt to each other
 - No more data silos (ownership of data is distributed)
 - You allocate resources on demand
- Execute Business Process -> Data Science
 - You think about experience you have made



What is Big Data?



- Three alternative perspectives
 - philosophical
 - business
 - technical
- (Ultimately, it is a buzz word for everybody.)





intelligence

experience

- What is more valuable, if you had to pick one?
 - experience or intelligence?
- Traditional (computer) science: logic!
 - understand the problem, build model / algorithm
 - answer question from implementation of model
- New twist in (computer) science: statistics!.
 - collect data
 - answer question from data (what did others do?)



Statistics vs. Logic?



- Problems:
 - Find a spouse?
 - Should Adam bite into the apple?
 - 1 + 1?
 - Cure for cancer?
 - How to treat a cough?
 - Should I give Matthias a loan?
 - Premium for life insurance?
 - When should my son come home?
 - Which book should I read next?
 - Translate from German to English.

What Type of solutions (statistic/logic)? What do you think?



Statistics vs. Logic?



- Problems:
 - Find a spouse?
 - Should Adam bite into the apple? If you believe...
 - 1 + 1?
 - Cure for cancer?
 - How to treat a cough?
 - Should I give Matthias a loan?
 - Premium for life insurance?
 - When should my son come home? No, but...
 - Which book should I read next? Yes (e.g. Amazon)
 - Translate from German to English. Yes (Google Transl.)

Is there a solution?

- I don't want to know!
- Yes (Definition)
- I don't know, maybe.
- Yes (Google Insight)
- Yes (e.g. Schufa)
- YES (e.g. Alliance)





- New approach to do science
 - Step 1: Collect data
 - Step 2: Generate Hypotheses
 - Step 3: Validate Hypotheses
 - Step 4: (Goto Step 1 or 2)
- Why is this a good approach?
 - it can be automated: no thinking, less error
- Why is this a bad approach?
 - how do you debug without a ground truth?



Is bigger = smarter?



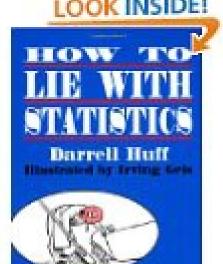
- Yes!
 - tolerate errors
 - discover the long tail and corner cases
 - machine learning works much better



Is bigger = smarter?



- Yes!
 - tolerate errors
 - discover the long tail and corner cases
 - machine learning works much better



• But!

- more data, more error (e.g., semantic heterogeneity)
- with enough data you can prove anything
- still need humans to ask right questions





- Google Translate
 - you collect snipets of translations
 - you match sentences to snipets
 - you continuously debug your system
- Why does it work?
 - there are tons of snipets on the Web
 - there is a ground truth that helps to debug system





- Which lane is fastest in a traffic jam?
 - you ask people where they go and whether happy
 - (maybe, you even use a GPS device)
 - you conclude that left lane is fastest
- Why is this stupid?
 - because there is no ground truth!
 - you will get a conclusion because Big Data always gives an answer. But, it does not make sense!
 - getting more data does not help either





- Step 1: You visit (and pay) "oracles"
 - they tell you which numbers to play
- Step 2: You visit (and pay) "interpreters"
 - they explain what oracles told you
- Step 3: After you lost, you visit (and pay) "analyst"
 - they explain why "oracles" and "interpreters" were right
- goto Step 1
- Lessons learned
 - life is try and error; trying keeps the system running

[Luciano de Crescenzo: Thus Spake Bellavista]



What is Big Data?



- Business Perspective
 - it is a new business model
- People pay with data
 - e.g. Facebook, Google, Twitter:
 - use service, give data
 - Google sells your data to advertisers
 - (you pay advertisers indirectly)
 - e.g., 23andMe, Amazon:
 - pay service + give data
 - sells data and uses data to improve service



Business Perspective



- Bank
 - keeps your money securely (kind of...)
 - puts your money at work (lends it to others), interest
 - you keep ownership of money and take it when needed
- Databank
 - keeps your data securely (kind of...)
 - puts your data at work: interest or better service
 - (you keep ownership of data: hopefully to come)





- You collect all data
 - the more the better -> statistical relevance, long tail
 - keeping all is cheaper than deciding what to keep
- You decide independently what to do with data
 - run experiments on data when question arises
- Huge difference to traditional information systems
 - design upfront what data to keep and why!!!
 - (e.g., waterfall model of software engineering!)





- Volume: data at rest
 - it is going to be a lot of data
- Speed: data in motion
 - it is going to arrive fast
- **Diversity: data in many formats**
 - it is going to come in different shapes
 - (e.g., different versions, different sources)
- Complexity: You want to do something interesting
 - SQL will not be enough





The 4 Vs of Big Data

- Volume: same as before
- Velocity: same as "speed"
- Variety: same as "diversity"
- Veracity: data in doubt
 - you do not know exactly what you have





Literature does not agree upon the # of Vs defining Big Data Examples:

• Laney 2001

Laney D. 3D data management: controlling data volume, velocity, and variety, META Group, Tech. Rep. 2001. http://blogs.gartner.com/doug-laney/files/2012/01/ad949-3D-Data-Management-Controlling-Data-Volume-Velocity-and-Variety.pdf.

talks about 3 Vs: volume, velocity, and variety

• later in Van Rijmenam 2014 and Borne 2014

van Rijmenam M. Why the 3v's are not sufficient to describe big data, BigData Startups, Tech. Rep. 2013. <u>http://www.bigdata-startups.com/3vs-sufficient-describe-big-data/</u>.

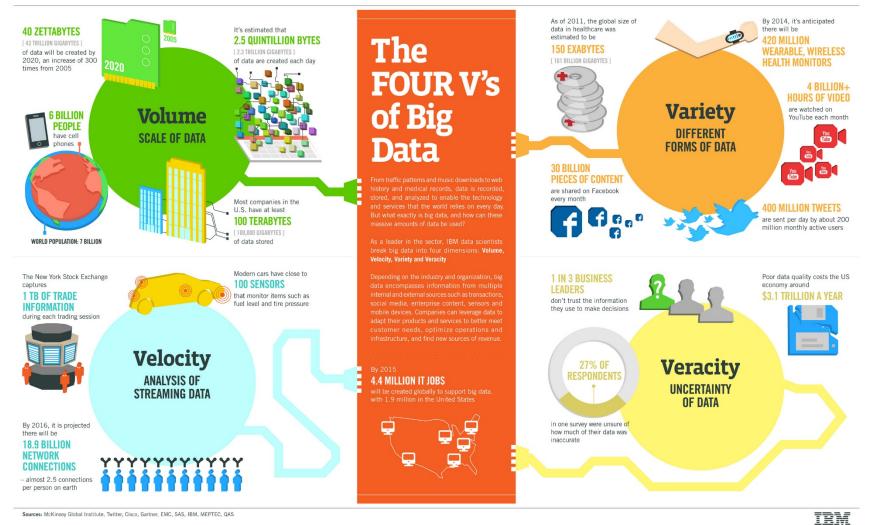
it is pointed out that 3Vs are insufficient.

- In addition to volume, velocity, and variety, further 7 Vs are identified:
- veracity, validity, value, variability, venue, vocabulary, and vagueness



Four Vs of Big Data





Sources: McKinsey Global Institute, Twitter, Cisco, Gartner, EMC, SAS, IBM, MEPTEC, QAS

Big Data Management and Analytics



- Intro
- What is Big Data?
- NoSQL Systems
- Hadoop / HDFS / MapReduce & Applications
- Spark
- Data Streams & Applications Storm, ...
- Text Data
- High-Dimensional Data
- Graph Data
- Uncertain Data



Volume

Velocity

Variety

Veracity







- Mega-trend: All data is digital, digitally born!
 - 70 years ago: computers for "+"
 - 15 years ago: disks cheaper than paper
 - 7 years ago: Internet has eyes and ears
- Because we can
 - 40 years of databases -> volume
 - 40 years of Moore's law -> complexity
 - 2000+ years of statistics -> it is only counting
 - enough optimisms that we get the rest done, too
- Because we reached dead end with logic (?)



Really?



- Yes!
 - all data is digitally born
 - storage capacity is increasing
 - counting is embarrassingly parallel



- Yes!
 - all data is digitally born
 - storage capacity is increasing
 - counting is embarrassingly parallel
- But,
 - data grows faster than energy on chip
 - value / cost tradeoff unknown
 - ownership of data unclear (aggregate vs. individual)

Really?





- a number of buzz words, some cool examples
 - you should survive any discussion with your boss
- motivation to come back next week
 - learn some of the technologies