

# Big Data

A general approach to process external multimedia datasets

David Mera

Laboratory of Data Intensive Systems and Applications (DISA)  
Masaryk University  
Brno, Czech Republic

7/10/2014

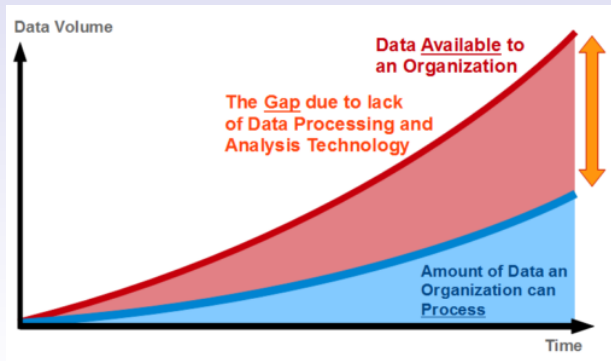
# Table of Contents

# Table of Contents

# Introduction

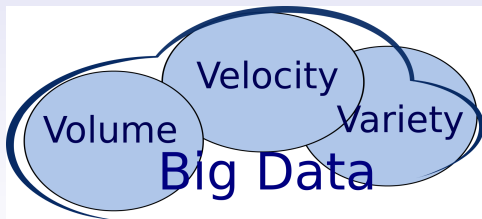
## Big Data

- Huge new datasets are constantly created.
- “90% of the data in the world today has been created in the last two years”, 2013 <sup>1</sup>
- Organizations have potential access to a wealth of information, but they do not know how to get value out of it



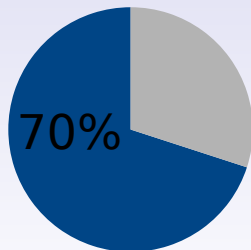
<sup>1</sup>Source: SINTEF. “Big Data - for better or worse”

- Big Data phenomenon
  - **Volume** refers to the vast amount of data generated every second
  - **Variety** refers to the different forms of data
  - **Velocity** refers to the speed at which new data are generated
  - Veracity refers to the reliability of the data
  - Value

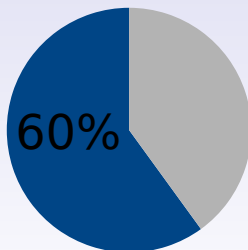


### ■ Multimedia Big Data

- 100 hours of video are uploaded to YouTube every minute
- 350 millions of photos are uploaded every day to Facebook (2012)
- Each day, 60 million photos are uploaded on Instagram
- ...



Non-Structured Data



Internet Traffic<sup>2</sup>

- Getting information from large volumes of multimedia data
  - Content-based retrieval techniques
  - Findability problem
    - Extraction of suitable features → Time-consuming task
- Feature extraction approaches
  - Sequential approach → not affordable
  - Distributed computing: Cluster computing, Grid computing
    - High computer skills
    - 'Ad-hoc' approaches → Low reusability.
    - Lack of handling failures
  - Distributed computing: Big data approaches
    - Batch data: Map-Reduce paradigm (Apache Hadoop)
    - Stream data: S4, Apache Storm.

# Table of Contents

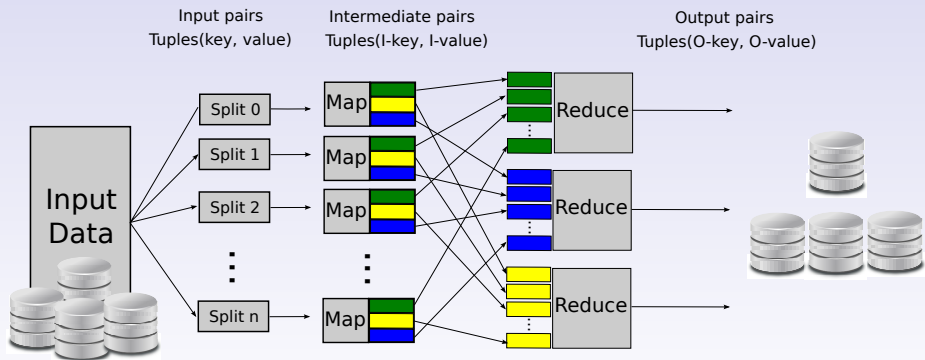


- Apache Hadoop characteristics (Map-Reduce paradigm)
  - Batch data processing system
  - Commodity computing
  - No specialized distributed-computing skills are required
  - Machine communication
  - Task scheduling
  - Scalability
  - Handling failures
  - Automatic partition of the input data

# Big Data processing frameworks

Hadoop

## Map-Reduce paradigm



- Weaknesses and limitations
  - Large files optimization
  - Batch data processing
  - Response time
  - Hard configuration process - iterative optimization
  - Lack of real-time processing
  - The parallelization level cannot be altered in running time

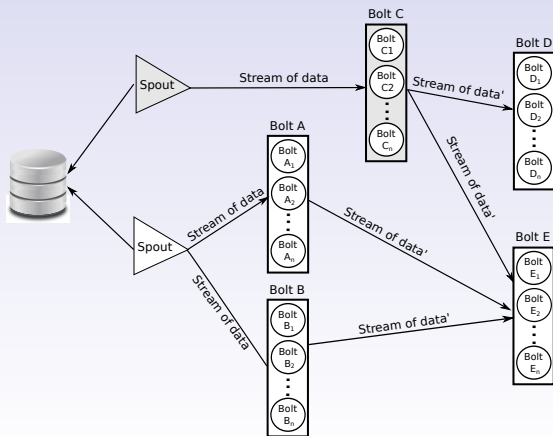
- Apache Storm characteristics
  - **Real-time processing system**
  - Commodity computing
  - No specialized distributed-computing skills are required
  - **Set of generic tools to build distributed graphs of computation**
  - Machine communication
  - Task scheduling
  - Scalability
  - Handling failures
  - **The parallelization can be adapted in processing time**

# Big Data processing frameworks

Apache Storm

## ■ Storm runs topologies

- Streams: unbounded sequence of tuples
- Spouts: source of streams
- Bolts: input streams → some processing → new streams



- Weaknesses and limitations
  - Lack of support for processing batch data
  - low-level framework
  - Pull mode
  - Specific scenario configurations

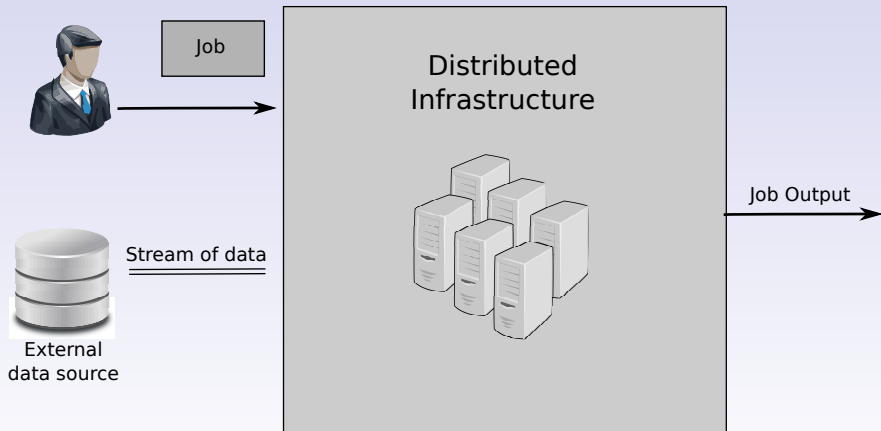
# Table of Contents

- Prototype goals
  - Efficient processing of huge external datasets
  - Heterogeneous data management
  - Processing of arbitrary functions
  - Infrastructure flexibility
  - Handling failures



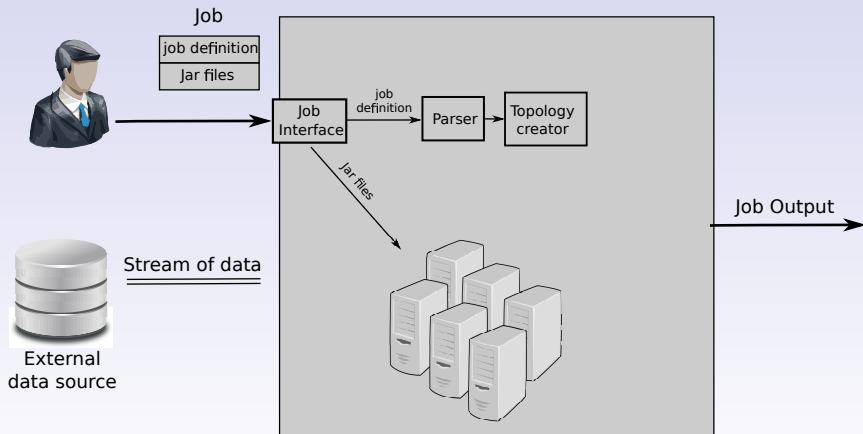
# Prototype

General overview



# Prototype

General overview



# Prototype

## Job definition

```
<job>
  <name>...</name>
  <datasource>...</datasource>

  <data save="bool">
    <operators>
      <operator>*
        <class>
          <name>...</name>
          <method>...</method>
        </class>
        <data save="bool">
          <operators>...</operators>
        </data>
      </operator>
    </operators>
  </data>
</Job>
```

# Prototype

## Job definition

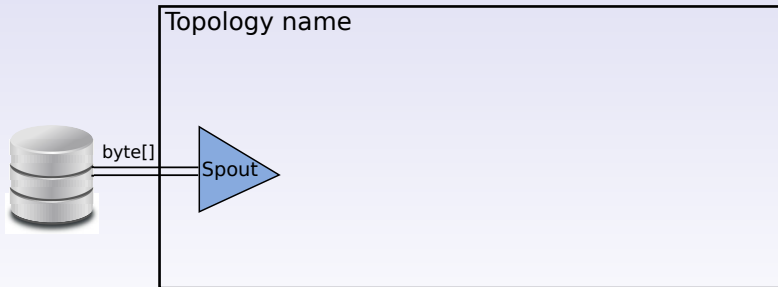
```
<job>  
  <name>...</name> -----> Topology name  
  <datasource>...</datasource>  
  
  <data save="bool">  
    <operators>  
      <operator>*  
        <class>  
          <name>...</name>  
          <method>...</method>  
        </class>  
  
        <data save="bool">  
          <operators>...</operators>  
        </data>  
      </operator>  
    </operators>  
  </data>  
</job>
```

# Prototype

## Job definition

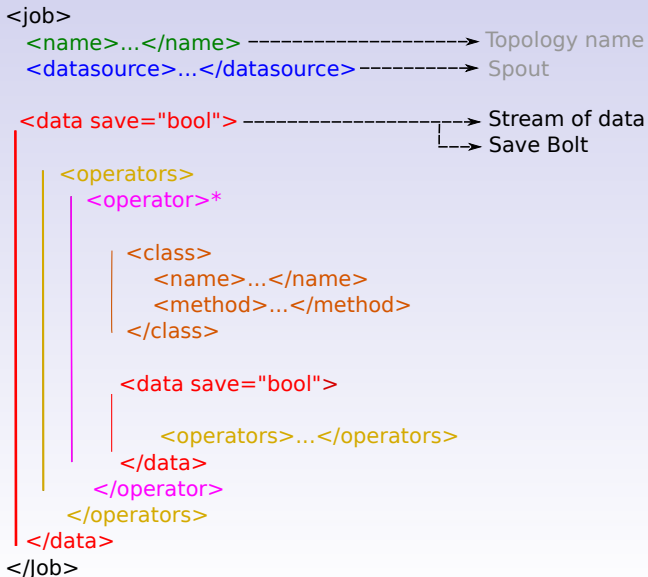
```
<job>  
  <name>...</name> -----> Topology name  
  <datasource>...</datasource> -----> Spout  
  
  <data save="bool">  
    <operators>  
      <operator>*  
        <class>  
          <name>...</name>  
          <method>...</method>  
        </class>  
  
        <data save="bool">  
          <operators>...</operators>  
        </data>  
      </operator>  
    </operators>  
  </data>  
</job>
```

- Spouts
  - Socket
  - Apache Kafka
    - Distributed messaging system



# Prototype

## Job definition



# Prototype

## Job definition

```
<job>
  <name>...</name> -----> Topology name
  <datasource>...</datasource> -----> Spout

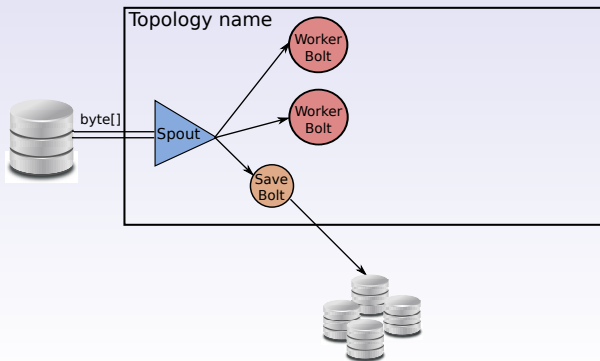
  <data save="bool"> -----> Stream of data
  |                               |-----> Save Bolt
  |
  |   <operators> -----> Stream processing
  |   |
  |   |   <operator>* -----> Operation
  |   |   |
  |   |   |   <class>
  |   |   |   |   <name>...</name> -----> Class name (inside Jar file)
  |   |   |   |   <method>...</method> -----> public byte[] methodName(byte[])
  |   |   |   |   </class>
  |   |   |   |
  |   |   |   |   <data save="bool">
  |   |   |   |   |
  |   |   |   |   |   <operators>...</operators>
  |   |   |   |   |   </data>
  |   |   |   |   |   </operator>
  |   |   |   |   |   </operators>
  |   |   |   |   </data>
  |   |   |   </operators>
  |   |   </data>
  </job>
```

Diagram illustrating the XML structure of a Job definition with annotations:

- `<job>`: Root element.
- `<name>...</name>`: Topology name.
- `<datasource>...</datasource>`: Spout.
- `<data save="bool">`: Stream of data.
- `<operators>`: Stream processing.
- `<operator>*`: Operation.
- `<class>`:
  - `<name>...</name>`: Class name (inside Jar file).
  - `<method>...</method>`: public byte[] methodName(byte[]).
- `</data>`: End of data stream.
- `</operator>`: End of operator.
- `</operators>`: End of operators.
- `</data>`: End of data stream.
- `</job>`: End of job definition.

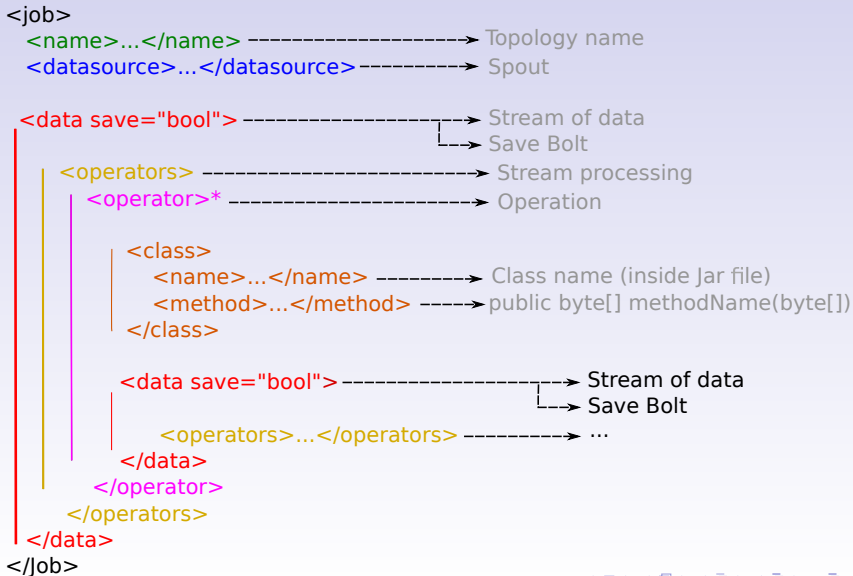


- Bolts
  - SaveBolts
    - Data storage into HDFS
    - Buffer → Hadoop SequenceFiles
  - WorkerBolt
    - Processing tuples
    - `public byte[] methodName(byte[])`



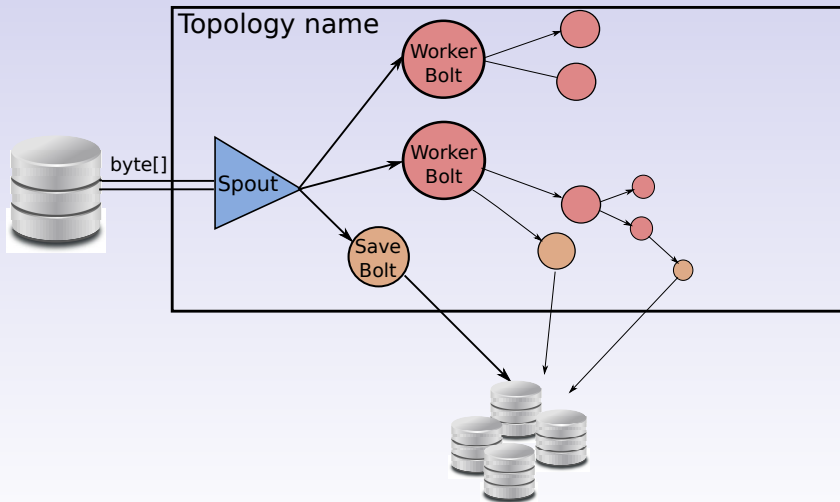
# Prototype

## Job definition



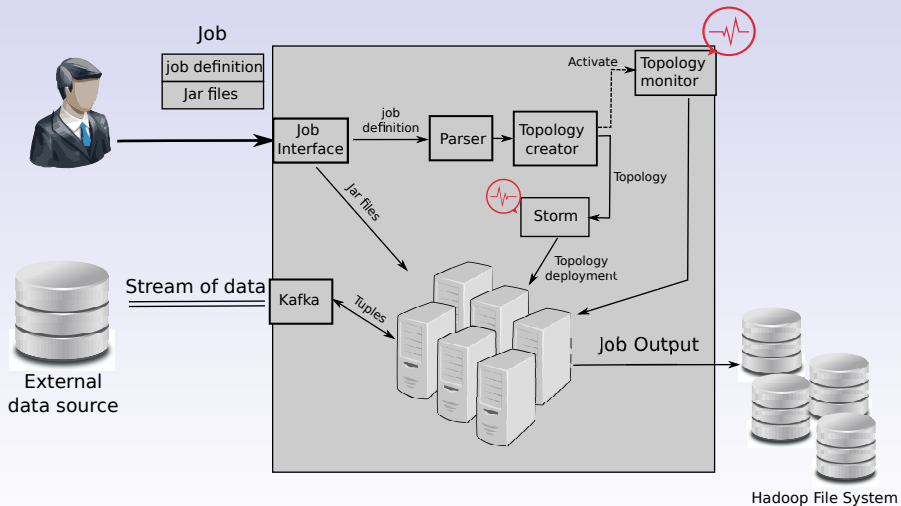
# Prototype

Job definition



# Prototype

## Job definition



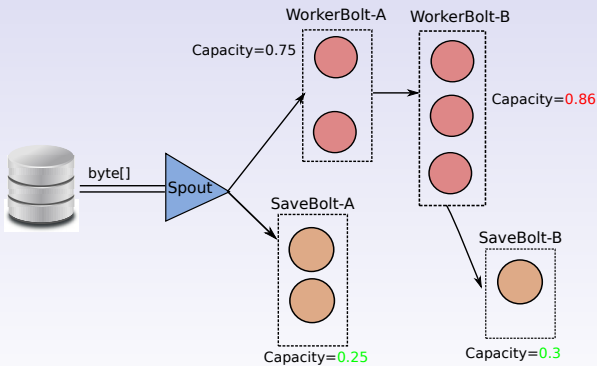
- Internal monitoring system → Max pending tuples parameter.
  - Topology starts with a low parameter value.
  - Every 'X' seconds the monitor checks the 'acked' tuples.
  - First iteration → the monitor increases the parameter value.
  - Next iterations:
    - Current 'acked' tuples > previous 'acked' tuples → Increasing parameter value.
    - Current 'acked' tuples < previous 'acked' tuples → Decreasing parameter value.
    - Current 'acked' tuples == previous 'acked' tuples → Doing nothing unless this scenario was repeated 'X' times → Increasing parameter value.

- External monitoring system
  - Administrator can add rules.  
*Rule = (metric, operator, value, action)*
  - The monitor gets topology metrics every 'X' seconds. Each bolt produces a set of metrics.
  - The monitors evaluates each rules using the bolt metrics
  - The monitor applies the rule action in every Bolt which has triggered it.

# Prototype

## Monitoring system - Example

- Rule1:(capacity,<,0.4,-1)
- Rule2:(capacity,>,0.8,+2)



# Table of Contents



## ■ Goals

- Efficient processing of huge external datasets
- Heterogeneous data management
- Processing of arbitrary functions
- Infrastructure flexibility
- Handling failures
- **Data relations management**
- **Efficient processing of huge internal datasets**

# Big Data

A general approach to process external multimedia datasets

Thank you for your attention!