## Spark Streaming

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### Presentation Plan

Spark

Spark Streaming

3 Example 1: Stream processing task

4 Example 2: Twitter

## What is Apache Spark™?

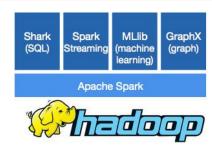
### Apache Spark™

- distributed computations system
- not only MapReduce applications
- supports in-memory operations (Resilient Distributed Dataset)
- may use HDFS

#### **APIs**

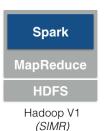
- Scala
- Java
- python

## Spark architecture









# Why do we want Spark Streaming?

Many applications require fast processing of lots of live data, for example:

- Fraud detection
- Financial market analysis
- On-line surveillance
- Early earthquakes detection

We need a system, which:

- Scales to hundreds of nodes
- Achieves second-scale latencies
- Recovers from node failures
- Integrates batch and online processing

State of the art

In 2012, existing frameworks couldn't do both:

- low-latency stream processing (ex. Storm)
- high latency batch processing (ex. Hadoop)

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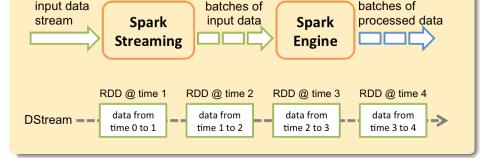
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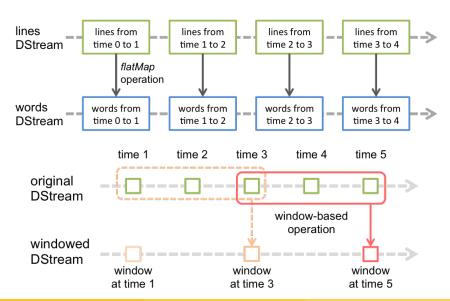
# What is Spark Streaming?

### Spark Streaming

- subproject of Apache Spark™
- allows for real-time distributed stream processing
- utilizes an idea called Discretized Stream (DStream)
- no python API, only Java and Scala



# What is Spark Streaming?



## First example I

### Data producer

Generates next integer every 100ms

### Data analyser

Counts all distinct numbers

## Data analyser code I

### Init spark stream

```
SparkConf sparkConf = new SparkConf().setAppName("Socket");

JavaStreamingContext ssc =
    new JavaStreamingContext(sparkConf, new Duration(1000));

JavaReceiverInputDStream < String > lines =
    ssc.socketTextStream(host, port));
```

### Data analyser code II

#### Transform stream

```
JavaDStream < String > words = lines
    .flatMap(line -> Lists.newArrayList("□".split(line)))
    .map(word -> Long.valueOf(word))
    .mapToPair(number -> {
        Set < Long > set = new HashSet < > ();
        set.add(number);
        return new Tuple2 < String , Set < Long >> ("distinct", set);
    })
    .updateStateByKey((values, state) -> {
        Set < Long > result = state.or(new HashSet < Long > ());
        for (Set < Long > set : values)
             result.addAll(set);
        return Optional.of(result);
    });
```

## Data analyser code III

#### Print stream

```
stream.foreachRDD(rdd -> {
    if (rdd.count() != 1) {
        System.out.println("Empty_RDD");
        return null;
    }
    List<Tuple2<String, Set<Long>>> collect = rdd.collect();
    Set<Long> set = collect.get(0)._2;
    System.out.println("Distinct:_\_" + set.size());
    return null;
});
```

## How to run Spark? I

### Running master on Linux

- Run sbin/start-master.sh
- Check in browser if http://localhost:8080 is available

### Running worker on Linux

- Get precompiled Spark 1.1.0 for Hadoop 1.x from /home/2012/m.okulewicz/spark and unpack it
- If necessary edit: conf/spark-env.sh and add location of JAVA\_HOME
- Run sbin/start-slave.sh 1 spark://phd01.phd.ipipan.waw.pl
- Check in browser if http://localhost:8081 is available and master points to phd01.phd.ipipan.waw.pl

# How to run Spark? II

### Running task on Linux

- Run:
  - ./bin/spark-submit
  - --class
  - pl.waw.ipipan.phd.mkopec.sparkReceiver.SocketReceiver
  - --master spark://phd01.phd.ipipan.waw.pl:7077
  - --executor-memory 20G
  - --total-executor-cores 100
  - /path/to/jar.jar localhost 9999 1000 1

# Example stream processing task

#### Data producer

Retrieves stream of tweets containing specified keywords

### Data analyser

Counts most frequent words in tweets

## Data analyser code I

### Init spark stream

## Data analyser code II

#### Transform stream

```
JavaPairDStream < String, Long > counts = twitterStream
    .flatMap(tweet -> Arrays.asList("u".split(tweet.getText())
    .filter(word -> {
         sleep(1000);
         return !StringUtils.isBlank(word);
    })
    .mapToPair(word -> new Tuple2<String, Long>(word, 1L))
    .reduceBvKev((x, y) \rightarrow x + y)
    .updateStateByKey((values, state) -> {
         Long total = state.or(OL);
         for (Long v : values)
             total += v;
         return Optional.of(total);
    });
```

## Data analyser code III

#### Print stream

```
stream.foreachRDD(rdd -> {
    if (rdd.count() == 0) {
        System.out.println("Empty RDD");
        return null;
    List < Tuple 2 < String, Long >> list = rdd.collect();
    Collections.sort(list, ...);
    System.out.println("|Top||10||words");
    for (Tuple2<String, Long> tuple :
        list.subList(0, Math.min(list.size(), 10)))
        System.out.println(tuple._1 + "\t" + tuple._2);
    return null;
});
```

### How to run this task?

### Running task on Linux

```
• Run:
```

```
./bin/spark-submit
```

```
--class
```

pl.waw.ipipan.phd.mkopec.sparkReceiver.TwitterReceiver

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/path/to/jar.jar localhost 1000 1 Polska, Poland

## Bibliography I



https://spark.apache.org/docs/0.9.0/streaming-programming guide.html. Streaming Programming Guide, 2014.



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Discretized Streams: Fault-tolerant Streaming Computation at Scale.

In Proceedings of the Twenty-Fourth ACM Symposium on Operating Systems Principles, SOSP '13, pages 423–438, New York, NY, USA, 2013. ACM.



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