# Data Analysis With Apache Flink



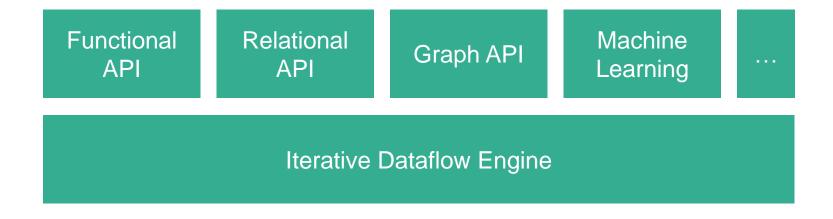
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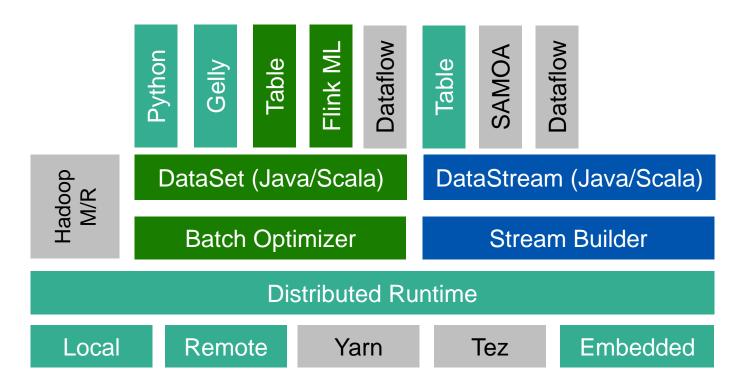
## What is Apache Flink?





## Apache Flink Stack





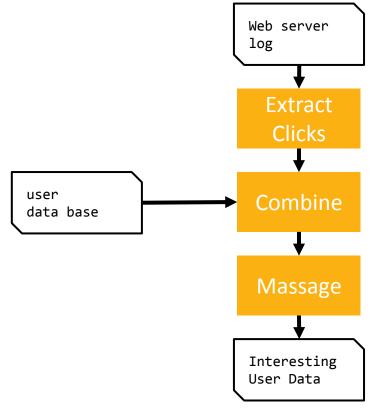
<sup>\*</sup>current Flink master + few PRs

## **Example Use Case: Log Analysis**

#### What Seems to be the Problem?



- Collect clicks from a webserver log
- Find interesting URLs
- Combine with user data



#### The Execution Environment



- Entry point for all Flink programs
- Creates DataSets from data sources

ExecutionEnvironment env = ExecutionEnvironment.getExecutionEnvironment();

### Getting at Those Clicks



post /foo/bar... 313 get /data/pic.jpg 128 post /bar/baz... 128 post /hello/there... 42

```
DataSet<String> log = env.readTextFile("hdfs:///log");
DataSet<Tuple2<String, Integer>> clicks = log.flatMap(
 (String line, Collector<Tuple2<String, Integer>> out) ->
  String[] parts = in.split("*magic regex*");
  if (isClick(parts)) {
   out.collect(new Tuple2<>(parts[1],Integer.parseInt(parts[2])));
```

#### The Table Environment



- Environment for dealing with Tables
- Converts between DataSet and Table

TableEnvironment tableEnv = new TableEnvironment();

## Counting those Clicks



```
Table clicksTable = tableEnv.toTable(clicks, "url, userId");

Table urlClickCounts = clicksTable
    .groupBy("url, userId")
    .select("url, userId, url.count as count");
```

## Getting the User Information



```
Table userInfo = tableEnv.toTable(..., "name, id, ...");

Table resultTable = urlClickCounts.join(userInfo)
.where("userId = id && count > 10")
.select("url, count, name, ...");
```

## The Final Step

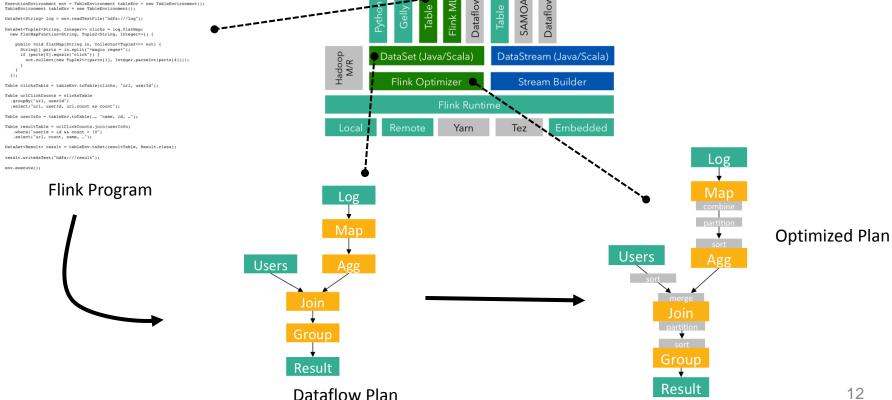


```
class Result {
 public String url;
 public int count;
 public String name;
DataSet<Result> set =
 tableEnv.toSet(resultTable, Result.class);
DataSet<Result> result =
 set.groupBy("url").reduceGroup(new ComplexOperation());
result.writeAsText("hdfs:///result");
env.execute();
```

## What happens under the hood?

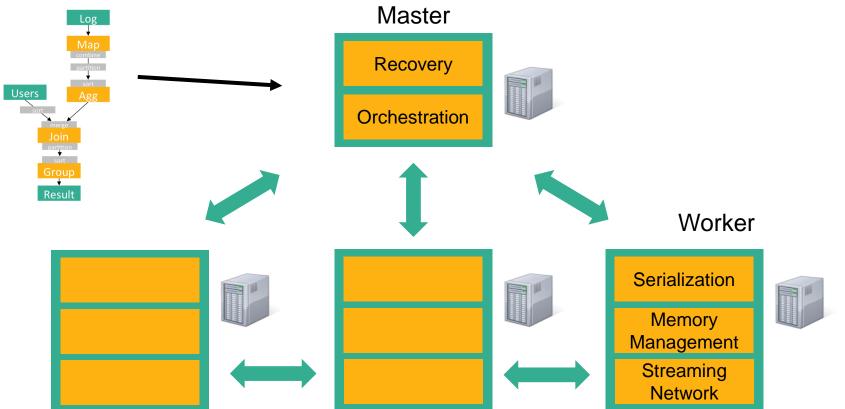
## From Program to Dataflow





#### Distributed Execution



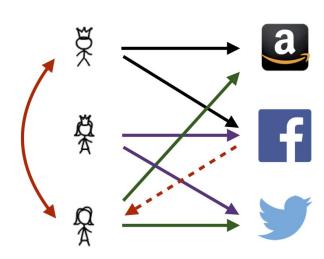


## Advanced Analysis: Website Recommendation

## Going Further



- Log analysis result:
   Which user visited how often which web site
- Which other websites might they like?
- Recommendation by collaborative filtering



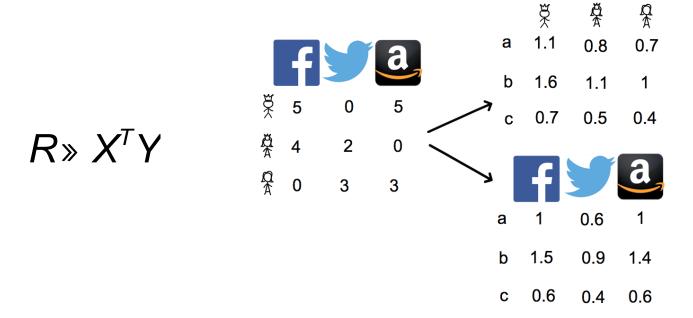
## Collaborative Filtering



- Recommend items based on users with similar preferences
- Latent factor models capture underlying characteristics of items and preferences of user
- Predicted preference:  $\hat{r}_{u,i} = \mathbf{x}_u^T \mathbf{y}_i$

#### Matrix Factorization



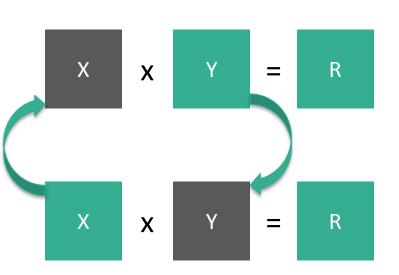


$$\min_{X,Y} \underset{r_{u,i} \to 0}{\overset{\circ}{\bigcirc}} \left( r_{u,i} - \mathbf{x}_{u}^{\mathsf{T}} \mathbf{y}_{i} \right)^{2} + / \underset{\overset{\circ}{\bigcirc}}{\overset{\circ}{\bigcirc}} \underset{u}{\overset{\circ}{\bigcirc}} n_{u} \| \mathbf{x}_{u} \|^{2} + \underset{i}{\overset{\circ}{\bigcirc}} n_{i} \| \mathbf{y}_{i} \|^{2} \underset{\overset{\circ}{\bigcirc}}{\overset{\circ}{\bigcirc}}$$

## Alternating least squares



- Iterative approximation
  - 1. Fix X and optimize Y
  - 2. Fix Y and optimize X
- Communication and computation intensive



## Matrix Factorization Pipeline

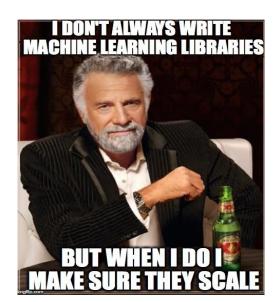


```
Clickstream
Data

Hashing
Feature
Extractor

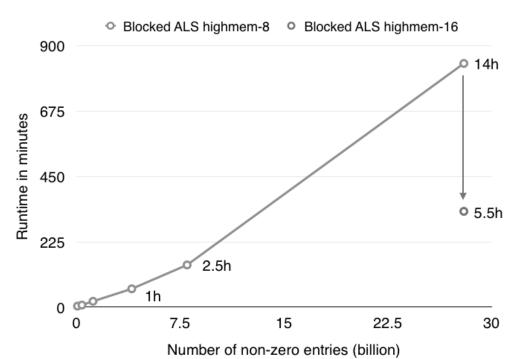
Hashing
Feature
Extractor
```

```
val featureExtractor = HashingFT()
val factorizer = ALS()
val pipeline = featureExtractor.chain(factorizer)
val clickstreamDS =
  env.readCsvFile[(String, String, Int)](clickStreamData)
val parameters = ParameterMap()
  .add(HashingFT.NumFeatures, 1000000)
  .add(ALS.Iterations, 10)
  .add(ALS.NumFactors, 50)
  .add(ALS.Lambda, 1.5)
val factorization = pipeline.fit(clickstreamDS, parameters)
```



#### Does it Scale?





Scale of Netflix or Spotify

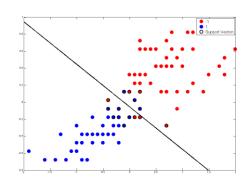


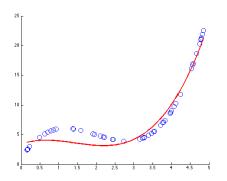
- 40 node GCE cluster, highmem-8
- 10 ALS iteration with 50 latent factors
- Based on Spark MLlib's implementation

#### What Else Can You Do?



- Classification using SVMs
  - Conversion goal prediction
- Clustering
  - Visitor segmentation
- Multiple linear regression
  - Visitor prediction





## Closing

#### What Have You Seen?



- Flink is a general-purpose analytics system
- Highly expressive Table API
- Advanced analysis with Flink's machine learning library
- Jobs are executed on powerful distributed dataflow engine

## Flink Roadmap for 2015



- Additions to Machine Learning library
- Streaming Machine Learning
- Support for interactive programs
- Optimization for Table API queries
- SQL on top of Table API



**– 2015 —** 

## Flink Forward

**BERLIN 12/13 OCT** •



flink.apache.org @ApacheFlink

## **Backup Slides**

#### WordCount in DataSet API



```
case class Word (word: String, frequency: Int)
val env = ExecutionEnvironment.getExecutionEnvironment()
val lines = env.readTextFile(...)
lines
   .flatMap {line => line.split(" ").map(word => Word(word,1))}
   .groupBy("word").sum("frequency")
   .print()
env.execute()
```

Java and Scala APIs offer the same functionality.

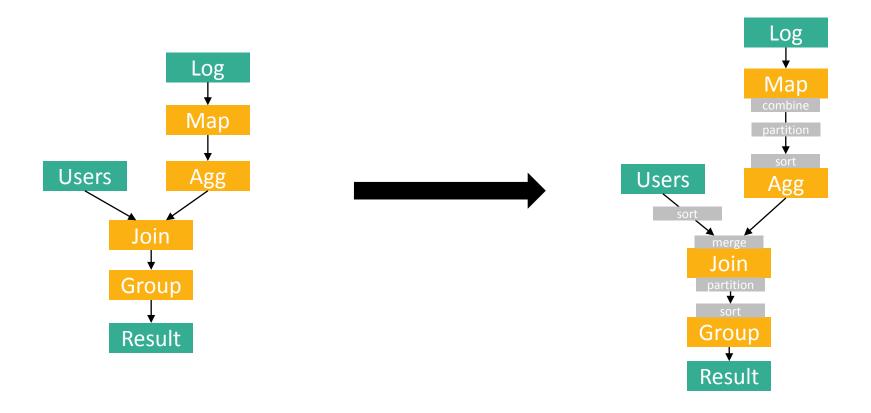
## Log Analysis Code



```
ExecutionEnvironment env = TableEnvironment tableEnv = new TableEnvironment();
TableEnvironment tableEnv = new TableEnvironment();
DataSet<String> log = env.readTextFile("hdfs:///log");
DataSet<Tuple2<String, Integer>> clicks = log.flatMap(
 new FlatMapFunction<String, Tuple2<String, Integer>>() {
  public void flatMap(String in, Collector<Tuple2<>> out) {
    String[] parts = in.split("*magic regex*");
    if (parts[0].equals("click")) {
     out.collect(new Tuple2<>(parts[1], Integer.parseInt(parts[4])));
 });
Table clicksTable = tableEnv.toTable(clicks, "url, userId");
Table urlClickCounts = clicksTable
 .groupBy("url, userId")
 .select("url, userld, url.count as count");
Table userInfo = tableEnv.toTable(..., "name, id, ...");
Table resultTable = urlClickCounts.join(userInfo)
 .where("userId = id && count > 10")
 .select("url, count, name, ...");
DataSet<Result> result = tableEnv.toSet(resultTable, Result.class);
result.writeAsText("hdfs:///result");
env.execute();
```

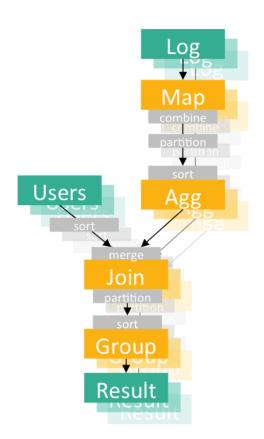
## Log Analysis Dataflow Graph





## Pipelined Execution





**Note:** Intermediate DataSets are not necessarily "created"!

Only 1 Stage (depending on join strategy)



Data transfer in-memory and disk if needed

#### API in a Nutshell



- Element-wise
  - map, flatMap, filter
- Group-wise
  - groupBy, reduce, reduceGroup, combineGroup, mapPartition, aggregate, distinct
- Binary
  - join, coGroup, union, cross
- Iterations
  - iterate, iterateDelta
- Physical re-organization
  - rebalance, partitionByHash, sortPartition
- Streaming
  - window, windowMap, coMap, ...