Welcome to the first Workshop on Big data Open Source Systems (BOSS)

September 4th, 2015 Co-located with VLDB 2015

Tilmann Rabl

Hands on Big Data

- 8 parallel tutorials
- 8 systems
 - Open source
 - Publicly available
- Presenters
 - System experts
- Hands on
 - This is not a demo!
- You can pick two!

But why?

- Initial idea: Malu Castellanos
- Mike Carey
 - Doing It On Big Data: a Tutorial/Workshop
 - Driving force
- Other people involved
 - Volker Markl
 - Norman Patton
 - Lipyeow Lim
 - Kerstin Forster
- Experiment
 - Tell us what you think
 - Email: rabl@tu-berlin.de



Presented Systems

- Apache AsterixDB
 Asteri
- Apache Flink



• Apache Reef



• Apache Singa



• Apache Spark



Padres



rasdaman

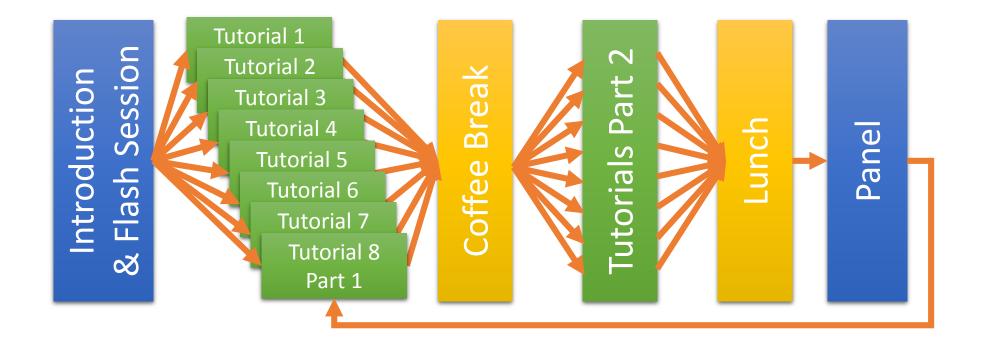


• SciDB

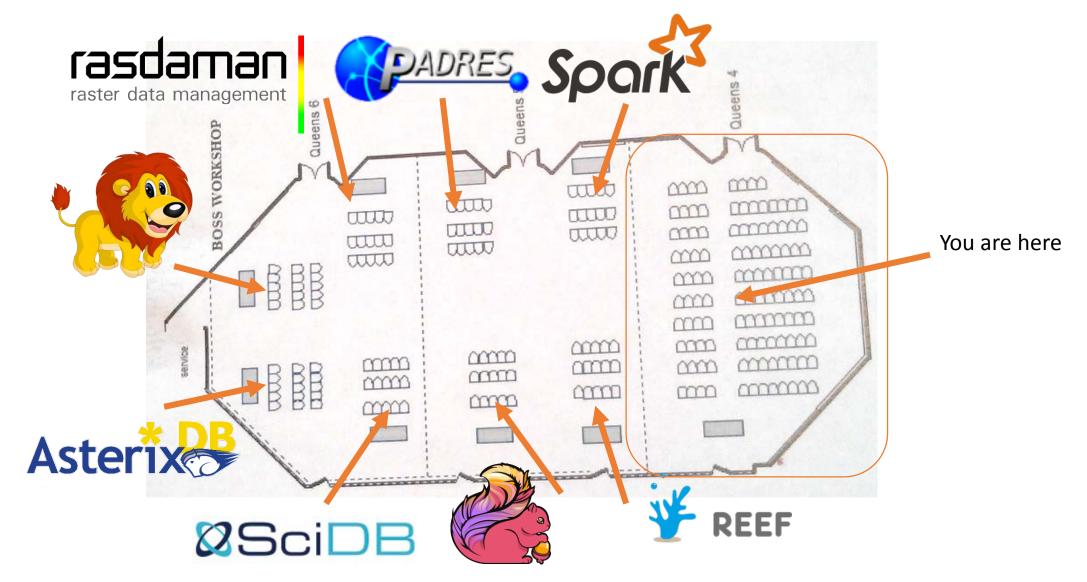


Massively Parallel Program

• Bulk Synchronous Parallel



Runtime Environment



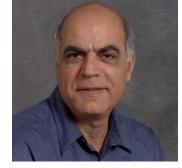
Panel – Big Data and Exascale

- Panel Chair
 - Chaitanya Baru, San Diego Supercomputing Center
- Panelists
 - Arie Shoshani, LBNL
 - Guy Lohmann, IBM
 - Mike Carey, UC Irvine
 - Paul G. Brown, Paradigm4
 - Peter Baumann, Jacobs University
 - Volker Markl, TU Berlin













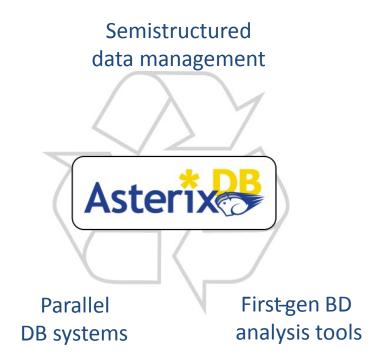


Apache AsterixDB (Incubating) Asterix

AsterixDB: "One Size Fits a Bunch!"

Wish-list:

- Able to manage data
- Flexible data model
- Full query capability
- Continuous data ingestion
- Efficient and robust parallel runtime
- Cost proportional to task at hand
- Support today's "Big Data data types"



Apache Flink

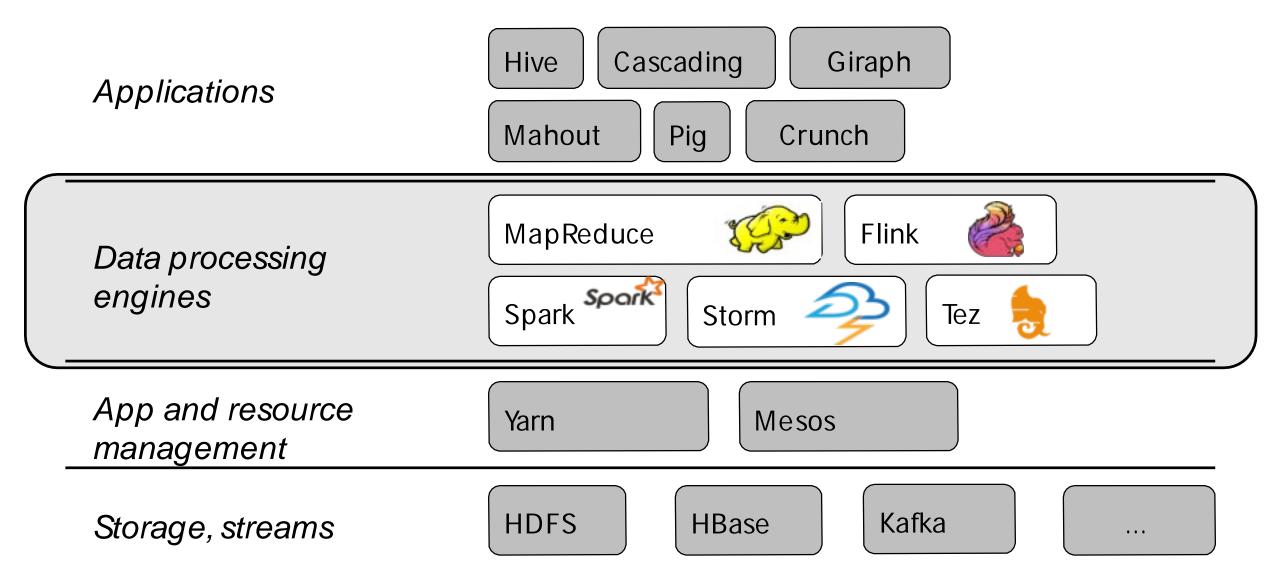




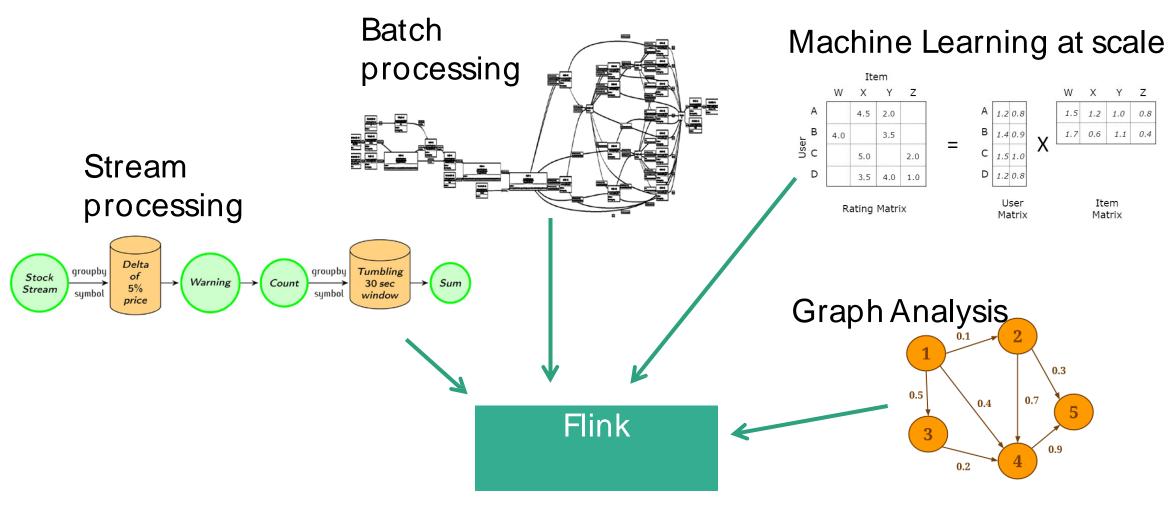
Apache Flink[™]: Stream and Batch processing at Scale

Marton Balassi(ELTE/SZTAKI, Hungary)Paris Carbone(KTH, Stockholm, Sweden)Gyula For a(SICS, Stockholm, Sweden)Vasia Kalavri(KTH, Stockholm, Sweden)Asterios Katsifodimos (TU Berlin, Germany)

What is Flink?



What can I do with Flink?



An engine that can **natively** support all these workloads.

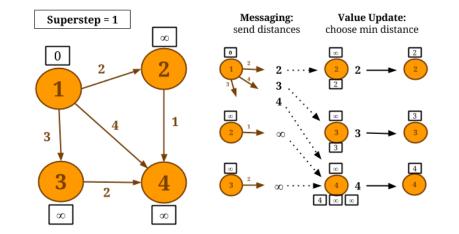
But what will I do with Flink today?

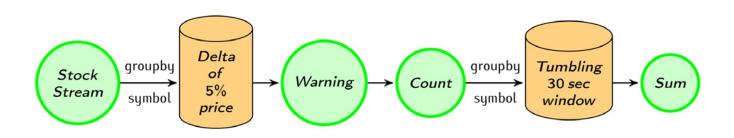
• Graph processing

- ETL on Datasets
- Graph creation & analysis

Stream Processing

- Rolling Aggregates
- Windows & Alerts





Agenda

Introduction

- 15' Overview
- 15' Gelly (Graph) API
- 30' Break

Graph Processing

• 20' DataSet/Gelly Hands-on

• Stream processing with Flink

- 10' DataStream API
- 15' Fault Tolerance Demo
- 45' Streaming Hands-on

Apache Reef





Deep Dive into Apache REEF (Incubating) **BOSS 2015**

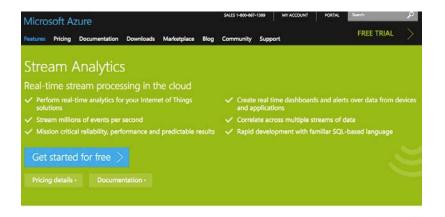
Byung-Gon Chun, Brian Cho (Seoul National University)

Sep. 4, 2015

A meta-framework that eases the development of Big Data applications atop resource managers such as YARN and Mesos

Interactive Query	Business Intelligence	Stream Processing	Machine Learning	Batch Processing
Data Processing Lib (REEF, Third-party)				
REEF				
Resource Manager				
Distributed File System				
 Reusable control plane for coordinating data plane tasks 				

- Adaptation layer for resource managers \checkmark
- Container and state reuse across tasks from heterogeneous frameworks
- Simple and safe configuration management
- Scalable local, remote event handling \checkmark
- Java and C# (.NET) support



Uncover Real-time Insights

Learn how Microsoft's stream analytics service in the cloud enables you to rapidly develop and deploy a low cost real-time analytics solution to uncover real-time insights from devices, sensors, astructure, and applications. It will enable various opportunities including Internet of Things (IoT) scenarios such as real-time remote management and monitoring or gaining insights from devices like obile phones or connected cars



In production use (Microsoft Azure)



Deep Dive into Apache REEF (Incubating) BOSS 2015 Sep. 4, 2015

Byung-Gon Chun, Brian Cho (Seoul National University)

<u>Tutorial</u>

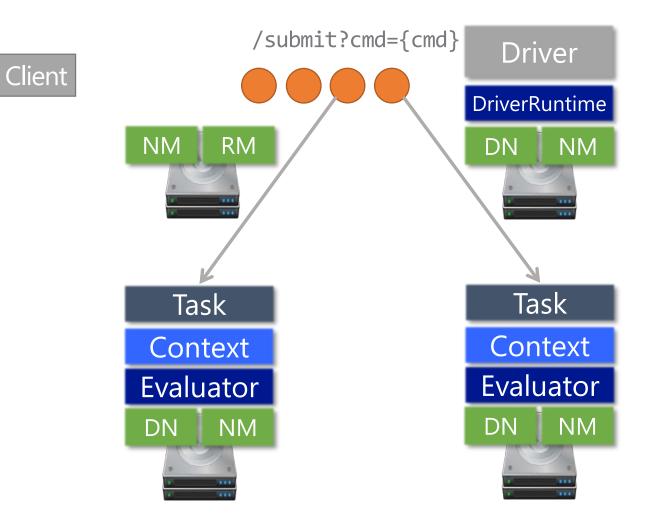
1. What is REEF?

2. Install REEF

- 3. Run your first REEF job: HelloREEF
- 4. Why would you want REEF?

5. Create your own Task Scheduler with REEF

Contact: Byung-Gon Chun bgchun@gmail.comBrian Chochobrian@gmail.com



Apache Singa





Apache SINGA



A General Distributed Deep Learning Platform

- Motivation
 - Deep learning is effective for classification tasks, e.g., image recognition
 - Training code is complex to write from the scratch
 - Training is time consuming, e.g., 10 days or weeks
- Goals
 - Easy to use
 - General to support popular deep learning models
 - Extensible for users to do customization, e.g., training new models
 - Scalable
 - Reduce training time with more computation resources, e.g. machines
 - Improve efficiency of one training iteration by synchronous training
 - Reduce total number of training iterations by asynchronous training

Apache Spark



Spark Tutorial

Reynold Xin @rxin Sep 4, 2015 @ VLDB BOSS 2015



Apache Spark

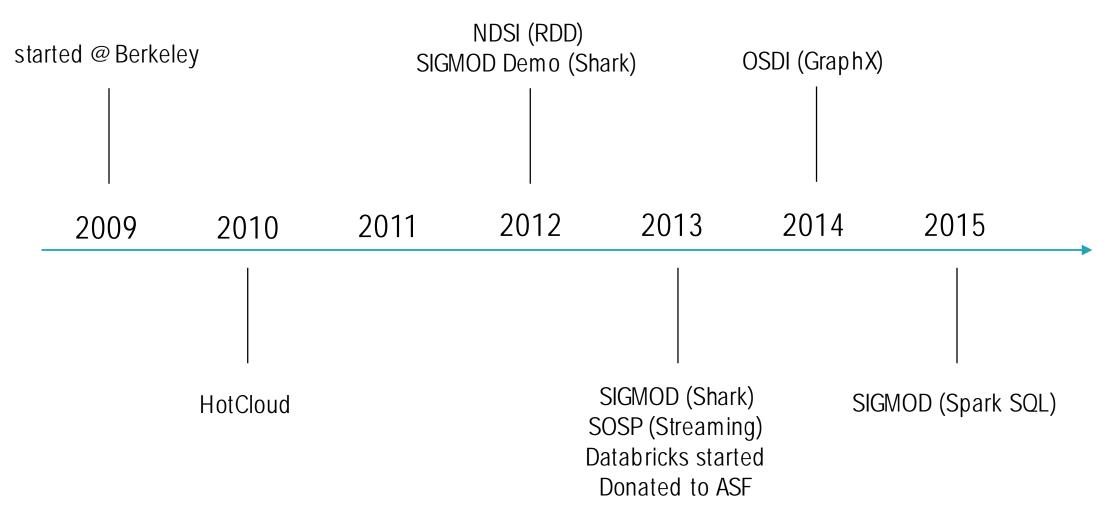
Fast & general distributed data processing engine, with APIs in SQL, Scala, Java, Python, and R

800+ contributors and many academic papers

Largest open source project in (big) data & at Apache



A Brief History







1000+ companies



Distributors + Apps

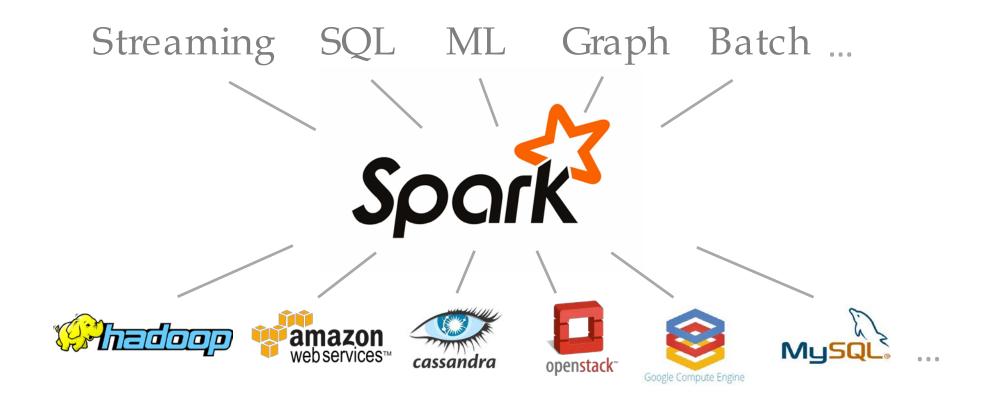
50+ companies





Our Goal for Spark

Unified engine across data workloads and platforms



databricks^{*}

Agenda Today

Spark 101: RDD Fundamentals

Spark 102: DataFrames

Spark 201: Understanding Spark Internals

(with exercises in Databricks notebooks)



PADRES

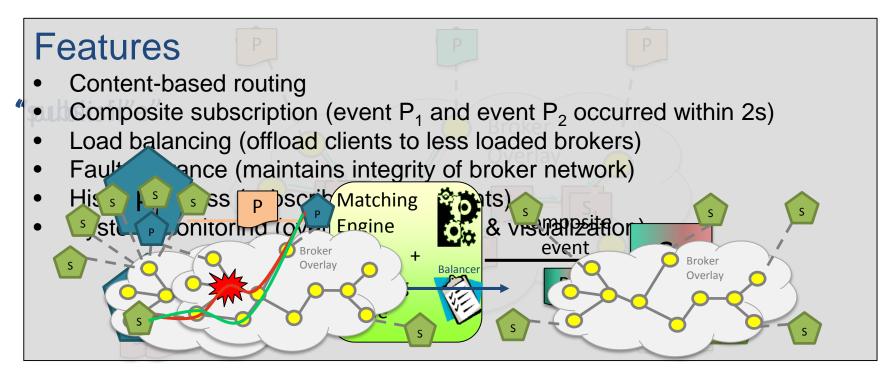




Presenter: <u>Kaiwen Zhang</u> University of Toronto

Pub/Sub is a communication paradigm / middleware

Communication between information producers (**publisher**) and consumers (**subscriber**) is mediated by a set of **brokers** (p2p overlay).



rasdaman

raster data management

rasdaman the Array Database

the pioneer Array DBMS: analytics on n-D dense/sparse arrays optimization & parallel QP on multicore, cloud, modern hw scalable from cubesat to datacenter federations seamless integration with R, python, ... operationally deployed on Petascale, basis for ISO Array SQL

www.rasdaman.org

SciDB







No cute animals ...

No 5 color marketing brochure ...

... just an ...

Open Source, Transactional, Massively Parallel, Array DBMS with A Scalable Analytic Query Engine.

