

Spark技术内幕

张安站 2015

什么是Spark?

• 分布式计算框架

● 基于内存的调度

兼容Hadoop生

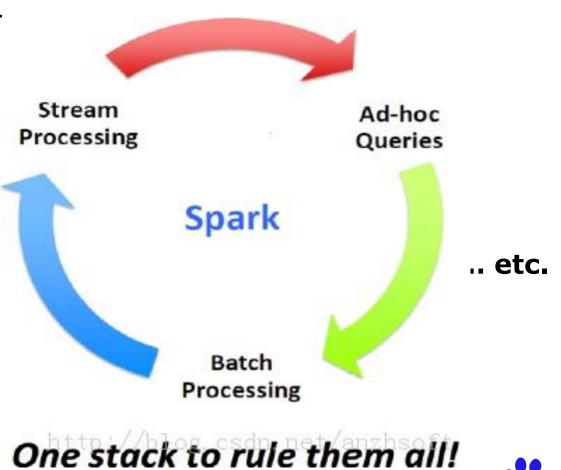
一数据存储格式

Works with

● 丰富的应用场景

- batch /strea

- SQL / 机器学



快速发展的生态环境

- 快速从Apache Incubator项目毕业成为Apache顶级项目
- 大数据解决方案提供商的支持
 - Cloudera / MapR / Hortonworks / Pivotal / SAP ...
 - 一华为 / 星环科技
- 应用
 - 一百度/阿里/腾讯/爱奇艺/优酷/京东...
- 2014年大数据领域最活跃的开源项目



核心组件

Spark SQL

Spark Streaming MLlib (machine learning) Graph) (graph)

Apache Spark



代码规模

Spark core: 64, 350 LOC

api: 5063 broadcast: 878

common: 7201 deploy: 8100

executor: 1249 metrics: 772

network: 3364 partial: 782

scheduler: 8334 serializer: 558

storage : 5251 ui: 3978

Shuffle:1356 Rdd:6405

Util:10126 Input:609

bagel: 335

mllib: 19724

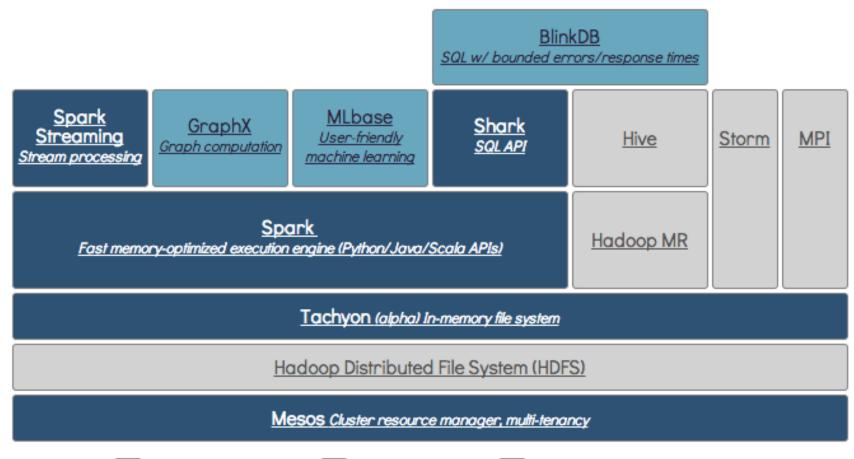
sql: 33875

examples: 7472

streaming: 11760 YARN: 3796 graphx: 5793



BDAS

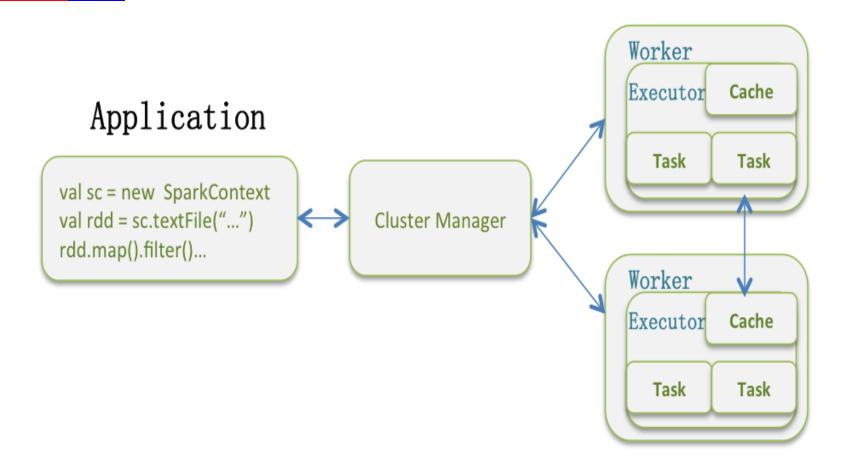


In Development

Related External Project

Supported Release

整体架构





Spark的目标

- 高效和通用的编程模型
 - 一相比MapReduce,更加适用于两种类型的应用
 - ●迭代算法类(机器学习,图计算)
 - 交互式数据挖掘
 - 一相似的编程接口
- 良好的用户体验
 - ●编程效率:基于Scala的核心模块,并提供 Java/python编程接口
 - •功能强大的API,丰富的操作算子
 - 交互式的解释执行接口(调试,学习)

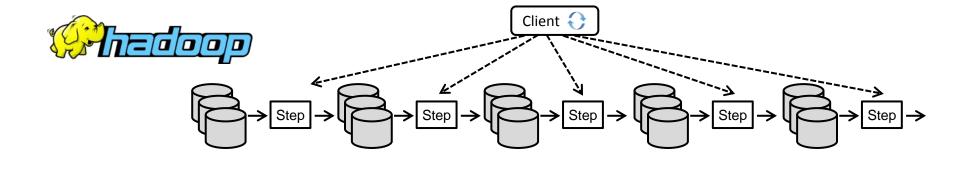


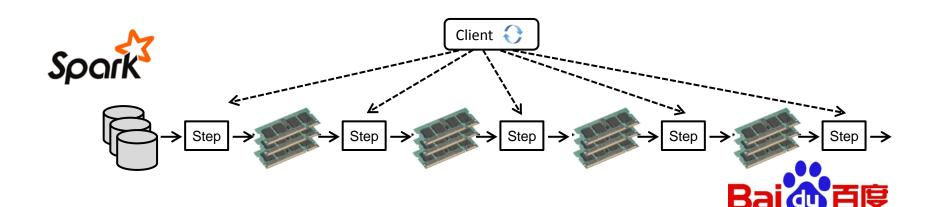
相似的编程接口

ssc.start()

```
val conf = new SparkConf()
val sc = new SparkContext(conf)
val lines = sc.textFile(args(1))
val words = lines.flatMap(_.split(" "))
val result = words.map(x => (x, 1)).reduceByKey(_ + _).collect()
```

仅仅是因为内存?





从Word Count开始

- 代码实现
- RDD
- 资源分配
- 任务划分
- 任务调度
- 任务执行
 - Shuffle的具体实现



Word Count的代码实现

```
Scala
                  Java
 Python
val file = spark.textFile("hdfs://...")
val counts = file.flatMap(line => line.split(" "))
                 .map(word => (word, 1))
                 reduceByKey(_ + _)
counts.saveAsTextFile("hdfs://...")
 Python
          Scala
                  Java
file = spark.textFile("hdfs://...")
counts = file.flatMap(lambda line: line.split(" ")) \
             .map(lambda word: (word, 1)) \
             reduceByKey(lambda a, b: a + b)
counts.saveAsTextFile("hdfs://...")
```



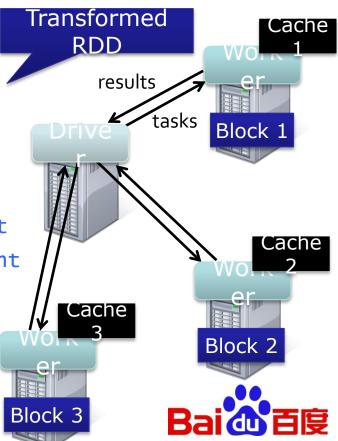
RDD

- Resilient distributed dataset, 弹性分布式数据集
- 不可变的, 按分区组织的数据对象
- 支持多种转换 + 动作
- 可以通过多种数据源创建RDD
- 缓存 + 检查点
- 容错,数据本地性,可扩展性



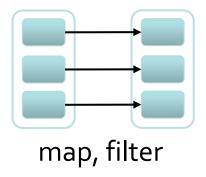
RDD的缓存

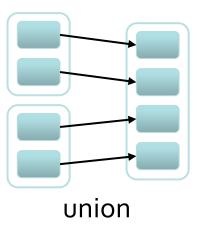
```
Base RDD
lines = spark.textFile("hdfs://...")
errors = lines.filter(_.startsWith("ERROR"))
messages = errors.map(_.split('\t')(2))
cachedMsgs = messages.cache()
                                      Action
cachedMsgs.filter(_.contains("wordseg")).count
cachedMsgs.filter(_.contains("wordrank")).count
```

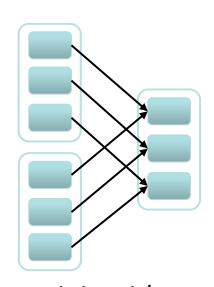


RDD的不同依赖

窄依赖:

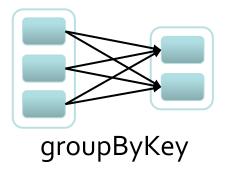


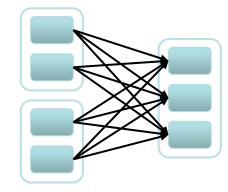




join with inputs copartitioned

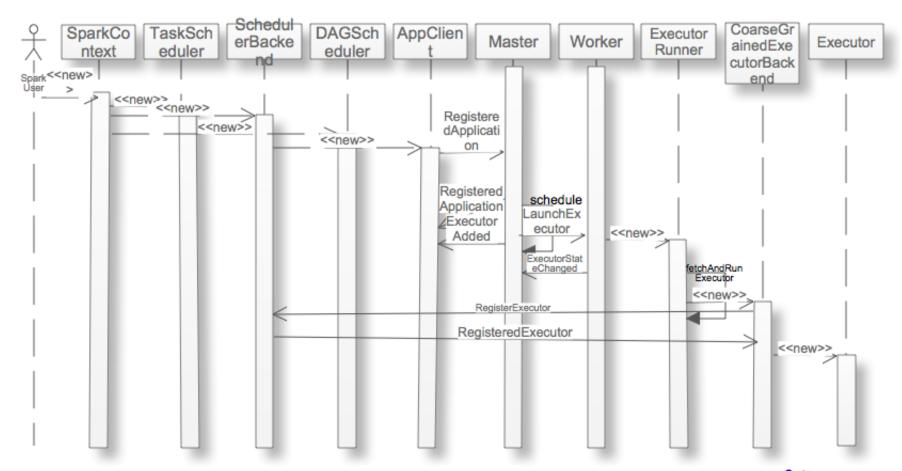
宽依赖:





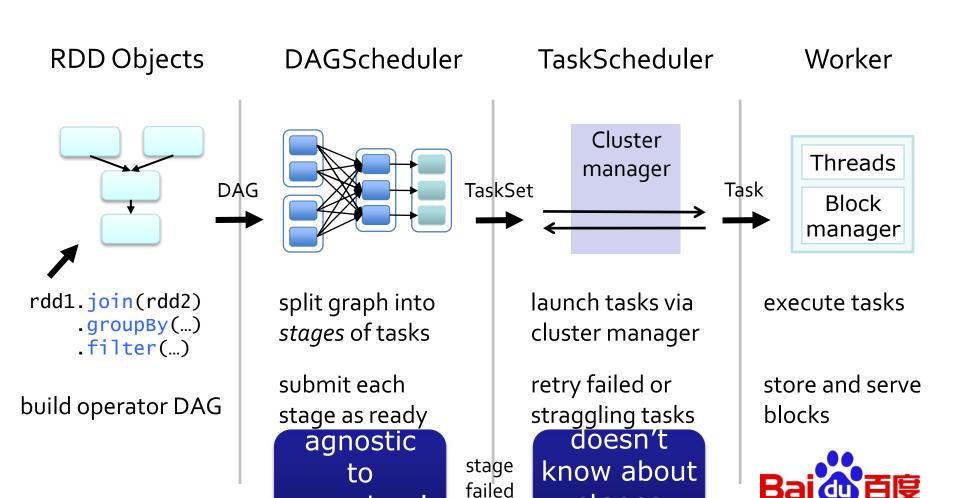
join with inputs not co-partitioned **Bai** 首度

资源的分配





任务调度

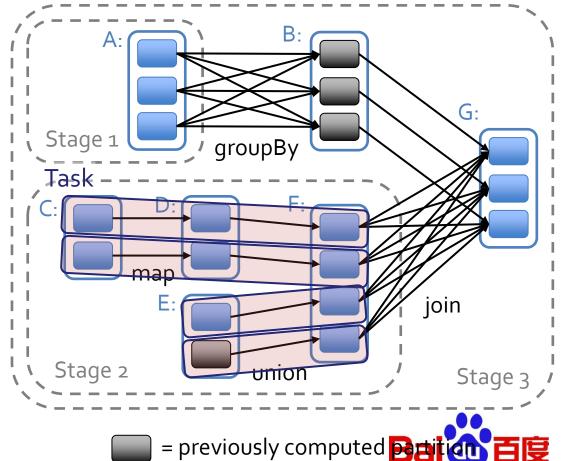


stages

operators!

DAG的划分

- 由DAGScheduler 完成DAG不同 Stage的划分
- 最大化pipeline
- 尽可能的减少 Shuffle







Task的生成

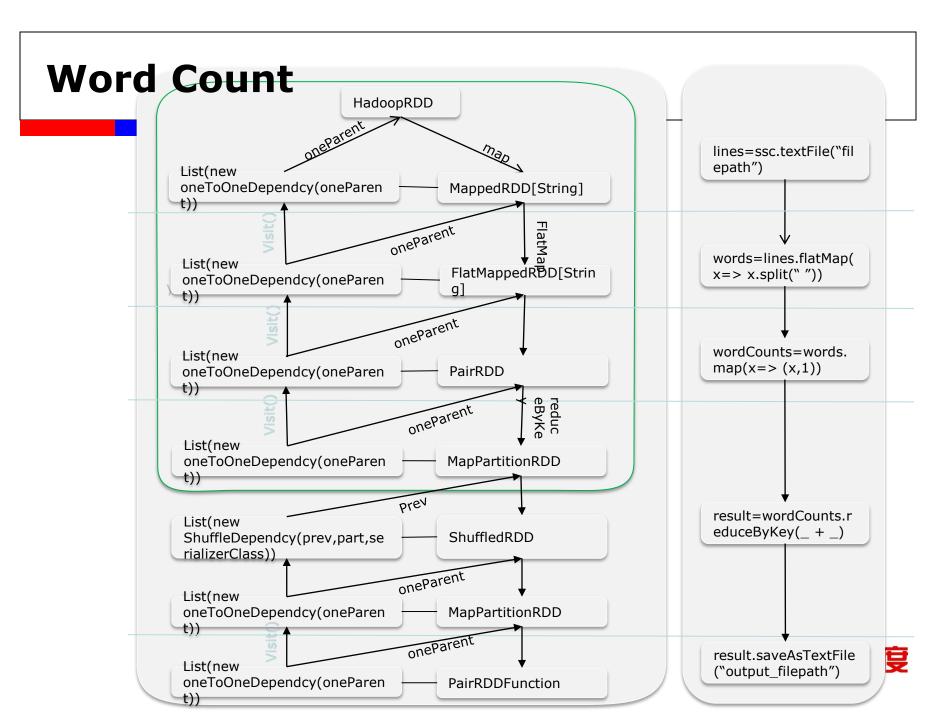
- 根据Final Stage的Partition生成ResultTask ,每个Partition对应于一个Task
- 其余的Stage的每个Partition生成一个 ShuffleMapTask



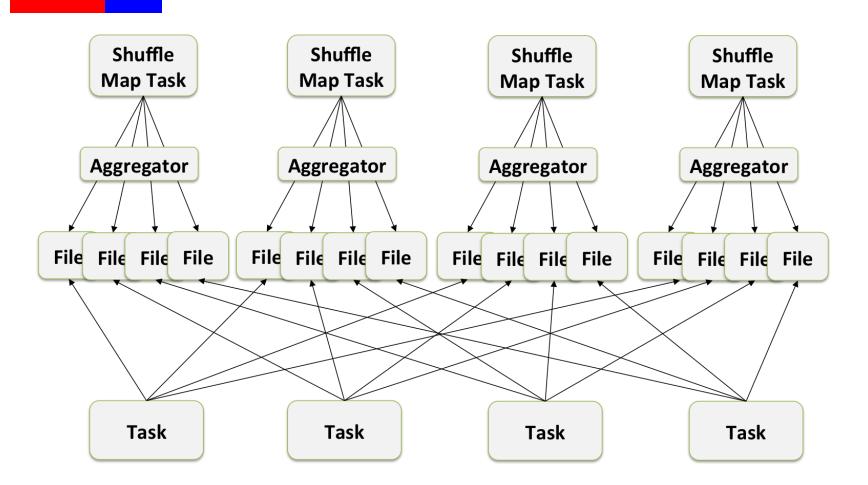
Task的执行

- TaskScheduler会将任务发送到Executor
- Executor开始每个Task的执行



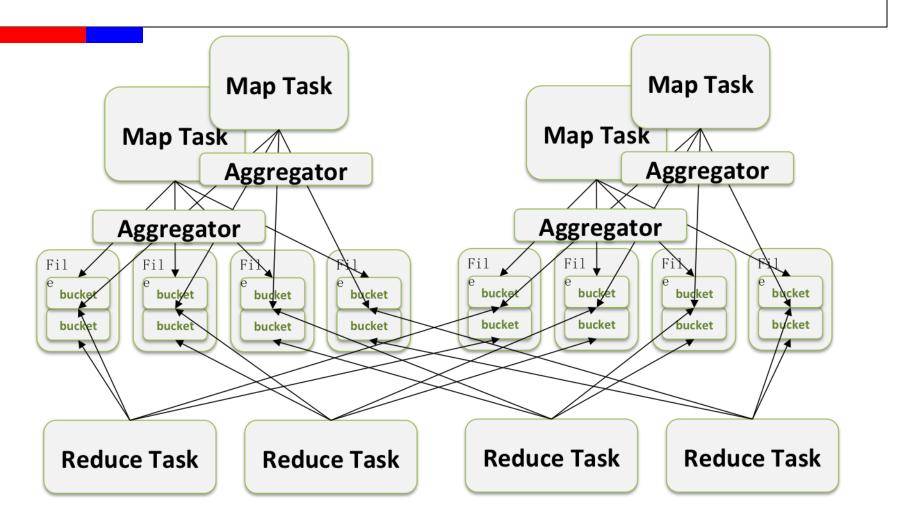


Base Shuffle Write



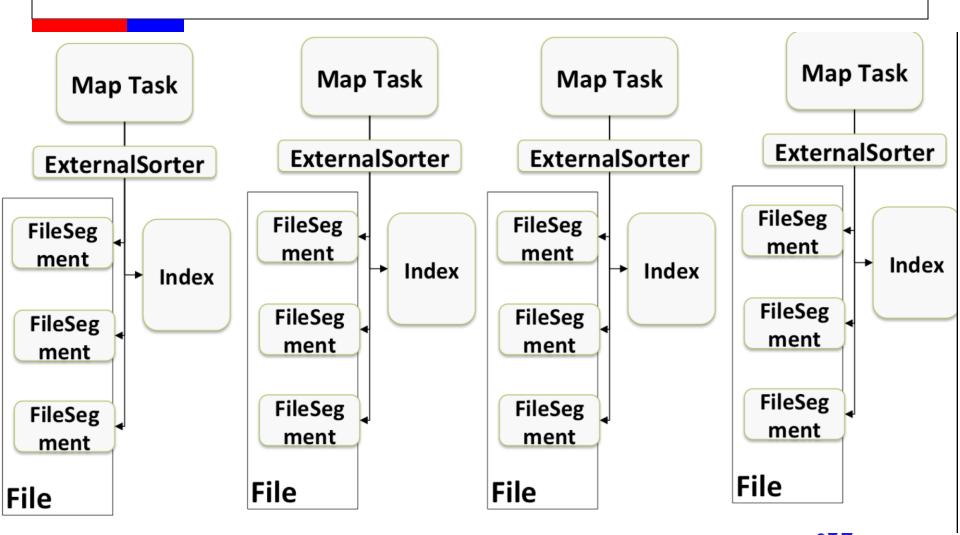


Consolidate Shuffle Write



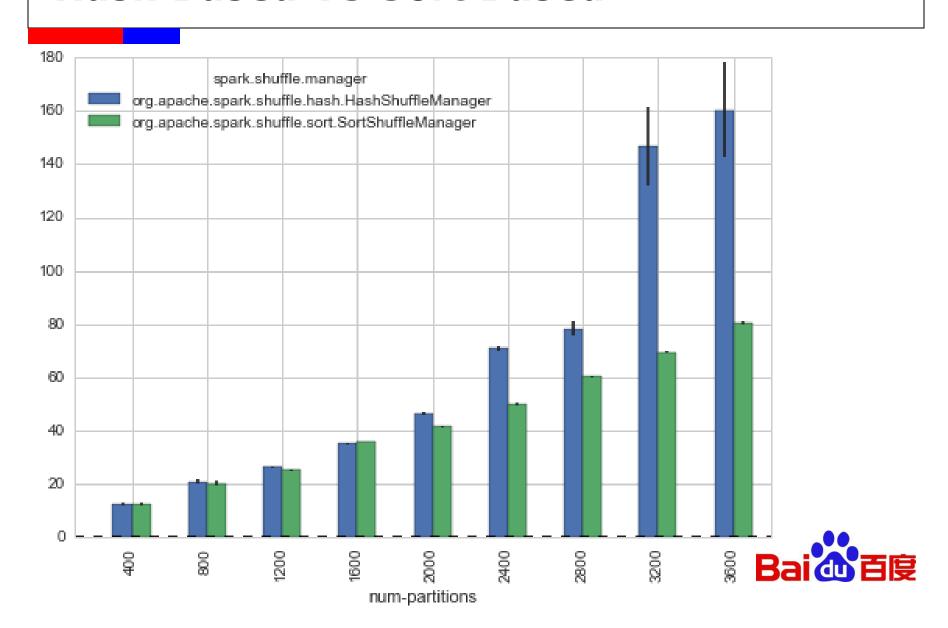


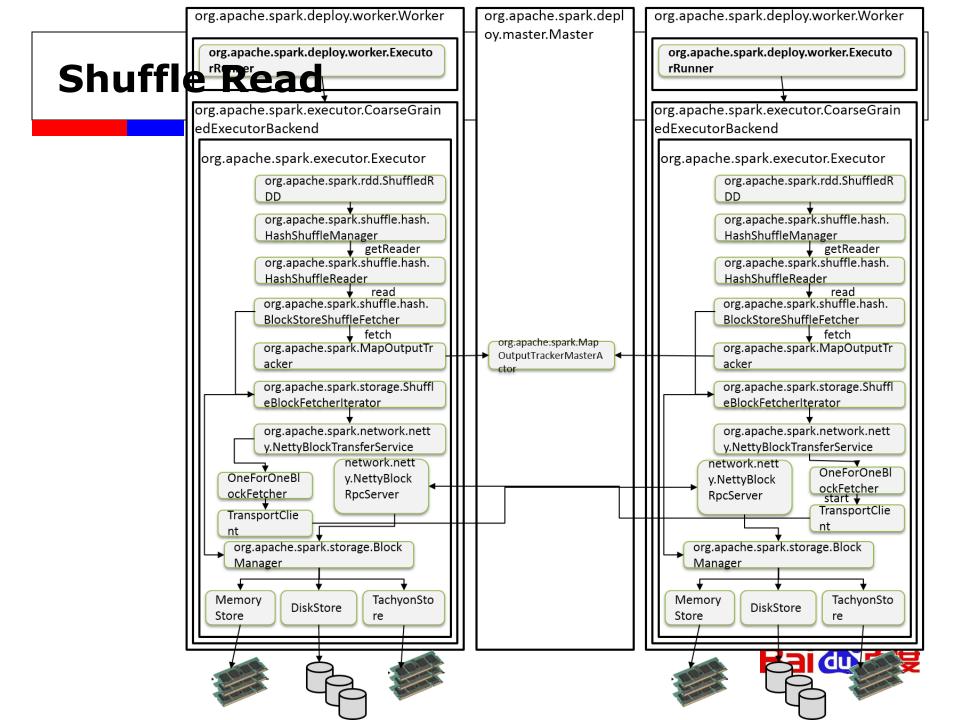
Sort Based Shuffle - Shuffle Writer



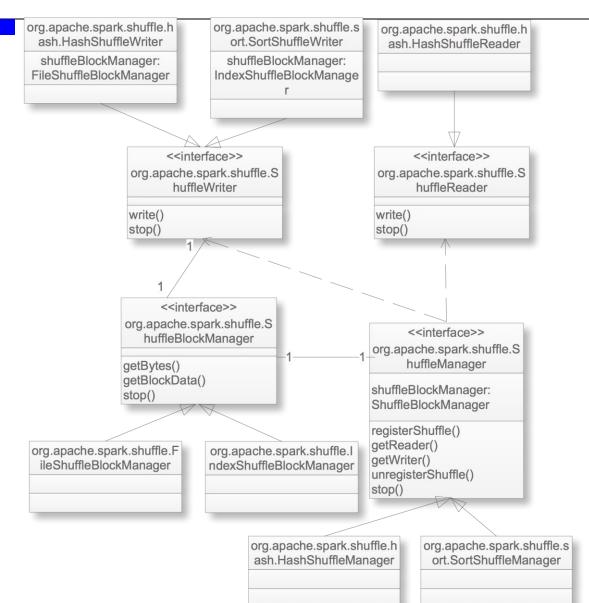


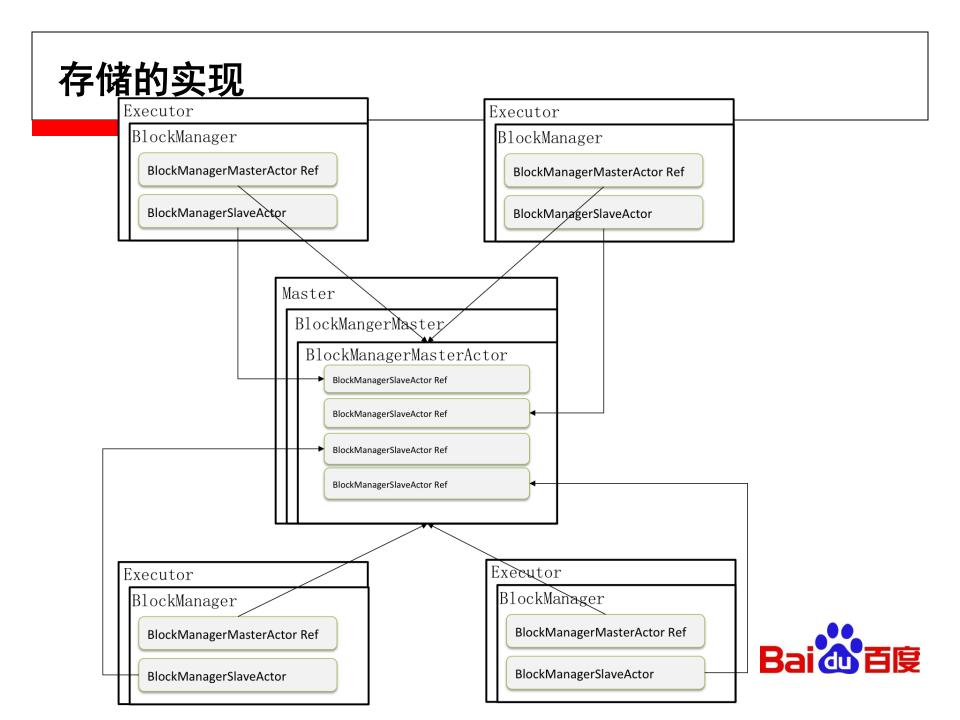
Hash Based VS Sort Based





Shuffle Pluggable 框架





Important But not Covered

- Scala
- Spark SQL
- Tachyon & RAMCloud
- Spark Best Practice (performance tuning)
- Spark in Industry
- Spark Ecosystem





