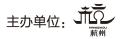




分布式流处理框架

——功能对比和性能评估







战略合作伙伴: (intel)

署名: 毛玮

职称: 技术专家









- MISC
- PerformanceBenchmark









Execution Model + Fault Tolerance Mechanism

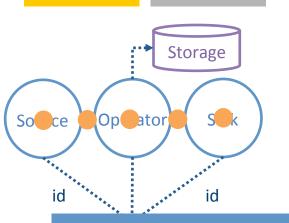






Continuous Streaming Ack per Record

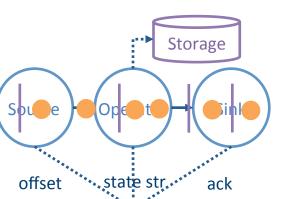
Storm Heron



Continuous Streaming

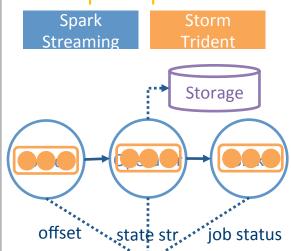
Checkpoint "per Batch"

Flink Gearpump



Micro-Batch

Checkpoint per Batch

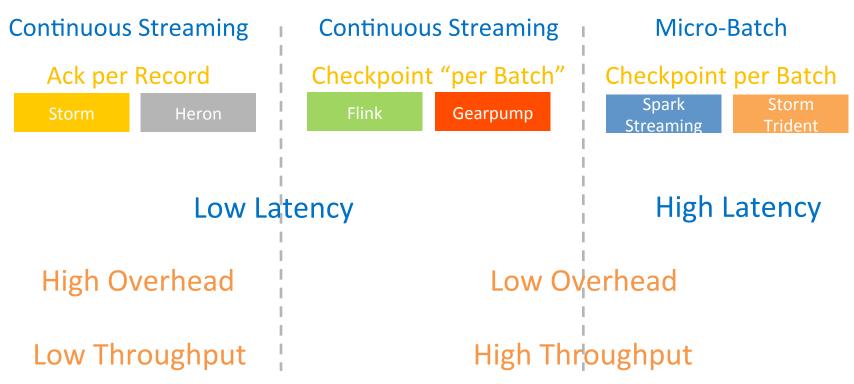


This is the **Critical** part, as it affects many features













Delivery Guarantee

Spark Flink Gearpump Heron Streaming At least once Exactly once Ackers know about if a • State is persisted in durable record is processed storage successfully or not. If it failed, replay it. Checkpoint is linked with state storage per Batch There is no state consistency guarantee.





Native State Operator

Storm

Heron

Yes*

- Storm:
 - √ KeyValueState

- Heron:
- X User Maintain

Flink

Gearpump

Yes

- Flink Java API:
 - ✓ ValueState
 - ✓ ListState
 - ✓ ReduceState
- Flink Scala API:
 - √ mapWithState
- Gearpump
 - ✓ persistState

Spark Streaming Storm Trident

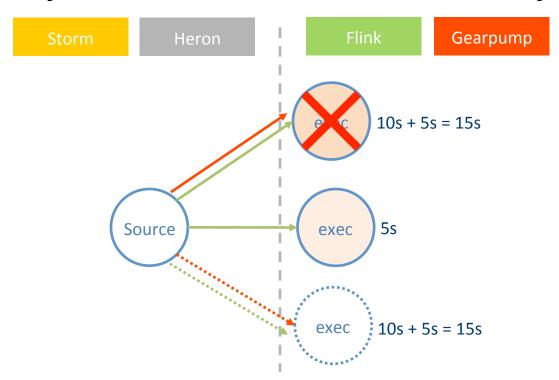
Yes

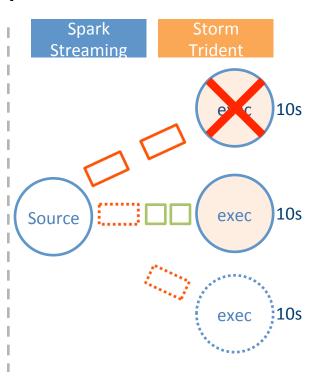
- Spark 1.5:
 - ✓ updateStateByKey
- Spark 1.6:
 - √ mapWithState
- Trident:
 - ✓ persistentAggregate
 - ✓ State





Dynamic Load Balance & Recovery Speed









API

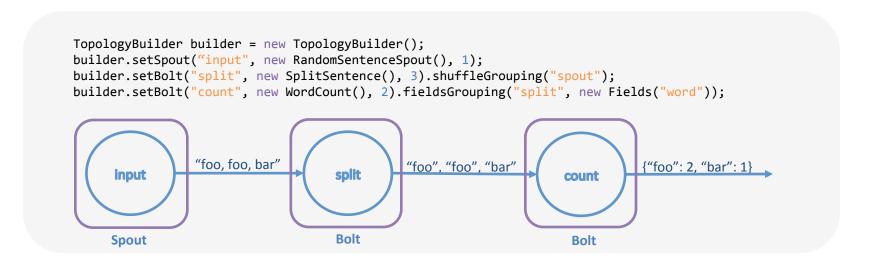




Storm Gearpump Heron

Compositional

- Highly customizable operator based on basic building blocks
- Manual topology definition and optimization







Declarative

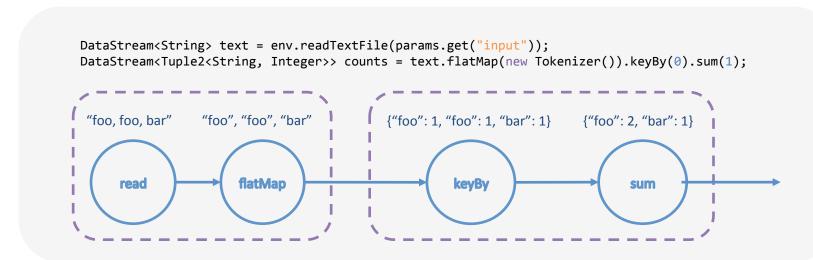
Streaming Storm Trident Flink

Spark

Higher order function as operators (filter, mapWithState...)

Gearpump

Logical plan optimization







Statistical

- Data scientist friendly
- Dynamic type

Spark
Streaming
Storm
Heron*

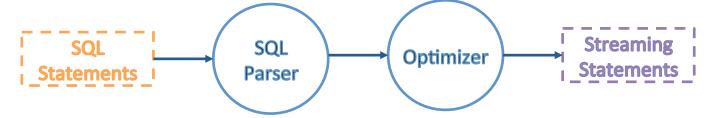
Python

```
lines = ssc.textFileStream(params.get("input"))
words = lines.flatMap(lambda line: line.split(","))
pairs = words.map(lambda word: (word, 1))
counts = pairs.reduceByKey(lambda x, y: x + y)
counts.saveAsTextFiles(params.get("output"))
```









Fusion Style

Spark
Streaming
Flink

```
InputDStream.transform((rdd: RDD[Order], time: Time) => {
   import sqlContext.implicits._
   rdd.toDF.registAsTempTable
   val SQL = "SELECT ID, UNIT_PRICE * QUANTITY
      AS TOTAL FROM ORDERS WHERE UNIT_PRICE * QUANTITY > 50"
   val largeOrderDF = sqlContext.sql(SQL)
   largeOrderDF.toRDD
})
```

Pure Style

Structured
Streaming*
Storm
Trident

CREATE EXTERNAL TABLE

ORDERS (ID INT PRIMARY KEY, UNIT_PRICE INT, QUANTITY INT) LOCATION 'kafka://localhost:2181/brokers?topic=orders' TBLPROPERTIES '{...}}'

INSERT INTO LARGE_ORDERS SELECT ID, UNIT_PRICE * QUANTITY AS TOTAL FROM ORDERS WHERE UNIT_PRICE * QUANTITY > 50

bin/storm sql XXXX.sql





Summary

	Compositional	Declarative	Python/R	SQL
Spark Streaming	X	\checkmark	√	√
Storm	√	X	√	NOT support aggregation,
Storm Trident	X	\checkmark	X	windowing and joining
Gearpump	V	\checkmark	X	X
Flink	Х	V	X	Support select, from, where, union
Heron	V	X	√*	X





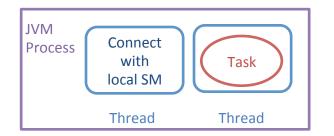
Runtime Model

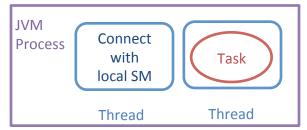




Single Task on Single Process

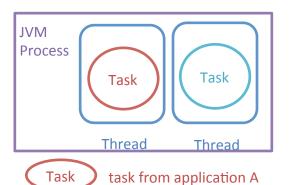
Heron

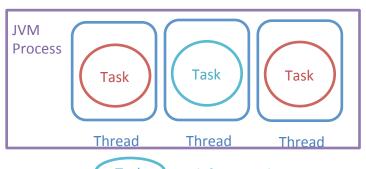




Multi Tasks of Multi Applications on Single Process

Flink



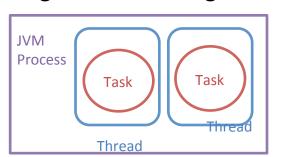


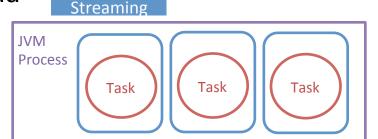




Thread

- Multi Tasks of Single application on Single Process
 - Single task on single thread

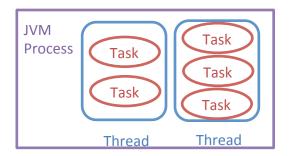


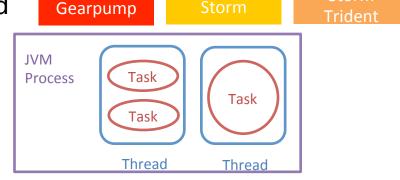


Spark

Thread

Multi tasks on single thread





Thread





MISC

- Window Support
 Out-of-order Processing
 Memory Management
- Resource Management
 Web UI
 Community Maturity

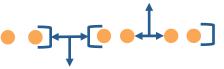




smaller than gap

Window Support

Sliding Window Session Window Sessi



Count Window

Output

session gap

	Sliding Window	Count Window	Session Window
Spark Streaming	√	X	X*
Storm	√	\checkmark	X
Storm Trident	V	V	X
Gearpump	√*	X	X
Flink	V	V	\checkmark
Heron	X	X	X





Out-of-order Processing

	Processing Time	Event Time	Watermark
Spark Streaming	V	√*	X*
Storm	\checkmark	\checkmark	√
Storm Trident	V	X	X
Gearpump	\checkmark	\checkmark	V
Flink	\checkmark	\checkmark	V
Heron	V	X	X





Memory Management

	JVM Manage	Self Manage on-heap	Self Manage off-heap
Spark Streaming	V	√*	√*
Flink	√	V	V
Storm	\checkmark	X	X
Gearpump	\checkmark	X	X
Heron	\checkmark	X	X





Resource Management

	Standalone	YARN	Mesos
Spark Streaming	√	V	√
Storm	V	√*	√*
Storm Trident	√	√*	√*
Gearpump	√	\checkmark	X
Flink	V	\checkmark	X
Heron	√	V	√





Web UI

	Submit Jobs	Cancel Jobs	Inspect Jobs	Show Statistics	Show Input Rate	Check Exceptions	Inspect Config	Alert
Spark Streaming	X	\checkmark	V	V	V	\checkmark	V	X
Storm	X	V	V	V	√*	\checkmark	V	X
Gearpump	V	\checkmark	V	V	√*	\checkmark	V	X
Flink	V	\checkmark	V	\checkmark	X	\checkmark	V	X
Heron	X	X	V	V	√*	V	V	X

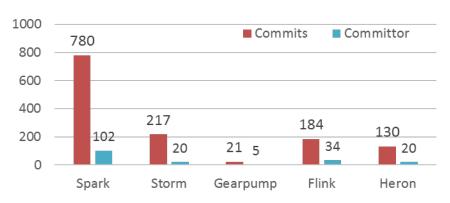




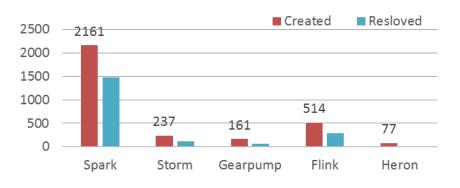
Community Maturity

	Initiation Time	Apache Top Project	Contri butors
Spark Streaming	2013	2014	926
Storm	2011	2014	219
Gearpump	2014	Incubator	21
Flink	2010	2015	208
Heron	2014	N/A	44

Past 1 Months Summary on GitHub



Past 3 Months Summary on JIRA







Performance Benchmark

HiBench 6.0





Test Philosophical

- Lazy Benchmarking
- Simple test case infer practical use case





The Setup

Name	Version
Java	1.8
Scala	2.11.7
Hadoop	2.6.2
Zookeeper	3.4.8
Kafka	0.8.2.2
Spark	1.6.1
Storm	1.0.1
Flink	1.0.3
Gearpump	0.8.1

- Heron require specific Operation System (Ubuntu / CentOS / Mac OS)
- Structured Streaming doesn' t support Kafka source yet (Spark 2.0)

Kafka Cluster

•CPU: 2 x Intel(R) Xeon(R) CPU

E5-2699 v3@ 2.30GHz

•Mem: 128 GB

•Disk: 8 x HDD (1TB)
•Network: 10 Gbps

Test Cluster

•CPU: 2 x Intel(R) Xeon(R) CPU

E5-2697 v2@ 2.70GHz

•Core: 20 / 24

•Mem: 80 / 128 GB
•Disk: 8 x HDD (1TB)

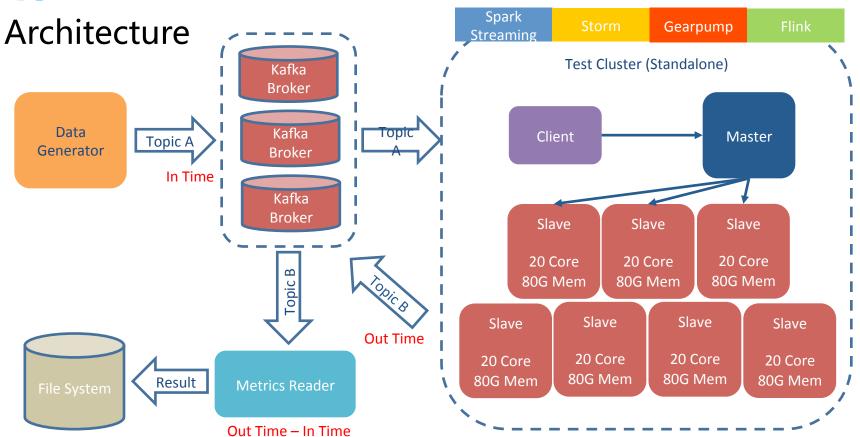
•Network: 10 Gbps















Framework Configuration

Framework	Related Configuration
Spark Streaming	7 Executor 140 Parallelism
Flink	7 TaskManager 140 Parallelism
Storm	28 Worker 140 KafkaSpout
Gearpump	28 Executors 140 KafkaSource





Raw Input Data

- Kafka Topic Partition: 140
- Size Per Message (configurable): 200 bytes
- Raw Input Message Example:

"0,227.209.164.46,nbizrgdziebsaecsecujfjcqtvnpcnxxwiopmddorcxnlijdizgoi, 1991-06-10,0.115967035,Mozilla/5.0 (iPhone; U; CPU like Mac OS X)AppleWebKit/420.1 (KHTML like Gecko) Version/3.0 Mobile/4A93Safari/419.3,YEM,snowdrops,1"

Strong Type: class UserVisit (ip, sessionId, browser)

Keep feeding data at specific rate for 5 minutes

5 minutes





Data Input Rate

Throughput	Message/Second	Kafka Producer Num
40KB/s	0.2K	1
400KB/s	2K	1
4MB/s	20K	1
40MB/s	200K	1
80MB/s	400K	1
400MB/s	2M	10
600MB/s	3M	15
800MB/s	4M	20





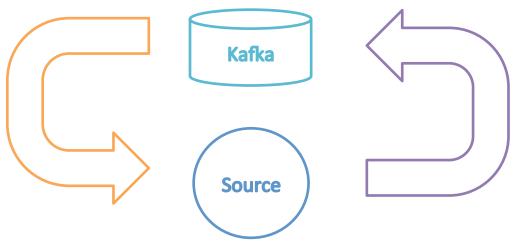
Let's start with the simplest case





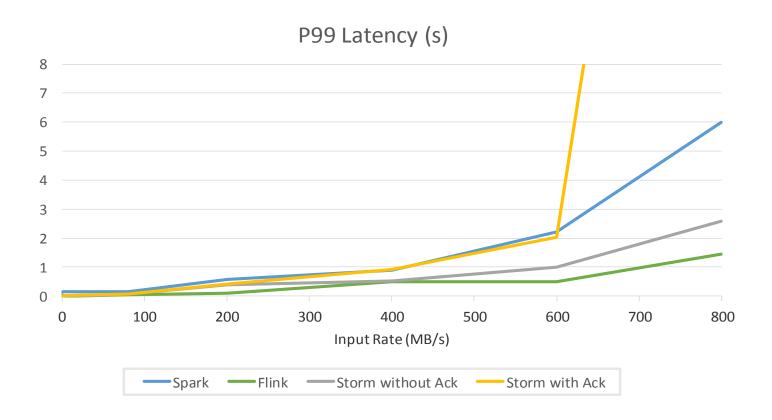
Test Case: Identity

 The application reads input data from Kafka and then writes result to Kafka immediately, there is no complex business logic involved.













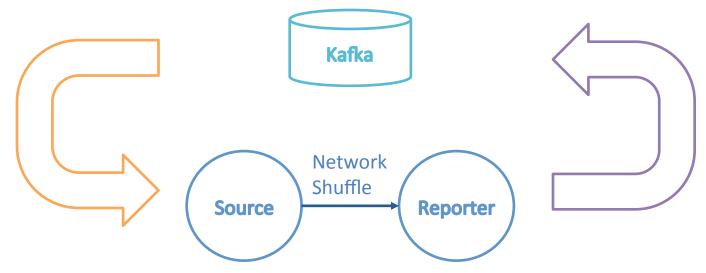
Q: What if source data are skew or even packed?





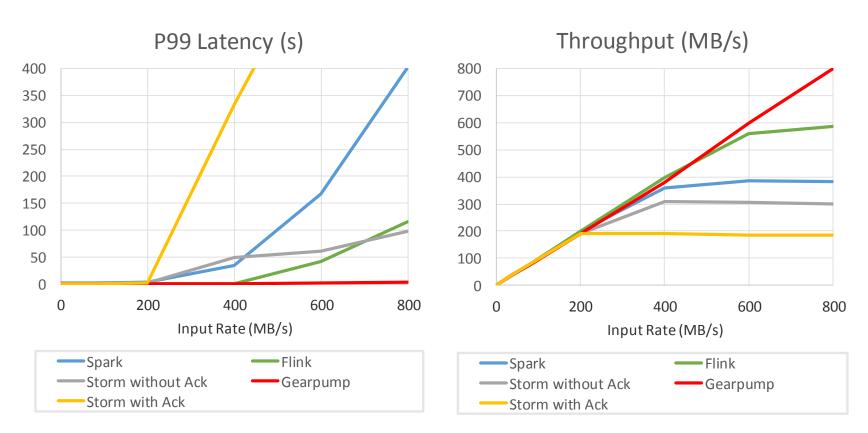
Test Case: Repartition

 Basically, this test case can stand for the efficiency of data shuffle.













- Flink and Storm has close performance and are better choices to meet sub-second SLA requirement if no repartition happened.
- Spark Streaming need to schedule task with additional context. Under tiny batch interval case, the overhead could be dramatic worse compared to other frameworks.
- According to our test, minimum Batch Interval of Spark is about 80ms (140 tasks per batch), otherwise task schedule delay will keep increasing
- Repartition is heavy for every framework, but usually it's unavoidable.
- Latency of Gearpump is still quite low even under 800MB/s input throughput.





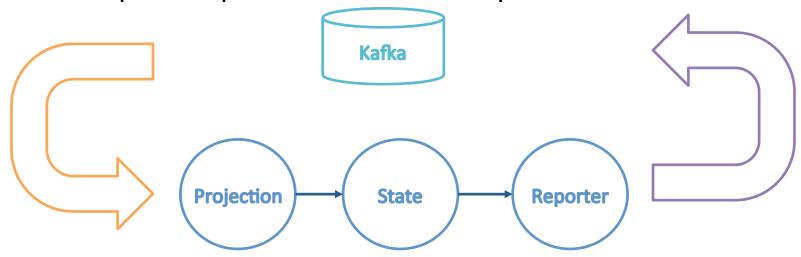
Q: What if I want to apply slightly complex logic which need to maintain entire state?





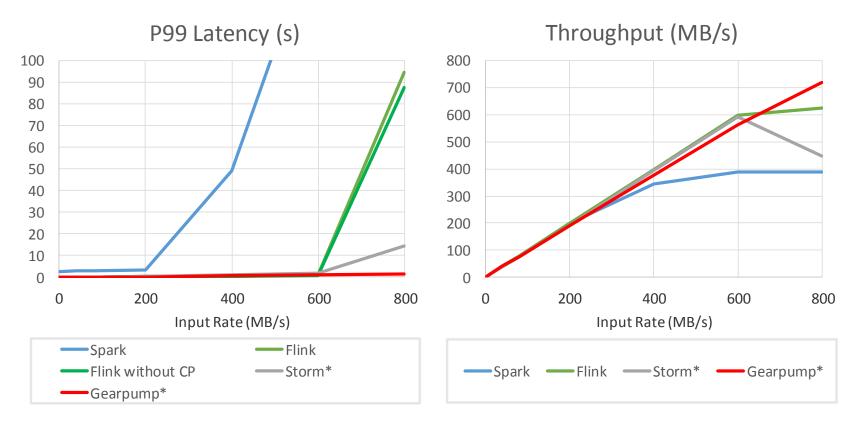
Test Case: Stateful WordCount

- Native state operator is supported by all frameworks we evaluated
- Stateful operator performance + Checkpoint/Acker cost













Observation

- Exactly-once semantics usually require state management and checkpoint. But better guarantees come at high cost.
- There is no obvious performance difference in Flink when switching fault tolerance on or off.
- Checkpoint mechanisms and storages play a critical role here.





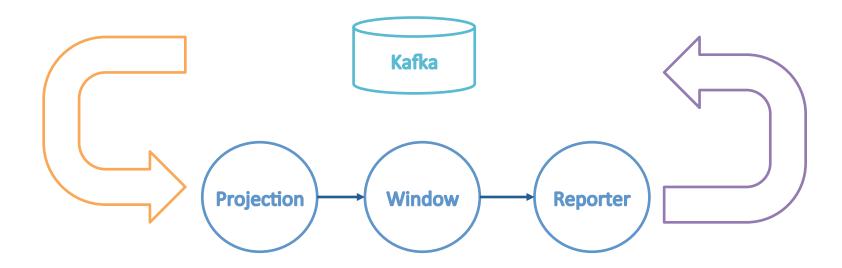
Q: How about Window Operation?





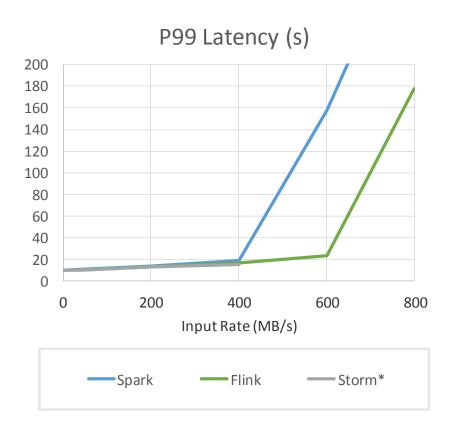
Test Case: Window Based Aggregation

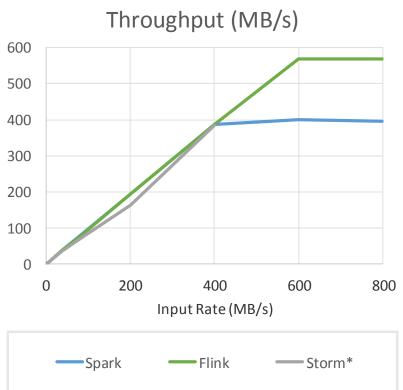
This test case manages a 10-seconds sliding window









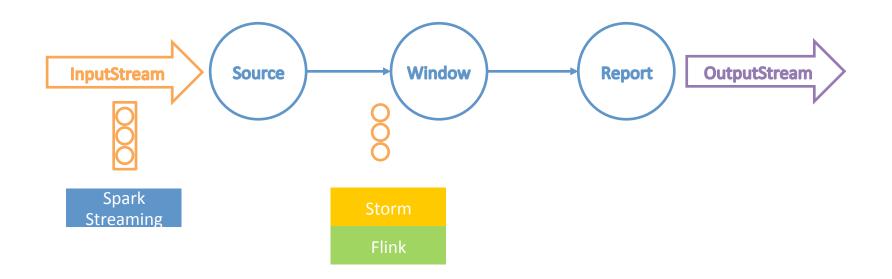






Observation

The native streaming execution model helps here







So which streaming framework should I use?





Do your own benchmark

• **HiBench**: a cross platforms micro-benchmark suite for big data (https://github.com/intel-hadoop/HiBench)

- Open Source since 2012
- Better streaming benchmark supporting will be included in next release [HiBench 6.0]



