



TACHYON

A Reliable Memory-Centric Distributed Storage System

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Scott Shenker, Ion Stoica

September, 2014, ISTC @ Portland

<http://www.istc-cc.cmu.edu/>



Outline

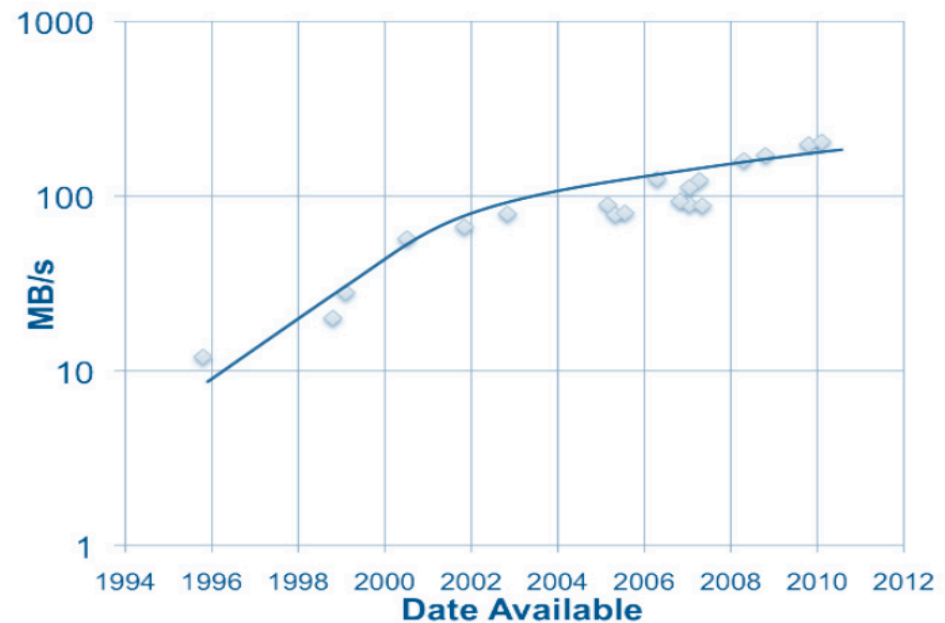
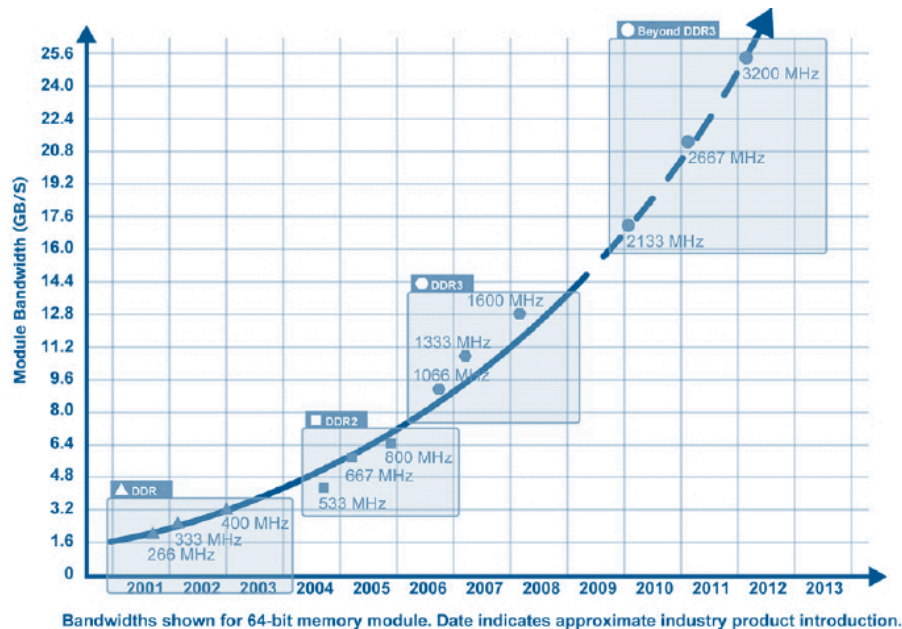
- Overview
 - Feature 1: Memory Centric Storage Architecture
 - Feature 2: Lineage in Storage
- Challenges
- Open Source
- Future

Outline

- **Overview**
 - Feature 1: Memory Centric Storage Architecture
 - Feature 2: Lineage in Storage
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Memory is King

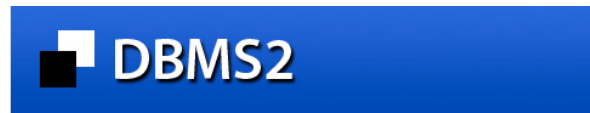
- RAM throughput increasing **exponentially**
- Disk throughput increasing **slowly**



Memory-locality key to interactive response time

Realized by many...

- Frameworks already leverage memory



April 7, 2012

Many kinds of memory-centric data management

I'm frequently asked to generalize in some way about in-memory or memory-centric data management. I can start:

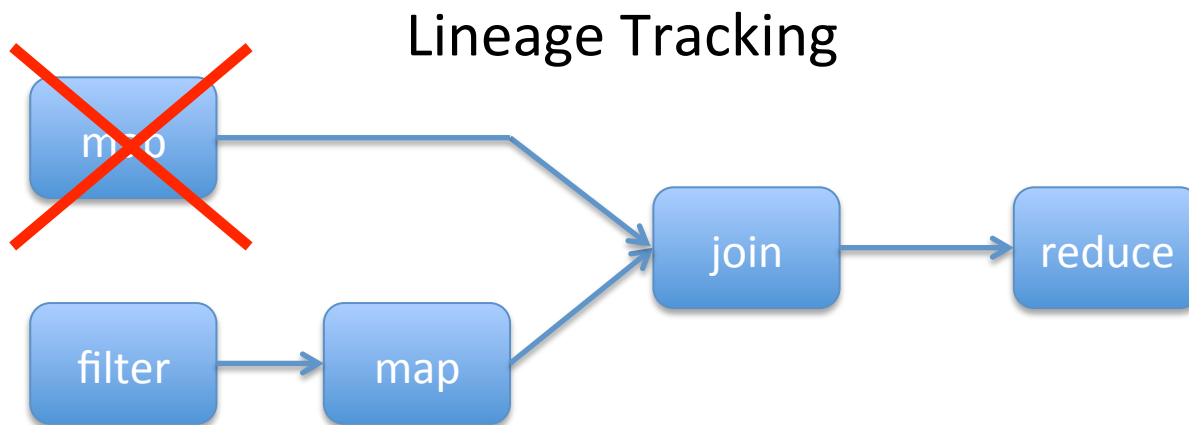
- The desire for human real-time interactive response naturally leads to



Problem solved?

An Example: Spark

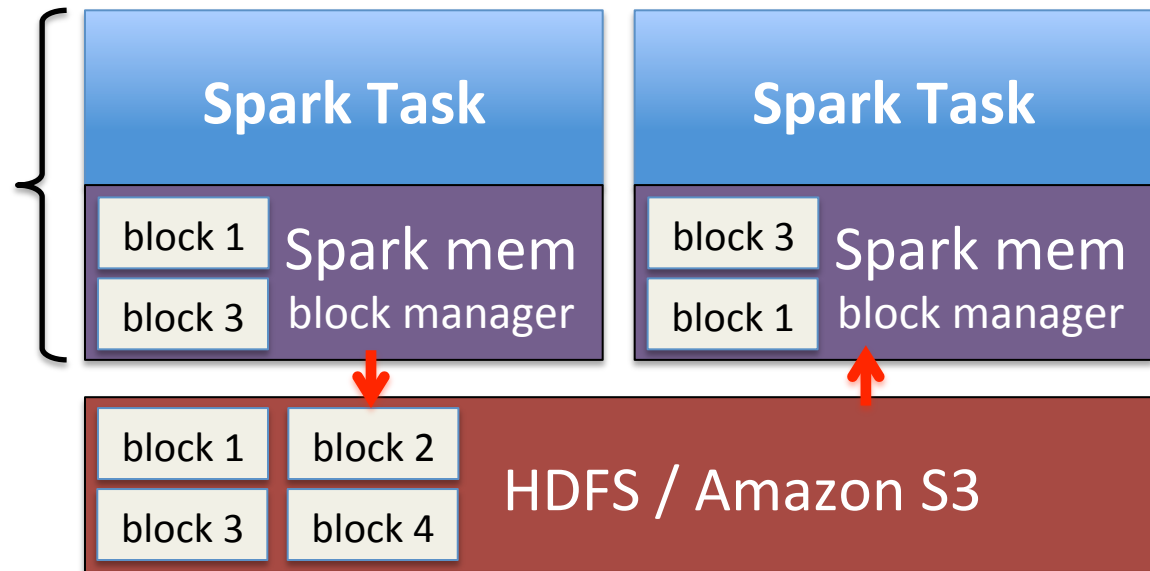
- Fast in-memory data processing framework
 - Keep **one** in-memory copy inside JVM
 - Track **lineage** of operations used to derive data
 - Upon failure, use lineage to recompute data



Issue 1

***Data Sharing is the bottleneck in analytics pipeline:
Slow writes to disk***

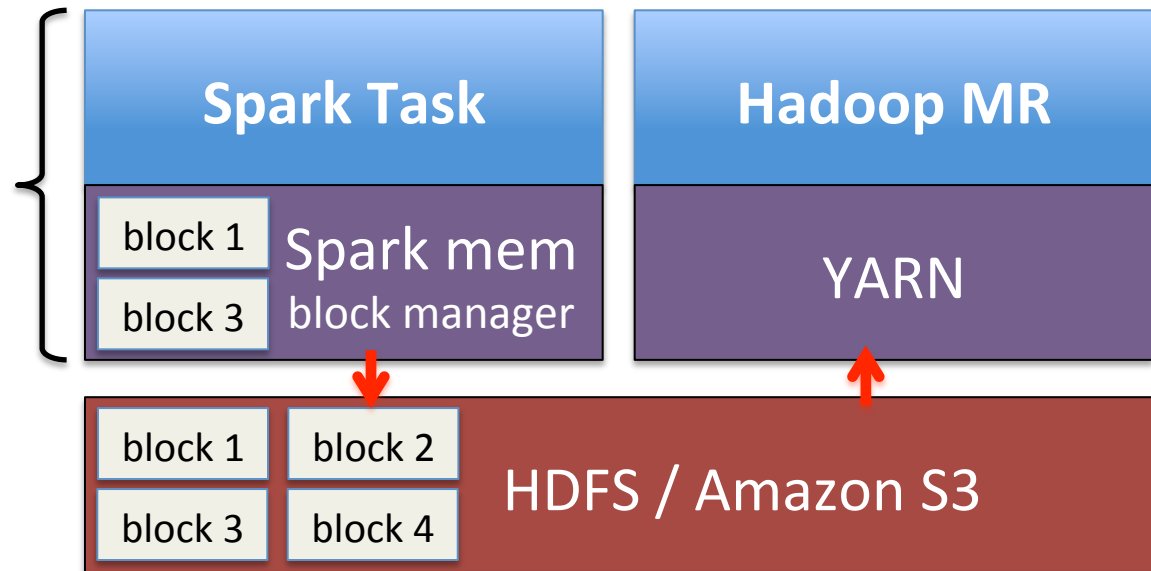
storage engine &
execution engine
same process
(slow writes)



Issue 1

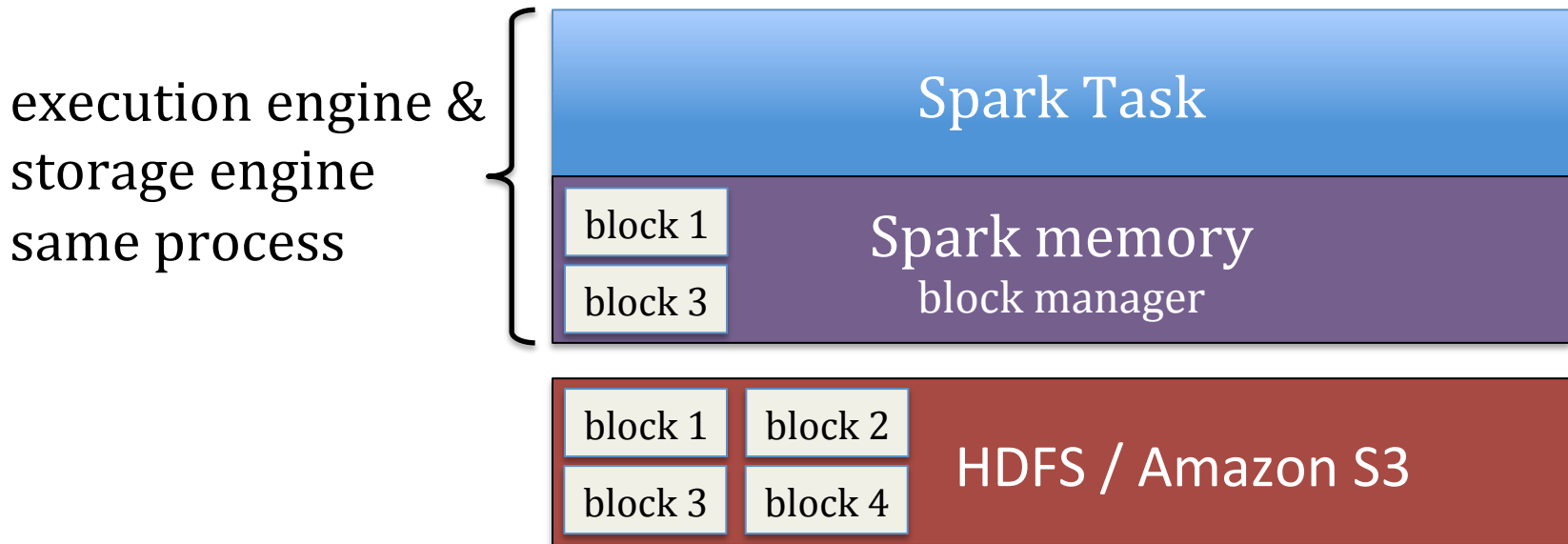
***Data Sharing is the bottleneck in analytics pipeline:
Slow writes to disk***

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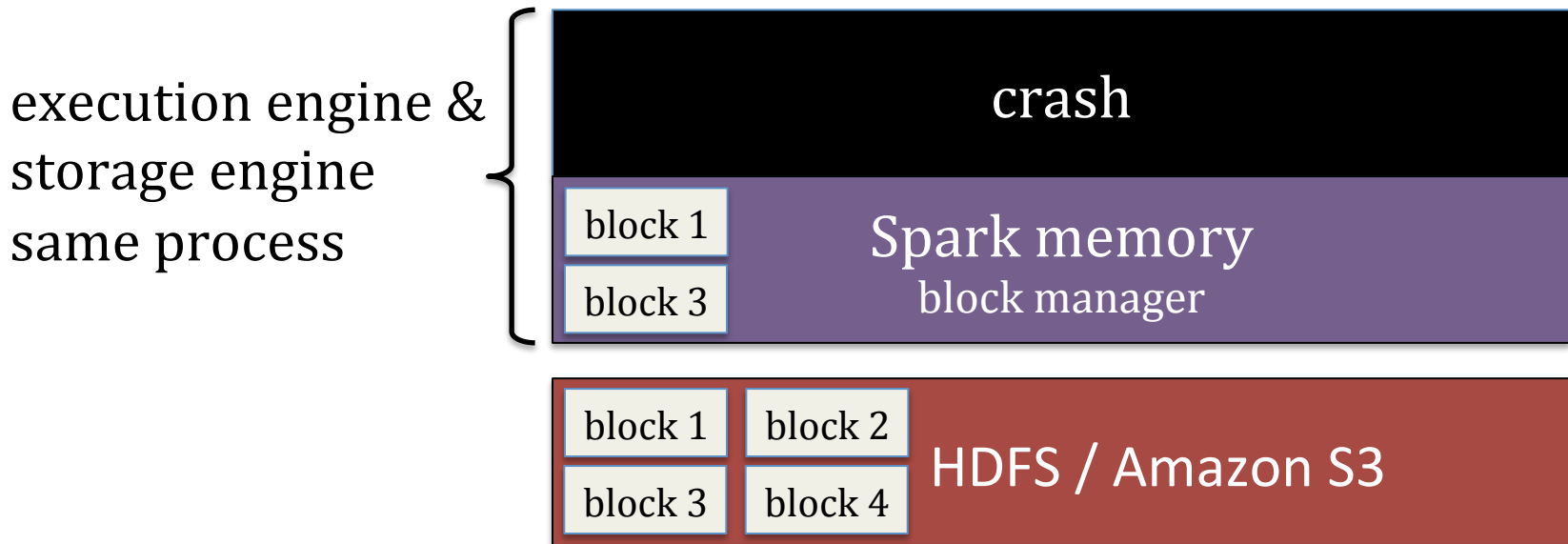
Issue 2

Cache loss when process crashes.



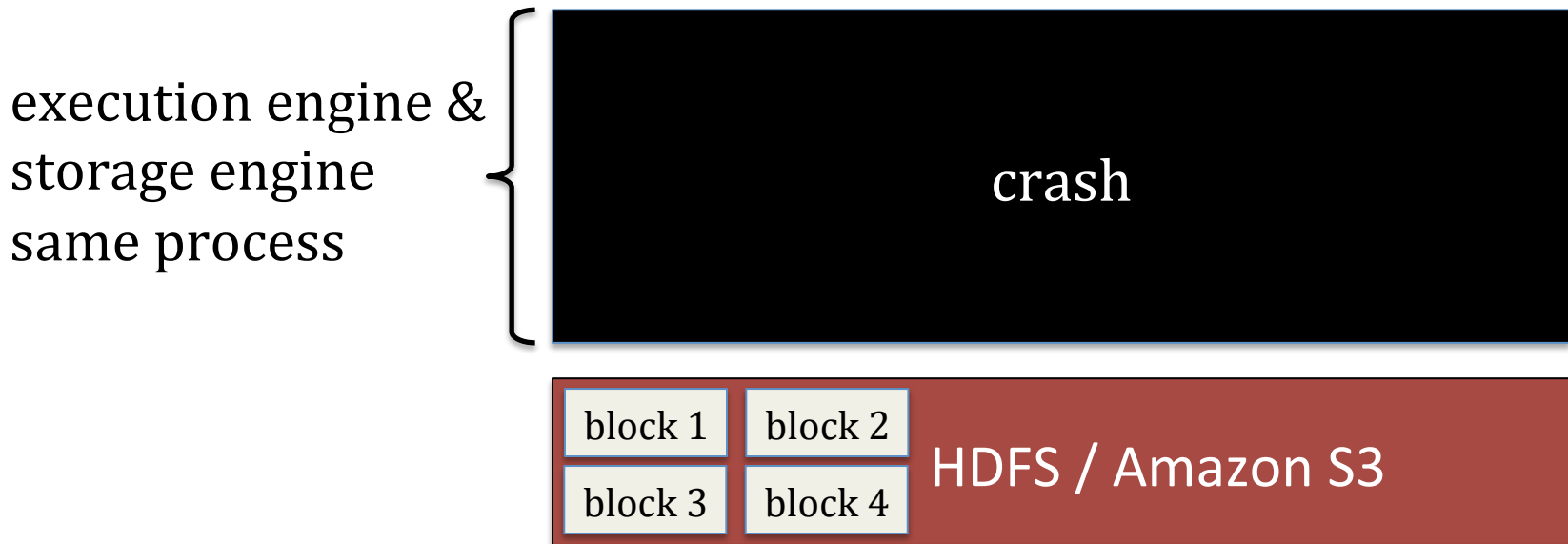
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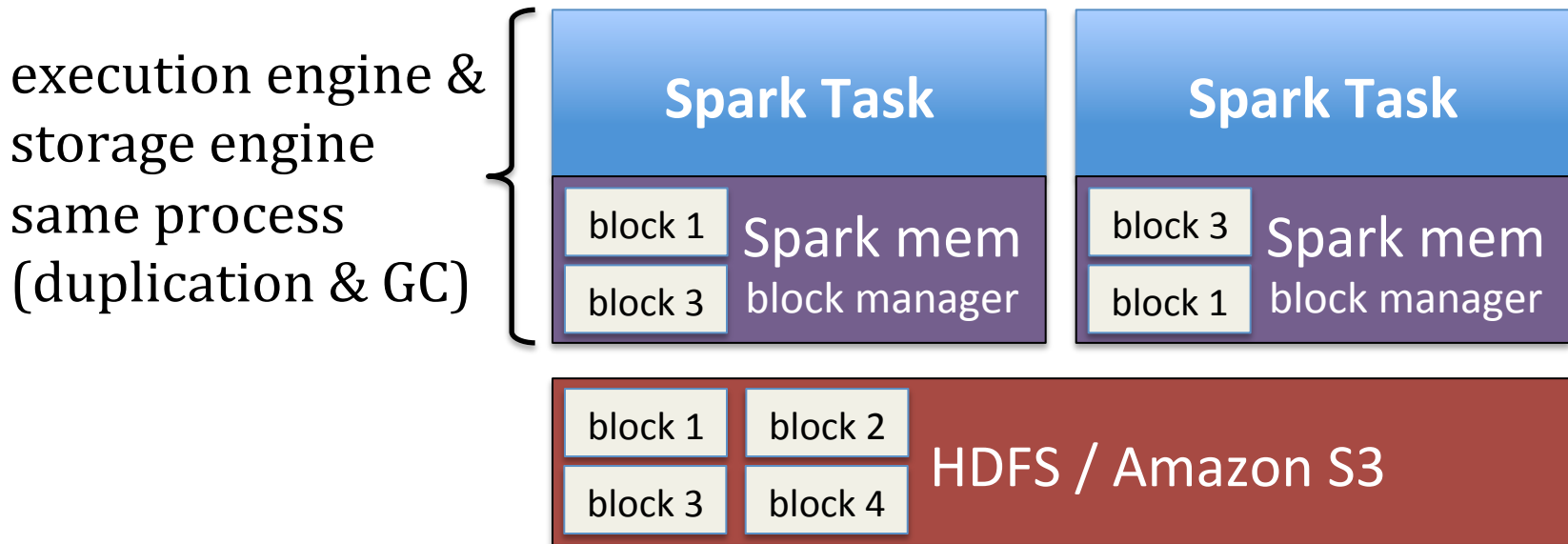
Issue 2

Cache loss when process crashes.



Issue 3

In-memory Data Duplication & Java Garbage Collection



Tachyon

Reliable data sharing at ***memory-speed***
within and across cluster frameworks/jobs

Solution Overview

Basic idea

- Feature 1: **memory-centric** storage architecture
- Feature 2: push **lineage** down to storage layer

Facts

- One data copy in memory
- Recomputation for fault-tolerance

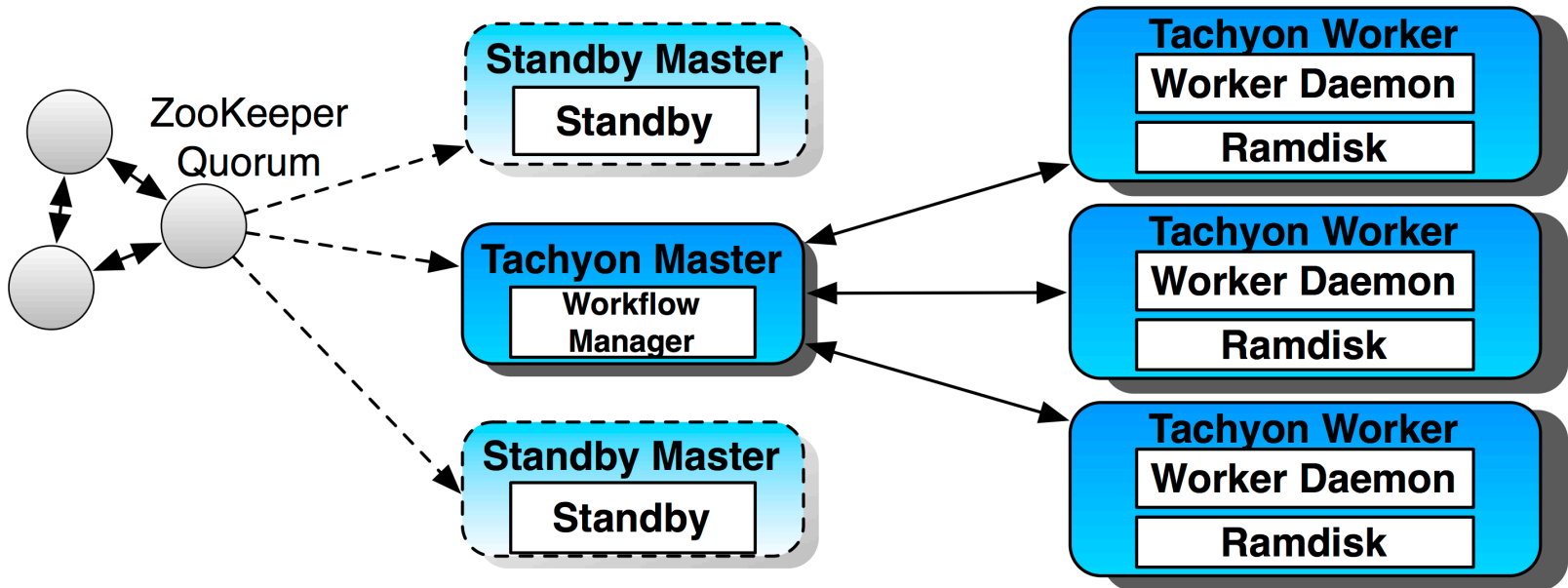
Stack

Computation Frameworks
(Spark, MapReduce, Impala, H2O, ...)

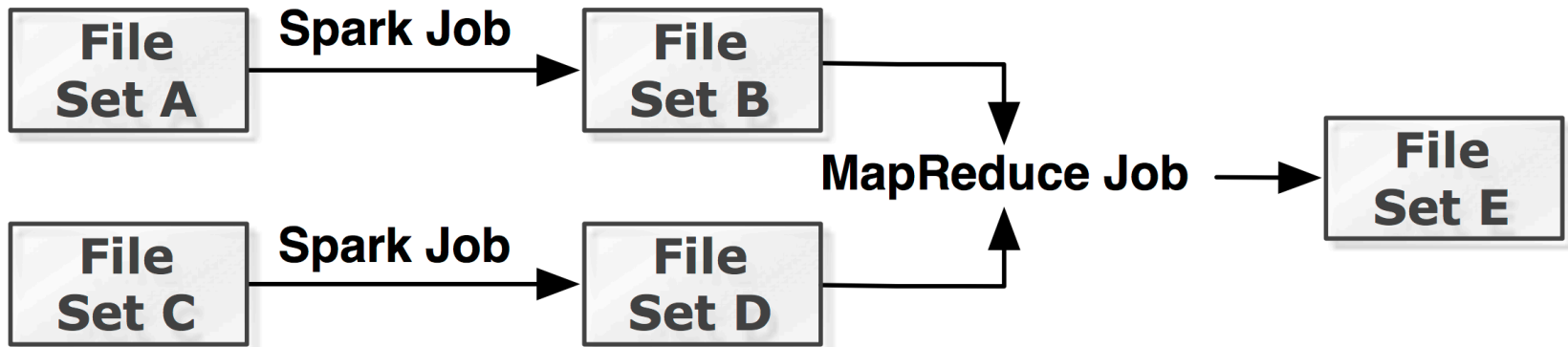
Tachyon

Existing Storage Systems
(HDFS, S3, GlusterFS, ...)

Memory-Centric Storage Architecture



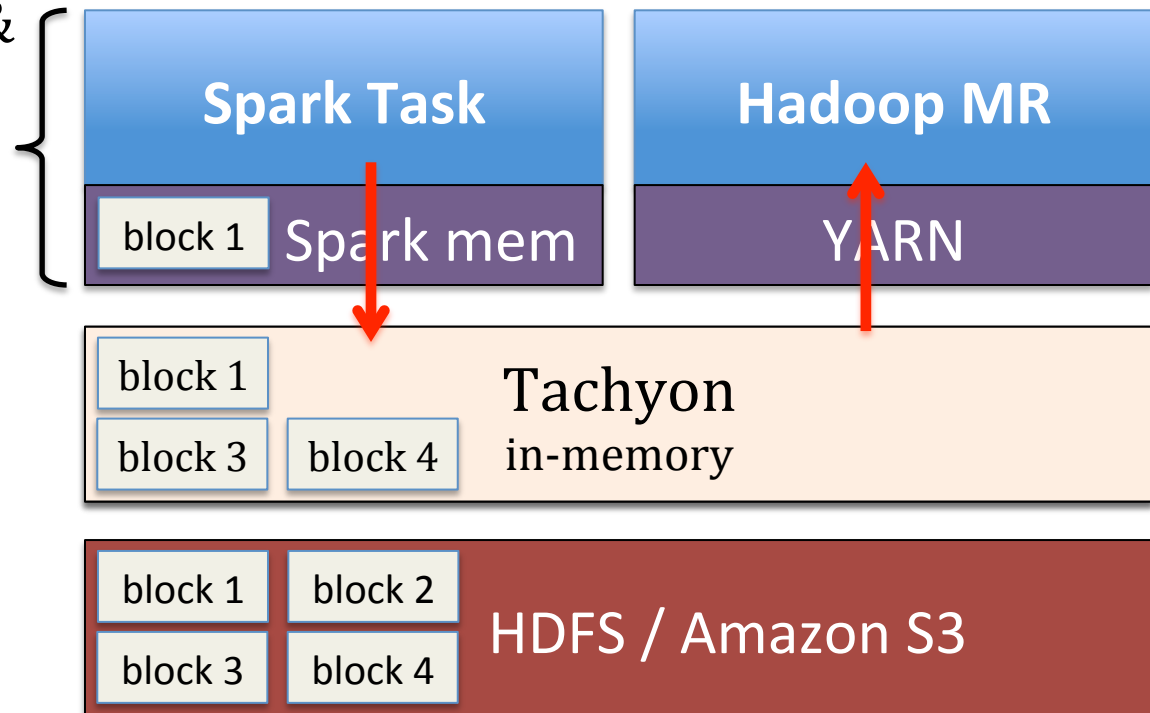
Lineage in Storage



Issue 1 revisited

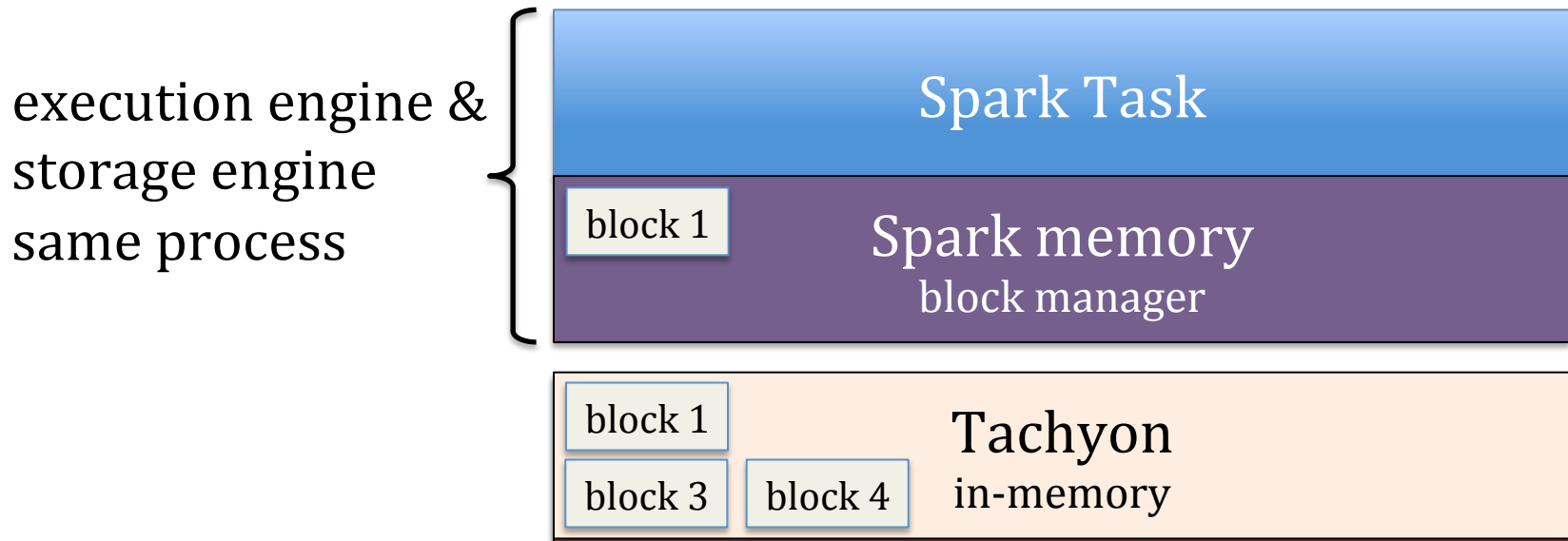
Memory-speed data sharing among jobs in different frameworks

execution engine &
storage engine
same process
(fast writes)



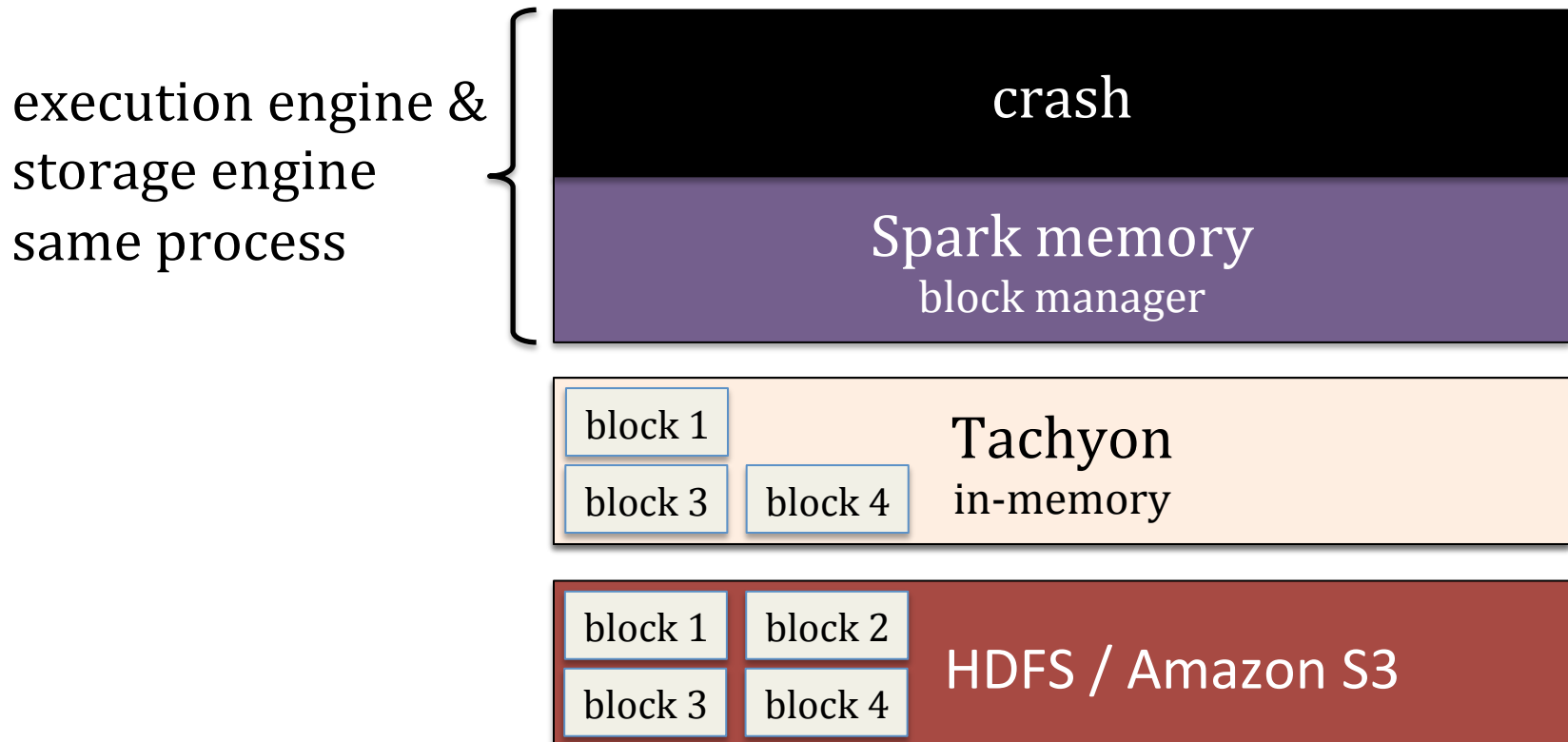
Issue 2 revisited

***Keep in-memory data safe,
even when a job crashes.***



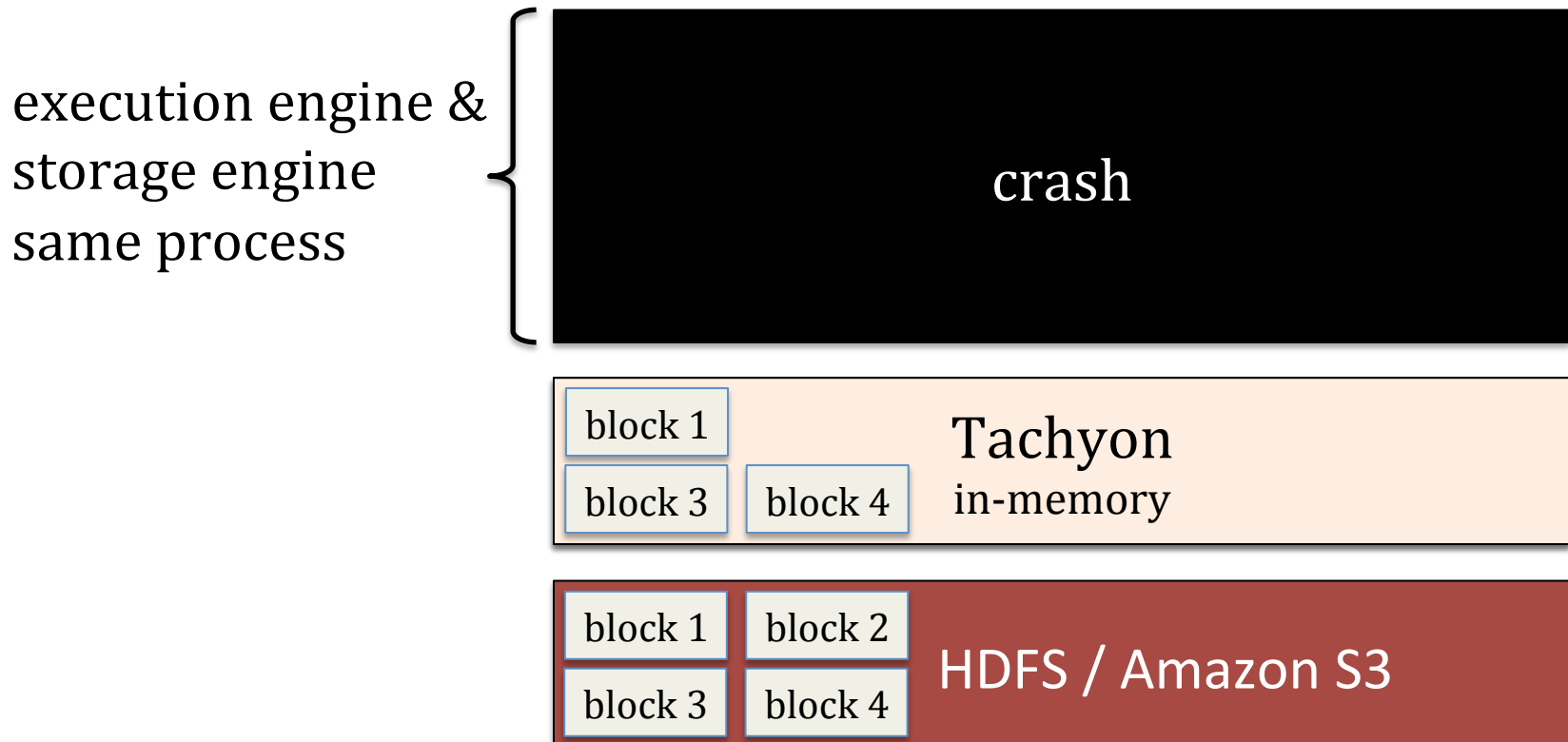
Issue 2 revisited

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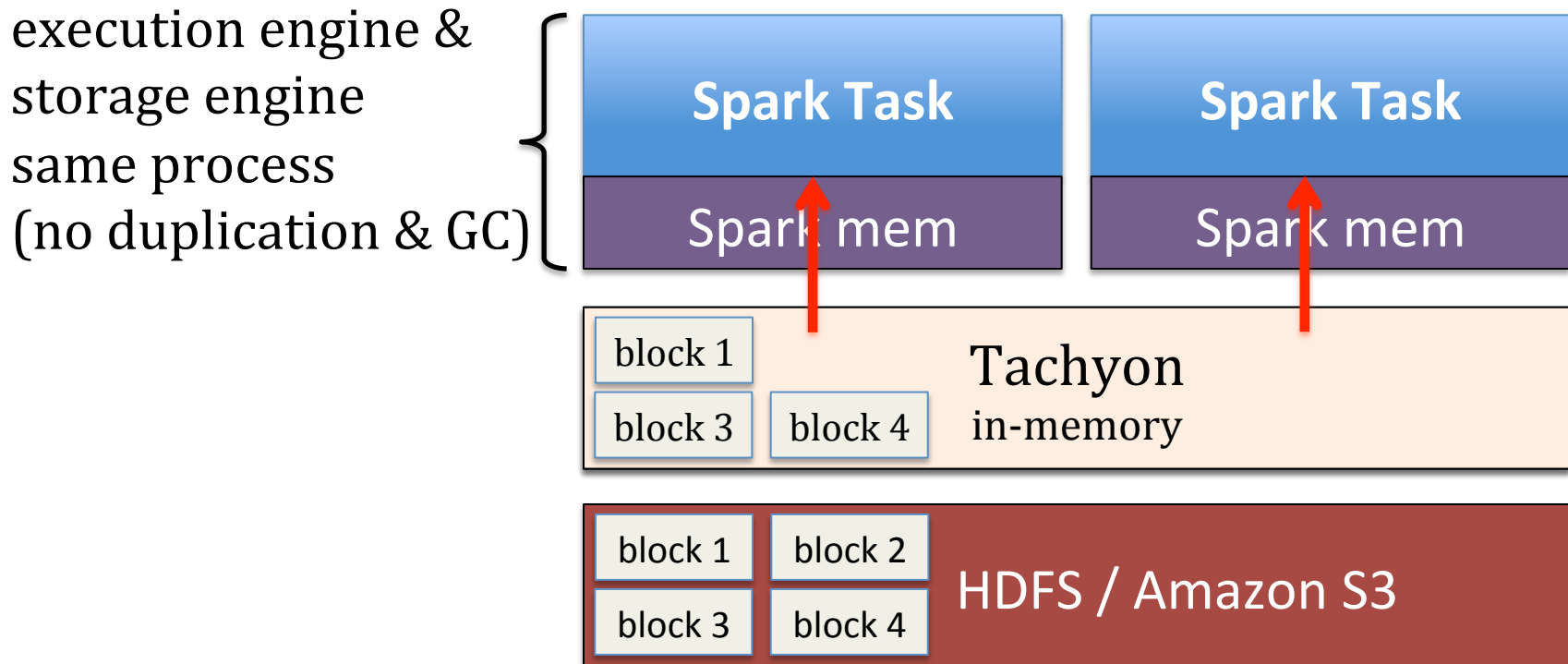
Issue 2 revisited

*Keep in-memory data safe,
even when a job crashes.*



Issue 3 revisited

***No in-memory data duplication,
much less GC***



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- **Challenges**
- Open Source
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Question 1: How long to get missing data back?



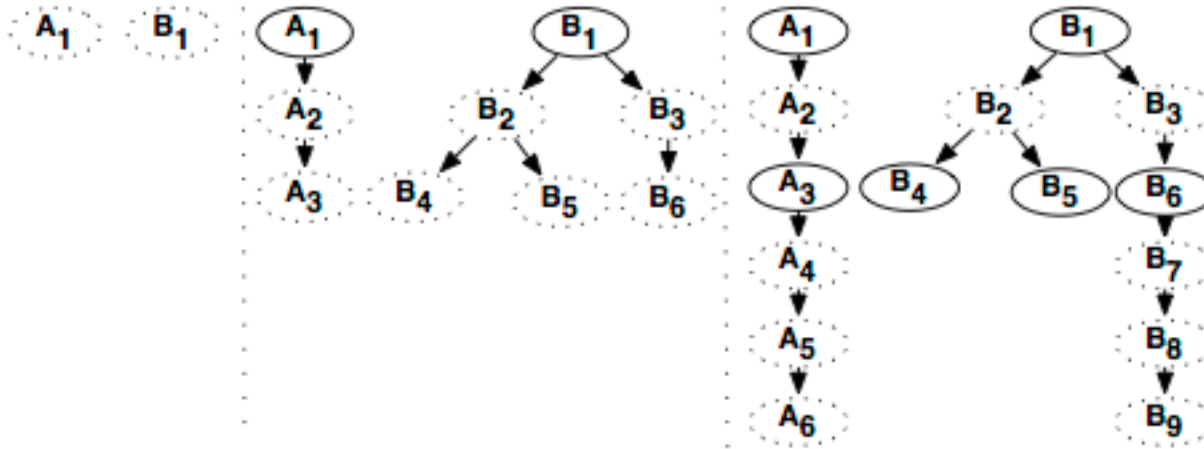
That server contains
the data computed
last month!



Lineage enables **Asynchronous Checkpointing**

Edge Algorithm

- Checkpoint leaves
- Checkpoint hot files
- Bounded Recovery Cost



Question 2: How to allocate recomputation resource?

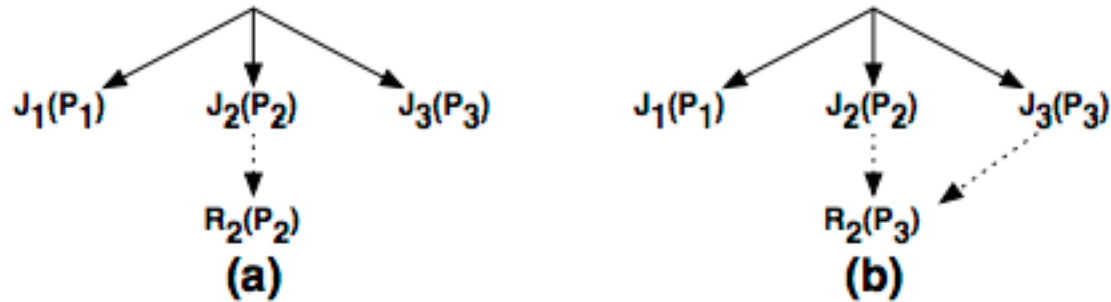


Would recomputation slow down my high priority jobs?
Priority Inversion?

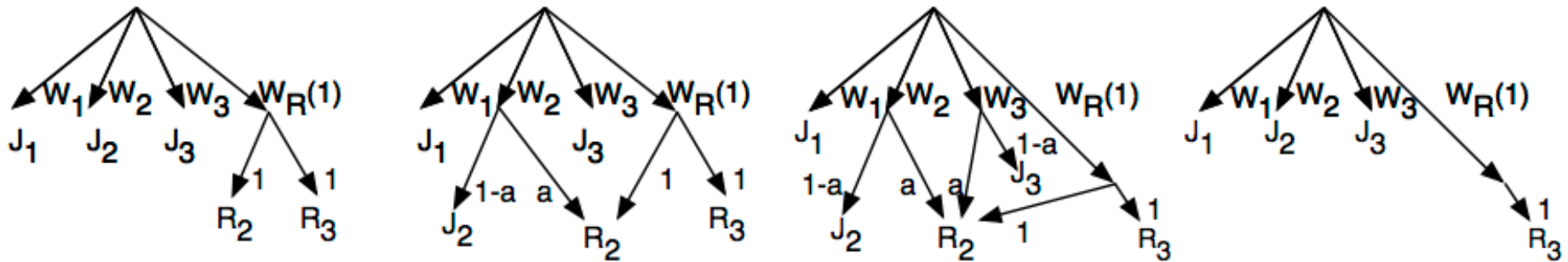


Recomputation Resource Allocation

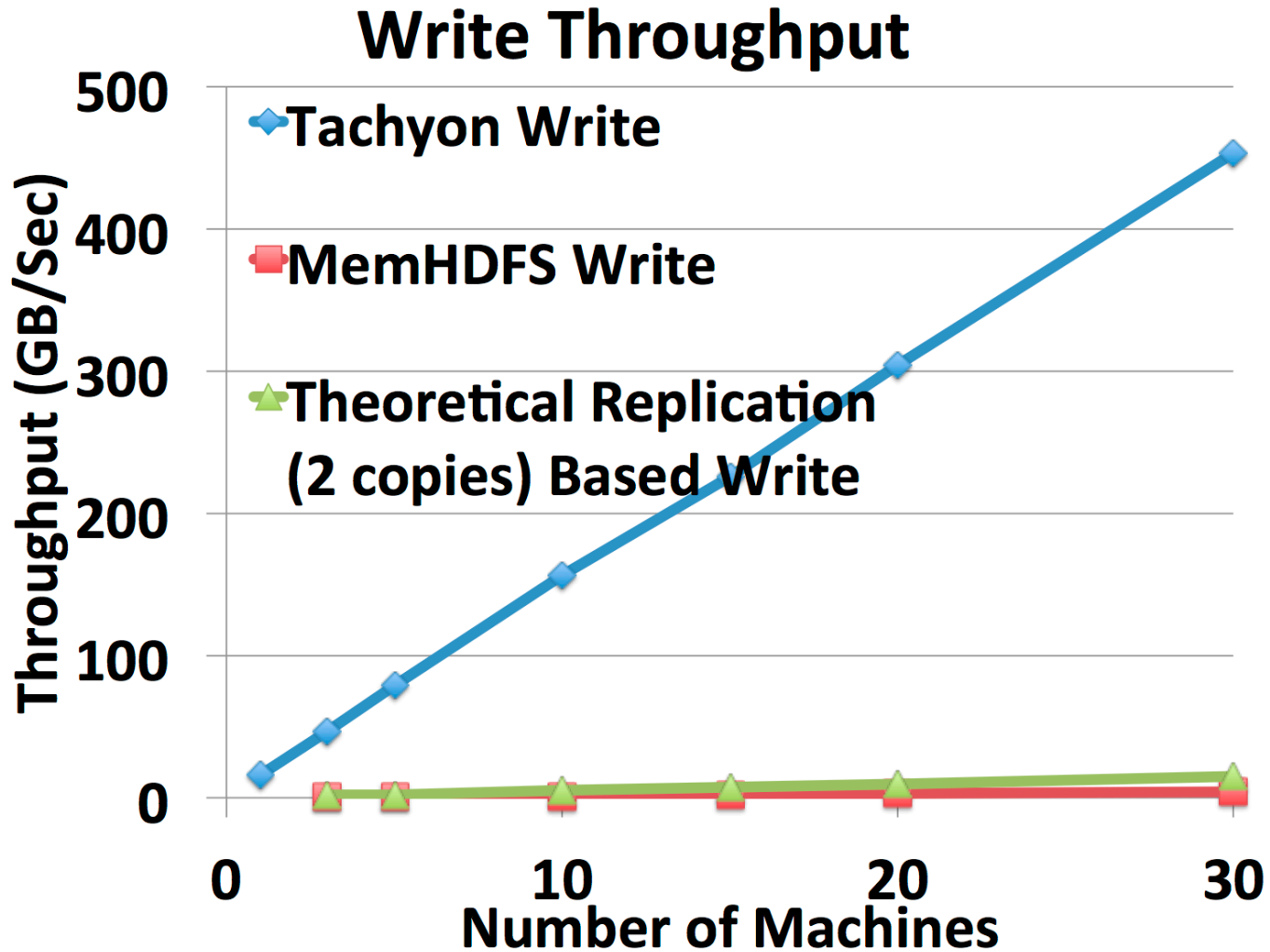
- Priority Based Scheduler



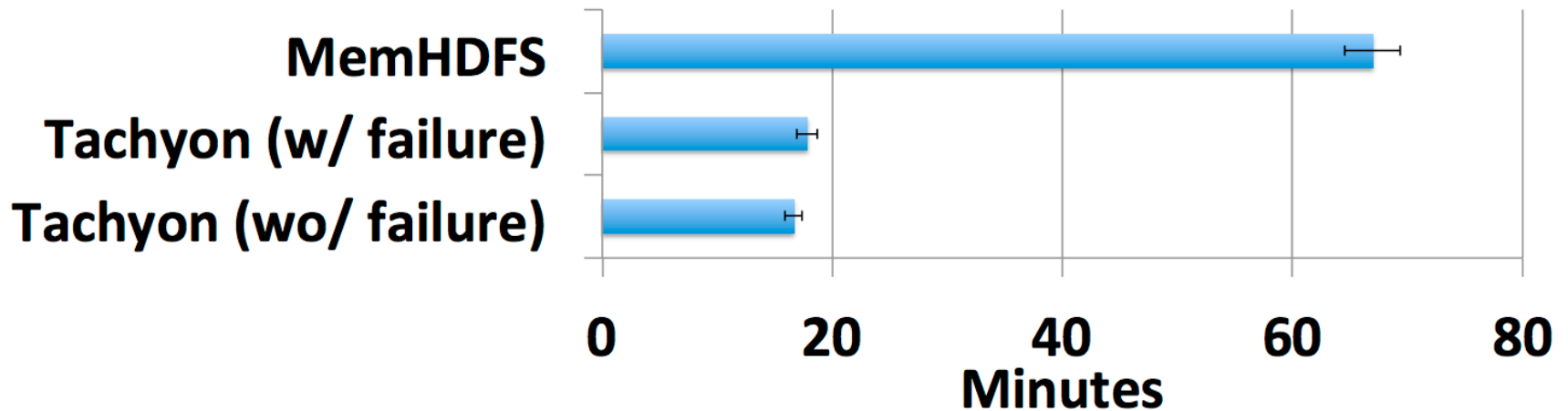
- Fair Sharing Based Scheduler



Comparison with in Memory HDFS



Workflow Improvement

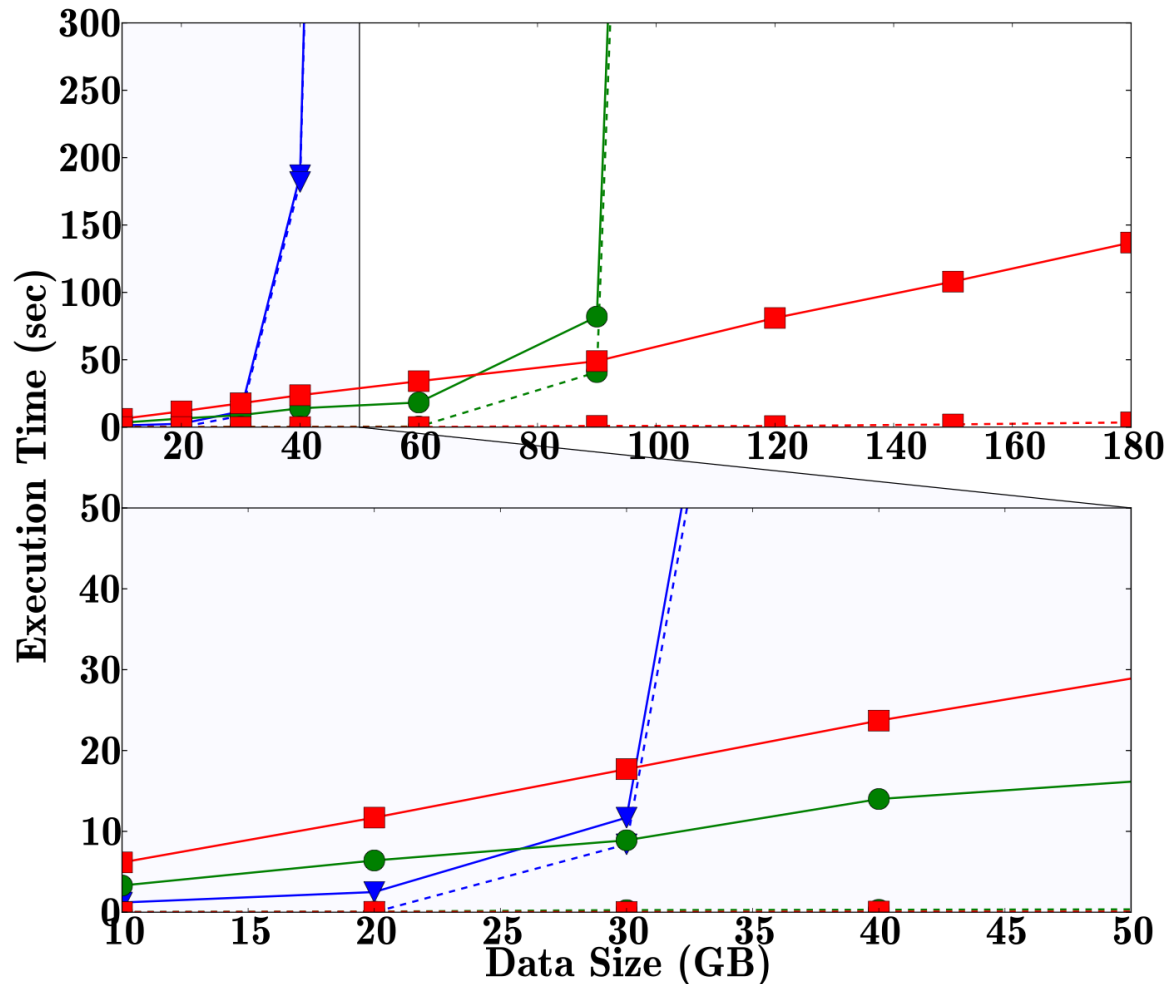


Performance comparison for realistic workflow. The workflow ran 4x faster on Tachyon than on MemHDFS. In case of node failure, applications in Tachyon still finishes 3.8x faster.

Further Improve Spark's Performance



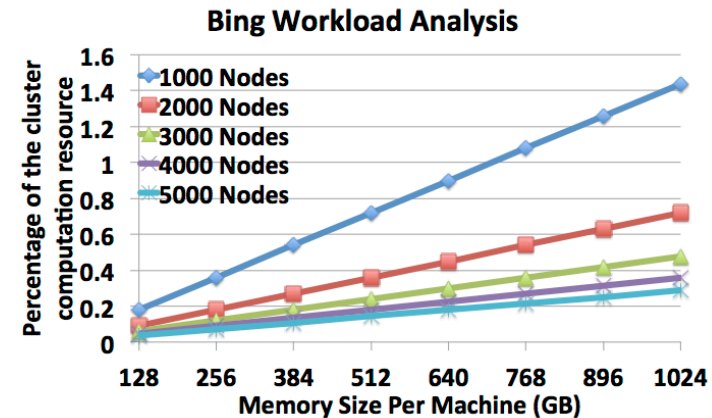
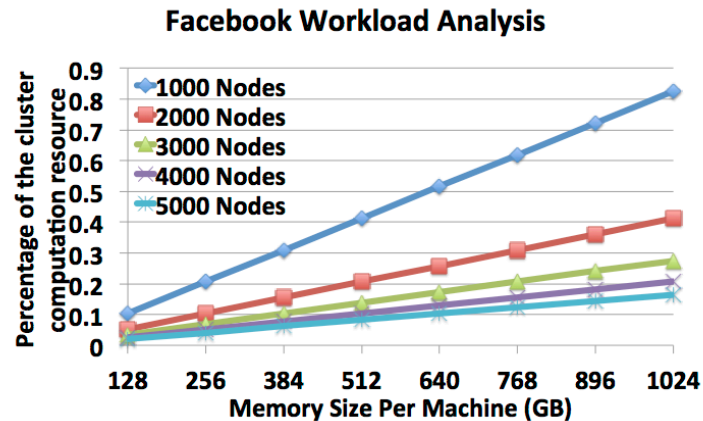
**Grep
Program**



Recomputation Resource Consumption

Bin	Tasks	% of Jobs	
		Facebook	Bing
1	1 - 10	85%	43%
2	11 - 50	4%	8%
3	51 - 150	8%	24%
4	151 - 500	2%	23%
5	> 500	1%	2%

Trace Summary



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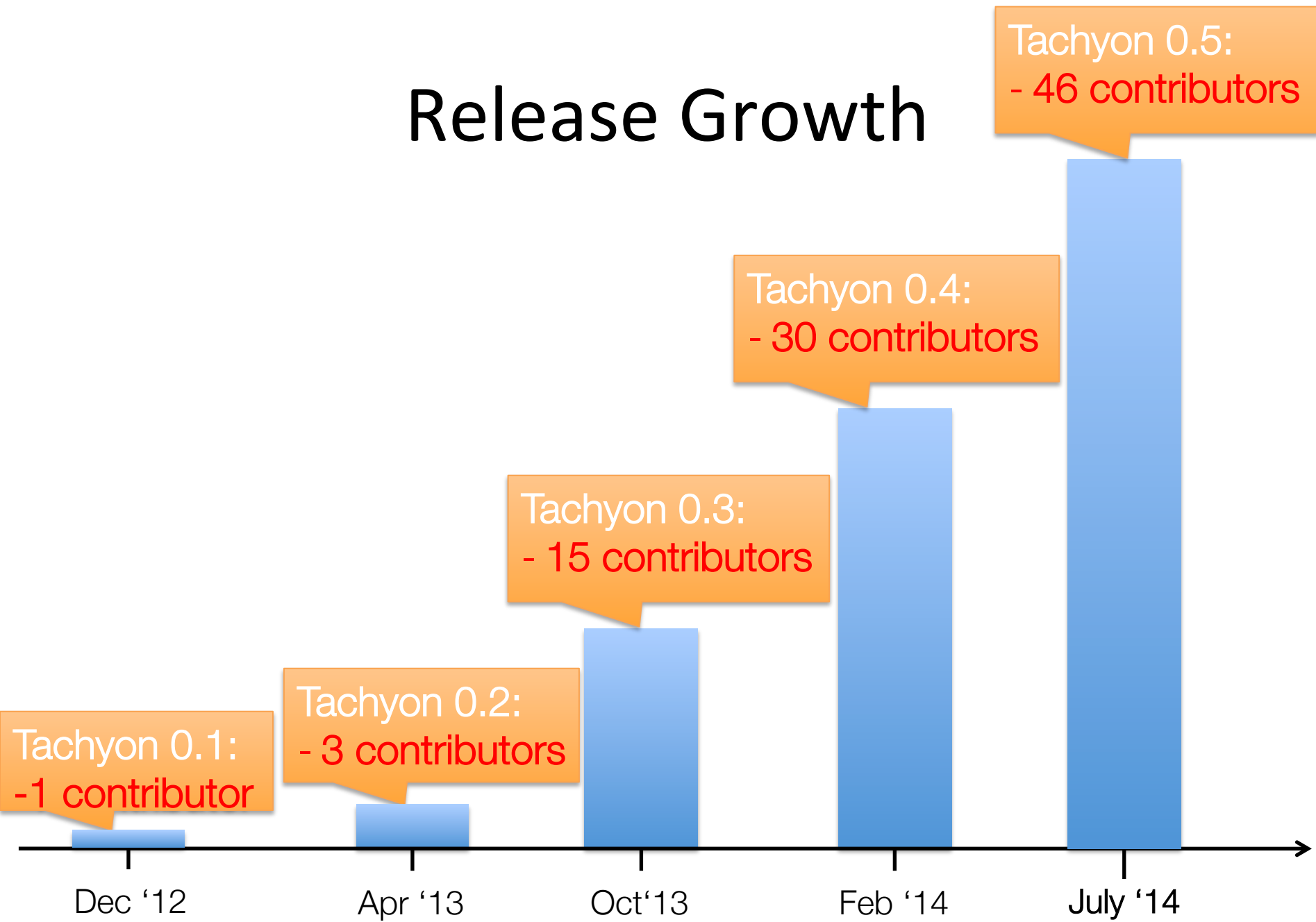
TACHYON Open Source Status

- Apache License 2.0, Version 0.5.0 (July 2014)

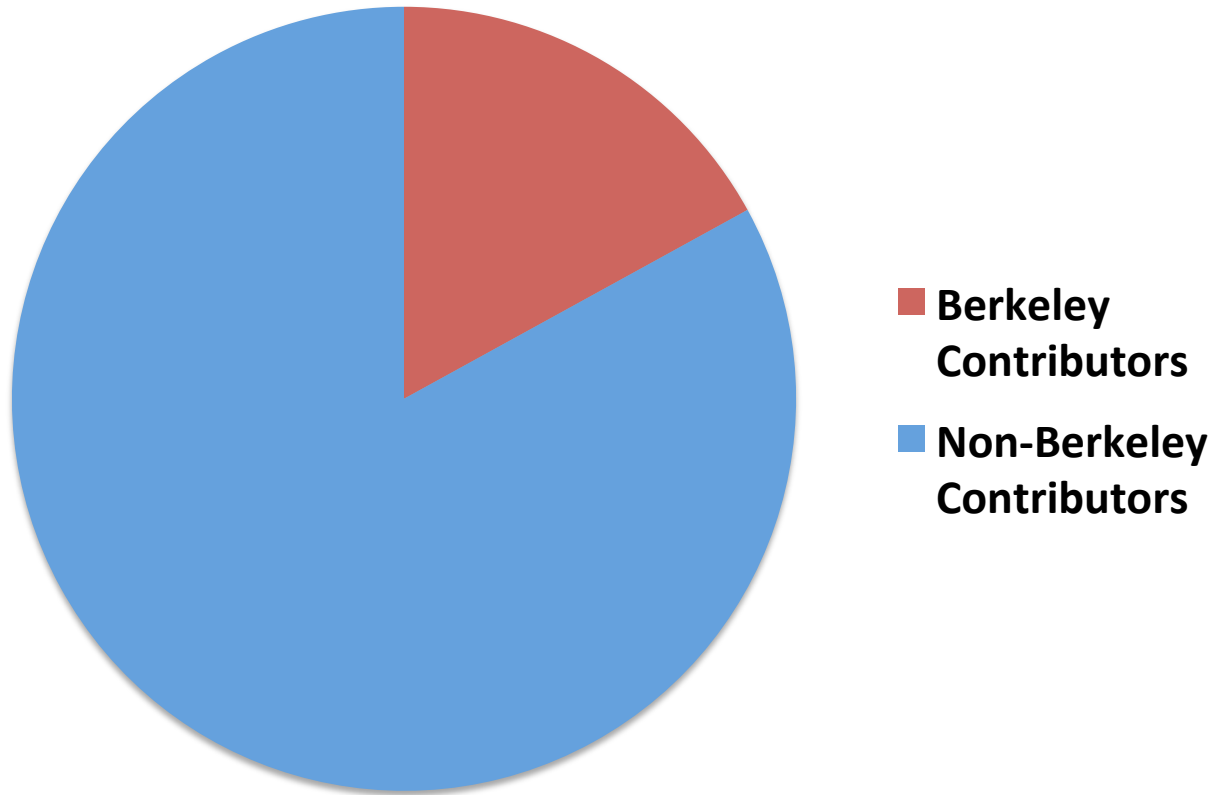


- Deployed at tens of companies
- 15+ Companies Contributing
- No code change for Spark and MapReduce applications.

Release Growth



Open Community



Thanks to our Code Contributors!

Aaron Davidson

Achal Soni

Ali Ghodsi

Andrew Ash

Anurag Khandelwal

Aslan Bekirov

Bill Zhao

Brad Childs

Calvin Jia

Chao Chen

Cheng Chang

Cheng Hao

Colin Patrick McCabe

David Capwell

David Zhu

Du Li

Fei Wang

Gerald Zhang

Grace Huang

Haoyuan Li

Henry Saputra

Hobin Yoon

Huamin Chen

Jey Kottalam

Joseph Tang

Juan Zhou

Lukasz Jastrzebski

Manu Goyal

Mark Hamstra

Mingfei Shi

Mubarak Seyed

Nick Lanham

Orcun Simsek

Pengfei Xuan

Qianhao Dong

Qifan Pu

Raymond Liu

Reynold Xin

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Sean Zhong

Seonghwan Moon

Shivaram Venkataraman

Srinivas Parayya

Tao Wang

Thu Kyaw

Timothy St. Clair

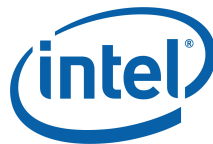
Vamsi Chitters

Xi Liu

Xiang Zhong

Xiaomin Zhang

Zhao Zhang



**Tachyon
is in Fedora 20**

Thanks to Redhat!

Commercially supported

by  **Atigeo**[™]

**and running in dozens of
their customers' clusters**

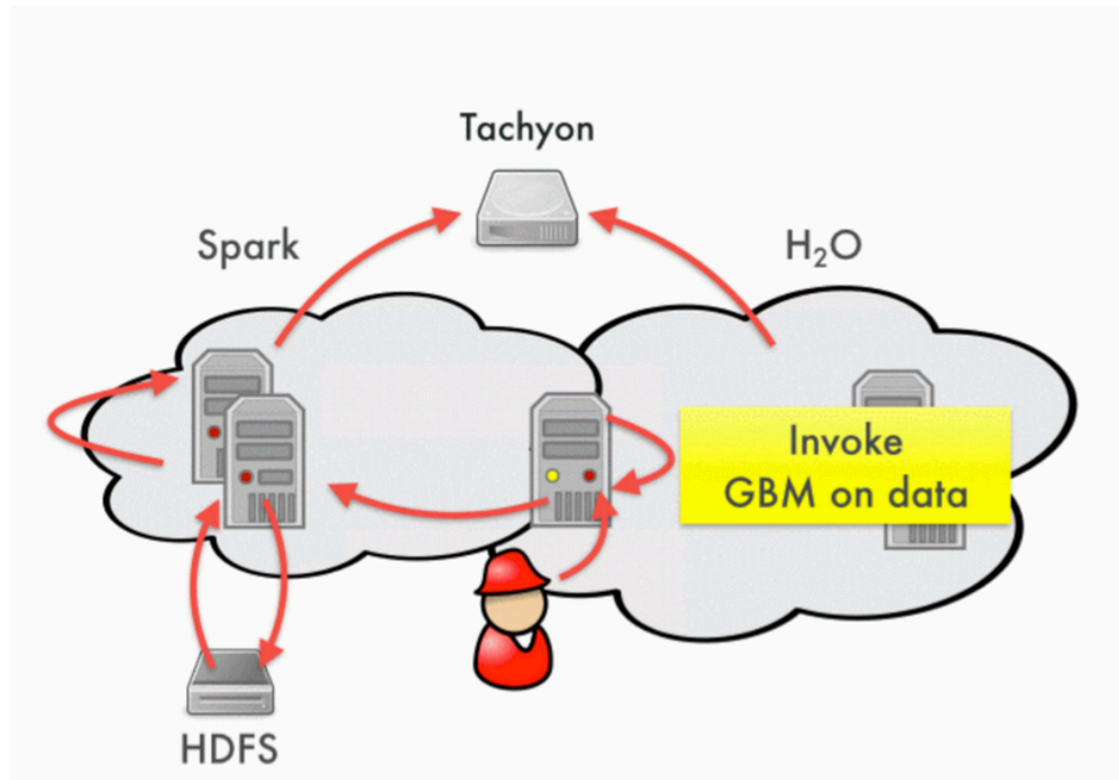
Tachyon is the
Default Off-Heap Storage
Solution for  Spark



TACHYON in




Today, data gets parsed and exchanged between Spark and H2O via **Tachyon**. Users can interactively query big data both via SQL and ML from within the same context.



Reaching wider communities: e.g. GlusterFS

blog.gluster.org/2014/08/glusterfs-and-tachyon/



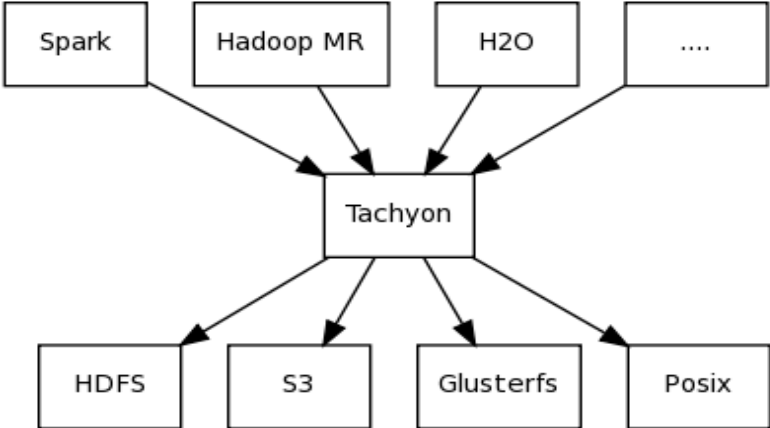
DOCUMENTATION CONTACT ABOUT

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by huamin on August 7, 2014 [← Return to Blog Home](#)

Glusterfs and Tachyon

Tachyon, an in-memory distributed filesystem, is among the most dynamic projects in big data analytics stack. It provides java io like API, support Apache Spark, and vastly improves Spark's performance under large data set.



```
graph TD; Spark[Spark] --> Tachyon[Tachyon]; HadoopMR[Hadoop MR] --> Tachyon; H2O[H2O] --> Tachyon; Ellipsis[....] --> Tachyon; Tachyon --> HDFS[HDFS]; Tachyon --> S3[S3]; Tachyon --> Glusterfs[Glusterfs]; Tachyon --> Posix[Posix];
```

Under Filesystem Choices (Big Data, Cloud, HPC, Enterprise)



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Short Term Roadmap (0.6 Release)

- Ceph Integration (Ceph Community)
- Hierarchical Local Storage (Intel)
- Performance Improvement (Yahoo)
- Multi-tenancy (AMPLab)
- Mesos Integration (Mesos Community)
- ***Many more*** from AMPLab and Industry Collaborators.

Features

- Memory Centric Storage Architecture
- Lineage in Storage (alpha)
- Hierarchical Local Storage
- Data Serving
- Scalable metadata management
- Different hardware
- More...
- Your Requirements?

Data Serving: An Example

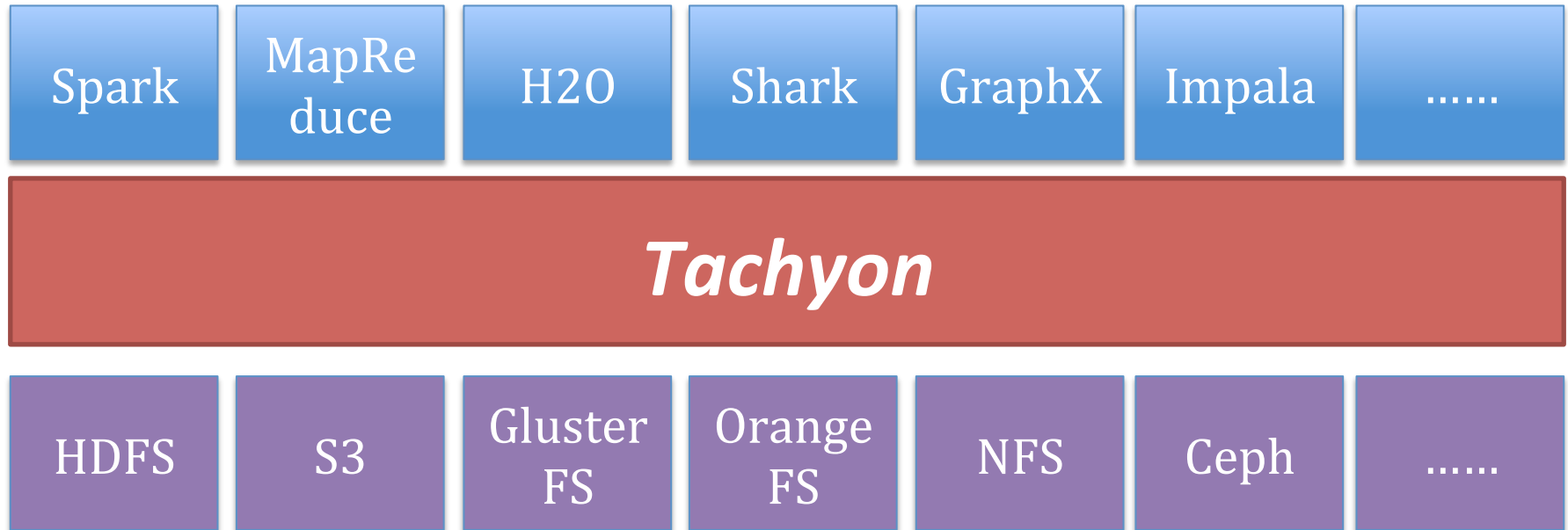
Data Analytics Pipeline:
Query the results of batch jobs.

What do we need?

Sequential I/O + Random Access!

Tachyon Goal?

Better Assist Other Components



Welcome Collaboration!

Thanks!

Questions?

- *More Information:*
 - Website: <http://tachyon-project.org>
 - Github: <https://github.com/amplab/tachyon>
 - Meetup: <http://www.meetup.com/Tachyon>
- Email: haoyuan@cs.berkeley.edu