

### Pfimbi: Accelerating Big Data Jobs Through Flow-Controlled Data Replication

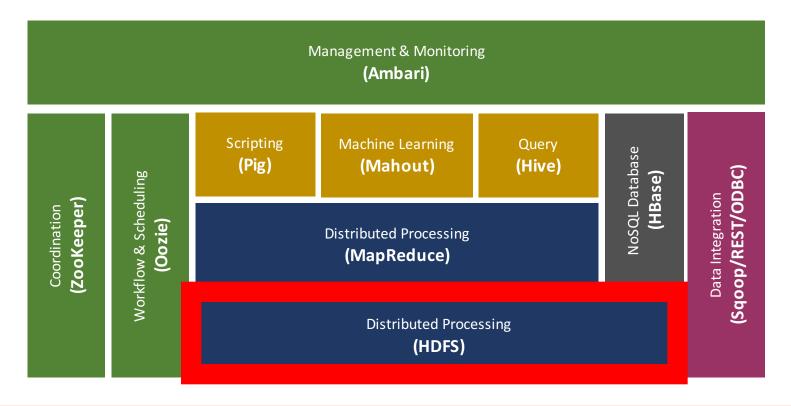
Simbarashe Dzinamarira\*

Florin Dinu<sup>△</sup> T. S. Eugene Ng\*

\*Rice University, △EPFL



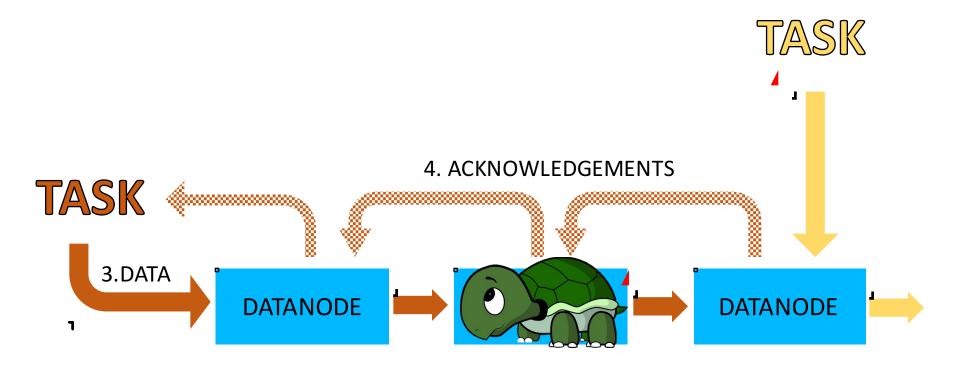
### DFSs have a critical role on the Big-Data landscape



- Rich ecosystem of distributed systems around Hadoop and Spark
- Predominantly use HDFS for persistent storage
- A performant HDFS benefits all these system



### Synchronous data replication in HDFS and its shortcomings

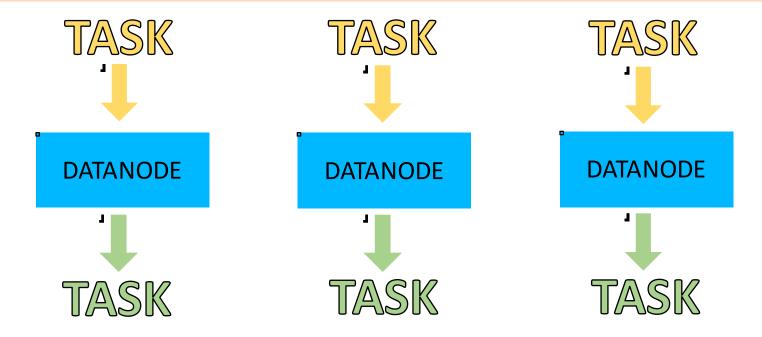


- Bottlenecks affect the whole pipeline
- Contention between primary writes and replication



# Synchronous replication seldom helps boost application performance

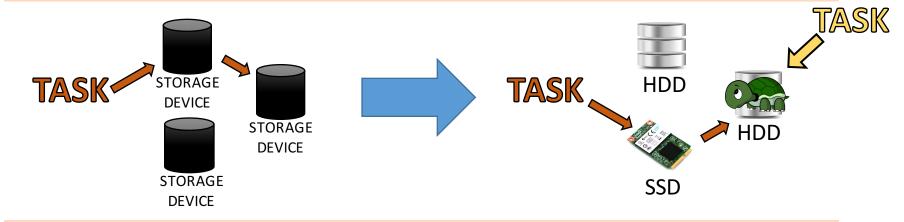
- In a study by Fetterly et al. only about 2% of data was read within 5 minutes of being written [TidyFS: USENIX ATC 2011]
- Fast networks reduce the cost of non-local reads
- There can be data locality without replication



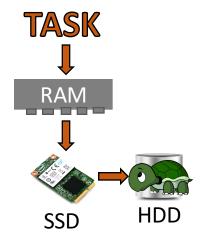


### Synchronous replication impedes industry efforts to improve HDFS

Heterogeneous storage

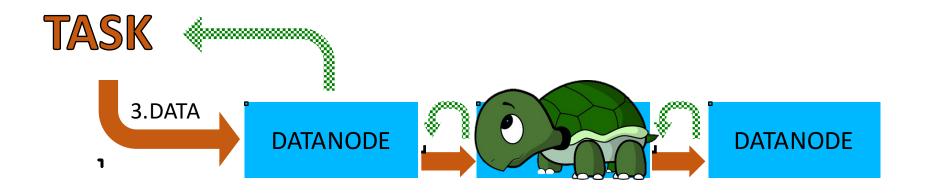


Memory as a storage medium





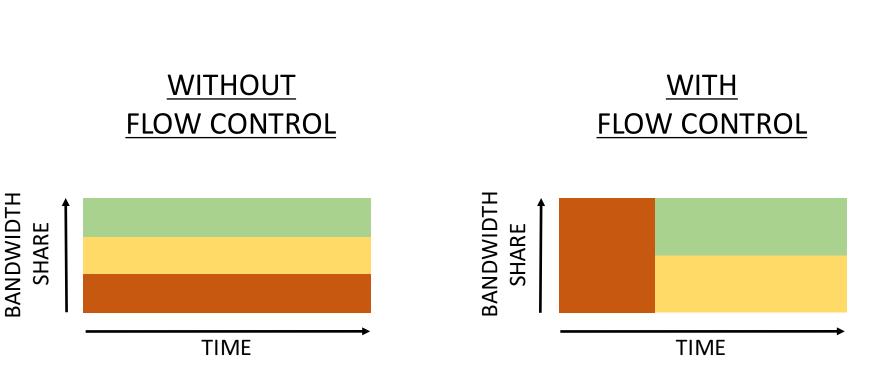
### Asynchronous replication relieves the effects of pipeline bottlenecks





### Beside asynchronous replication, we need flow control to manage contention







# Pfimbi effectively supports flow controlled asynchronous replication

Allows diverse flow control policies

Cleanly separates mechanisms from policies

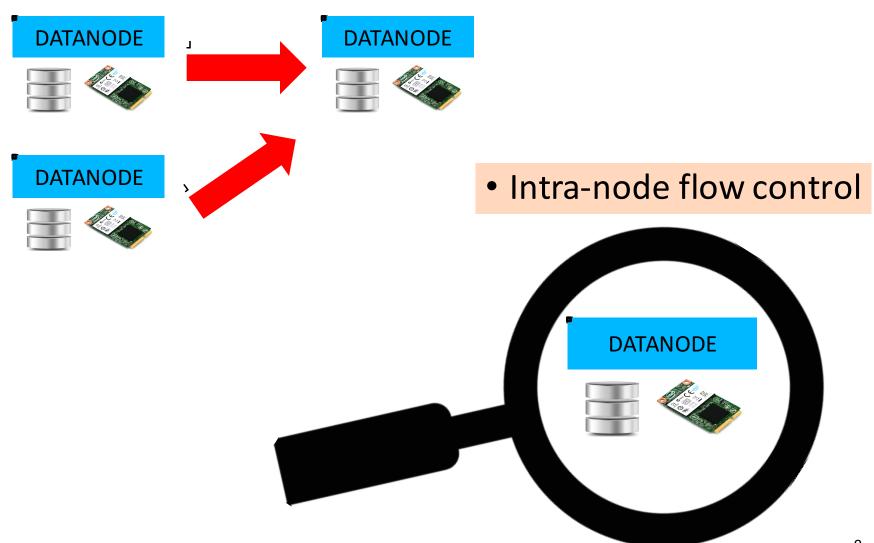
Isolates primary writes from replication

Avoids IO underutilization



#### Pfimbi Overview

• Inter-node flow control

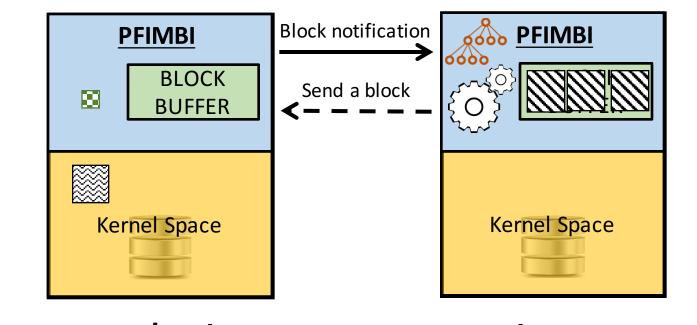




#### Inter-node flow control

- Client API: (# of replicas, # of synchronous replicas)
- Timely transfer of replicas to ensure high utilization
- Flexible policies for sharing bandwidth





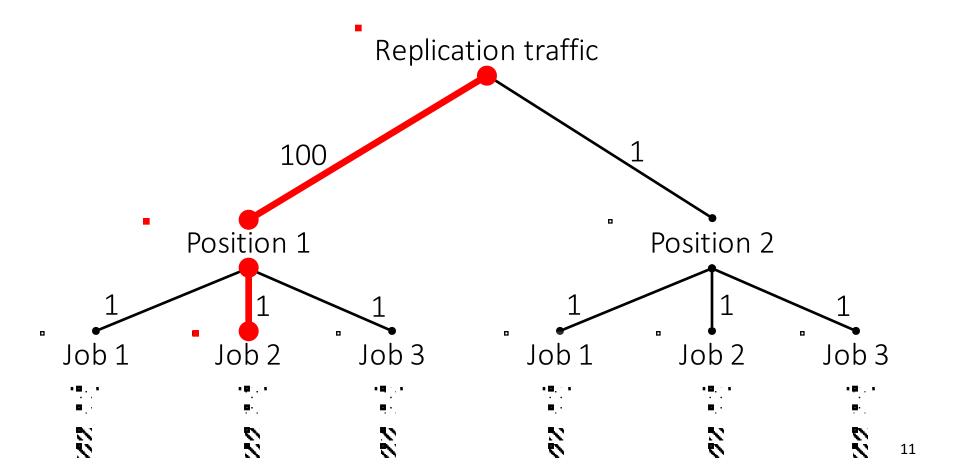
Synchronous

Asynchronous



# Hierarchical flow control allows Pfimbi to implement many IO policies

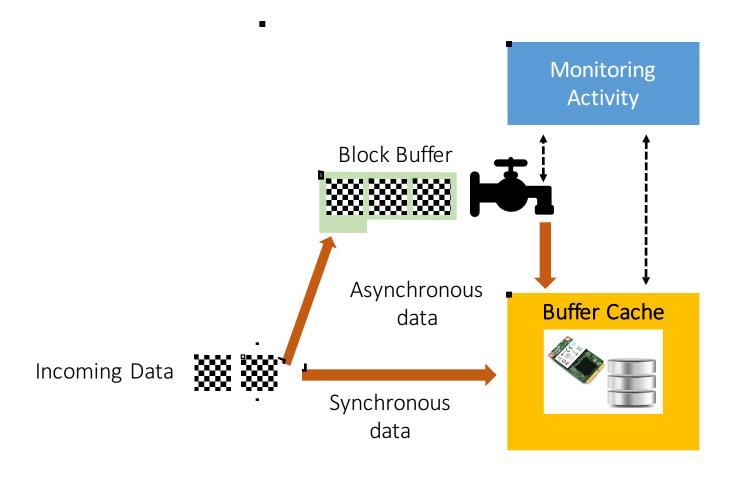
- Example 1 : prioritize replicas earlier in the pipeline
- Example 2 : fair sharing of bandwidth between jobs





#### Intra-node flow control

- Isolate synchronous data from asynchronous data
- Avoid IO underutilization





# Intra-node flow control Pfimbi's strategy

- Keep the disk fully utilized
- Limit the amount of replication data in the buffer cache

Buffer cache

Threshold for asynchronous replication :  $\mathbf{T} + \mathbf{\delta}$ 

OS threshold for flushing buffered data: T

	Typical Values
T	10% of RAM (~13GB)
δ	500MB
Buffer Cache	20% of RAM (~26GB)



### Additional topics that are discussed in detail in the paper

- Other activity metrics and their shortcomings
- Consistency
  - We maintain read and write consistency
- Failure handling
  - Same mechanism as in HDFS to recover from failures
- Scalability
  - Pfimbi's flow control is distributed

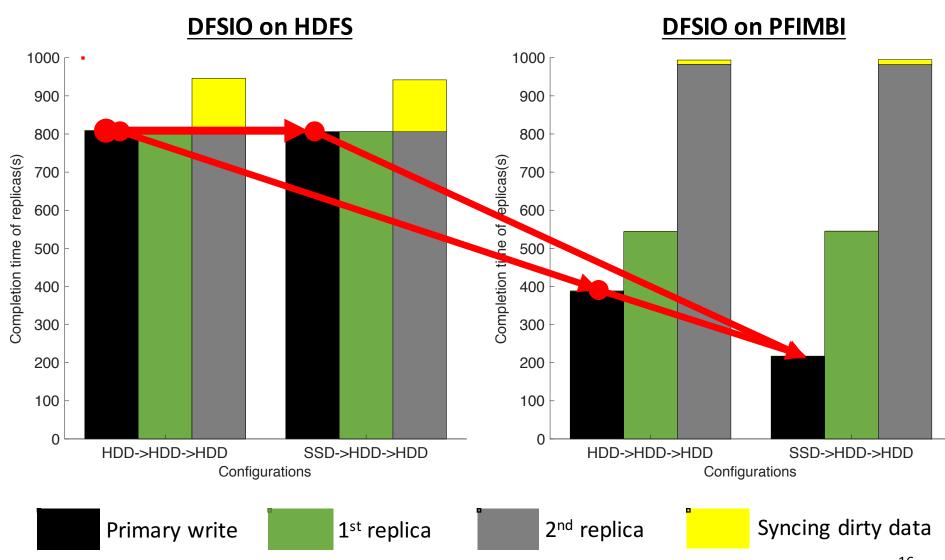


#### **Evaluation**

- 30 worker nodes
  - NodeManagers collocated with DataNodes
- 1 Master node
  - ResourceManager collocated with NameNode
- Storage
  - 2TB HDD
  - 200GB SSD
  - 128GB DRAM

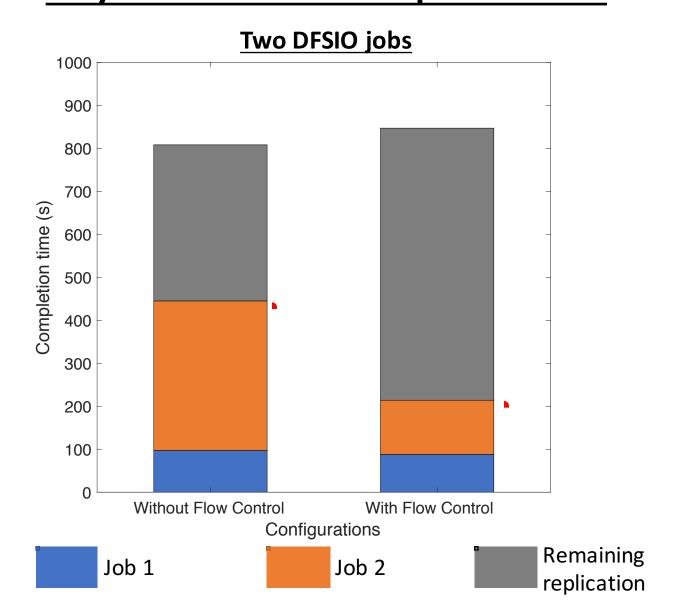


# Pfimbi improves job runtime and exploits SSDs well



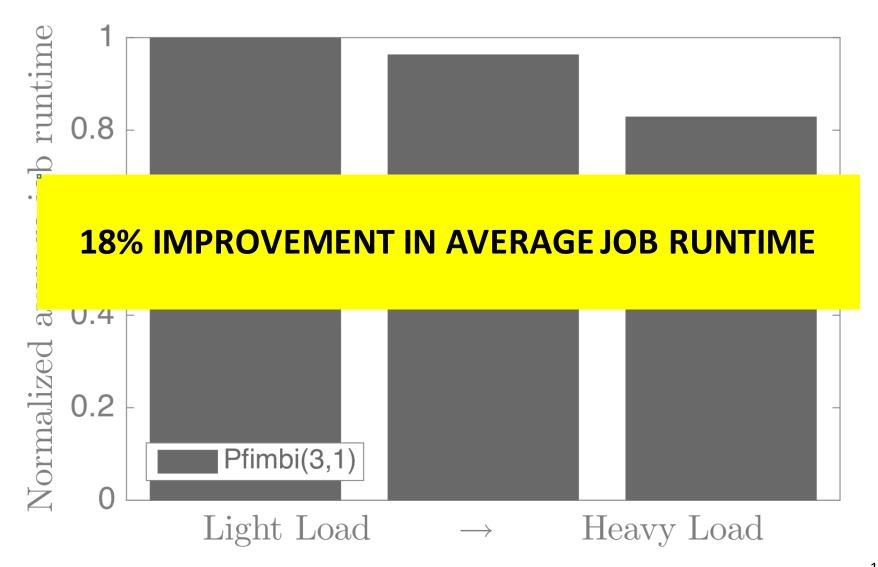


### Necessity of flow control when doing asynchronous replication



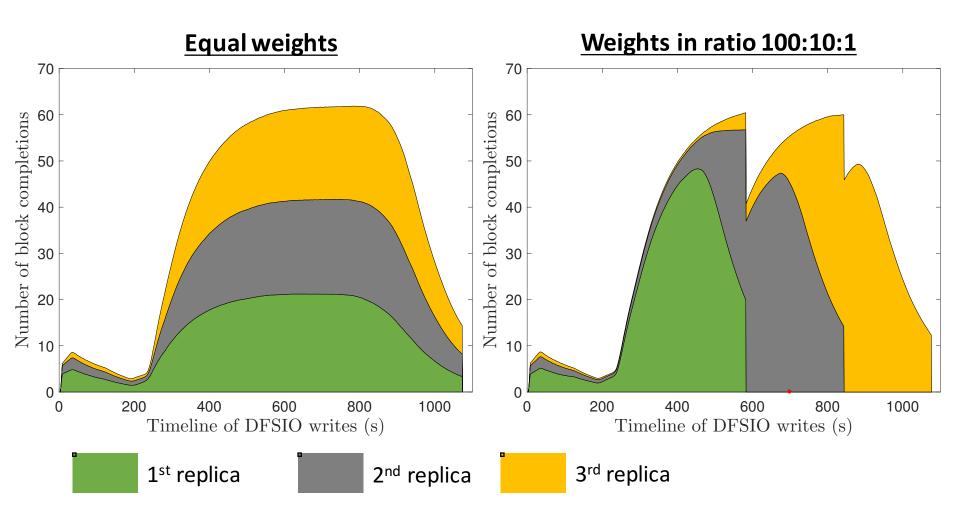


# Pfimbi performs well for a mix of different jobs: SWIM workload





### Policy Example: Pfimbi can flexibly divide bandwidth between replica positions





#### Related Work

- Sinbad [SIGCOMM 2013]
  - Flexible endpoint to reduce network congestion
  - Does not eliminate contention within nodes
- TidyFS [USENIX ATC 2011]
  - Asynchronous replication
  - No flow control leads to arbitrary contention
- Retro [NSDI 2015]
  - Fairness and prioritization using rate control
  - Synchronous replication



#### Conclusion

- Pfimbi effectively supports flow controlled asynchronous replication
  - Successfully balances managing contention and maintaining high utilization
  - Expressive and backward compatible with HDFS