

# Performance Improvement of MapReduce Applications using Flame-MR

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Performance Results

- Big Data has been adopted by many organizations
- Hadoop is one of the most used frameworks
- Limited performance
  - Redundant memory copies
  - Disk overhead
- Existing alternatives must rewrite applications

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## Our proposal: Flame-MR

- Completely new event-drivent architecture
- Transparent performance improvement of Hadoop applications
- In-memory computing
- Overlapping of data movement and computation

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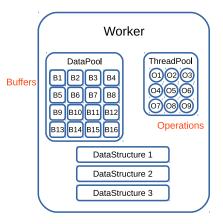
Conclusions & Future Work

- Event-driven architecture
- Efficient memory management
- In-memory sort and merge algorithms
- Support for iterative workloads
- Full compatibility with Hadoop

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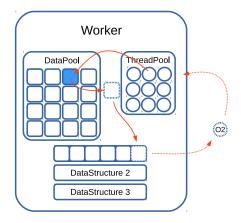
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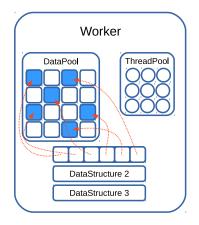
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## Testbed configuration

- Evaluations conducted on Amazon EC2
  - 33 i2.4×large instances
  - Interconnected via GbE
  - Instance characteristics
    - 2  $\times$  8-core Intel Xeon E5-2670 v2 2.6 GHz
    - 122 GB RAM
    - 4 × 800 GB SSD
- Experiments automated by the Big Data Evaluator tool (BDEv)
  - Configuration of the frameworks
  - Generation of input datasets
  - Collection of results
  - Available at http://bdev.des.udc.es

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## Frameworks

- Hadoop 2.7.2
- Flame-MR 1.0

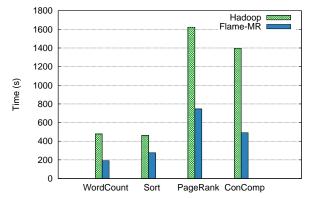
## Benchmarks

Benchmark	Characterization	Input data size
WordCount	CPU bound	500 GB
Sort	I/O bound	500 GB
Connected Components	Iterative (5 iter.)	40 GB
PageRank	Iterative (5 iter.)	40 GB

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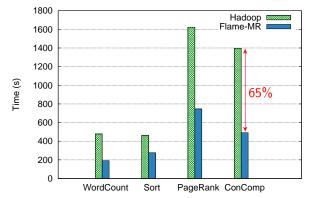
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# Conclusions & Future Work

## Conclusions

- Flame-MR improves transparently the performance of Hadoop
- Results show high performance improvements
  - Up to 65%
- Publicly available at http://flamemr.des.udc.es

### Future work

- Development of new features
  - Automatic load balancing
- Evaluation of Flame-MR using real-world use cases

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Acknowledgments









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