Sparkling: Identification of Task Skew and Speculative Partition of Data for Spark Applications

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Peilong Li (*), Xiaobing Huang (+), Tian Zhao (+),
Yan Luo (*) and Yu Cao (*)
(*) University of Massachusetts Lowell,
(+) University of Wisconsin Milwaukee
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Motivations

Motivations	Improvements
Spark web UI and other instrumentation tools (Ganglia etc.) → X Why an application does not yield expected performance	Sparkling web tool provides: ✓ Better execution visualization ✓ In-depth statistical analysis
X Spark may generate unbalanced task (data/ computational skew).	Two partitioning methods: ✓ Decremented partitioning ✓ Application-aware partitioning

Outline

- Part I. Sparkling Web Tool
- Part II. Mitigate Data Skew in Spark Application

Part I. Sparkling Web Tool

- Sparkling enhances Spark application development:
 - More detailed metrics
 - Task view and executor view
 - Overall performance with statistics
 - Help avoid "data skew" and "small task"

Sparkling: Spark Data Web Analyzer

Task Timeline

Executor Timeline

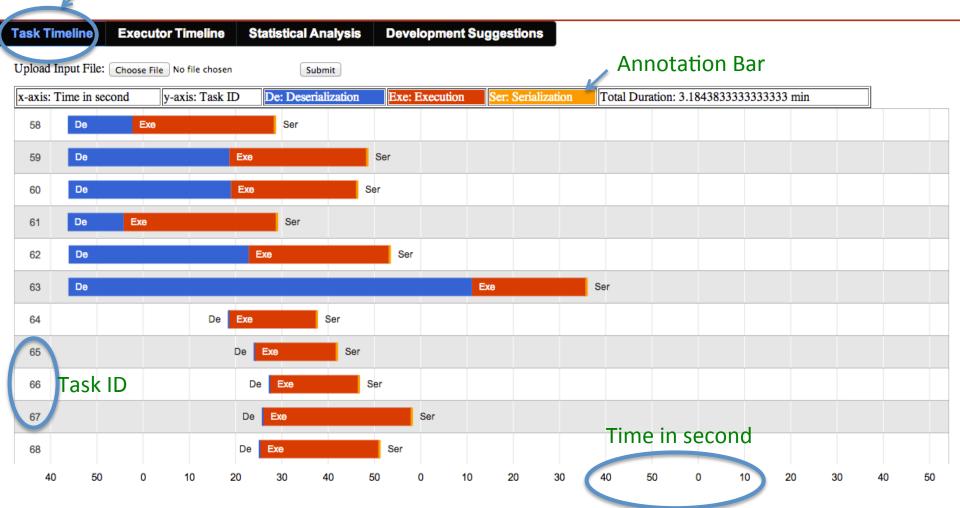
Statistical Analysis

Development Suggestions

Upload Input File: chCheck out demo page: http://pr01.uml.edu

Task Timeline View

Click this tab



Executor Timeline View

Click this tab



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Time in second

Statistical Analysis

	Item	Max Time (ms)	Min Time (ms)	Average Time (ms)	Standard Deviation (ms)
1	Deserialization Stage	45329	9	1423.2756944444445	6924.886105639271
2	Execution Stage	51332	0	1869.0159722222222	4788.835964934546
3	Serialization Stage	149	0	8.395138888888888	22.471620466036303
4	Scheduler Delay	47130	19	2199.002777777777	7196.321615357318

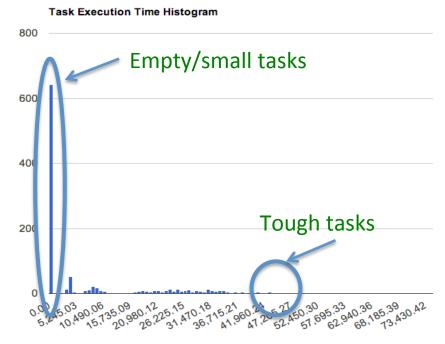
Item	Empty Task Percentage (%)	
1 Empty Task (Small Task)	46.875	





Serialization Time Histogram

Scheduler Delay Histogram



- In contrast with Spark web UI, the table shown above provides in-depth analysis on statistical properties of the app behavior.
- Histogram use case:
 - Find a high percentage of "empty/small tasks".
 - Find tough tasks -> avoid data skew.

Duration in ms

Part II. Mitigate Data Skew in Spark Application

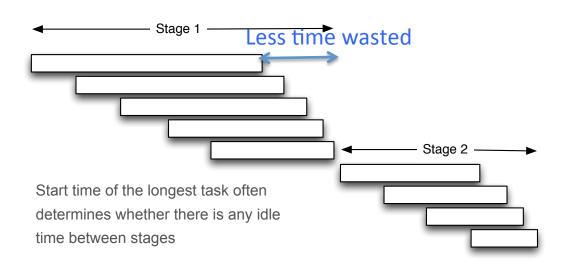
- Spark speculative mechanism: handles stragglers
- Biomedical multimedia processing application
 computational skew.
- Spark naïve partitioning: divide workload into slices with equal head-count.

Problem

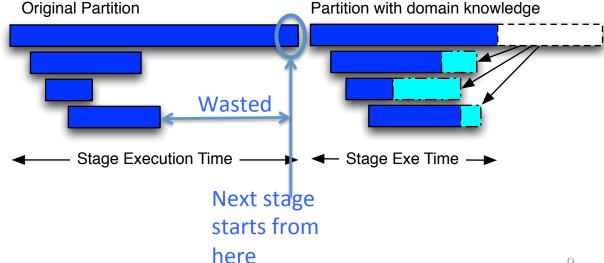
What if there are unbalanced workload on each slice?

Two Ways to Mitigate Data Skew

Decremented workload:



 Partition RDD with domain knowledge:



Our Test Benchmark

- PIR Benchmark [1]
 - Image Query and Transmedia query application (SIFT, LDA).
 - Dataset: Wikipedia articles (2866 multimedia documents: images + texts)[2]
- Domain knowledge
 - Edge Pixel Percentage (EPP) = # of Edge Pixels / Total # of Pixels
- Why use EPP
 - Image/Transmedia query app use local feature.
 - EPP is a good stand for local feature complexity.

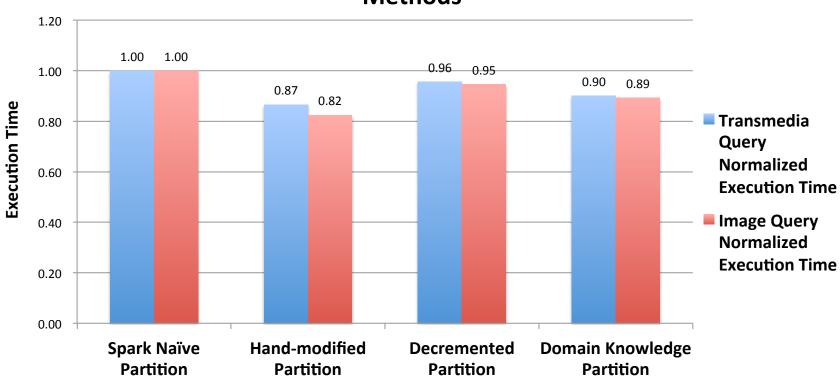
Reference:

[1] Xiaobing Huang; Tian Zhao; Yu Cao, "PIR: A Domain Specific Language for Multimedia Retrieval," Multimedia (ISM), 2013 IEEE International Symposium on , vol., no., pp.359,363, 9-11 Dec. 2013

 $^{7/6}2^{1/4}$ UCSD Statistical Visual Computing Lab. http://www.svcl.ucsd.edu/projects/crossmodal/

Experiment Results

Image and Transmedia Query Apps With Different Partition Methods



Conclusion

- Sparkling web tool provides more insight for developers on how to improve app performance by decreasing data skew and small tasks.
- Our proposed automatic data skew mitigation algorithm shows a 11% increase of performance on our biomedical benchmarks.