A Moldable Online Scheduling Algorithm and Its Application to Parallel Short Sequence Mapping

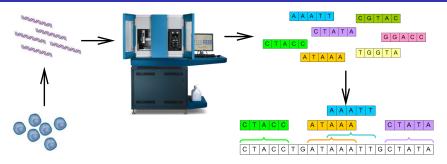
Erik Saule, Doruk Bozdağ, Umit V. Catalyurek

Department of Biomedical Informatics, The Ohio State University {esaule,bozdagd,umit}@bmi.osu.edu

Scheduling for Large Scale Systems, May 2009

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Motivation



Sequencing

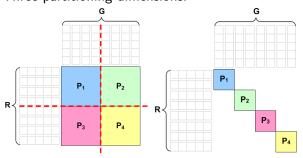
- Next generation sequencing instruments (SOLiD, Solexa, 454) can sequence up to 1 billion bases a day
 - Hundreds of millions of 35-50 base reads

Mapping

- Map reads to a reference genome efficiently (Human genome: 3Gb)
- Sequential mapping takes about a day
- Need fast, parallel algorithms that can handle mismatches

Parallel Short Sequence Mapping[Bozdag et al., IPDPS 09]

Three partitioning dimensions:



$$P(m_g, m_r, m_s) = c_{gs} \frac{G}{m_g} + c_g \frac{G}{m_g m_s} + c_{rs} \frac{R}{m_r} + (c_r + c_c \frac{G}{m_g m_s}) \frac{R}{m_r m_s}$$

Partitioning on m processors is finding minimum $P(m_g,m_r,m_s)$ such that $m_g m_r m_s \leq m$

This talk

A cost efficient approach

To reduce cost, Ohio SuperComputing Center is building a bioscience dedicated cluster. It will host a Short Sequence Mapping service.

- Laboratories submits mapping request over the network.
- The service computes the mapping using the parallel algorithm.
- And sends the result back.

This talk

How to schedule the mapping request?

This talk

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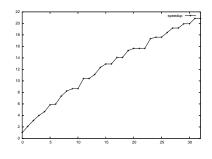
Outline of the Talk

- Introduction
- 2 A Moldable Scheduling Problem
- 3 Deadline Based Online Scheduler (DBOS)
- 4 Experiments
- Conclusion

Parallel Short Sequence Mapping

The important facts:

- can adapt to different number of processor
- good runtime prediction function
- no super linear speed up
- non convex speedup function (steps)
- no preemption



Moldable Scheduling

Instance

- m processors
- n tasks
- Task i arrives at r_i
- The execution of i on j processors takes $p_{i,j}$ time units



Solution

- Task i is executed on π_i processors
- Task i starts at σ_i
- Task *i* finishes at $C_i = \sigma_i + p_{i,\pi_i}$

Objective Function

Flow time

The flow time is the time spent in the system per a task $F_i = C_i - r_i$.

- Does not take task size into account.
- Optimizing the maximum flow time is unfair to small tasks.
- Optimizing the average flow time should starve large tasks.

Stretch [Bender et al. SoDA 98]

The stretch is the flow time normalized by the processing time of the task $s_i = \frac{C_i - r_i}{\rho_{i,1}}$.

- It provides a better fairness between tasks.
- Optimizing maximum stretch avoids starvation.

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Stretch [Bender et al. SoDA 98]

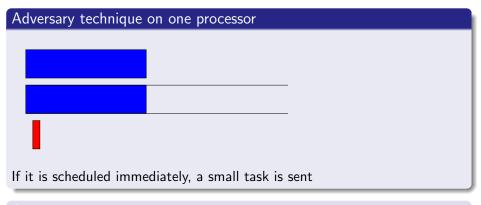
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Adversary technique on one processor	
A large task enters in the system	

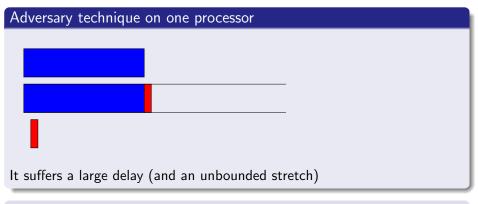
On several processors

There are similar techniques on several processors but there are more complicated and thus less prone to appear in practice.



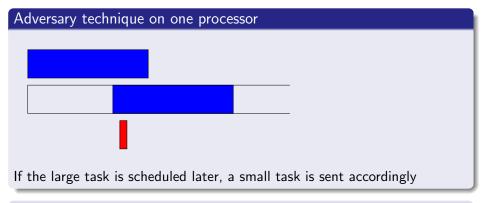
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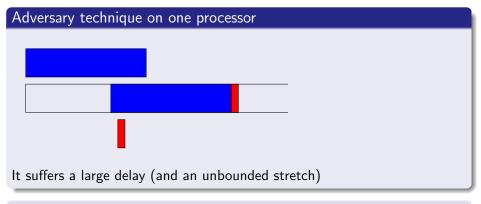
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Adversary technique on one processor



It suffers a large delay (and an unbounded stretch)

On several processors

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Principle of the Deadline Based Online Scheduler (DBOS)

- All tasks running concurrently should get the same stretch to maximize efficiency
- Using the optimal maximum stretch as an instant measure of the load
- Aim at a more efficient schedule than the optimal instant maximum stretch one to deal with still-to-arrive tasks

The DBOS Algorithm

Targeting a maximum stretch S

Task *i* must complete before the deadline $D_i = r_i + p_{i,1}S$.

Moldable Earliest Deadline First (MEDF)

- Considers task in deadline order.
- Allocates the minimum number of processors to each task to completes before the deadline.
- Schedules the task as soon as possible without moving any other task.

$DBOS(\rho)$

- Estimate the optimal maximum stretch S^* using a binary search.
- The deadline problem is solved by MEDF.
- Build a schedule of good efficiency of stretch ρS^* .

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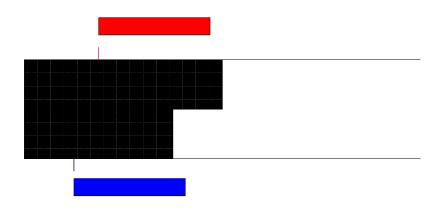
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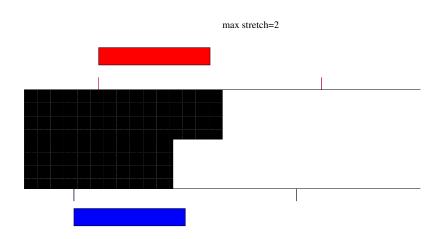
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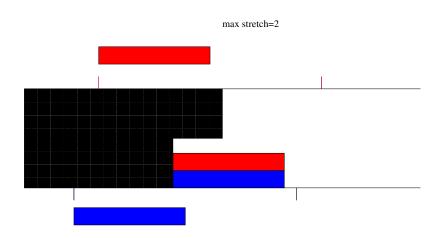
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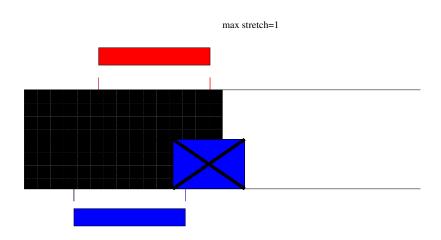
A system with two pending tasks



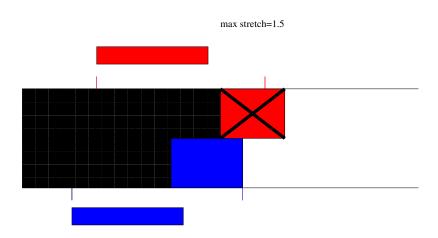
Deadlines induced by a stretch of 2



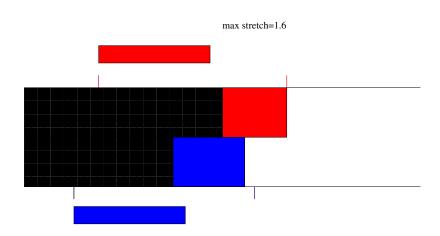
A maximum stretch of 2 is reachable



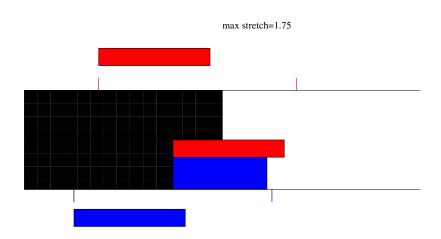
But 1 is not



Neither 1.5



The optimal stretch is 1.6



The online parameter $\rho=1.1$ leaves much more space (thanks to MEDF).

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An Iterative Process [Sabin et al, JSSPP 06]

The algorithm

- $\forall i, \pi_i \leftarrow 1, mark[i] \leftarrow false$
- $\sigma \leftarrow schedule(\pi)$
- while $\exists i \mid mark[i] = false$
 - Get unmarked i such that $p_{i,\pi_i} p_{i,\pi_i+1}$ is maximal and positive
 - $\pi_i \leftarrow \pi_i + 1$
 - $\sigma' \leftarrow schedule(\pi)$
 - if $avgflow(\sigma') < avgflow(\sigma)$
 - $\sigma \leftarrow \sigma'$
 - else
 - $\pi_i \leftarrow \pi_i + 1$; $mark[i] \leftarrow true$

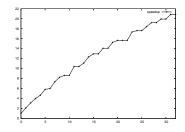
schedule

schedule is a conservative backfilling algorithm. Unspecified, we used FCFS.

An Iterative Process [Sabin et al, JSSPP 06]

Properties

- Optimizing flow time
- Claimed to outperform fair share
- Parameter-less



Improvement

If the speedup function is non convex or has steps. The algorithm gets stuck. (It was originally tested with a model where the speedup is convex) Modification:

- Get unmarked i and k such that $(p_{i,\pi_i}-p_{i,\pi_i+k})/k$ is maximal and positive
- $\pi_i \leftarrow \pi_i + k$

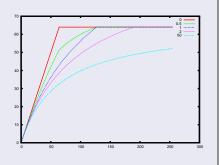
First Experimental Setting

Goal: assess performance on a well known setting

Downey model

Two parameters:

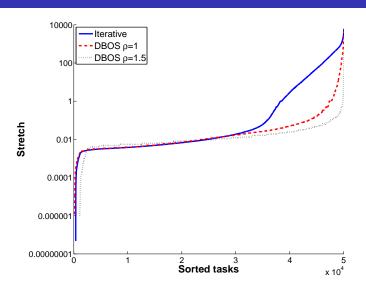
- Average parallelism (64)
- Distance to linear speedup



Generation

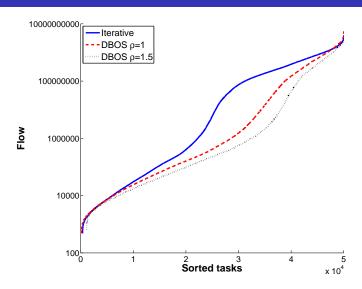
- 512 processors
- First 5000 tasks of SDSC Par 96 (From the Feitelson archive)
- Sequential time : total execution time
- Average parallelism : between number of used processor and 512
- Distance to linear speedup : between 0 and 2

Downey model results



DBOS generates less tasks with high stretch.

Downey model results



DBOS leads to better flow time. Iterative could be improved.

Second Experimental Setting

Goal: test case reflecting the cluster usage

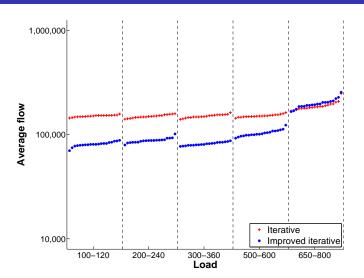
Generation

- 512 processors
- Each task corresponds to one lab studying one genome
- Speedup according to the runtime prediction function
- 2000 tasks are uniformly distributed in an time interval
- Changing the span of the interval to control the load

Real data

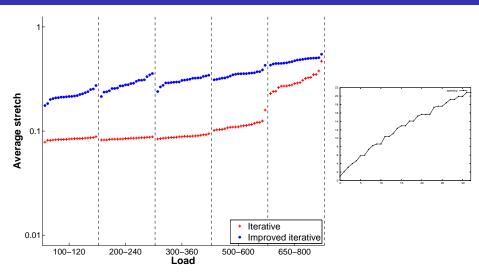
		Genome	Size
		E. Coli	4.6 million
Sequencing machine	Reads	Yeast	15 million
454 GS FLX Genome Analyzer	1 million	A. Thaliana	100 million
Solexa IG sequencer	200 million	Mosquito	280 million
SOLiD system	400 million	Rice	465 million
		Chicken	1.2 billion
		Human	3.4 billion

Mapping: Improvement on Iterative (flow)



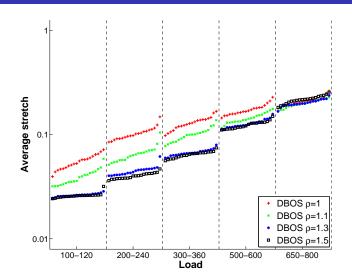
The improvement really improves. The iterative got stuck.

Mapping: Improvement on Iterative (stretch)



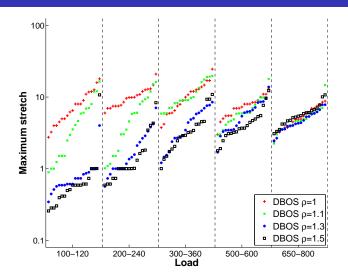
Getting stuck is good for stretch since it avoids interrupting tasks. They are just lucky.

Mapping: the online parameter (average stretch)



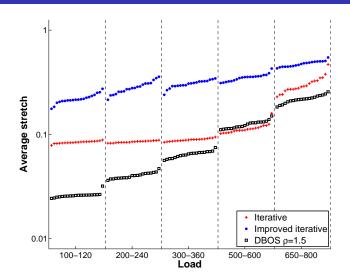
Quickly drops with ρ . Step at $\rho = 1.3$.

Mapping: the online parameter (maximum stretch)



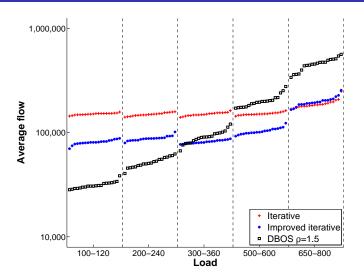
Max stretch is kept at a reasonable level. The online parameter ρ is very helpful here.

Mapping: DBOS vs Iterative (average stretch)



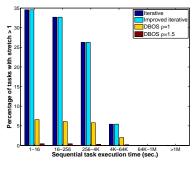
DBOS leads to much better stretch (even when iterative got stuck).

Mapping: DBOS vs Iterative (average flow)

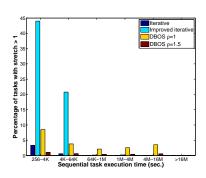


Confirm there is room for improvement for Iterative. DBOS is not bad.

Mapping: DBOS vs Iterative (Fairness Issues)



Downey model



Short Sequence Mapping

The Iterative algorithm leads to high stretch for a lot of the smaller tasks. DBOS has better performance and less fairness issues thanks to stretch optimization.

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The end

Conclusion

- A cluster dedicated to bioscience will be built.
- To provide fairness stretch should be considered instead of flow time.
- An scheduling algorithm is proposed to optimize stretch and avoid worst case online scenario.
- Which performs well on Short Sequence Mapping application.

Perspective

- Investigate other way to avoid worst case scenarios.
- Study more simple algorithm to get reference points.
- Build the service!

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