

Online and Offline List Batching

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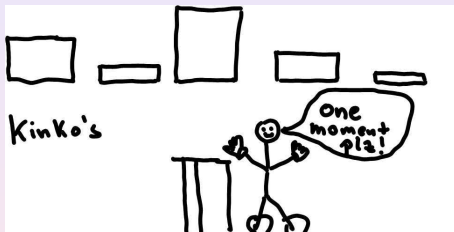
Osaka Kinko's



List Batching

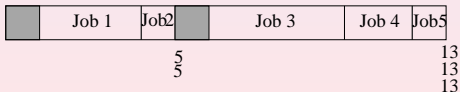
n jobs are given to be processed in batches

Job 1 Job 2 Job 3 Job 4 Job 5



all jobs in a batch finish at the same time

there is a setup time to get a batch started



the object is to minimize the average completion time

List Batching, continued...

Jobs with processing requirements p_1, p_2, \dots, p_n are given and have to be processed in that order.

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- The completion time C_i of job i is the completion time of its batch.



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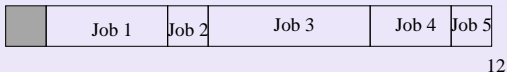
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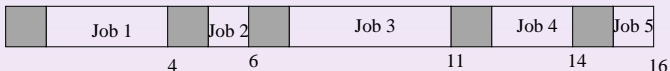
Jobs are given to the machine in batches. Every batch has a setup time of 1.

- The completion time C_i of job i is the completion time of its batch.
- The object is to batch the jobs in such a way that $\sum C_i$ is minimized.

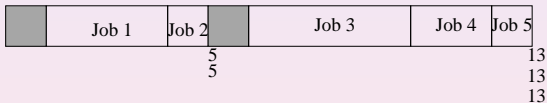
Our Example



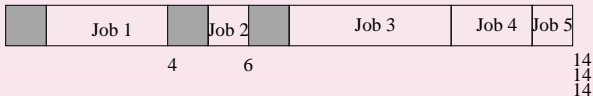
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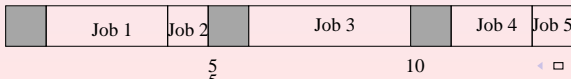
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49



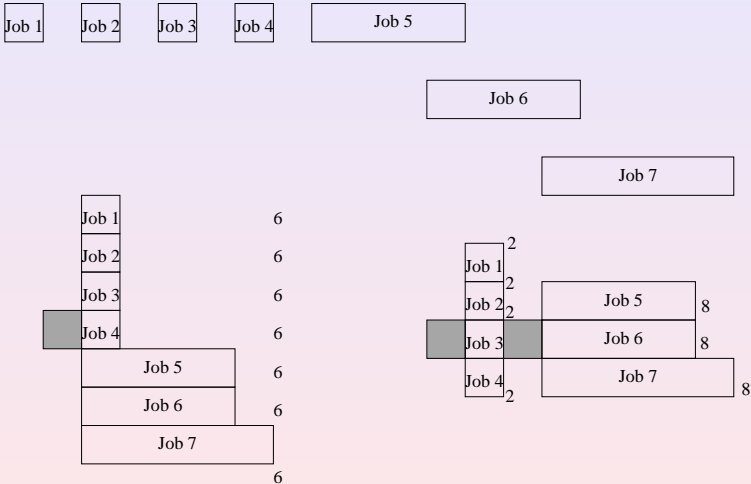
52



48



Sor far List s-Batching, but here is also List p-Batching



History

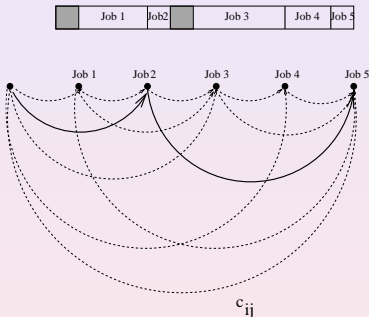
Large body of work on offline batching, i.e.

- [Coffman, Yannakakis, Magazine, Santos, 1990]
- [Albers, Brucker, 1993]
- [Brucker, Gladky, Hoogeveen, Kovalyov, Pots, Tautenhahn, Velde, 1998]



List s-Batching, Offline

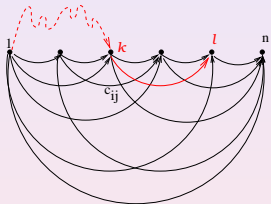
The **offline list s-batching problem** can be reduced to a path problem¹[AB92]:



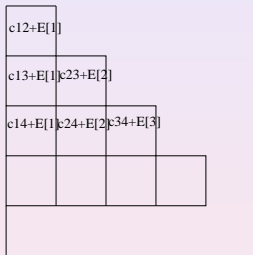
- $c_{ij} = (n - i)(s + P_j - P_i)$ with $P_i = \sum_{\ell=0}^i p_\ell$

¹List p-Batching has a similar reduction

A Simple Dynamic Program



$$E[1]=0$$

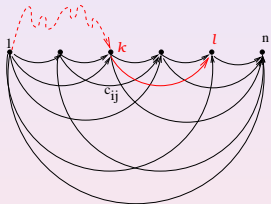


$E[l]$ = the shortest path from 1 to l

$$E[l] = \min_{1 \leq k < l} \{E[k] + c_{kl}\}$$

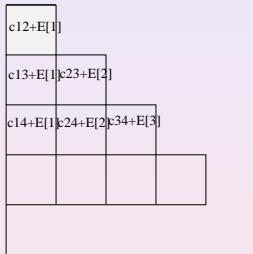
$$O(n^2)$$

A Simple Dynamic Program



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$$E[2]$$

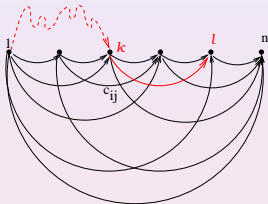


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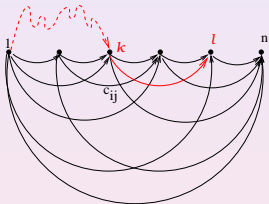
$c_{12}+E[1]$			
$c_{13}+E[1]$	$c_{23}+E[2]$		
$c_{14}+E[1]$	$c_{24}+E[2]$	$c_{34}+E[3]$	

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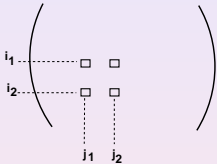
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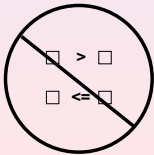
$$O(n^2)$$

How to do this in $O(n \log n)$

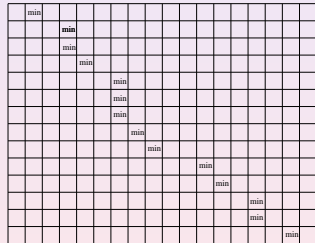
Monge Property



$$C_{i_1 j_1} + C_{i_2 j_2} \leq C_{i_2 j_1} + C_{i_1 j_2}$$

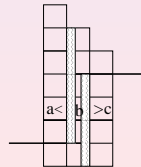
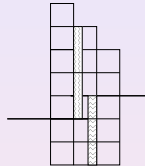
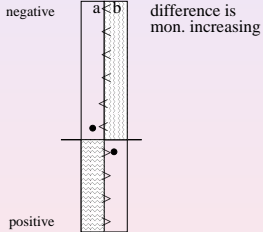


Totally Monotone



Various Inferences

Entire columns can be eliminated in $O(\log n)$ time:



The Online Protocol of the Dynamic Program

[LS91]

Protocol:

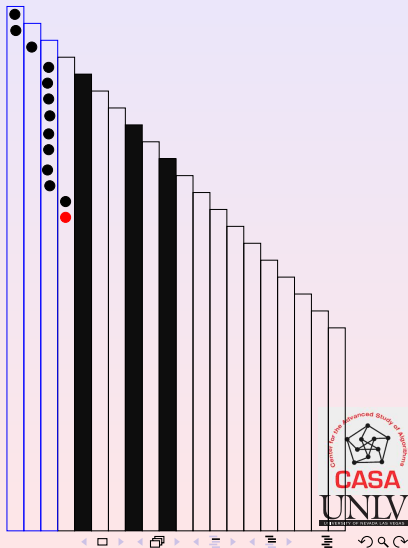
Once the minimum of the

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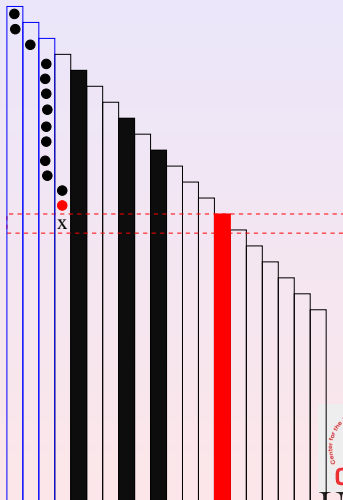
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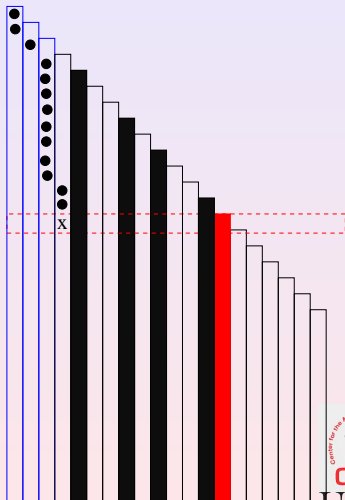
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Algorithm is $O(n \log n)$

The Hire/Fire/Retire Algorithm can be implemented in $O(n \log n)$

- Potential: number of rows + number of columns.
- Retire eliminates a column, not-retire eliminates a row, fire eliminates a column, not-fire eliminates a row.

$O(n)$ Algorithms:

- [LARSH 91]
- [Albers, Brucker 93]



A closed form for $p_i = s = 1$

Theorem ([BELN 04])

$$\text{optcost}[n] = \frac{m(m+1)(m+2)(3m+5)}{24} + k(n + m - k + 1) + \frac{k(k+1)}{2}$$

for $n = \frac{m(m+1)}{2} + k$

The optimal size of the first batch

$$= \begin{cases} m & \text{if } k = 0 \\ m \text{ or } m + 1 & \text{if } 0 < k < m + 1 \\ m + 1 & \text{if } k = m + 1 \end{cases}$$

Online List Batching

- Jobs J_1, J_2, \dots arrive one by one over a list.



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- Jobs J_1, J_2, \dots arrive one by one over a list.
- Job J_i must be scheduled before a new job is seen, and even before knowing whether current is the last job.
- For job J_i an online Algorithm must decide whether to
 - “batch”: to make J_i the first job of a new batch
 - “not to batch”: to add J_i to the current batch.



Competitiveness

- A measure of the performance that compares the decision made online with the optimal offline solution for the same problem.

For any sequence of jobs $\rho = \{J_1, J_2, \dots\}$

$cost_{\mathcal{A}}(\rho)$: cost of the schedule produced by \mathcal{A} for ρ
 $cost_{opt}(\rho)$ is the minimum cost of any schedule for ρ

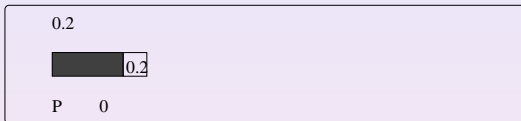
We say that \mathcal{A} is **C-competitive** if for each sequence ρ we have

$$cost_{\mathcal{A}}(\rho) \leq C \cdot cost_{opt}(\rho)$$



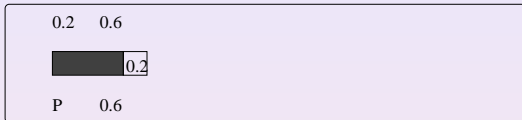
Algorithm $\text{PSEUDOBATCH}(B)$

- $\text{PSEUDOBATCH}(B)$ maintains a variable P which will be the sum of the processing times of a set of recent jobs.
- When J_1 is received, P is set to 0. After receiving each subsequent J_i , $\text{PSEUDOBATCH}(B)$ first adds p_i to P .
- If $P > B$, $\text{PSEUDOBATCH}(B)$ batches and also sets P to zero.

Algorithm PSEUDOBATCH(B)

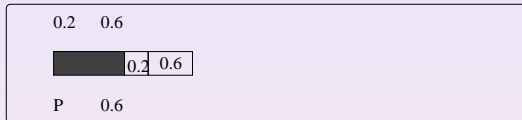
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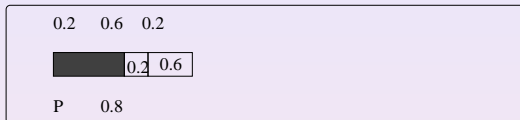
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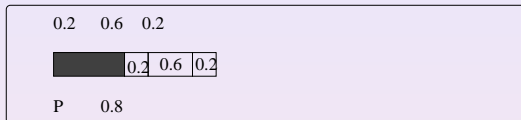
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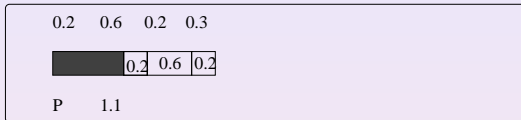
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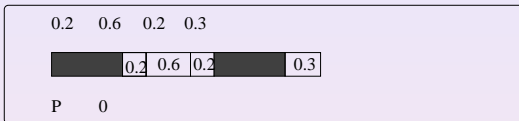
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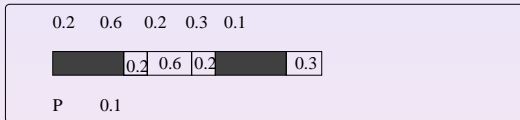
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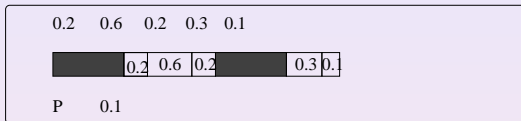
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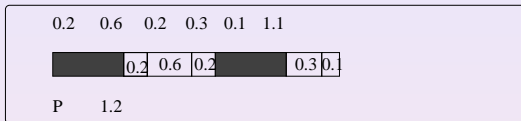
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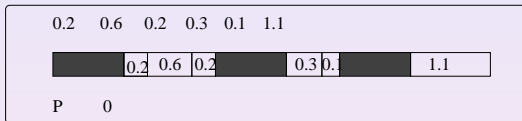
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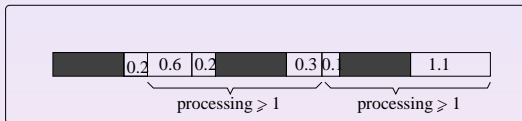
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- For PSEUDOBATCH(1): $C_i \leq \#batches + S_i + 1$
- $\#batches \leq 1 + S_i$
- Thus $C_i \leq 2 + 2S_i$, which implies the result.



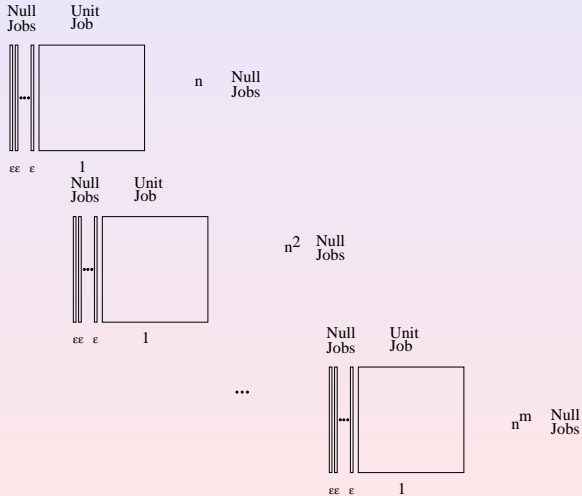
PSEDOBATCH(1) is Optimal

Theorem ([BELN 04])

The competitiveness of any deterministic online algorithm for the list s -batch problem is at least 2.

- Construct an adversary such that any deterministic algorithm will perform “poorly”.
- Adversary uses **Null Jobs**.
- Null Jobs are jobs with “arbitrarily” small processing times.

Lower Bound Adversary



Proof Sketch

Proof.

- Let m be a large integer; the sequence ends
 - a: the first time \mathcal{A} does not batch,
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- case b** similar...



Small jobs are needed...

- The next result shows that the exact competitiveness of 2 relies on the fact that the jobs may be arbitrarily small.
- In fact, if there is a positive lower bound on the size of the jobs, it is possible to construct an algorithm with competitiveness less than two.

Theorem ([BELN 04])

If the processing time of every job is at least p , then $\mathcal{A} = \text{PSEUDO BATCH}(\sqrt{p+1})$ is C -competitive, where

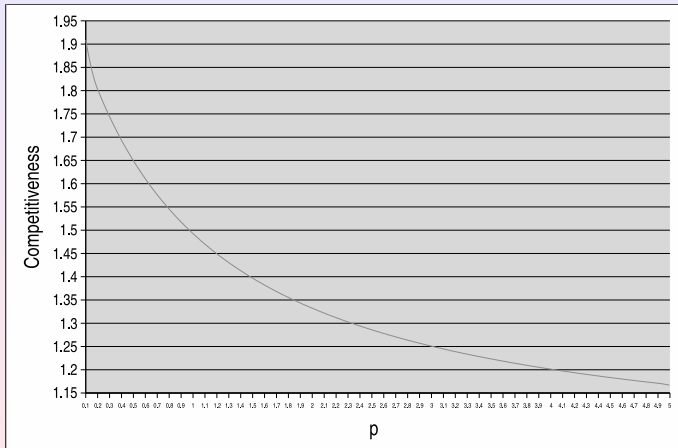
$$C = \min \left(\frac{1 + \sqrt{p+1}}{\sqrt{p+1}}, \frac{p+1}{p} \right).$$



CASA

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If jobs are at least p ...



The uniform case of $p_i = s = 1$

Define \mathcal{D} to be the online algorithm which batches after jobs: 2, 5, 9, 13, 18, 23, 29, 35, 41, 48, 54, 61, 68, 76, 84, 91, 100, 108, 117, 126, 135, 145, 156, 167, 179, 192, 206, 221, 238, 257, 278, 302, 329, 361, 397, 439, 488, 545, 612, 690, 781, 888, 1013, 1159, 1329, 1528, 1760, and $2000+40i$ for all $i \geq 0$.

Algorithm was found by computer

Theorem ([BELN 04])

\mathcal{D} is $\frac{619}{583}$ -competitive, and no online algorithm the list batching problem restricted to unit job sizes has competitiveness smaller than $\frac{619}{583}$.



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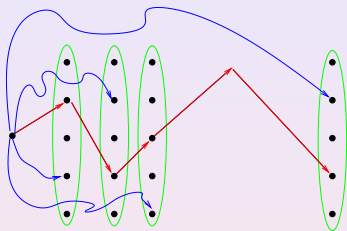
The case pf $p_i = s = 1$; upper bound

- It is easy to show that $\frac{619}{583}$ is an upper bound on the competitive ratio of the algorithms if there are more than 2000 jobs.
- The ratio is only tight when there are fewer than 2000 jobs.
- The algorithm was found by computer simulation.



The case of $p_i = s = 1$; Algorithm

Minimum Competitiveness Layered Graph Problem

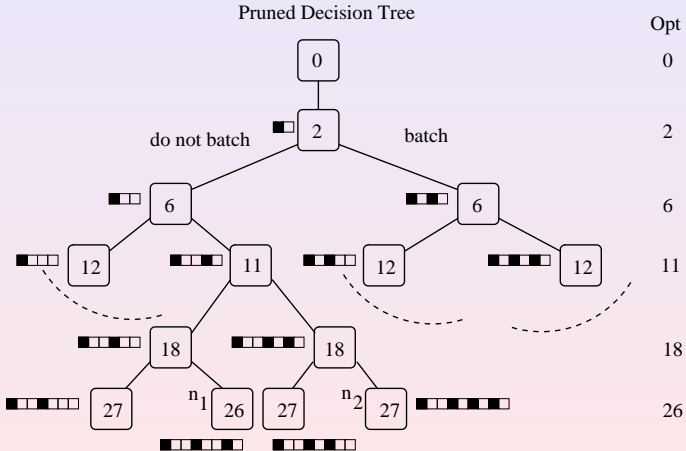


→ optimal paths
 → competitive path

nodes in layer k :
 k jobs requested
 $m = \# \text{batches}$
 $b = \# \text{jobs in current batch}$

- Schedule are combined into classes.
- A class has schedules where there are m batches, the last batch contains b jobs, and k jobs have been requested.

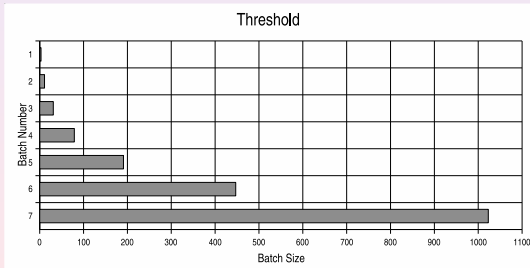
The case of $p_i = s = 1$; Lower Bound Proof



Competitiveness for p-Batching

THRESHOLD:

batches for the ℓ^{th} time whenever the processing requirement of the next job $\geq (\ell + 1)2^\ell - 1$,
i.e. 3, 11, 31, 79, ...



Competitiveness for p-Batching

Theorem ([BELN 04])

THRESHOLD is 4-competitive. No deterministic online algorithm for the list p-batch problem can have competitiveness less than 4.

Proof.

Lower bound proof a bit subtle.... □

Open Problems: Weighted Batching

Problem

Given n jobs with

- 1 processing times p_1, \dots, p_n
- 2 non-negative weights $w_1 \dots w_n$.
- 3 offline

Find an order and s -batching that minimizes $\sum w_i C_i$.

- Problem is NP-hard.
- Sort jobs in order of “priorities” $\frac{w_i}{p_i}$ then PSEUDOBATCH(1) is a 2-approximation.

PTAS ?