Approximating the Throughput by Coolest First Scheduling

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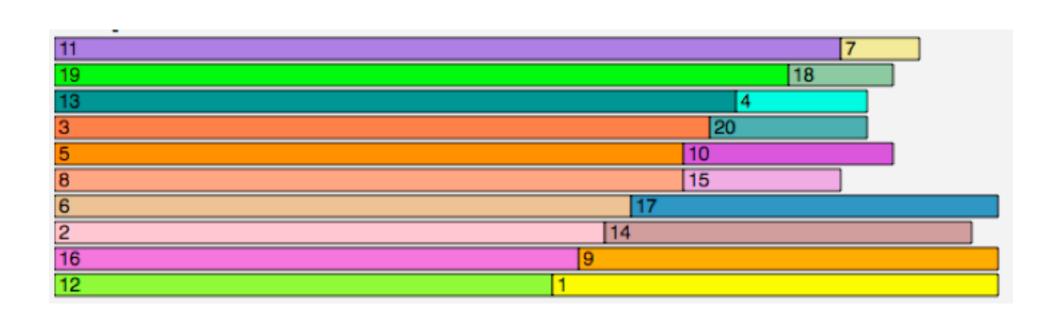
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Motivation



When the CPU temperature exceeds some threshold

- some models freeze until CPU cooled down enough
- some models just start melting

The model

jobs have

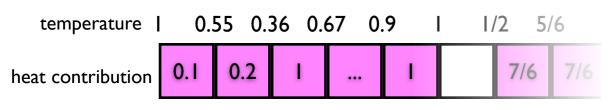
- unit processing time
- distinct heat contribution in [0,2]

CPU temperature

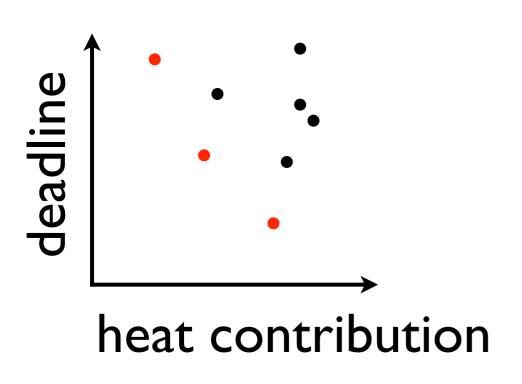
- is initially I
- should never exceed I

goal

- schedule a maximum number of jobs before (common) deadline D
- problem is NP-hard



What was known



more general on-line setting

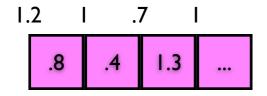
- jobs arrive online and have distinct deadlines, heat-contributions
- goal: maximize number of scheduled jobs respecting deadlines

[Chrobak, D, Hurand, Robert'08]

- scheduling any non-dominated job yields a 2-competitive on-line algorithm
- how good are these algorithms in the offline setting?

Minimizing makespan

HottestFirst



CoolestFirst 1.2 .8 .8 .4 .85 .4 .8 idle 1.3

offline setting

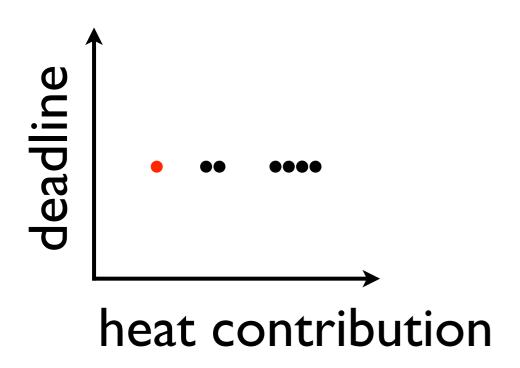
- all jobs are available at time 0 and have no deadline
- goal: produce a schedule respecting temperature threshold with smallest makespan

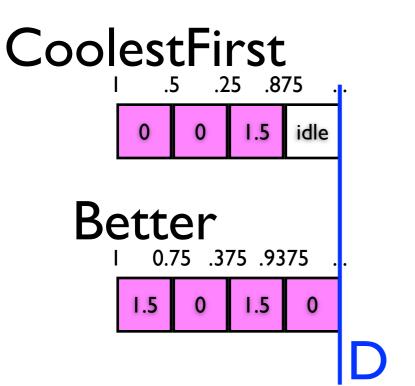
[Yang, Zhou, Chrobak, Zhang'08]

[Bampis, Letsios, Lucarelli, Markakis, Milis' 12]

- HottestFirst: schedule at any moment the job with the largest heat contribution among jobs that would respect the temperature threshold
- seems to be a good heuristic to keep temperature low at later point and so avoid idle times

Maximizing throughput





offline setting

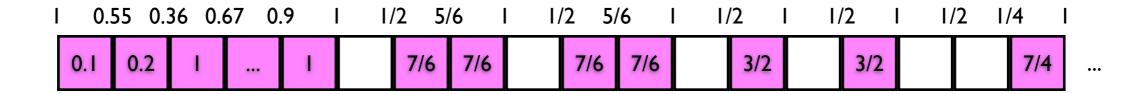
- all jobs are available at time 0
- all jobs have the same deadline D
- goal: schedule as many jobs before deadline D

Only the coolest job is non-dominated

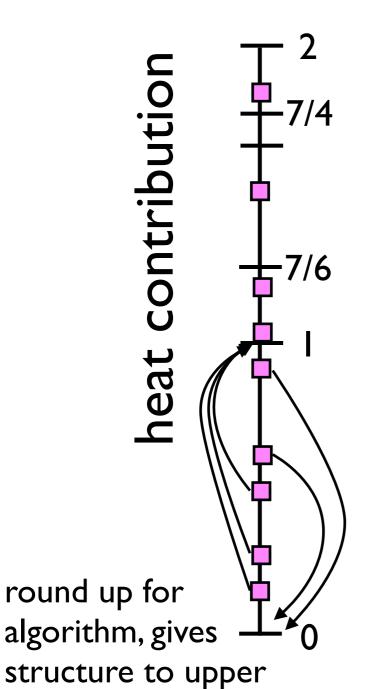
- CoolestFirst: schedule at any moment the job with the smallest heat contribution among allowed jobs
- It would be more clever to give priority to jobs with heat contribution ∈(1,2] since jobs from [0,1] can be scheduled anytime. But then we don't know how to analyse.

Analyze Coolest First

- Input: n jobs with different heat contributions ∈ [0,2] and a common deadline D
- CoolestFirst: At any time schedule the one with smallest heat contribution among the jobs that would respect the temperature threshold
- typical behavior: schedule get's less and less dense



Method of analysis



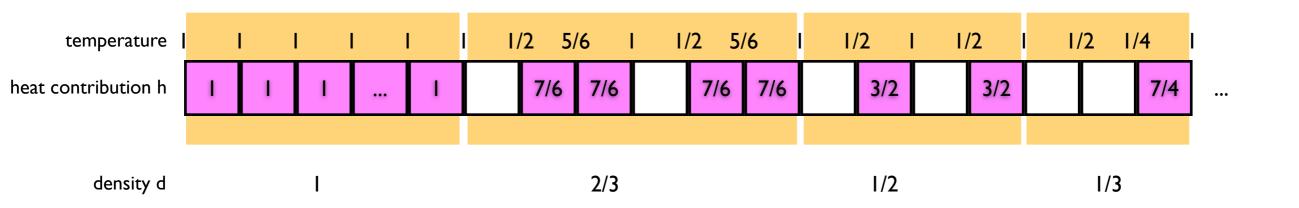
bound objective

- Assume optimum schedules all jobs
- Let D be the optimal makespan
- Partition [0,2] into job classes
- This pessimistic rounding permits to lower bound approximation ratio

round down for optimum, gives structure to lower bound D

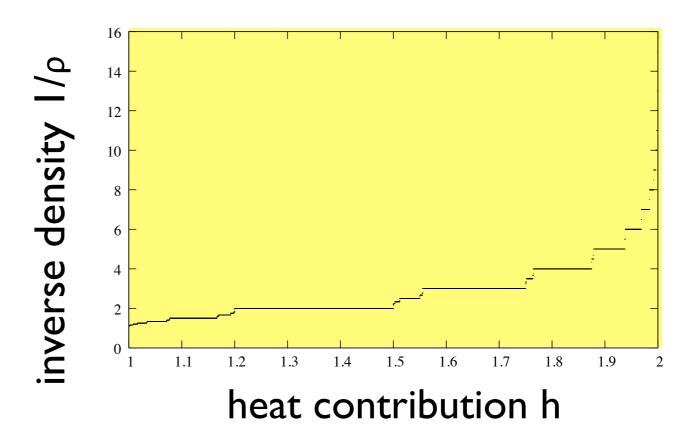
Density

- the output for the rounded instance, consist of blocks of jobs of same heat contribution h
- every block has some density ρ :=average #jobs per slot
- what is the relation between **h** and ρ ?



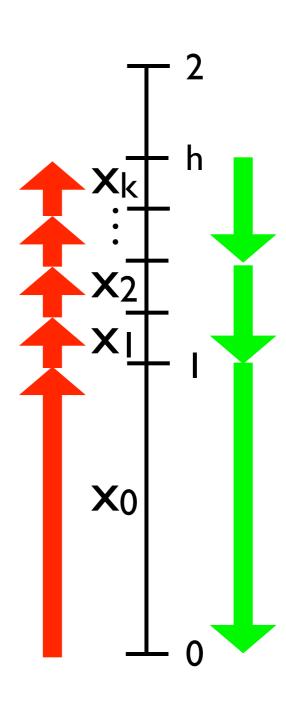
Relation between h and p

- experiments indicate it's monotone
 (as one would expect)
- confirmed by a proof:
 ∀rational ρ ∃hρ:
 G(hρ) has density ρ and hρ is decreasing in ρ



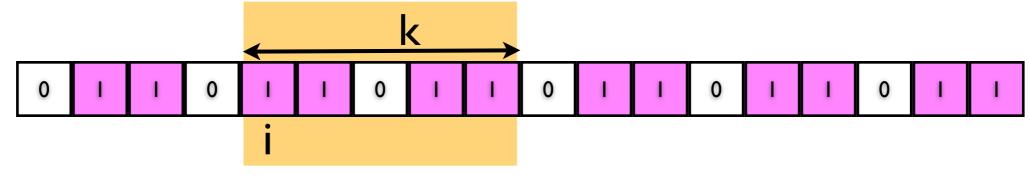
Structure of the analysis

- Suppose last job scheduled by CoolestFirst has heat contribution h
- Partition [0,2] into s classes
- Round heat contributions up for algorithm
- For every q=1,...,k-1 round jobs down for the optimum
- Find worst case instance described by number of jobs from each class $x_1,...,x_k$ boils down in solving a linear program
- We provide a solution to the dual linear program, bounding the primal
- For the worst choice of h, this shows
 approximation ratio of CoolestFirst ≥0.72
- In contrast worst case example has ratio=0.75



Properties of sequence

- G:h → infinite 01-sequence w describing output of CoolestFirst on infinite h-jobs
- exists ρ , such that for all i,k sliding window property floor(ρk) $\leq \sum_{j=i}^{i+k-1} w_j \leq ceil(\rho k)$

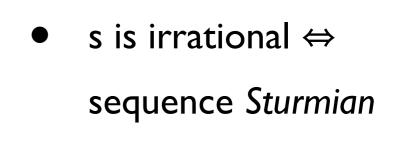


- is periodic (if h is rational)
- sounds familiar?

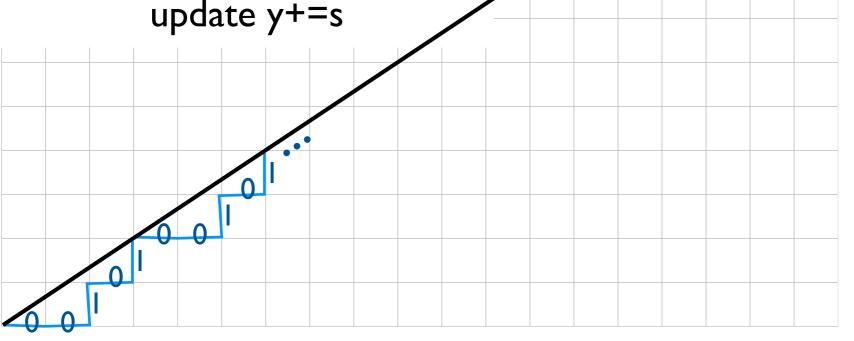
Discretization of line

discrete line (slope s, offset o)

- init y=o
- repeat:
 - if y≥ I, produce I,
 update y-= I
 - if y<1, produce 0, update y+=s



- s is rational ⇔sequence is periodic
- sliding window property



Seem so different

discrete line (slope s, offset o)

- init y=o
- repeat:
 - if y≥ I, produce I, update y-= I
 - if y<1, produce 0, update y+=s
- different slopes produce different sequences
- add

ok, but all we need $\forall \rho \exists h_{\rho}$ Greedy(I,h_{\rho}) produces density ρ , and $\rho < \rho$ ' implies h_{\rho} > h_{\rho'}

Greedy (init temperature t, h)

- init t
- repeat:
 - if t+h≤2, produce I,update t=(t+h)/2
 - if t+h>2, produce 0, update t/=2
- several h produce same sequence
- add and divide

$\forall \rho \exists h_{\rho}$



20/20 10/20 5/20 18/20 9/20 20/20

31/20		31/20
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density p/q

2/5

slope p/(q-p)

2/3

produces w*

(00101)*

 $\bullet \qquad \underline{\mathbf{w}} := \sum_{i=0..q-1} 2^i \, \mathbf{w}_i$

4+16=20

• $h=(2^q-1)/\underline{w}$

- 31/20
- CircShift($x_1x_2...x_k$) := $x_2...x_kx_1$
- Circ(x) := closure of {x} under CircShift
- Properties ∀u∈Circ(w)

00101 01010

- u≤w reverse lexicographically: <u>u≤w</u>
- $ul \le lw$ iff u starts with l

procedure (w)

- init u=w
- repeat:
 - if u starts with I, produce I
 - else produce 0
 - update u=CircShift(u)

equivalences (let t=u/w)

```
\begin{array}{rl} t+h & \leq 2 \\ \underline{u}+2^{q}-1 & \leq 2\underline{w} \\ \underline{u}-1 & \leq 0\underline{w} \\ \underline{u} & \leq \underline{l}\underline{w} \\ u & \text{starts with } I \end{array}
```

update

- if u starts with 0 then CircShift(u)=u/2
- if u starts with I then u=1v and $(\underline{u}+2^{q}-1)/2=\underline{v}+2^{q-1}=\underline{v}$ $=\underline{CircShift}(u)$

conclusion

 CoolestFirst(h) produces same sequence as DiscreteLine(p/q)

Future directions

- Use this technique to analyse scheduling problems with different renewable resources
- Develop a new technique to analyse the Better Algorithm (priority to jobs ∈[0,1])
- Come up with better bounds for the optimum schedule by rounding jobs ∈[0,1] as well