

*Online reassignment models in*  
~~*suchglinde*~~  
*scheduling*

*(on two related machines)*

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# My scientific origin

- Msc: ELTE (Eötvös Loránd Univ.), Budapest, Hungary, 1987
- PhD: Szeged, Hungary, 2009
  - from the hands of J. Csirik,
  - supervisor: B. Vizvari,
  - basically on the common works with Yong He
- Yong He, Hangzhou, 7 papers
- Zsolt Tuza, Leah Epstein, Xin Han,...
- I work at Univ. of Pannonia, Hungary, near Balaton (biggest lake in middle Europe)



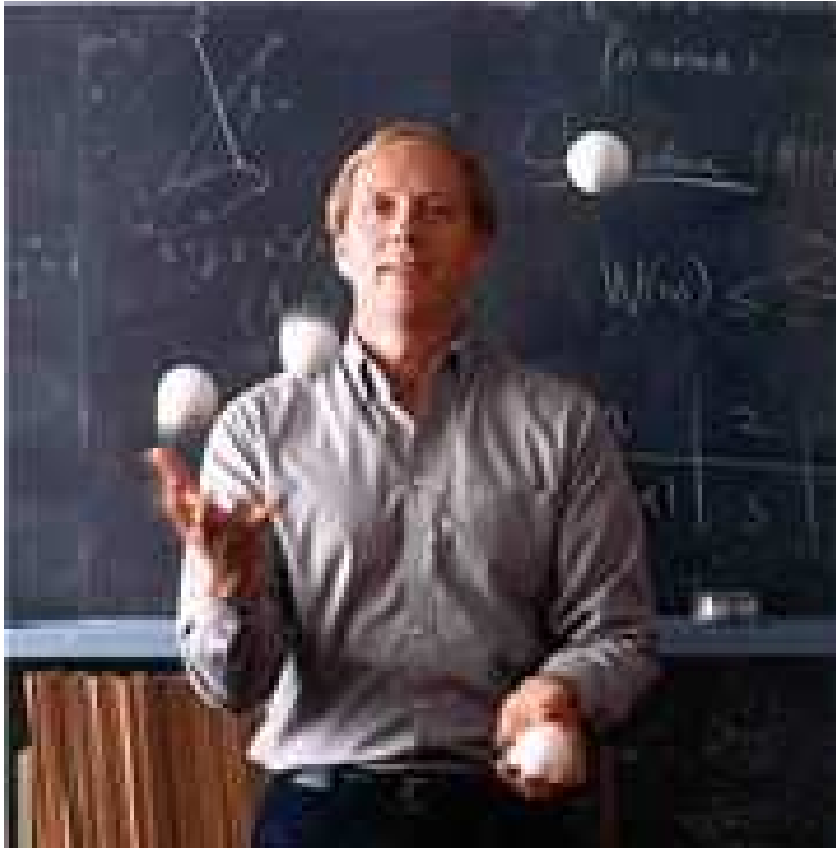
Hangzhou, (a bit left from Sanghai), and the West Lake  
„Above is Heaven, here is Hangzhou”

# My numbers:

Erdős-number=2	Dosa – Zs. Tuza – P. Erdos
Bezdek-number=3	Dosa – Tuza – - T.I.Zamfirescu – K. Bezdek
Deza-number=4	Dosa - He - Frank K. Hwang – - Samuel Onn – Antoine Deza
Lorea-number=4	Dosa – He - G. Woeginger – J. Urrutia – Jesus A. De Lorea
Mitchell-number=3	Dosa- He – G. Woeginger – - Joseph S.B. Mitchell
Ye-number=4	Dosa- He – En Yu Yao – Jie Sun – Yinyu Ye

# some words about scheduling

## the father: Ronald Graham

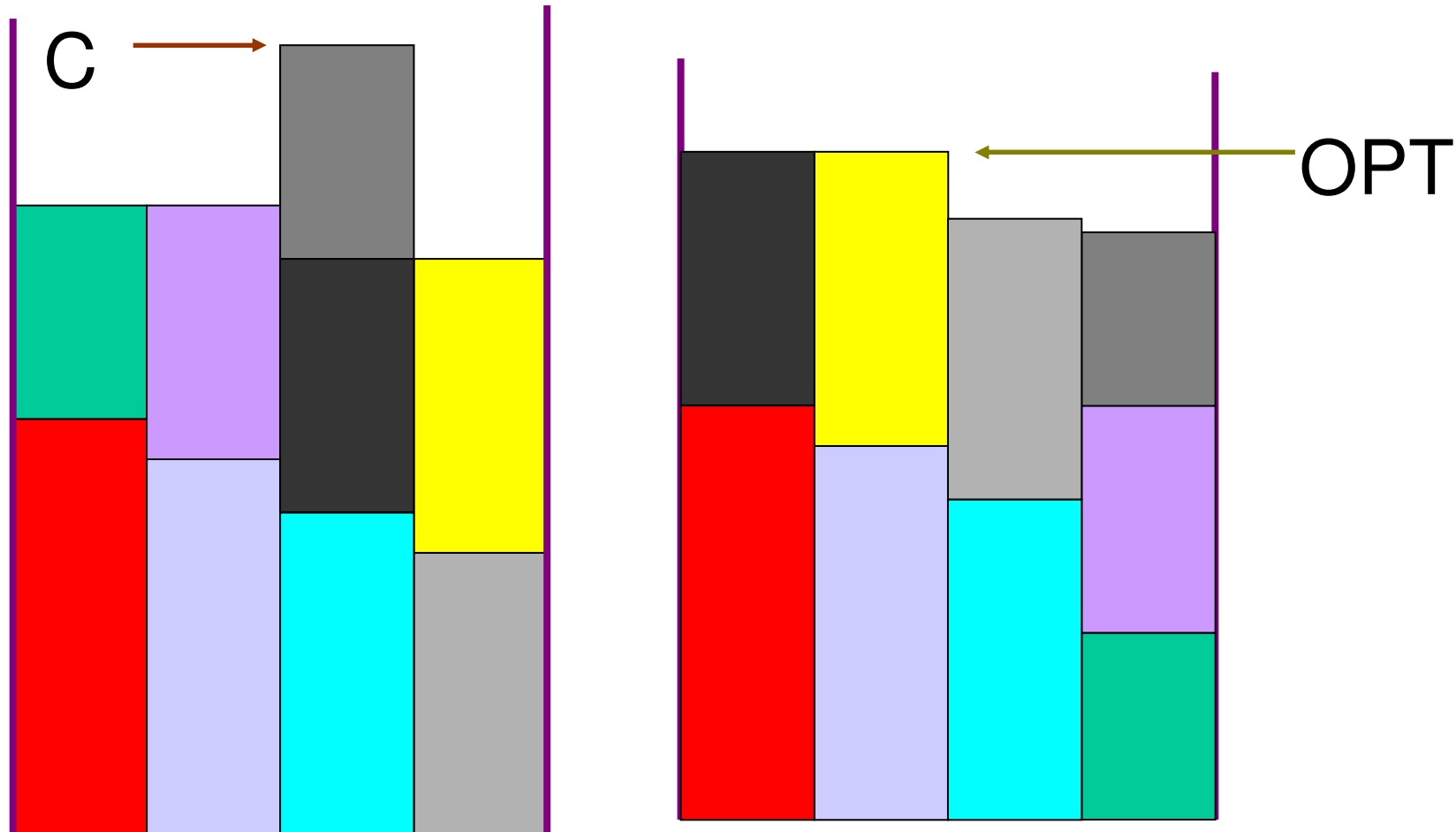


with balls



with his wife and Erdős

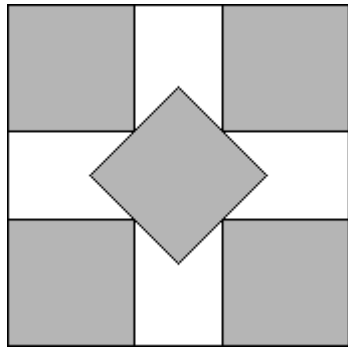
# The beginning: Graham's algorithms LS (and its ordered version, LPT), 1966



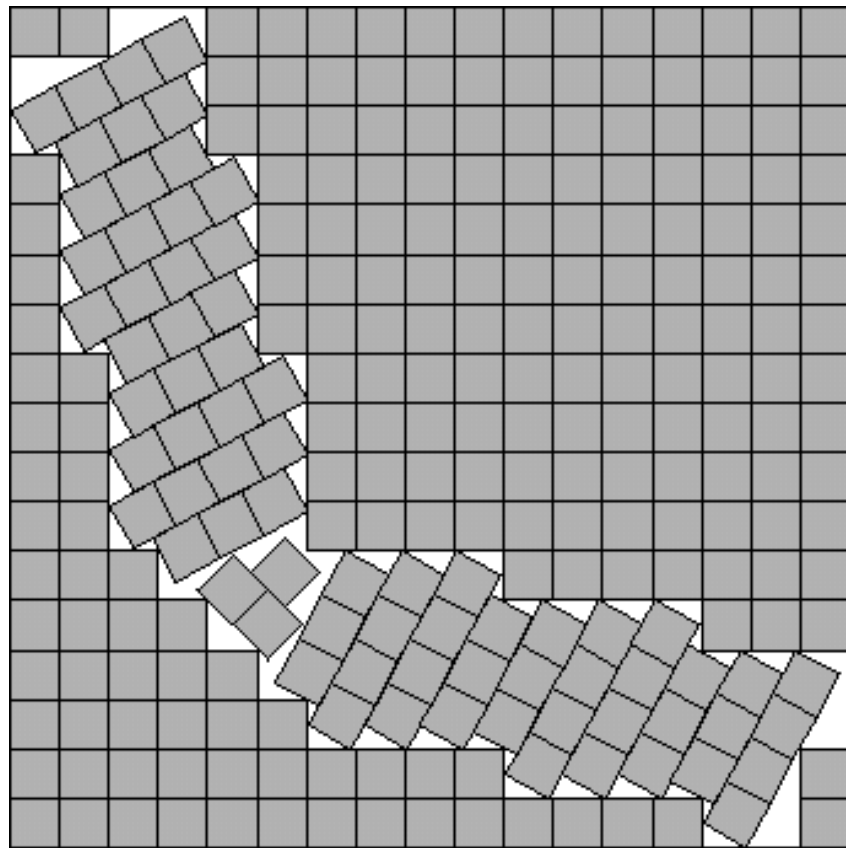
C (=maximum completion time = makespan) → min

a 40 years old problem of Graham and Erdős:  $k$  equal squares should be packed into the smallest (big) square

$k=5$

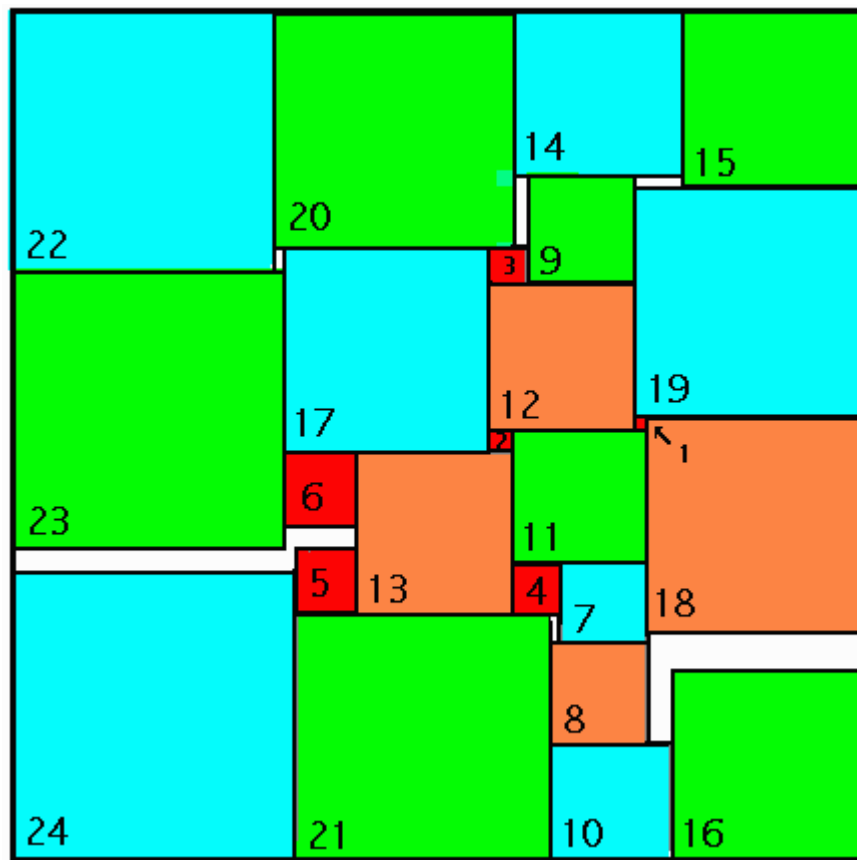


$k=272$ , they fit  
into  $17 \times 17$





Or: there is one square with size  $k$ ,  
for  $1 \leq k \leq 24$  (for example)

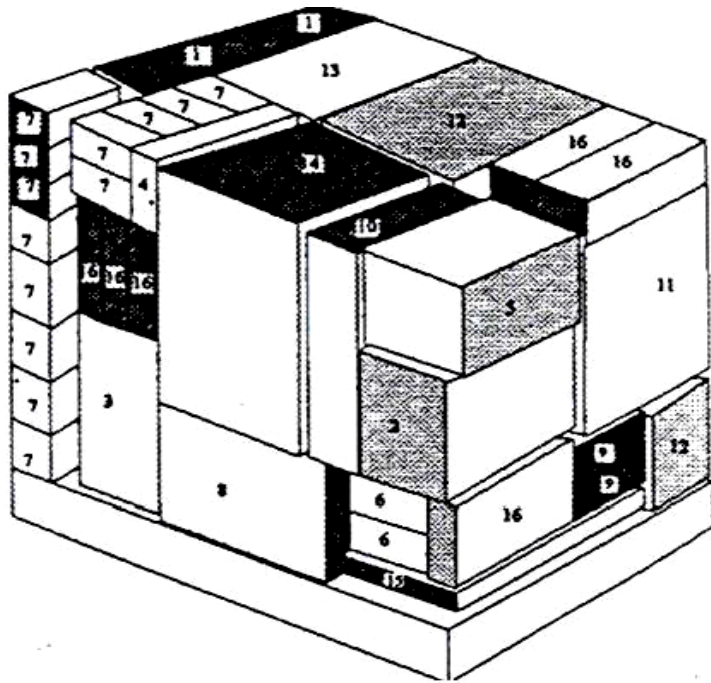


This is 71 x 71,  
(by M. Hujter)

is there smaller  
big square?



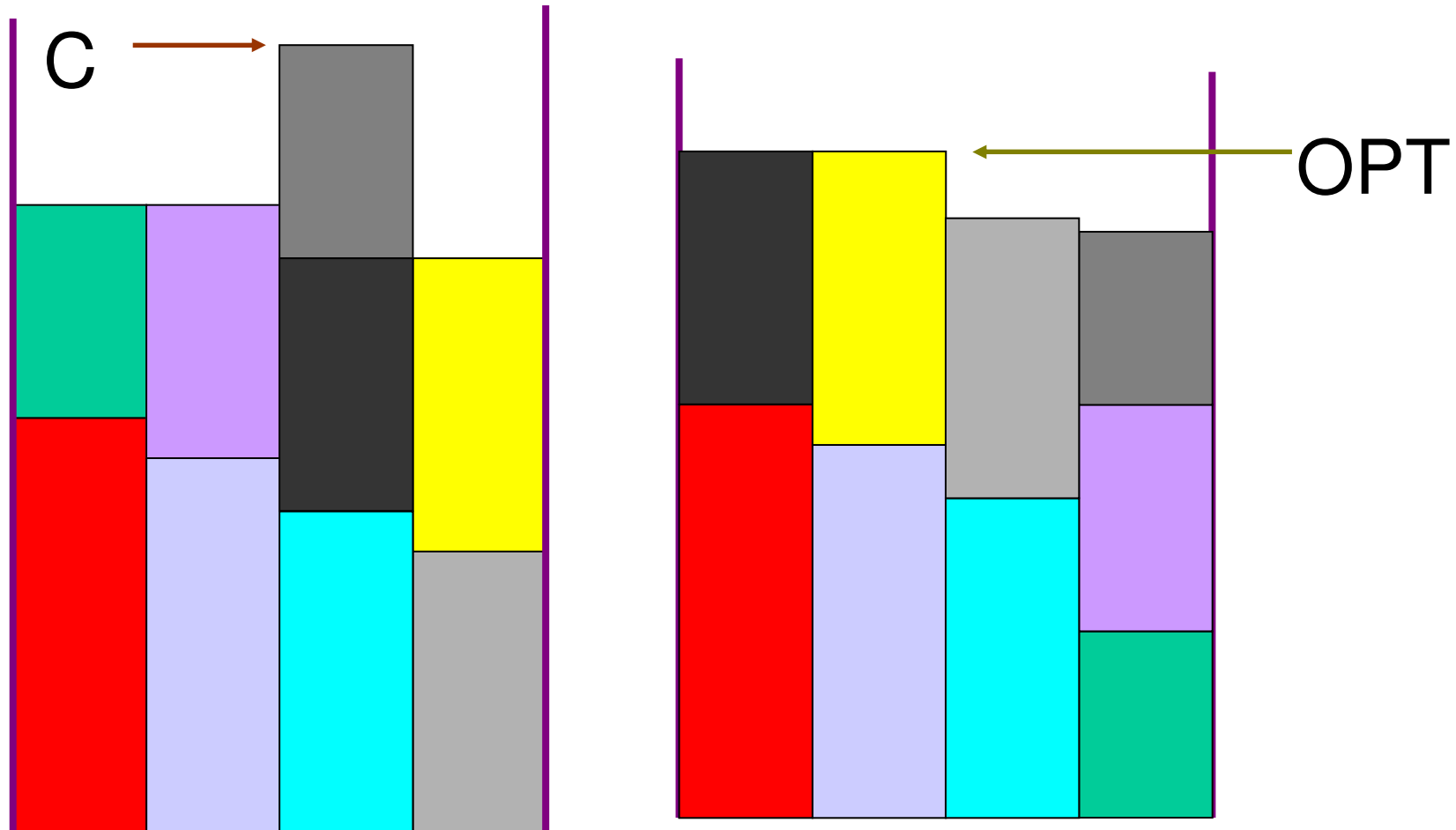
a three dimensional packing problem:



computer aided packing plan, A. Mészáros

So: scheduling (or packing) can be seen as a special kind of „discrete geometry” (in visualization)

LS (or other algorithm) again:  
How **bad** can be  $\text{Alg}/\text{OPT}$ ?



# How big (bad) can be $ALG/OPT$ =? (approx. or competitive ratio)

- LS with decreasing sizes:  $\approx 4/3$

(LPT/OPT is not worse, not bigger than  $\approx 4/3$ )

- LS generally: not worse than  $2 - 1/m \approx 2$   
offline: we know everything about the input

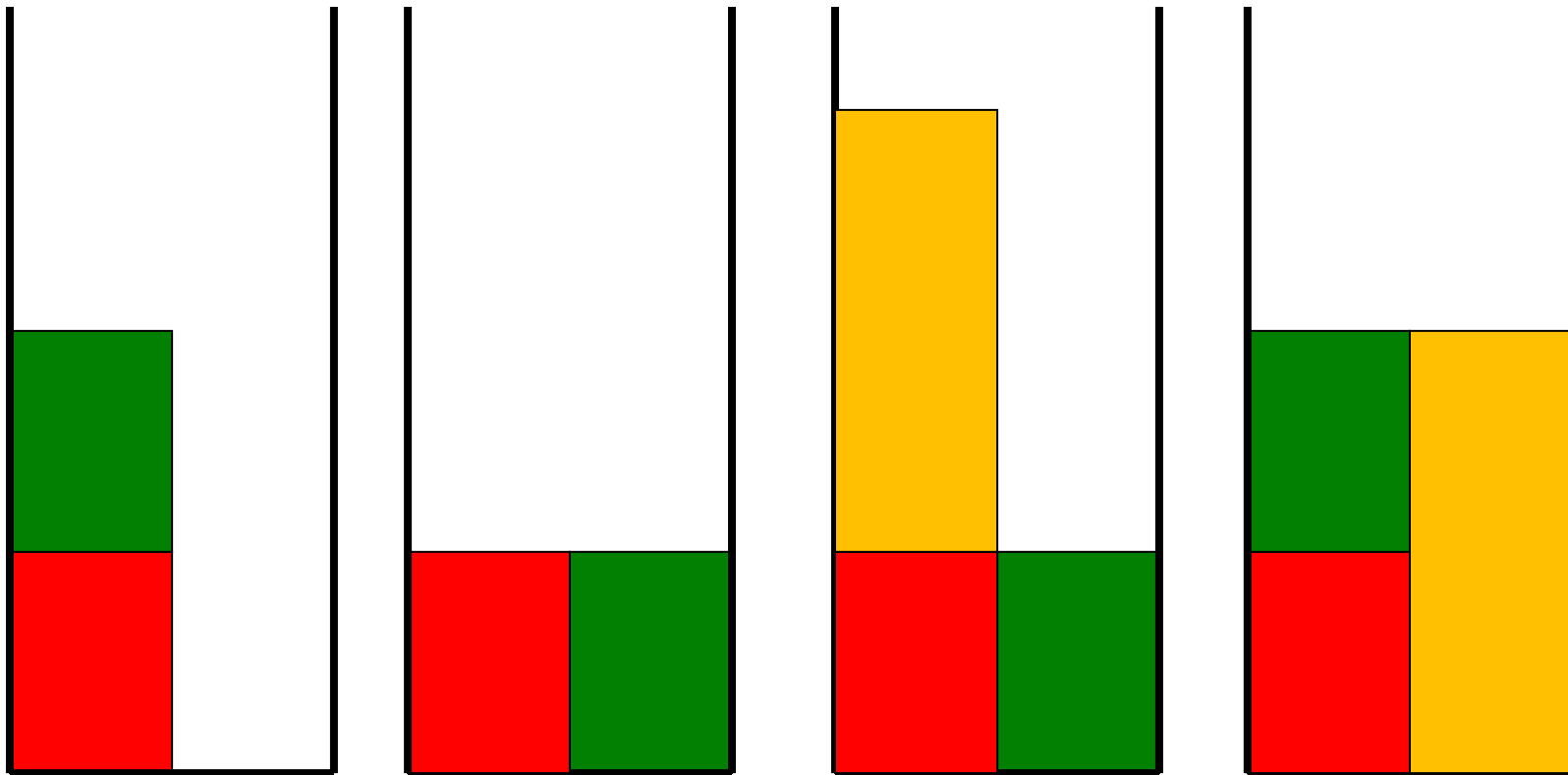
online: we know nothing in advance (but we must make decisions without knowledge of the future)

semi online: we know some things, but not everything (between offline and online)

# offline scheduling (of parallel machines)

- getting optimal solution needs exponential many time (NP-hardness)
- solution within  $(1+\epsilon) \times \text{OPT}$  can be got in pol. (but still much) time, (approx. scheme)
- there are fast heuristics

# online scheduling: lower bounds (by adversaries) and optimality



# online scheduling (of parallel machines)

- Graham's algorithm is **optimal** for two machines, ( $C/OPT$  is at most  $3/2$ )
- LS is also **optimal** for  $m=3$ , (Faigle, Kern, Turán, 1989)
- Four machines ( $m=4$ ) ?
- We **do not know** optimal algorithm for  $m=4$ !  
but there exists better than LS  
(for ex. Woeginger, Galambos, 1993)

# semi online scheduling

The first paper:

- Kellerer, Kotov, Speranza and Tuza (97), three models ( $m=2$ , makespan is minimized, **plus**):
- We know the total size of jobs (also by G. Zhang)
- or we can use a buffer of size  $K$
- or we can make two schedule, and finally chose the better one
- For all of them: opt. Comp ratio  $3/2 \rightarrow 4/3$
- Further models: Decreasing sizes, we know the value of OPT, etc., for example:

**REASSIGNMENT** (a bit later)



# Reassignment in „real life”

- Seasons (winter, spring,...)
- Working places, living places (Toronto is very nice!)
- in politics: in every 4 years
- in economics: „big fish eats small fish”
- in personal habits: young man likes icecream,  
older man likes beer, stew, (more beer)...

## in online models:

- our decisions must be made just in time.  
(just when something happens we must make our decision)
- we cannot change our decisions later

**reassignment:** (some kind of semi online):

- We have some time to think, (we are allowed to delay decisions = reordering buffer), or
- we can change (a bit) later our decisions

# rearrangement in bin-packing:

G. Galambos and G. J. Woeginger, Repacking helps in bounded space on-line bin-packing, Computing, Volume 49, Number 4,

- on-line bounded-space bin-packing problem where:
- repacking the items within the active bins is allowed.
- the 1.69103 lower bound of Lee and Lee for the worst case ratios of bounded-space approximation algorithms still applies.
- A polynomial time approximation algorithm is presented, that reaches the best possible worst case ratio matching the Lee and Lee lower bound while using only *three* active bins.
- (and other papers)

# rearrangement in scheduling:

- the first model considering reassignment is among the first semi online models:

having a (reordering) buffer of size  $K$

- Kellerer, Kotov, Speranza, Tuza: Semi online algorithms for the partition problem, *Op.Res Letters* 21 (1997), 235-242.

Results: -three semi online versions of  $P2 || C_{\max}$ ,  
- three optimal algorithms with comp. ratio  $C=4/3$ ,  
( the pure online comp. ratio is  $C=3/2$  (LS) )

the buffer model is also treated in:

- G.C. Zhang, A simple semi-online algorithm for  $P2 || C_{\max}$  with a buffer. *Information Processing Letters*, 61 (1997), 145-148.

there are also some (new?) models:

- bounded migration by

N. Sivadasan, P. Sanders, M. Skutella, Online scheduling with bounded migration, Math. Oper.Res. 34 (2) (2009) 481-498.

(for any job there are two parameters, the size and a rearrangement parameter, and also given a global rearrangement factor  $\beta$ ,...a **different type** of rearr.

so we do not deal with this model now,  
on the other hand it is in fact interesting)

- REAR(K), (rearr. at any time) defined by

G. Dosa, Y. Wang, X. Han, H. Guo, Online scheduling with rearrangement on two related machines, Theoretical Computer Science, 412(8-10): 642-653 (2011)

- and three further models of

Z. Tan, S. Yu, Online scheduling with reassignment, Oper.Res.Lett. 36(2) 2008, 250-254., as :

- K jobs can be rearranged at the end of the sequence (*after the sequence ended*)
  - <sub>1</sub> The last job of *any machine* can be moved to the other machine (*after the sequence ended*)
  - <sub>1</sub> The last K jobs of the *sequence* can be moved (*after the sequence ended*)

now we consider only three models,  
only in case of 2 uniform machines  
(i.e., the second machine is faster)

- **BUFF(K), the buffer problem**

G. Dosa, L. Epstein, Online scheduling with a buffer on related machines, J.Comb.Optim. 20(2) 2010, 161-179.

- **REAR(K), at any time (many times)**

G. Dosa, Y. Wang, X. Han, H. Guo, Online scheduling with rearrangement on two related machines, Theoretical Computer Science, 412(8-10): 642-653 (2011)

- **REND(K), rearrangement only at the end**

A. Benko, X. Chen, G. Dosa, X. Han, Online scheduling with bounded rearrangement at the end, TCS, 2011.

(the model is defined originally by Tan, 2008)



without reassignment and buffer  
(pure online case, i.e.  $K=0$ )

In this case, the tight competitive ratio is

$$(2s+1)/(s+1) \text{ if } 1 < s < 1.618 = (\text{sqrt}(5)+1)/2$$

and

$$(s+1)/s \text{ if } 1.618 < s$$

(L. Epstein, J. Noga, S. S. Seiden, J. Sgall, and G. J. Woeginger,  
Randomized Online Scheduling on Two Uniform Machines,  
*Journal of Scheduling*, 4(2):71–92, 2001.)

## *comparison between the models*

- (we know that at least formally)  
     $\text{REAR}(K)$  is more flexible than  $\text{BUFF}(K)$   
    (=allows at least as good comp. ratio)

**BUT**

- Is  $\text{REND}(K)$  more flexible than  $\text{BUFF}(K)$  ?
- Is  $\text{REND}(K)$  more flexible than  $\text{REAR}(K)$  ?

since *!!!* in  $\text{REND}(K)$

- we are allowed to make rearrangement only once
- but we know before the rearrangement that the sequence is over  
    (We will see that any of the three models allows better comp. ratio than the pure only model)

# Results 1. **if $s \geq 2$ :**

- everything is quite simple
- $K=1$  is enough  
( $K=1$  is better than  $K=0$ , but  $K>1$  do not give much than  $K=1$ )
- The tight ratio is  $C=(s+2)/(s+1)$   
uniformly, **for all three models!**  
(for  $K=0$  the tight ratio is  $(s+1)/s$  )

## Results 2. **if $1 \leq s \leq 2$** , and $K=2$

- this case is not very hard (but not simple)
- $K=2$  is enough  
( $K=2$  is better than  $K=0$ , but  $K>2$  does not give better than  $K=2$ )
- The tight ratio is

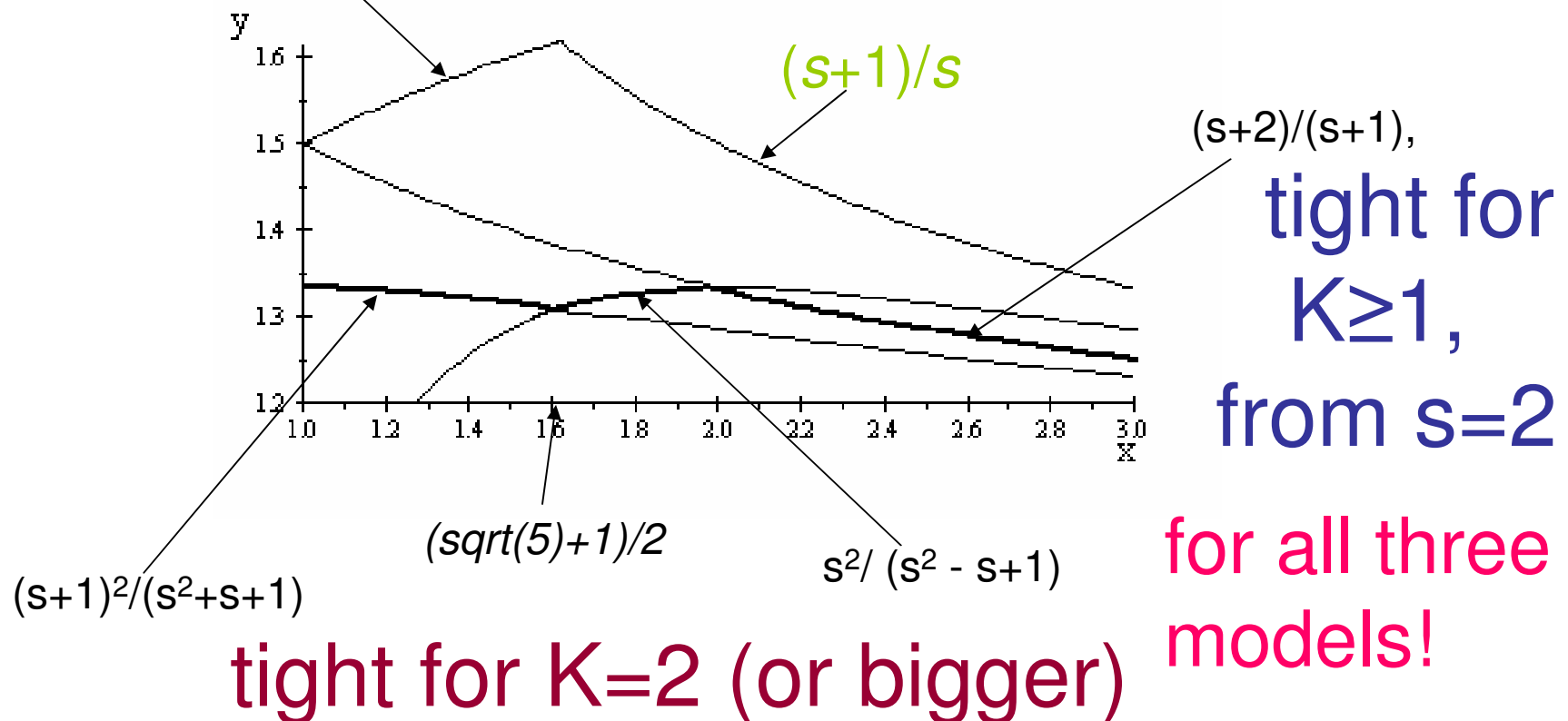
$$C = (s+1)^2 / (s^2 + s + 1) \quad \text{if } s \leq 1.61 = (\sqrt{5} + 1) / 2$$

$$C = s^2 / (s^2 - s + 1) \quad \text{if } s \geq 1.61$$

uniformly, **for all three models!**

so, the tight ratios:

$(2s+1)/(s+1)$  pure online ( $K=0$ )

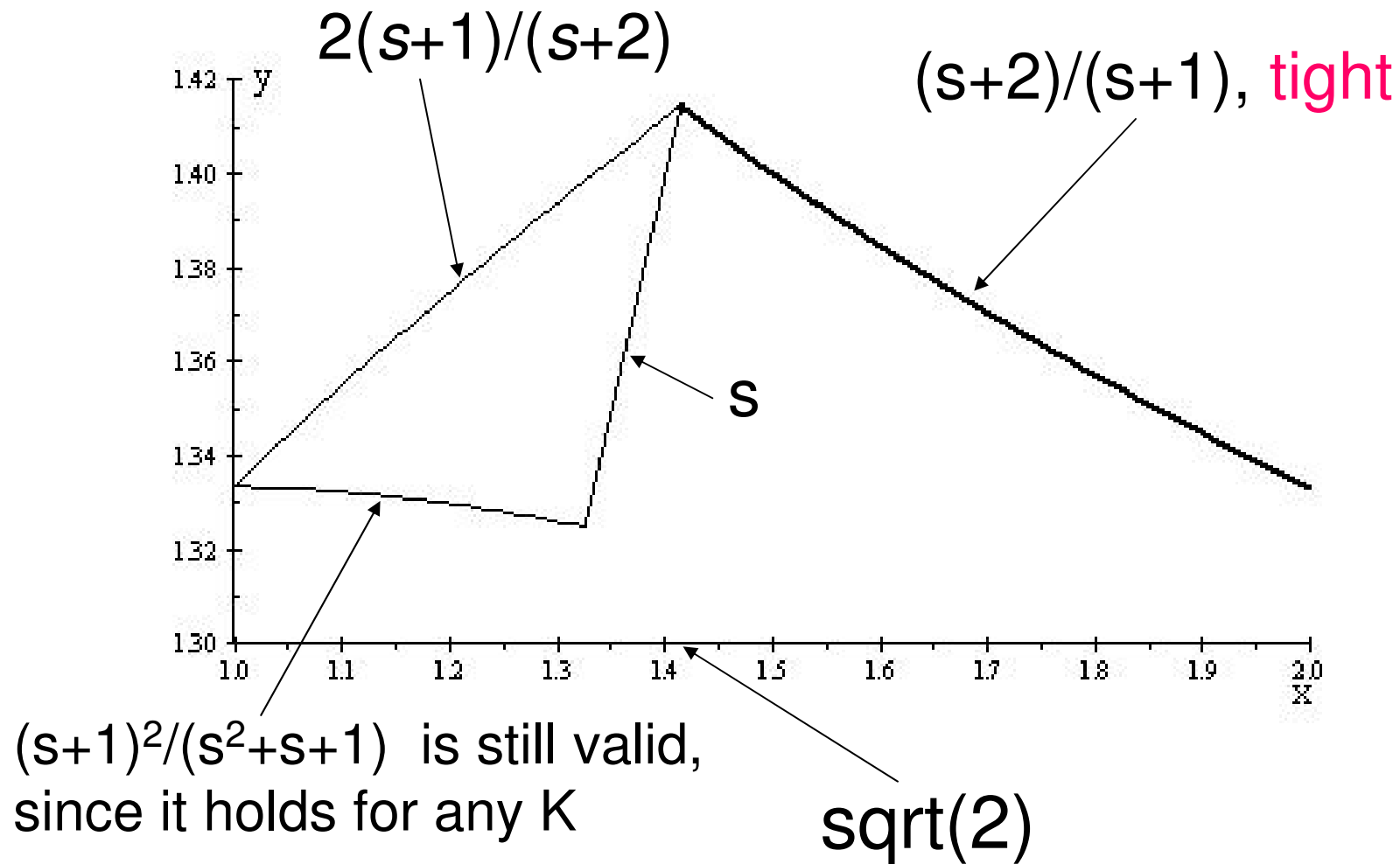


## Results 3. the hard case:

$$1 \leq s \leq 2, \text{ and } K=1$$

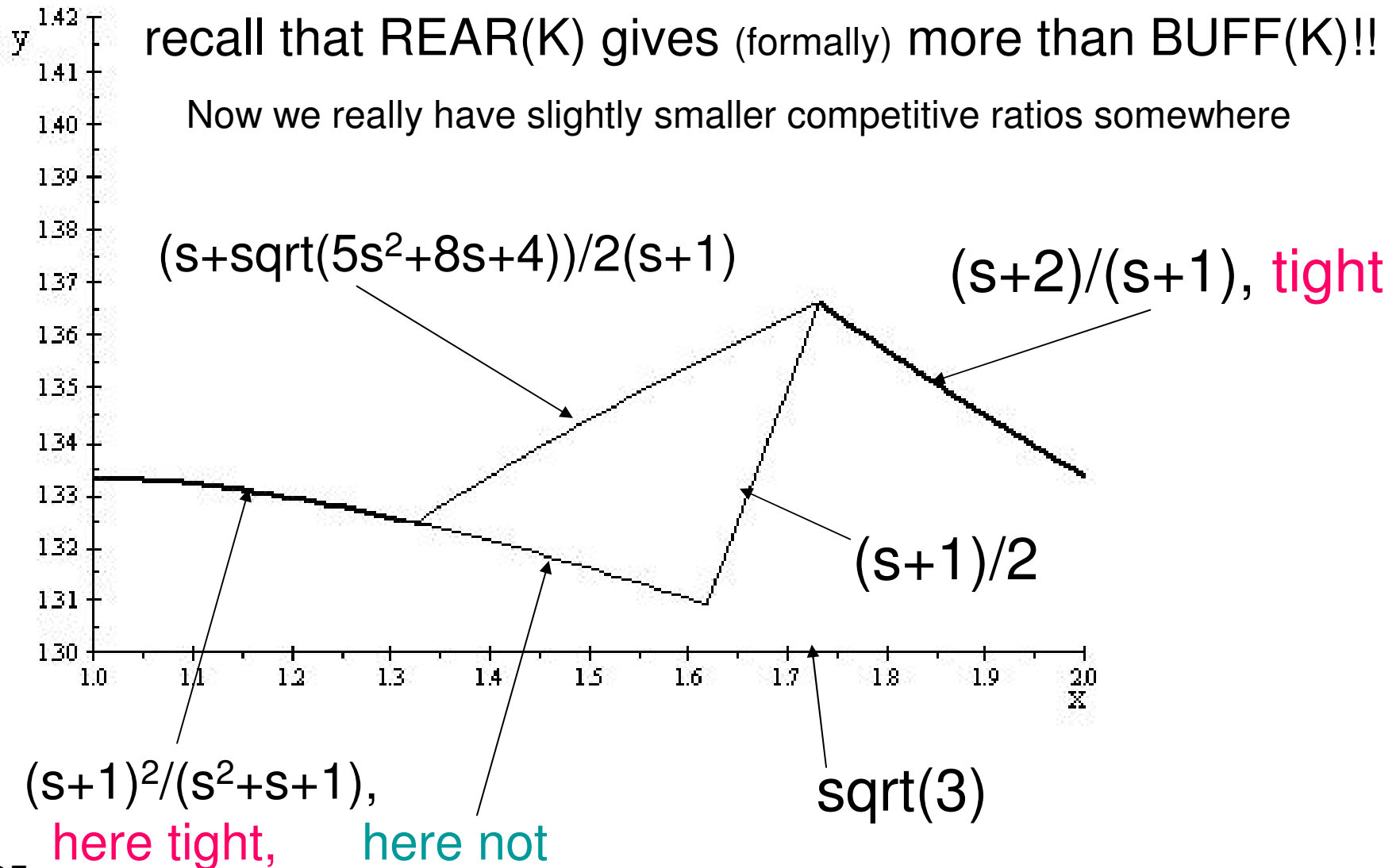
- we actually do have results (in all three models), but it seems hard to get the tight ratios for the whole interval (in any of the three models)
- here the models differ from each other (regarding competitive ratio)
- we still have the best results so far (for all three models)

what we could prove, **BUFF(1)**:





# what we could prove, REAR(1):



## The third model: Two simple algorithms for $\text{REND}(1)$ , for both:

- 1 We use the (classical online) lower bound **LB**:  
 $\max\{\text{total size}/(s+1), \text{biggest job}/s\}$
- 1 We also use the allowed competitive ratio  **$C(s)$**
- 1 A schedule is **feasible**, if the desired competitive ratio is not violated, **where the (increased) makespan is compared to the lower bound LB**  
(the makespan is not bigger than  $C$ -times LB)

# ALG<sup>1</sup> for REND(1)

- <sup>1</sup> Assign the incoming job to M1 if so the (temporary) schedule **is** feasible, otherwise to M2.
- <sup>1</sup> REASSIGNMENT: If the final schedule is not feasible, move **a job** from M2 to M1 so that the makespan decreases as much as possible
  - <sup>1</sup> It is  $(s+2)/(s+1)$ -comp. for any  $s \geq 1$
  - <sup>1</sup> so for  $s \geq 2$  it is optimal.

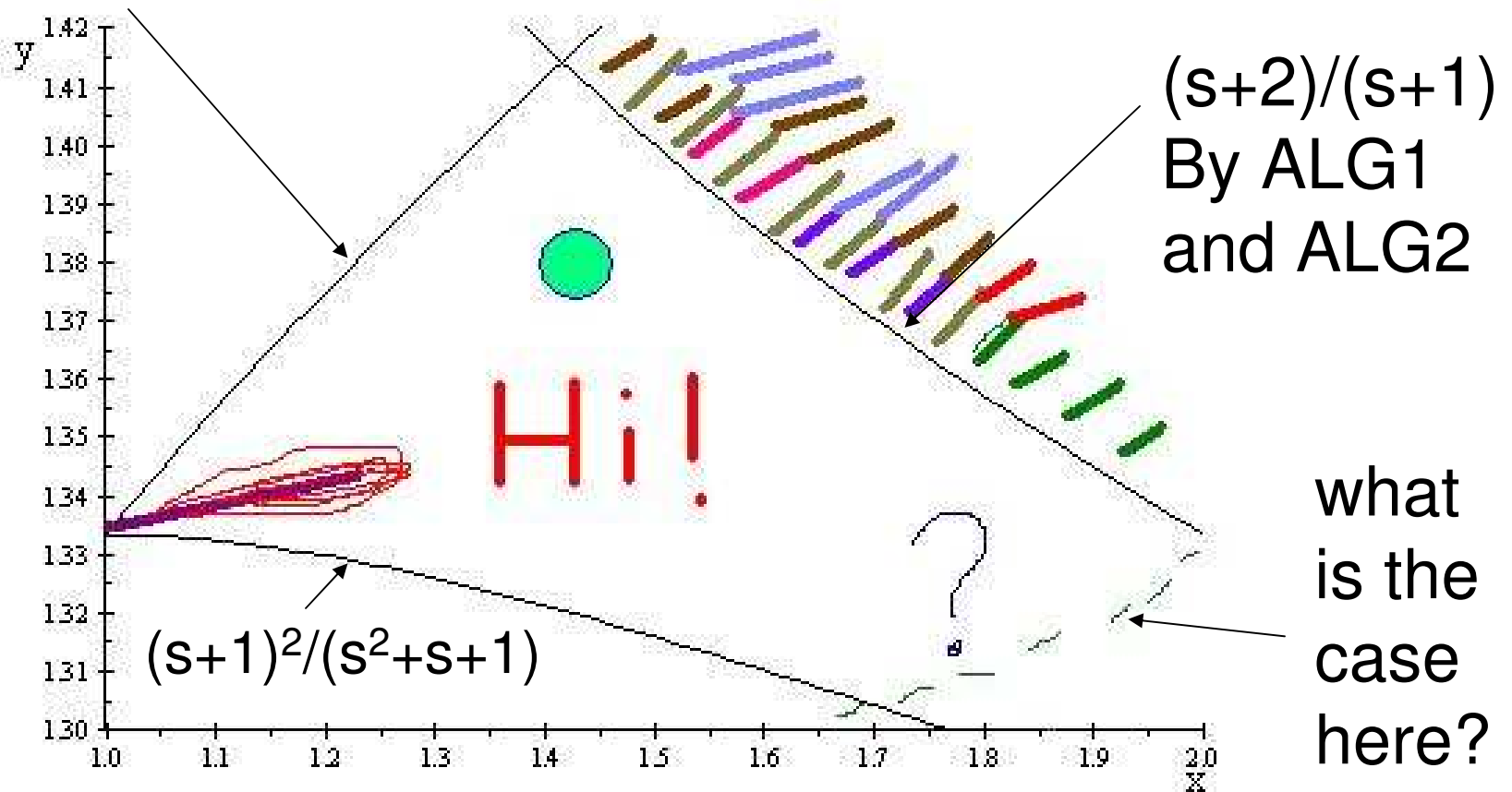
# ALG<sup>2</sup> for REND(1)

- <sup>1</sup> Assign the incoming job to M1 **if** so the (temporary) schedule is **not** feasible, but moving the biggest job from M1 to M2 the schedule becomes feasible, **otherwise** to M2.
- <sup>1</sup> REASSIGNMENT: If the final schedule is not feasible, move the biggest **job** from M1 to M2 if so the makespan decreases
- <sup>1</sup> It is  $(s+2)/(s+1)$ -competitive only if  $1 \leq s \leq 2$ , **but**
- <sup>1</sup> We know in advance what job will be moved
- <sup>1</sup> And it is also  $2(s+1)/(s+2)$ -competitive

# what we could prove, **REND(1)**:

(only preliminary results so far, but these are the best results)

$2(s+1)/(s+2)$  by ALG2



the same upper bounds as in **BUFF(1)**! **good lower bounds? really challenging task,**  
previous constructions do not work!  
but now we nowhere know whether they are tight or not

# Algorithms (some statistics)

- buffer model  $\text{BUFF}(K)$ :
  - 4 algorithms on 9 pages
- rearrangement at any time  $\text{REAR}(K)$ 
  - 1 algorithm, the analysis is on 5 pages through 5 Lemmas, 8 Observations and 3 Claims
- rearrangement at the end  $\text{REND}(K)$ 
  - 3 algorithms, the proofs are on  $\approx 20$  pages through 11 Lemmas, 11 Observations, many Cases

~~YANK THOU!~~ THANK YOU!