# The price of anarchy on uniformly related machines revisited\*

Leah Epstein<sup> $\dagger$ </sup> Rob van Stee<sup> $\ddagger$ </sup>

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#### Abstract

Recent interest in Nash equilibria led to a study of the *price of anarchy* (POA) and the *strong price of anarchy* (SPOA) for scheduling problems. The two measures express the worst case ratio between the cost of an equilibrium (a pure Nash equilibrium, and a strong equilibrium, respectively) to the cost of a social optimum.

The atomic players are the jobs, and the delay of a job is the completion time of the machine running it, also called the load of this machine. The social goal is to minimize the maximum delay of any job, while the selfish goal of each job is to minimize its own delay, that is, the delay of the machine running it.

We consider scheduling on uniformly related machines. While previous studies either consider identical speed machines or an arbitrary number of speeds, focusing on the number of machines as a parameter, we consider the situation in which the number of different speeds is small. We reveal a linear dependence between the number of speeds and the POA. For a set of machine of at most p speeds, the POA turns out to be exactly p + 1. The growth of the POA for large numbers of related machines is therefore a direct result of the large number of potential speeds. We further consider a well known structure of processors, where all machines are of the same speed except for one possibly faster machine. We investigate the POA as a function of both the speed ratio between the fastest machine and the number of slow machines.

# **1** Introduction

A Nash equilibrium [22] is a state in non-cooperative games which is stable in the sense that no agent can gain from unilaterally switching strategies. Many "solution concepts" are used to study the behavior of selfish agents in non-cooperative games.

A strong equilibrium is a pure Nash equilibrium, in which not only single players cannot benefit from changing their strategy (to a different pure strategy), but no non-empty subset of players can form a coalition, where a coalition means that all of them can change their strategies together, and all gain from the change (see [2, 1, 6]).

Following recent interest of computer scientists in game theory [23, 16, 17, 26], we study pure Nash equilibria and strong equilibria for a scheduling problem on uniformly related machines. We next define the problem and pure equilibria for scheduling problems.

Scheduling on uniformly related machines is a basic assignment problem. In such problems, a set of jobs  $J = \{j_1, j_2, \ldots, j_n\}$  is to be assigned to a set of m machines  $M = \{M_1, \ldots, M_m\}$ , where machine  $M_i$  has a speed  $s_i$ . The size of job  $j_k$  is denoted by  $w_k$  and it is equal to its running time

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<sup>&</sup>lt;sup>†</sup>Department of Mathematics, University of Haifa, 31905 Haifa, Israel. lea@math.haifa.ac.il.

<sup>&</sup>lt;sup>‡</sup>Max-Planck-Institut für Informatik, Saarbrücken, Germany. vanstee@mpi-inf.mpg.de. Research supported by German Research Foundation (DFG).

on a unit speed machine. Moreover, the running time of this job on a machine of speed s is  $\frac{w_k}{s}$ . An assignment or schedule is a function  $\mathcal{A} : J \to M$ . The completion time of machine  $M_i$ , which is also called the *delay* or *load* of this machine, is  $\sum_{k:\mathcal{A}(j_k)=M_i} \frac{w_k}{s_i}$ . The cost, or the *social cost* of a schedule is the maximum delay of any machine, i.e., the makespan. We see jobs as atomic players, thus we use terms such as choice and benefit for these players.

A schedule is a *Nash equilibrium* if there exists no job that can decrease its delay by migrating to a different machine unilaterally. More precisely, consider an assignment  $\mathcal{A} : J \to \{M_1, \ldots, M_m\}$ . The class of schedules  $\mathcal{C}$  contains all schedules  $\mathcal{A}'$  that differ from  $\mathcal{A}$  only in the assignment of a single job. That is,  $\mathcal{A}' \in \mathcal{C}$  if there exists a job  $j_k \in J$  such that  $\mathcal{A}'(j_\ell) = \mathcal{A}(j_\ell)$  for all  $J_\ell \in J$ ,  $J_\ell \neq J_k$ , and  $\mathcal{A}'(j_k) \neq \mathcal{A}(j_k)$ . We say that  $\mathcal{A}$  is a (pure) Nash equilibrium if for any job  $j_k$ , the delay of  $j_k$  in any schedule  $\mathcal{A}' \in \mathcal{C}$ , for which  $\mathcal{A}'(j_k) \neq \mathcal{A}(j_k)$ , is no smaller than its delay in  $\mathcal{A}$ . Pure Nash equilibria do not necessary exist for all games (as opposed to mixed Nash equilibria). It is known that for scheduling games of this type, a pure Nash equilibrium always exists [12, 8].

A schedule is a *strong equilibrium* if there exists no (non-empty) subset of jobs, such that if all jobs in this set migrate to different machines of their choice simultaneously, this results in a smaller delay for each and every one of them. More precisely, given a schedule  $\mathcal{A}$ , we can define a class of schedules  $\tilde{\mathcal{C}}$  which contains all sets of schedules  $\mathcal{C}_K$ , where  $K \subseteq J$ ,  $K \neq \emptyset$ . For any  $\mathcal{A}' \in \mathcal{C}_K$ , and  $\ell \notin K$ , we have  $\mathcal{A}'(j_\ell) = \mathcal{A}(j_\ell)$  whereas for  $\ell \in K$ , we have  $\mathcal{A}'(j_\ell) \neq \mathcal{A}(j_\ell)$ .  $\mathcal{A}$  is a strong equilibrium if for any  $K \neq \emptyset$ , and any  $\mathcal{A}' \in \mathcal{C}_K$ , there exists at least of job  $j_k \in K$  whose delay in  $\mathcal{C}_K$  is no smaller than its delay in  $\mathcal{A}$ . A strong equilibrium is always a pure Nash equilibrium (by definition). Strong equilibria do not necessarily exist. Andelman, Feldman and Mansour [1] were the first to study strong equilibria in the context of scheduling and proved that scheduling games (of a more general form) admit strong equilibria. More general classes of congestion games which admit strong equilibria were studied in [14, 28].

In this paper, we study the price of anarchy (POA) and the strong price of anarchy (SPOA) for scheduling on uniformly related machines.

In our scheduling model, the *coordination ratio*, or *price of anarchy* (POA) (see [25]) is the worst case ratio between the cost of a pure Nash equilibrium and the cost (i.e., maximum delay or makespan) of an optimal schedule. Such an optimal schedule as well as its cost are denoted by OPT. The *strong price of anarchy* (SPOA) is defined similarly, but only strong equilibria are considered. Therefore we refer to the pure price of anarchy by POA and when we discuss the mixed price of anarchy we call it the mixed POA. Note that a pure equilibrium is a special case of mixed equilibria.

It is noted in a series of papers (e.g., [16, 21, 24, 5, 4]) that the model which we study is a simplification of problems arising in real networks, that seems appropriate for describing basic problems in networks.

A number of papers studied equilibria for scheduling on uniformly related machines [16, 21, 5, 9, 10]. Chumaj and Vöcking [5] showed that the POA is  $\Theta(\frac{\log m}{\log \log m})$  (and  $\Theta(\frac{\log m}{\log \log \log m})$  for mixed strategies). Feldmann et al. [9] proved that the POA for m = 2 and m = 3 is  $\frac{\sqrt{4m-3}+1}{2}$  which equals  $\phi = \frac{\sqrt{5}+1}{2}$  for two machines and 2 for three machines. In [7], the exact POA and SPOA for two machines is found as a function of the machine speeds. The two measures given different results for the interval (1.618, 2.247) of speeds ratios between the two machines, and identical results otherwise. As for the mixed POA, it was shown in [16] that it is at least  $1 + \frac{s}{s+1}$  for  $s \le \phi$ . Recently, Fiat et al. [10] showed that the SPOA for this model is  $\Theta(\frac{\log m}{(\log \log m)^2})$ .

For *m* identical machines (i.e., the case where all speed are equal), the POA is  $\frac{2m}{m+1}$ , which can be deduced from the results of [11] (the upper bound) and [27] (the lower bound). It was shown in [1] that the SPOA has the same value as the POA for every *m*. Note, however, that the mixed POA is non constant already in this case, and equals  $\Theta(\frac{\log m}{\log \log m})$ , where the lower bound was shown by Koutsoupias and Papadimitriou [16] and the upper bound by Chumaj and Vöcking [5] and independently by Koutsoupias,

Mavronicolas and Spirakis [15]. Tight bounds of  $\frac{3}{2}$  on the mixed POA for two identical machines were shown by [16].

It can be seen that the POA and SPOA were studied mainly as a function of the number of machines. Another relevant parameter for uniformly related machines is the number of different speeds. A natural question is whether the POA and SPOA grow as the number of machines increases even if the number of different speeds is constant, or whether it is actually the number of speeds that needs to increase. Previous results, and in particular, the POA for identical machines, already hint that the second option is the right one. We prove this property formally, specifically, we show that the POA for inputs with at most p different speeds is exactly p + 1. We note that it can be deduced from [10] that the SPOA for inputs with at most p different speeds is  $\Omega(\frac{p}{\log p})$  (and O(p) by our result), therefore the SPOA is quite close to the POA and it is it influenced by the number of different speeds as well. We further focus on a well known architecture of machines, which consists of a single "fast machine" of speed  $s \ge 1$  together with m - 1 unit speed machines. Such a structure, where one processor is fast, and all others are identical, is natural, and was studied in [20, 13, 3, 19, 18].

We give a complete analysis of the exact POA as a function of the speed of the faster machine, s, and the number of identical machines, m' = m - 1. We believe that our comprehensive analysis would contribute to a deeper understanding of the POA as a function of several parameters, rather than as a function of the number of machines as a single parameter. Our results imply that the worst case POA (the supremum POA over all values of s and m) for this special case of two different speeds is already 3. We conclude the paper by showing that the worst case SPOA for this variant is strictly smaller than the POA, already in this special case, but it is still strictly larger than the SPOA for m identical machines.

### **2** A tight bound on the POA for *p* speeds

In this section, we consider the general case of a machine set with a fixed number of different speeds, and show that the POA is linearly dependent on the number of speeds, namely, it is p + 1 if there are p different speeds. We use ingredients of the proofs in [5], focusing on the load in different groups of machines. We assume that p > 1, since p = 1 is simply the case of identical machines, for which a tight bound is known [11, 27, 1].

**Lemma 1** The price of anarchy on m related machines that have at most p different speeds is at most p + 1.

**Proof** Consider a job assignment to machines, denoted by  $\mathcal{A}$ , that satisfies the conditions of a Nash equilibrium. Let  $\sigma_1 \geq \cdots \geq \sigma_p$  be a sorted list of the speeds. We define the speed class  $\ell$  as the subset of machines with speed  $\sigma_{\ell}$ . We assume that machines are numbered by  $1, \ldots, m$ , and their speeds  $s_1, \ldots, s_m$  are sorted by non-increasing speed (i.e.,  $s_1 \geq s_2 \geq \ldots \geq s_m$ ). Moreover, we assume that the machines of each speed class are sorted by non-increasing load in  $\mathcal{A}$ . Let T be the maximum load over all machines and scale the instance so that OPT = 1. Assume T > 1, otherwise we are done. Note that since some machine has a load that exceeds 1, then there must exist at least one machine whose load is strictly smaller than 1.

Let C be the load of the least loaded machine of speed class 1, by the order defined above, that is, a machine r of speed  $s_r = \sigma_1$  such that  $s_{r+1} = \sigma_2$ . We claim that  $C \ge T - 1$ . If the maximum load is achieved on this machine, then we have C = T and we are done. Otherwise, let k be a machine of load T. For a given job j of the instance, an optimal solution (which has makepsan 1) runs j on one of the machines, which we denote by  $i_j$ . Therefore we have that its size satisfies  $w_j \le s_{i_j} \le \sigma_1$  and thus  $\frac{w_j}{\sigma_1} \le 1$ . Since moving a job from machine k to machine r is not beneficial, for such a job we have  $T \le C + \frac{w_j}{\sigma_1} \le C + 1$ . This proves the claim. If  $C \le 1$  then  $T \le 2 . Therefore we assume$ <math>C > 1. We introduce additional notations. Let  $C' = \lceil C \rceil \geq 2$ . We define  $J_1, \ldots, J_{C'-1}$  and  $I_1, \ldots, I_{C'-1}$ which are indices of machines. We let  $I_i$  be the first machine (in the sorted order above) with a load which is strictly smaller than C' - i, and  $J_i = I_i - 1$ . We show that all  $I_i$ 's are well defined and the values  $J_i$  are actual indices of machines (i.e.,  $J_i \geq 1$  for  $i \geq 1$ ). Since machine r has load C and by definition C' < C + 1, we have that machine r has load C > C' - 1. By the ordering of machines, machines  $1, \ldots, r - 1$  have a load of at least C' - 1 as well. By the definition of the indices  $I_i$ , we have  $I_1 \geq r + 1$  and thus  $J_1 \geq r \geq 1$ . Moreover,  $I_i \geq I_{i-1}$  for all  $2 \leq i \leq C' - 1$ , thus  $J_i \geq 1$  for all  $i \geq 1$ , as claimed. Since  $I_{C'-1}$  is the first machine with a load smaller than C' + 1 - C' = 1, this last index must exist, since some machine must have load less than 1. Note that  $I_{C'}$  is not defined and cannot exist since this would imply a machine of load less than 0.

By definition, the load of machines  $1, \ldots, J_i$  is at least C' - i. We now claim that the speed of  $I_i$  is no larger than  $\sigma_{i+1}$  for  $i = 1, \ldots, C' - 1$ . We prove this by induction. For i = 1 we showed that  $I_1 \ge r + 1$ , so its speed is at most  $\sigma_2$ . For other values of i, we prove that the speed of  $I_i$  is strictly smaller than the speed of  $I_{i-1}$ . Let s' be the speed of  $I_{i-1}$ . All machines up to  $J_{i-1}$  have load of at least C' - (i - 1) = C' + 1 - i > 1 since  $i \le C' - 1$ . Recall that  $I_i \ge r + 1$  for  $i \ge 1$ . We showed that in  $\mathcal{A}$ , machines  $1, \ldots, J_{i-1}$  are loaded by more than 1. Thus in this schedule they must have a job that OPT schedules on one of the machines  $I_{i-1}, \ldots, m$ . Denote such a job and its size by a. The machine that runs it in  $\mathcal{A}$  has load of at least C' + 1 - i. Let y be the machine to which a is assigned in OPT. We have  $a \le s_y \le s'$  and  $J_{i-1} < I_{i-1} \le I_i$ . If the speed of machine  $I_i$  is s' as well, moving job a to  $I_i$  will result in load of less than (C' - i) + 1, which would be a contradiction to  $\mathcal{A}$  being a Nash equilibrium, since the load of the machine running a in  $\mathcal{A}$  is larger.

From this claim it follows that the speed of  $I_{C'-1}$  is at most  $\sigma_{C'}$ , i.e.,  $C' \leq p$  (since  $\sigma_p$  is the smallest speed). We conclude that  $T \leq C + 1 \leq C' + 1 \leq p + 1$ .

**Lemma 2** The price of anarchy on m related machines that have at most p different speeds is at least p + 1.

**Proof** Let  $\varepsilon > 0$  be such that  $1/\varepsilon$  is an even integer. We consider a set of machines with speeds in the set  $\{2^{p-1}, 2^{p-2}, \ldots, 1\}$  for some integer  $p \ge 2$ . There are  $N_i$  machines of speed  $2^i$ , where  $N_i$  will be determined later. In OPT, each machine of speed  $2^i$  has a job of size  $(1 - \varepsilon)2^i$ , for  $i \ge 1$ .  $4N_1$  of the machines of speed 1 have a single job of size  $1 - \varepsilon$  and the rest have sand (throughout the paper, we use the common term sand to describe arbitrarily small jobs) of total size 1. We will define  $N_0$  to be large enough to ensure  $N_0 > 4N_1$ . Therefore OPT = 1.

In the Nash equilibrium that we define, there is one machine of speed  $2^{p-1}$  which contains p + 1 jobs of size  $(1 - \varepsilon)2^{p-1}$ . We let  $N_{p-1} = p + 1$ . Each one of the other machines of speed  $2^{p-1}$  contains 2p jobs of size  $(1 - \varepsilon)2^{p-2}$ . We let  $N_{p-2} = 2p(N_{p-1} - 1) = 2p^2$ . For  $1 \le i \le p - 2$ , each machine of speed  $2^i$  in the Nash equilibrium contains 2(i + 1) jobs of size  $(1 - \varepsilon)2^{i-1}$ . Therefore, for these values of *i* (except for i = 1),  $N_{i-1} = 2(i + 1)N_i$ . We let  $N_0 = 4N_1/\varepsilon$ . Thus if in the Nash equilibrium, each machine of speed 1 has a total of  $1 - \varepsilon$  of sand, and in OPT, each machine except  $4N_1$  machines have a total of 1 of sand, we get that the amount of sand is constant;  $4N_1/\varepsilon - 4N_1 = (1 - \varepsilon)4N_1/\varepsilon$ .

Moreover, the load of a machine of speed  $2^i$  is  $(1 - \varepsilon)(i + 1)$ , except for one machine of speed  $2^{p-1}$  which has a load of  $(1 - \varepsilon)(p + 1)$ .

To show that this is indeed a Nash equilibrium. We do not need to consider cases in which jobs move to faster machines, since they are more loaded. We first consider the case where a job of size  $(1-\varepsilon)2^{p-1}$  moves from the machine of speed  $2^{p-1}$  that contains all jobs of this size, to a machine of some speed  $2^j$   $(j \le p-1)$ . It increases the load of the target machine by  $(1-\varepsilon)2^{p-1-j}$ . The load of this machine was  $(1-\varepsilon)(j+1)$ , so we need to show  $(1-\varepsilon)(j+1+2^{p-1-j}) \ge (1-\varepsilon)(p+1)$  or  $2^{p-1-j} \ge p-j$ . It is enough to show  $2^{t-1} \ge t$  for  $t \ge 1$ . This is easily shown by induction.

We now consider a job of size  $(1 - \varepsilon)2^i$  moving from a machine of speed  $2^{i+1}$  to a machine of speed  $2^j$ , where  $j \leq i$ . The load of the target machine increases by  $(1 - \varepsilon)2^{i-j}$ . The load there was



Figure 1: Definitions for Section 3, the structure of the Nash equilibrium which is examined.

 $(1-\varepsilon)(j+1)$  so we need to show  $2^{i-j}+j+1 \ge i+2$  for  $i-j \ge 0$ . Taking t=i-j+1, we get  $2^{t-1} \ge t$  as before.

The previous two lemmas together imply the following Theorem.

**Theorem 1** The price of anarchy on m related machines that have at most p different speeds is exactly p + 1.

Note that the SPOA increases rapidly as a function of the number of speeds as well. The lower bound construction of Fiat et al. [10] uses a parameter  $\ell$ , such that the SPOA is  $\Omega(\ell)$  and the number of speeds is  $\Theta(\ell \log \ell)$ . This implies a lower bound of  $\Omega(\frac{p}{\log p})$  on the SPOA for instances with at most p different speeds.

### **3** One fast machine: the POA

Recall that the configuration of processors that we consider here consists of m' = m - 1 identical slow machines of speed 1, and one fast machine of speed s, where  $m' \ge 2$ . Note that the case m' = 1 is fully covered in [7], for which case the POA is equal to  $1 + \frac{s}{s+2}$  for  $1 \le s \le \sqrt{2} \approx 1.4142$ , it is equal to s for  $\sqrt{2} \le s \le \frac{1+\sqrt{5}}{2} \approx 1.618$  and to  $1 + \frac{1}{s}$  otherwise.

We scale all sizes of jobs in the instances which we consider so that OPT = 1. We can therefore assume that the sum of jobs sizes is at most s + m'. Moreover, in an optimal schedule, all slow machines contain only jobs that are no larger than 1, and the largest job of any instance is no larger than s.

We assume that we are given a specific schedule with is a pure Nash equilibrium and study its properties. The price of anarchy is determined by the highest possible load of any machine. Obviously, if there is a machine with load above 1, there must also be a machine with load less than 1. To prove upper bounds we consider two basic cases; the price of anarchy is either determined by the fast machine, or by some other machine. In this schedule, denote the load on the fast machine by x, and the number of jobs there by t. Additionally, let y be the highest load of any slow machine, let  $M_y$  be a slow machine with this load, and let z be the smallest load of any slow machine. See Figure 1.

If x > 1 then the total size of jobs on the fast machine is xs > s and therefore this machine must contain at least one job that is of size no larger than 1 (which is assigned to a different machine in an optimal schedule).

We will give a closed formula for the POA for all possible combinations of s and m. We begin with the case  $1 \le s \le 2$  in Section 3.1, and in Section 3.2 we derive a condition (Equation (5)) under which the POA is equal to a global upper bound for it, denoted by GLOBMAX. We show that condition (5) is satisfied for s large enough, where the threshold value depends on m'. In particular, (5) holds for  $s \ge 4.562$  and any m' (Theorem 5).

Finally, in Section 3.3 we derive formulas for the POA in the remaining cases (Theorem 7), where (5) does not hold. Additionally, in Section 3.4, we show that  $\lim_{m'\to\infty} \text{POA} = \min(s, 2s/(s-1))$  for  $s \leq 4.562$ .

### **3.1** The exact POA for $1 \le s \le 2$

We define

FASTMAX = 
$$\frac{2m' + s}{m' + s} = 2 - \frac{s}{m' + s}$$
  
SMALLJOBS(t) =  $\frac{1 + \frac{s}{m'}}{1 + \frac{s}{m'} - \frac{s}{t}} = \frac{t(m' + s)}{t(m' + s) - m's}$ 

Recall that there are t jobs on the fast machine. We prove in the following lemma that FASTMAX is a general upper bound for x, and that SMALLJOBS(t) is an upper bound for x, in the case where  $t \ge xs$  (thus, the jobs have average size of at most 1). In the rest of the section, we answer the question of when each of these bounds is tight and when these bounds correspond to the POA. A general upper bound for x is given in Lemma 5, and upper bounds for y are given in Lemma 6. These bounds are combined in Theorem 2, which gives a closed formula for the POA for  $1 \le s \le 2$ .

Some of the lemmas and observations in this section hold not only for  $s \le 2$ , and are used in other sections as well. When this is the case, we state it explicitly (these are Lemma 3, Lemma 6 and Observation 1). Otherwise we may assume  $s \le 2$ .

**Lemma 3** If x > 1, then  $x \leq \text{FASTMAX}$ . If in addition  $t \geq xs$ , then  $x \leq \text{SMALLJOBS}(t)$ . This holds for any  $s \geq 1$ .

**Proof** The average load on the slow machines is at most

$$\frac{s+m'-xs}{m'} = 1 - (x-1)\frac{s}{m'}.$$
(1)

Since x > 1, and the optimal makespan is 1, there exists a job of size at most 1 on the fast machine. This job does not reduce its delay by moving to the least loaded slow machine. If it moves, the load on the machine that it moves to becomes at most  $2 - (x - 1)\frac{s}{m'}$ . Therefore, this value must be at least x. This implies  $x(1 + \frac{s}{m'}) \le 2 + \frac{s}{m'}$ , and therefore  $x \le \text{FASTMAX}$ .

Since there are t jobs on the fast machine, the average size of jobs there is xs/t, so among these jobs there is at least one job of size at most xs/t. This constraint does not add new information unless t > xs, we therefore assume  $t \ge xs$ , and therefore t > s. Once again, since this job does not benefit from moving to the least loaded slow machine, using (1), we find

$$x \le 1 - (x - 1)\frac{s}{m'} + \frac{xs}{t}$$

which implies  $x(1 + \frac{s}{m'} - \frac{s}{t}) \le 1 + \frac{s}{m'}$ , and therefore  $x \le \text{SMALLJOBS}(t)$  (since by t > s, we have t(m'+s) - m's > ts > 0).

**Lemma 4** Assume that t(m' + s) - m's > 0. We have  $SMALLJOBS(t) \leq FASTMAX$  if and only if  $\frac{s}{t}FASTMAX \leq 1$ .

**Proof** We have  $\text{SMALLJOBS}(t) = 1 + \frac{m's}{t(m'+s)-m's}$ . If this is at most  $2 - \frac{s}{m'+s}$ , then

$$\frac{m's}{t(m'+s) - m's} \le 1 - \frac{s}{m'+s} = \frac{m'}{m'+s}$$

and therefore

$$s \le \frac{t(m'+s) - m's}{m'+s} = t - \frac{m's}{m'+s},$$

so  $s + \frac{m's}{m'+s} \le t$  and

$$\mathsf{FASTMAX} = 2 - \frac{s}{m' + s} = 1 + \frac{m'}{m' + s} \le \frac{t}{s},$$

which implies  $\frac{s}{t}$  FASTMAX  $\leq 1$ . It can be seen that we have in fact an equivalence as long as t(m' + s) - m's > 0 (and m' > 0, s > 0 hold as well).

**Proof** We assume x > 1, otherwise the claim holds trivially. The first term is an upper bound by Lemma 3. If SMALLJOBS $(t) \le$  FASTMAX, then we have  $\frac{s}{t}$ FASTMAX  $\le 1$  by Lemma 4. Since  $x \le$  FASTMAX, Lemma 3 implies that  $x \le$  SMALLJOBS(t).

**Lemma 6** If there is only one job on  $M_y$ , then  $y \leq s$ . If there are at least two jobs, then  $y \leq 2z$  and

$$y \le \frac{2(m'+s)}{m'+2s} \; .$$

This holds for any  $s \ge 1$ .

**Proof** The first bound follows as there cannot be a job larger than s if the optimal makespan is 1.

Suppose there are at least two jobs and y > 2z. The smallest job on  $M_y$  has a size of at most y/2 and (using  $m' \ge 2$ ) it can reduce its delay by moving to a machine with a load of z where the load will be at most z + y/2 < y as a result. Thus this is not an equilibrium, which leads to a contradiction.

Therefore  $z \ge y/2$ . Since none of the jobs on  $M_y$  can improve by moving to the fast machine, we find  $y \le x + y/(2s)$  or  $x \ge \frac{2s-1}{2s}y$ . Since the total size of jobs is at most m' + s, this implies

$$(m'-1)\frac{y}{2} + \frac{2s-1}{2}y + y \le m' + s,$$

which gives

$$y(\frac{m'-1}{2} + \frac{2s-1}{2} + 1) = \frac{y}{2}(m'-1+2s-1+2) = \frac{y}{2}(m'+2s) \le m'+s,$$

which implies the desired bound.

**Corollary 1** If s < 2, then POA < 2.

**Proof** Follows immediately from Lemma 5 and Lemma 6. In fact,  $x \leq \text{FASTMAX} < 2$  and if y > s, then  $y \leq \frac{2m'+2s}{m'+2s} < 2$ .

**Observation 1** 
$$\frac{2(m'+s)}{m'+2s} < \frac{2m'+s}{m'+s} = \text{FASTMAX}$$
 for all positive m' and s.

**Proof** Since the denominators are positive, if it enough to prove  $2(m'+s)^2 < (2m'+s)(m'+2s)$ , which is equivalent to sm' > 0.

**Observation 2**  $\frac{2(m'+s)}{m'+2s} < \frac{3(m'+s)}{3(m'+s)-m's} = \text{SMALLJOBS}(3)$  for  $\frac{3}{2} < s \le 2$  and all m'.

**Proof** For  $s \le 2$ , we have 3(m'+s) > m's, therefore the denominators are positive. It is enough to prove 6(m'+s) - 2m's < 3m'+6s, which is equivalent to  $s > \frac{3}{2}$ .

**Theorem 2** For  $s \leq 2$  and  $m' \geq 2$ , we have

POA = max 
$$\left(\min\left(\text{SMALLJOBS}(2), \text{FASTMAX}, 1 + \frac{1}{s}\right), \\ \min\left(\text{SMALLJOBS}(3), \text{FASTMAX}\right), \frac{2(m'+s)}{m'+2s}, s\right)$$

**Proof** The four terms represent the following situations in order: two jobs on the fast machine, at least three jobs on the fast machine, at least two jobs on  $M_y$ , one job of size s on  $M_y$ .

It is easy to see that this covers all the relevant possibilities: if there is only one job on the fast machine, then the makespan would not be achieved on the fast machine but on  $M_y$  since  $x \leq 1$ . Therefore the upper bound will follow from showing the relevant upper bound on y. We consider the structure of an example which achieves the POA and analyze it.

In the examples for the lower bound, if the POA is achieved on the fast machine, all other machines will contain sand. In such a case, each machine will receive the same amount of sand, which in all cases would be less than 1. This already ensures that none of these jobs can improve their delay moving to the fast machine (where the load will be more than 1). Thus we only need to check that the jobs on the fast machine cannot benefit from moving.

The cases which need to be considered are the following.

- 1. The POA is achieved on the fast machine, where there are two jobs. To prove the upper bound, we note that the first two terms in the minimum are implied by Lemma 5. The last term follows because the total size of any two jobs is at most s + 1 if the optimal makespan is 1. We now show matching lower bounds using suitable instances for all three terms in the minimum.
  - (a) If the minimum is SMALLJOBS(2), we use SMALLJOBS(2)  $\leq$  FASTMAX to show that it is possible to enforce x = SMALLJOBS(2). We have  $\frac{s}{2}$ SMALLJOBS(2)  $\leq \frac{s}{2}$ FASTMAX  $\leq$  1 by Lemma 4. Therefore, consider the following instance. There are two jobs of size SMALLJOBS(2)  $\cdot \frac{s}{2} \leq 1$  which are running on the fast machine, i.e., t = 2. The total amount of sand is  $m' + s s \cdot$  SMALLJOBS(2) > 0 (since  $m' \geq 2$ ). Each slow machine has sand, where the amount of sand on each slow machine is  $1 \frac{s^2}{2(m'+s)-m's}$ . The optimal makespan is 1, by putting each large job on one machine, and adding sand to achieve an equal load on the machines. This schedule is an equilibrium since by moving a large job to a slow machine we get a delay of  $1 \frac{s^2}{2(m'+s)-m's} + \frac{2(m'+s)}{2(m'+s)-m's} \cdot \frac{s}{2} = \frac{2(m'+s)}{2(m'+s)-m's} = x$ .
  - (b) If the minimum is FASTMAX, we use FASTMAX  $\leq 1 + \frac{1}{s}$  to enforce x = FASTMAX. By Lemma 4,  $\frac{s}{2}$ FASTMAX  $\geq 1$ . Consider the following instance. There is one job of size 1 on the fast machine and one job of size sFASTMAX  $-1 \geq 1$ . Each slow machine has sand, where the amount of sand on each slow machine is (m' + s - sFASTMAX $)/m' = (m' + s - \frac{s(2m' + s)}{m' + s})/m' = \frac{m'}{m' + s}$ . This is an equilibrium since already moving the smaller job from the fast machine to a slow machine results in a load of  $1 + \frac{m'}{m' + s} = \text{FASTMAX}$ . If sFASTMAX  $-1 \leq s$ , then the optimal makespan is 1, since the fast machine runs this job, one slow machine runs the job of size 1, and the sand is distributed so that the machines are balanced. The condition on the size of the largest job holds since FASTMAX  $\leq 1 + \frac{1}{s}$  in this case.
  - (c) In the last case, where the minimum is  $1 + \frac{1}{s}$ , we show how to enforce  $x = 1 + \frac{1}{s}$ . In the instance, there is one job of size 1 on the fast machine and one job of size s. Each slow machine has an amount of (m' + s s 1)/m' = 1 1/m' of sand. This is an equilibrium since using  $1 + 1/s \leq \text{FASTMAX}$  in this case, we get  $m's \geq m' + s$  and therefore  $2 \frac{1}{m'} \geq 1 + \frac{1}{s}$ , which means that already the job of size 1 does not benefit from moving to a slow machine. In an optimal assignment, the fast machine runs the job of size s, one slow machine runs the job of size 1, and the sand is spread evenly between the other slow machines.
- 2. There are at least three jobs on the fast machine. First note that SMALLJOBS(t) is decreasing in t, so the upper bound follows directly from Lemma 5. There are two cases depending on the term for which the minimum is achieved. In both cases, each slow machine will receive an identical amount of sand, which is again chosen such that the total size of all the jobs is m' + s.
  - (a) If the minimum is SMALLJOBS(3), we enforce x = SMALLJOBS(3). There are 3 jobs of size SMALLJOBS(3)  $\cdot \frac{s}{3}$  on the fast machine. This is an equilibrium (this can be proved very similarly to the proof of case 1a). By Lemma 4, the jobs have size at most 1 since

SMALLJOBS(3)  $\leq$  FASTMAX. Since  $m' \geq 2$ , the optimal makespan is 1: these three jobs can each be assigned to separate machines, which can then be filled up with sand.

- (b) If the minimum is FASTMAX, we enforce x = FASTMAX. There are two jobs of size 1 on the fast machine and one job of size sFASTMAX 2. The size of the second job is at least 1 since xs/3 > 1 by Lemma 4 if we take x = FASTMAX, and at most s since  $\text{FASTMAX} \le 2 \le 1 + 2/s$  for  $s \le 2$ . In an optimal schedule, for  $m' \ge 2$  the jobs of size 1 can be assigned to two slow machines, and the larger job to the fast machine. This is an equilibrium (the proof of case 1b, including the calculation of the amount of sand on slow machines, holds here as well).
- 3. The POA is achieved on a slow machine with two jobs. The upper bound follows from Lemma 6. Comparing this case to the previous one, by Observations 1 and 2, we have  $s \leq 3/2$ . We show how to enforce  $y = \frac{2(m'+s)}{m'+2s}$ , let y denote this value. Note that this function is monotonically increasing in m'. To prove the lower bound, we consider a schedule with two jobs of size y/2 < 1 on one slow machine  $M_y$ , each other slow machine has y/2 of sand.

If  $m' \leq 3$ , the fast machine has one job of size  $\frac{2s-1}{2}y$ . For  $m' \geq 4$ , the fast machine has two jobs of size  $\frac{2s-1}{4}y \leq 1$ . In both cases, its load is  $\frac{2s-1}{2s}y \leq \frac{3}{4}y \leq \frac{3}{2}$ .

We show that the optimal makespan of this instance is 1 in all cases. Each job that is not part of the sand is put on a separate machine. We will show that there is at most one job that is larger than 1. If there is such a job, it is put on the fast machine. The sand is added to the machines in a balanced way. The jobs have a total size of  $\frac{y}{2}(2 + (m' - 1) + (2s - 1)) = m' + s$ . For  $m' \leq 3$ , we need to show that  $\frac{2s-1}{2}y \leq s$ . This holds because  $\frac{2(m'+s)}{m'+2s}(s-\frac{1}{2}) \leq \frac{6+2s}{3+2s}(s-\frac{1}{2}) \leq s \Leftrightarrow (6+2s)(s-\frac{1}{2}) \leq s(3+2s) \Leftrightarrow 5s-3 \leq 3s \Leftrightarrow s \leq \frac{3}{2}$ . For  $m' \geq 4$ ,  $\frac{2s-1}{4}y \leq \frac{1}{2}y < 1$  for  $s \leq 3/2$ .

The assignment is an equilibrium: the job(s) on the fast machine can not improve by moving, since already for the case where two jobs are assigned to this machine,  $\frac{2s-1}{4}y + y/2 = \frac{2s+1}{4}y \ge \frac{2s-1}{2s}y$  for s > 1. Since  $\frac{2s-1}{2s}y + \frac{y}{2s} = y$ , the jobs on  $M_y$  do not improve by moving to the fast machine. They also cannot benefit from moving to another slow machine. The sand cannot improve since the load on the fast machine  $\frac{(2s-1)y}{2s} \ge \frac{y}{2}$  for  $s \ge 1$ .

4. The POA is achieved on a slow machine with one job. The upper bound follows from Lemma 6. We show how to enforce a load of y = s on M<sub>y</sub> using the following schedule. There is one job of size s on a slow machine M<sub>y</sub>. Let t = [s(s − 1)]. There are t jobs of size s(s − 1)/t ≤ 1 on the fast machine, so that the load there is s − 1. The remaining machines have load m'-s(s-1)/m'-1 = 1 - s(s-1)-1/m'-1 = < 1, which we denote by z'. This load consists entirely of sand.</p>

If z' > s-1, redistribute the sand among all machines besides  $M_y$  (i.e. including the fast machine) so that the load on these machines is equal. This is done by moving some sand to the fast machine. It is clear that if this redistribution takes place, we have an equilibrium: we only need to check whether jobs can benefit by moving to or from  $M_y$ . But the job on  $M_y$  cannot improve since the load on the fast machine is at least s-1, and s-1+s/s=s. No job can improve by moving to  $M_y$  because  $M_y$  has the highest load.

Consider the case where  $z' \leq s - 1$ . Then clearly, none of the sand jobs can improve. Suppose  $s \leq \phi$ . Then t = 1. The single job of size s(s - 1) on the fast machine does not improve if it moves to  $M_y$  (load is higher) or to another slow machine (the load would become be at least s(s-1) > s - 1).

Suppose  $\phi < s \le 2$ . Then t = 2. We need to check that  $z' + s(s-1)/2 \ge s-1$ , so that the jobs on the fast machine do not improve by moving to a slow machine. This holds for m' = 2: we get

 $1 - (s(s-1)-1) + s(s-1)/2 \ge s-1 \Leftrightarrow -s(s-1)/2 \ge s-3 \Leftrightarrow -s^2 + s \ge 2s-6 \Leftrightarrow s^2 + s - 6 \le 0 \Leftrightarrow (s+3)(s-2) \le 0$ , which holds for  $s \le 2$ . Then it clearly also holds for larger m', because z' is increasing in m'.

This shows that the maximum can indeed be achieved in all four cases, and thus the bounds are tight.  $\Box$ Corollary 2 For s = 2, POA = 2 for all  $m' \ge 2$ .

**Proof** All the upper bounds in Theorem 2 are at most 2 for s = 2 and any  $m' \ge 2$ , and the bound s is equal to 2.

#### **3.2** The exact POA for sufficiently large values of s

In this section and the next one, we will prove that there exist instances which achieve the POA with several distinct properties. In each case, as soon as we have proved such a statement, we will restrict our attention to instances which have these properties in the remainder of the text.

We first show that  $\lim_{m'\to\infty} POA = s$  for 2 < s < 3 in Theorem 3. Lemma 7 shows that POA = FASTMAX in case the POA is achieved on the fast machine and  $s \ge 2$ . The rest of the text then deals with the case where the POA is achieved on the slow machine. We first derive structural properties for equilibria and an upper bound for y, the maximum load on a slow machine, in Lemmas 8 to 11, and then show that the POA is upper bounded by

GLOBMAX = 
$$\frac{s + 2m' - 1}{s + (m' - 1)(s - 1)/s}$$
. (2)

for all  $s \ge 2$  in Lemmas 12 and 13. We then derive bounds for GLOBMAX in Lemmas 14 and 15, and use these bounds to determine when POA = GLOBMAX holds in Theorems 4 and 5.

**Theorem 3** For 2 < s < 3 and large enough m', POA = s.

**Proof** Fix  $\varepsilon \in (0, 1]$ . We will show a lower bound of s on the POA for  $s = 3 - \varepsilon$  and sufficiently large m'. Consider the following schedule.

There is one job of size s which is scheduled on a dedicated slow machine. There are six jobs of size s(s-1)/6 which are on the fast machine, so its load is s-1. The remaining m'-1 machines have sand, specifically, each machine has an amount of  $(s-1) \cdot (1-\frac{s}{6}) < 1$  which is less than 1 for s < 3. The amount of sand per machine ensures that none of the six jobs on the fast machine improves by moving to a slow machine: if such a job moves there, it adds s(s-1)/6 to the load, making the total load exactly s - 1.

We need to make sure that the total size of all the jobs we use is not more than m' + s. This implies

$$m' + s \ge s + s(s-1) + (m'-1)(s-1)\left(1 - \frac{s}{6}\right)$$
  
$$\Leftrightarrow \quad m' \ge \frac{7s^2 - 13s + 6}{s^2 - 7s + 12} = \frac{(7s-6)(s-1)}{(3-s)(4-s)} = \frac{(15-7\varepsilon)(2-\varepsilon)}{\varepsilon(1+\varepsilon)}$$

For any  $\varepsilon > 0$ , this value is bounded from above. Since x < 2 by Lemma 3 and  $y \le s$  by Lemma 6 (the second bound there is at most  $2 \le s$ ), this proves the theorem.

**Lemma 7** For  $s \ge 2$ , if the POA is achieved on the fast machine then POA = FASTMAX.

**Proof** If the POA is achieved on the fast machine then we have POA  $\leq$  FASTMAX by Lemma 3. For the lower bound we present an instance where the fast machine has a load of FASTMAX. We place a total size of jobs of w = sFASTMAX on the fast machine. We do this by assigning  $\lceil s(1 - s/(m' + s)) \rceil \ge 1$  jobs of size 1 to this machine, as well as a job of size  $q = w - \lceil s(1 - s/(m' + s)) \rceil$  (where  $q = s(2 - s/(m' + s)) - \lceil s(1 - s/(m' + s)) \rceil \le s$  and  $q \ge s(2 - s/(m' + s)) - (s(1 - s/(m' + s)) + 1) = s - 1 \ge 1$ 

by the assumption  $s \ge 2$ ). We get x = FASTMAX, and a total size of sand jobs of s + m' - xs, thus each slow machine receives 1 - (x - 1)s/m' of sand. This situation is similar to the one in the proof of Theorem 2 (Case 1b), since the loads of machines are similar, and the smallest job on the fast machine is of size 1, and we already saw that this is an equilibrium.

To show that the optimal makespan is 1, we need to show that  $\lceil s(1 - s/(m' + s)) \rceil \le m'$  (i.e., we can assign all jobs on the fast machine but one to slow machines) and  $q \le s$  (which we already showed). The first inequality holds because  $s(1 - s/(m' + s)) = sm'/(m' + s) = m' \cdot \frac{s}{m' + s} < m'$ , so  $\lceil s(1 - s/(m' + s)) \rceil \le m'$ .

The question now becomes *when* the POA is achieved on the fast machine. To answer this question, we begin by deriving upper bounds on the load on any slow machine, and structural properties for worst-case equilibria in the following four lemmas. Eventually, we will find that the POA is in fact always achieved on a slow machine for  $s \ge 2$ .

**Lemma 8** For  $s \ge 2$ , consider an instance in which the POA is achieved, and it is achieved on a slow machine, i.e., POA = y. In this case, there is one job on  $M_y$ . Moreover, all additional slow machines contain sand.

**Proof** If there are at least two jobs on  $M_y$ , then by Lemma 6 and Observation 1, y < FASTMAX. But then y < POA by Lemma 7.

Let  $\Delta$  be the total size of jobs assigned to slow machines, excluding  $M_y$ . Replace these jobs by sand and distribute it evenly. The same process is applied on the jobs in the optimal solution. The only case which the resulting schedule is not an equilibrium is the case where the fast machine has a smaller load than the resulting load of the slow machines, in this case, the jobs on it are replaced by sand as well, and the sand is redistributed so that all machines, except for  $M_y$ , have equal load.

**Lemma 9** For any equilibrium instance, there exists an instance which is an equilibrium with the same loads on all machines, such that the fast machine has at most one job which is on the fast machine in the optimal solution. Specifically, it has at most one job larger than 1.

**Proof** If there are multiple such jobs, we can merge them into one job with size the total size of these jobs. This does not affect the optimal makespan, or the makespan of the schedule. Larger jobs can only benefit less from moving, thus the schedule is still an equilibrium if it was before. Regarding the second statement, clearly all jobs larger than 1 must be on the fast machine in an optimal solution with makespan 1.

Lemma 10 Any schedule which is in equilibrium satisfies

$$y \le \frac{xs}{s-1} \,. \tag{3}$$

Moreover, if  $M_y$  has a single job, this is a sufficient condition for this job not to benefit from moving.

**Proof** Consider  $M_y$ . This machine has a job of size at most y, which does not benefit from moving to the fast machine. Therefore  $y \le x + \frac{y}{s}$ , which implies the upper bound (3). If there is a single job of size y, then this is not only a necessary condition but also a sufficient condition.

**Lemma 11** For  $s \ge 2$  and  $m' \ge 2$ , if there exists an an equilibrium schedule where the POA is achieved, and it is achieved on a slow machine, then the POA is achieved in an instance with  $t \ge 2$ .

Note: this holds even after possibly merging some jobs as in the proof of Lemma 9.

**Proof** Suppose there is at most one job on the fast machine. The total size of the jobs on the fast machine and  $M_y$  (together) is then at most s + 1. This means that  $xs + y \le s + 1$ , or  $xs \le s + 1 - y$ . But then Lemma 10 implies  $y \le \frac{s+1-y}{s-1}$ , and therefore  $y(1 + \frac{1}{s-1}) \le \frac{s+1}{s-1}$ , or  $y \le \frac{s+1}{s}$ . But this value is smaller than 2 - s/(m' + s) for  $s \ge 2$  and  $m' \ge 2$ . To prove this we note that 2 - s/(m' + s) is increasing in m'. For m' = 2, it is equal to  $2 - s/(2 + s) = 1 + \frac{2}{2+s}$ . However  $\frac{2}{2+s} \ge \frac{1}{s}$  for  $s \ge 2$ .  $\Box$ 

The next lemma relates FASTMAX to GLOBMAX, allowing us to prove a general upper bound for the POA in Lemma 13.

**Lemma 12** We have FASTMAX < GLOBMAX for all  $s \ge 1$ ,  $m' \ge 2$ . **Proof** We have FASTMAX =  $\frac{2m'+s}{m'+s}$ . Working out the desired inequality gives

$$\frac{(2m'+s)s}{sm'+s^2} < \frac{(2m'+s-1)s}{s^2+(m'-1)(s-1)}$$

which holds if and only if

$$(2m'+s)(s^2+m's-m'-s+1) < (sm'+s^2)(2m'+s-1).$$

Working this out gives

 $2m's^{2} + 2m'^{2}s - 2m'^{2} - 2sm' + 2m' + s^{3} + m's^{2} - m's - s^{2} + s < 2m'^{2}s + m's^{2} - m's + 2m's^{2} + s^{3} - s^{2}$ This holds if  $2m' + s < 2sm' + 2m'^{2}$ , which is true because s < 2sm' and  $2m' < 2m'^{2}$ .

**Lemma 13** For  $s \ge 2$ , POA  $\le$  GLOBMAX.

**Proof** By Lemma 12 and Lemma 7, the claim holds if the POA is achieved on the fast machine. Therefore, suppose it is achieved on a slow machine  $M_y$  in some schedule. Denote the load there by y > 1. Then by Lemma 10, the load on the fast machine is at least  $x = y \cdot \frac{s-1}{s}$ , so the work there is y(s-1). By Lemma 9, the fast machine has at most one job larger than 1. By Lemma 11, the fast machine has at least two jobs, such that at least one of them is scheduled on a slow machine in an optimal schedule. Therefore, there is at least one job of size at most 1 on the fast machine. If this instance is an equilibrium, the load on each slow machine must then be at least x - 1. Finally, the total size of all the jobs must be at most m' + s. This implies

$$y\left(1+(s-1)+(m'-1)\frac{s-1}{s}\right)-(m'-1)\le s+m'$$
(4)

which holds if and only if  $y \le \frac{s+2m'-1}{s+(m'-1)(s-1)/s} = \text{GLOBMAX}$ . This proves the lemma.  $\Box$ We wish to find out when POA = GLOBMAX holds exactly. To give a condition for this, we first

We wish to find out when POA = GLOBMAX holds exactly. To give a condition for this, we first study the value GLOBMAX further in Lemmas 14 and 15.

**Lemma 14** We have  $GLOBMAX \leq (s + m')/s$  for all  $m' \geq 1$  and  $s \geq 2$ .

**Proof** We need to show that

$$\frac{s+2m'-1}{s+(m'-1)(s-1)/s} - \frac{m'}{s} \le 1$$

or

$$\frac{s^3 + 2m's^2 - s^2 - m'(s^2 + (m'-1)(s-1))}{s^3 + s(m'-1)(s-1)} \le 1$$

which holds if

$$m's^2 - s^2 - m'(m'-1)(s-1) \le s(m'-1)(s-1)$$

or (by dividing by  $m' - 1 \ge 1$  for  $m' \ge 2$ )

$$s^{2} - m'(s - 1) \le s(s - 1) = s^{2} - s.$$

Thus it is sufficient if  $m'(s-1) \ge s$ , or  $(s-1)/s = 1 - 1/s \ge 1/m'$ , or  $s \ge \frac{m'}{m'-1}$ . However, we have  $s \ge 2 \ge \frac{m'}{m'-1}$  by our assumption and because  $m' \ge 2$ , and therefore the claim holds.  $\Box$ Lemma 15 We have  $\frac{s-1}{s}$ GLOBMAX > 1 for all  $m' \ge 2$  and s > 2. **Proof** We can write the desired inequality as 1/GLOBMAX + 1/s < 1. Thus we need to show that

$$\frac{s^2 + (m'-1)(s-1)}{s^2 + 2m's - s} + \frac{1}{s} = \frac{s^2 + (m'-1)(s-1) + (s+2m'-1)}{s^2 + 2m's - s} < 1.$$

This holds if and only if

$$(m'-1)(s-1) + s + 2m' - 1 = m's + m' < 2m's - s.$$

But m' < s(m'-1) holds for  $s > \frac{m'}{m'-1}$  and then certainly for s > 2 (using  $m' \ge 2$ ).  $\Box$ **Theorem 4** Let  $m' \ge 2$  and s > 2. Then

$$(s-1)\mathsf{GLOBMAX} - \lceil s(\mathsf{GLOBMAX} - 1) \rceil \ge 1 \tag{5}$$

*implies that* POA = GLOBMAX.

**Proof** First note that by the condition (5),  $s(\text{GLOBMAX} - 1) \leq \lceil s(\text{GLOBMAX} - 1) \rceil \leq (s - 1)\text{GLOBMAX} - 1$  and so  $\text{GLOBMAX} \leq s - 1$ .

We present a class of instances, where POA = GLOBMAX, as long as (5) holds. Note that POA > GLOBMAX is impossible by Lemma 13. Place one job of size y = GLOBMAX on a slow machine. Set  $x = \frac{s-1}{s} \cdot y$ . Place

$$k = \lceil s(\text{GLOBMAX} - 1) \rceil$$

jobs of size 1 on the fast machine, together with a job of size

$$q = (s-1)\text{GLOBMAX} - k$$
  

$$\leq (s-1)\text{GLOBMAX} - s(\text{GLOBMAX} - 1) = s - \text{GLOBMAX}.$$
(6)

Then the total work assigned to the fast machine is (s - 1)GLOBMAX = xs, as desired.

On each empty slow machine, place x - 1 of sand. This is more than 0 by Lemma 15. We now have constructed an equilibrium, which can be verified as follows. Note that since y > x > x - 1, we only need to check that no job can improve by moving away from  $M_y$  or from the fast machine to a slow machine with load x - 1. The first part follows from Lemma 10 and the fact that there is only one job on  $M_y$  (so that job cannot improve by moving to another slow machine), and the second part holds as long as all the jobs on the fast machine (in particular, the job of size q) have size at least 1. This is exactly the condition (5).

We still need to verify that the optimal makespan of this instance is 1. First of all, the total size of all the jobs must be at most m' + s. This follows because (4) holds for y = GLOBMAX, and our loads are exactly the loads described in Lemma 13. Furthermore, we need to show that  $k \le m'$ . This holds as long as  $s(\text{GLOBMAX} - 1) \le m'$ , or  $\text{GLOBMAX} \le 1 + m'/s$ . This is true by Lemma 14. Since finally  $q \le s - \text{GLOBMAX}$  by (6), an optimal schedule with makespan 1 exists, since the jobs of size q and GLOBMAX can be placed together on the fast machine.

**Theorem 5** For  $s \ge \frac{5+\sqrt{17}}{2} \approx 4.562$ , we have POA = GLOBMAX.

**Proof** We give a condition which ensures that (5) holds. We know that (s - 1)GLOBMAX –  $\lceil s(\text{GLOBMAX} - 1) \rceil \ge (s - 1)$ GLOBMAX – s(GLOBMAX - 1) - 1 = s - GLOBMAX - 1. Thus, it suffices to have  $s - \text{GLOBMAX} - 1 \ge 1$ , or

$$GLOBMAX \leq s - 2,$$

in order to ensure (5). GLOBMAX is monotonically increasing in m' (we have  $\partial$ GLOBMAX $/\partial m' = s(s^2+1)/(s^2+(m'-1)(s-1))^2 > 0$ ) and tends to 2s/(s-1) for  $m' \to \infty$ . We have  $2s/(s-1) \le s-2$  for  $s \ge 4.562$ . The result now follows from Theorem 4.

In the following table, for several values of m' the minimum value of s is given such that we can be certain that POA = GLOBMAX for all speeds of at least s, rounded to three decimal places.

In all of these cases, (5) holds. Indeed, we noted in the proof of Theorem 5 that as m' grows large relative to s, GLOBMAX tends to 2/((s-1)/s) = 2s/(s-1). Then s(GLOBMAX-1) = s(2s-s+1)/(s-1) = s(s+1)/(s-1) and (s-1)GLOBMAX = 2s. For these values, inequality (5) holds for  $s \ge 4$ , so for large m', the bound for s above which POA = GLOBMAX tends to 4.

Using a computer program, it can be found that in fact POA = GLOBMAX for  $s \ge 4.365$  for all m', and that the value of m' for which the bound on s is maximized is 31. There are also several values of m' for which POA = GLOBMAX in non-contiguous intervals. The smallest value of m' for which this happens is m' = 14.

#### **3.3** The POA for intermediate values of *s*

Theorem 4 gives us a condition under which POA = GLOBMAX. What happens if this condition is not satisfied? We certainly still have the upper bound from Lemma 7 for the case where the POA is achieved on the fast machine. In this section, we therefore focus on the case that the POA is achieved on  $M_y$  (and thus y > 1). We assume that the modification of Lemma 9 was already applied on the schedule. We first show an additional structural property of worst-case equilibria in Lemma 16. We then give an upper bound for y which depends on the number of jobs t on the fast machine in Lemma 17. This raises the question of which value of t should be selected to get the highest possible POA. We first define a crucial value  $t^*$  in Definition 2, and examine this value in the remainder of Section 3.3.1. Section 3.3.2 then answers the question of how to select t based on  $t^*$ . The results are summarized in Theorem 7.

**Lemma 16** For  $s \ge 2$ , consider an instance where the POA is achieved on  $M_y$ . In this case, there exists an instance which achieves the POA, where no two jobs on the fast machine have a total size of at most 1.

**Proof** We first show that  $t \le m' + 1$ . Suppose this does not hold. By Lemma 9, there are at least  $t - 1 \ge m' + 1$  jobs on the fast machine which are on slow machines in the optimal solution, so there are at least two jobs from the same slow machine. Now these two jobs can be merged without affecting the equilibrium or the optimal makespan.

For the second statement, if there do exist two such jobs, we merge them into one larger job. Since  $t \le m' + 1$ , this leaves at most m' jobs on the fast machine, all of which have size at most 1. Thus we can assign each such job to its own slow machine, and the other jobs as in the previous case.

#### **3.3.1** The bound MAXSMALL(t) and the value $t^*$

#### **Definition 1** Let

MaxSmall(t) = 
$$\frac{s+m'}{s+(m'-1)(s-1)(1-s/t)/s}$$
.

We prove in the following lemma that MAXSMALL(t) is an upper bound for the load on  $M_y$  in case there are t jobs on the fast machine, and  $t \ge s$ .

**Lemma 17** If y > 1, and there are  $t \ge s$  jobs on the fast machine, then  $y \le MAXSMALL(t)$ .

**Proof** For an equilibrium, we require  $x \ge \frac{s-1}{s} \cdot y$  by Lemma 10. The load on any slow machine must be at least x - xs/t = x(1 - s/t), since xs/t is an upper bound on the size of the smallest job on the fast machine if there are t jobs on that machine. This implies

$$y\left(s+(m'-1)\frac{s-1}{s}\left(1-\frac{s}{t}\right)\right) \le s+m' \tag{8}$$

which together with  $t \ge s$  proves the upper bound.

**Observation 3** Let  $s \ge 2$ ,  $m' \ge 2$ . If MAXSMALL(t') > 0 for some  $t' \in \mathbb{R}^+$ , then for all real t > t' we have that 0 < MAXSMALL(t) < MAXSMALL(t') and that MAXSMALL(t) is continuous.

This observation holds because the condition MAXSMALL(t') > 0 is equivalent to the condition that its denominator be positive. But the denominator is strictly increasing in t for all t > 0, and hence there can be no discontinuity for t > t'.

Given Observation 3, we would like to choose t as small as possible in order to get a high POA. However, if it is too small, we find that one of the conditions t < s or t < xs will start to hold, in which case Lemma 17 does not give us a useful bound (in the proof, we use xs/t as an upper bound for the size of the smallest job on the fast machine; if t < xs, we have the stronger bound of 1).

We therefore define the following value, which will give us an initial upper bound for the POA (Lemma 21). Later, we will deal with the cases where there are fewer jobs on the fast machine.

### **Definition 2** Let $t^*$ be the minimum value of $t \in \mathbb{N}$ such that $t \ge (s-1)MAXSMALL(t) > 0$ .

We will see in the following that MAXSMALL( $t^*$ ) is the highest load on a slow machine that can be achieved using at least  $t^*$  jobs on the fast machine, while with fewer jobs on the fast machine, we get smaller bounds. Before we move on to the proofs of these statements, we first prove some useful properties of the value  $t^*$  that we will need later. In view of the condition in Lemma 17, we first derive a lower bound for  $t^*$ .

Solving the equation t = (s - 1)MAXSMALL(t) for  $t \in \mathbb{R}$  gives

$$t_1 = \frac{s^2 + 2(m'-1)(s-1) - 1}{s^2 + (m'-1)(s-1)} \cdot s.$$
(9)

**Lemma 18** We have  $t^* = \lfloor t_1 \rfloor > s$  for any  $m' \ge 2$  and s > 2.

**Proof** The fraction in the right hand side of (9) is strictly more than 1 for any  $m' \ge 2$  and s > 2, therefore  $t_1 > s$ . In particular, this implies  $t_1 > 0$ , and therefore MAXSMALL $(t_1) = t_1/(s-1) > 0$ . Since MAXSMALL(t) is decreasing and continuous as long as it is positive, we conclude  $t^* = \lceil t_1 \rceil > s$ : for  $t \le \lceil t_1 \rceil - 1$ ,  $t \in \mathbb{N}$ , we must have either MAXSMALL $(t) \le 0$  or t < (s-1)MAXSMALL(t).  $\Box$ 

This lemma shows that for  $t \ge t^*$ , the second condition in Lemma 17 is always satisfied. We show two additional bounds involving  $t^*$ , which will restrict the number of cases that we need to consider.

Lemma 19  $t^* - 1 < (s - 1)MAXSMALL(t^* - 1).$ 

**Proof** Given the definition of  $t^*$  and Observation 3, it is sufficient to show that MAXSMALL $(t^*-1) > 0$ . This holds if the denominator is positive. Solving s + (m'-1)(s-1)(1-s/t)/s = 0 for  $t \in \mathbb{R}$  gives

$$t_2 = \frac{(m'-1)(s-1)}{s^2 + (m'-1)(s-1)} \cdot s.$$

For any  $t > t_2$ , we have MAXSMALL(t) > 0 and then MAXSMALL(t) is continuous, decreasing in t, and positive by Observation 3. Thus if  $t^* > t_2 + 1$ , we have MAXSMALL $(t^* - 1) > 0$  as desired. Given Lemma 18, it is sufficient to show  $t_1 > t_2 + 1$ . Note that the denominator of  $t_2$  is equal to that of  $t_1$ . Thus we need to verify

$$(s^2 + 2(m'-1)(s-1) - 1)s > (m'-1)(s-1)s + (s^2 + (m'-1)(s-1))$$
  
or 
$$(s^2 + (m'-1)(s-1) - 1)s > s^2 + (m'-1)(s-1)$$

But this last inequality holds for s > 2 and  $m \ge 2$ , because then  $(s^2 + (m' - 1)(s - 1) - 1)s > 2(s^2 + (m' - 1)(s - 1) - 1) > (s^2 + (m' - 1)(s - 1) - 1) + (s^2 + 1 - 1) > s^2 + (m' - 1)(s - 1)$ . Lemma 20 If (5) does not hold, then  $t^* \le m' + 1$ . **Proof** We have that  $t_1$  is monotonically strictly increasing in s and in m for  $s \ge 2$  and  $m \ge 2$ : we can write it as

$$t_1 = s \cdot \left(1 - \frac{1}{s^2 + (m'-1)(s-1)}\right) + \frac{1}{\frac{s}{(m'-1)(s-1)} + \frac{1}{s}},$$

from which both assertions follow easily. Therefore  $t^* = \lfloor t_1 \rfloor$  is monotonically increasing in s and m.

If (5) does not hold, then  $s \le 4.56$  by Lemma 13. From (9) it is clear that  $t_1 < 2s$  and therefore  $t^* \le \lceil 2s \rceil$ . Thus the claim holds for  $m' + 1 \ge 10$ , or  $m' \ge 9$ . For  $m' = 2, \ldots, 8$ , we use the values from Table (7). Thus only the interval  $s \in [2, 3.9]$  remains to be checked, so we are done for m' = 8 and m' = 7, because  $t^* \le 8$  for  $s \le 4$ . For m' = 6, we know s < 3.5, so we are done for that value as well. For m' = 5 and s = 4, we find  $t^* = \lceil 39/7 \rceil = 6 = m' + 1$ , implying that if m' = 5, we have  $t^* \le 6$  for all s for which (5) does not hold (since then  $s \le 4$ , and  $t_1$  and  $t^*$  are increasing in s).

For the remaining values, we have the following results. We only treat the case m' = 2, the other two cases can be solved similarly. For m' = 2, we have  $s(\text{GLOBMAX} - 1) = (2s^2 + s)/(s^2 + s - 1) < (2s^2 + 2s - 2)/(s^2 + s - 1) = 2$  for all s > 2, and also clearly s(GLOBMAX - 1) > 1. Thus the values of s, such that s > 2, for which (5) does not hold, satisfy (s - 1)GLOBMAX - 2 < 1, or

$$(s-1)(s^2+3s)/(s^2+s-1) < 3.$$
<sup>(10)</sup>

On the other hand, for m' = 2 we have  $t_1 = \frac{s^2 + 2(s-1) - 1}{s^2 + (s-1)}s$ . We know that as long as  $t_1 \leq 3$ , we also have  $t^* \leq 3$ . We have  $t_1 \leq 3$  if

$$\frac{s^2 + 2(s-1) - 1}{s^2 + (s-1)}s = 3.$$
(11)

It is now easy to see that (11) follows from (10). Since  $t_1$  is monotonically increasing in s, this shows that  $t^* \leq 3$  as long as (5) does not hold.

#### **3.3.2** On the value of t which maximizes the POA

This section deals with the question: how should we select t, i.e. how many jobs should there be on the fast machine in order to get the highest possible POA. Lemma 21 deals with the case where there are at least  $t^* \leq m'$  jobs on the fast machine, and shows that the worst case (highest POA) is if there are exactly  $t^*$  jobs. Lemma 25 deals with the case where there are less than  $t^*$  jobs on the fast machine, and shows that the worst case is if there are exactly  $t^* - 1$  jobs.

Finally, Lemma 28 (in general, the text below (13)) deals with the case where  $t^* = m' + 1$ , which requires separate attention, and gives a new upper bound for the poa for this case.

**Lemma 21** If  $t^* \leq m'$ , then there is an equilibrium instance with  $t^*$  jobs on the fast machine and  $y = \min(s, \text{MAXSMALL}(t^*))$ . If y > 1, then we have  $y \leq \min(s, \text{MAXSMALL}(t^*))$  for all equilibria with at least  $t^*$  jobs on the fast machine.

**Proof** To show existence, let  $y = \min(s, \text{MAXSMALL}(t^*))$ . Place a job of size y on a slow machine,  $t^*$  jobs of size  $y(s-1)/t^* \leq 1$  on the fast machine and  $z = y\frac{s-1}{s}(1-s/t^*) \geq 0$  of sand on the empty slow machines. The claimed inequalities in the previous line follow from the definition of  $t^*$  and the fact that  $t^* > s$  (Lemma 18). Then this is an equilibrium with POA = y. The job of size y does not benefit from moving by Lemma 10, and no job on the fast machine benefits from moving because  $z + y(s-1)/t^* = y(s-1)/s$ . Since  $y > y(s-1)/s > y\frac{s-1}{s}(1-s/t^*)$ , the sand also does not benefit from moving.

The total size of all the jobs is at most m' + s since  $y \leq MAXSMALL(t^*)$  so that y satisfies (8). Since we also have  $y \leq s$ , this shows that the optimal makespan is 1 as long as  $t^* \leq m'$ , because we can then assign each job which is on the fast machine to its own slow machine in the optimal solution. With exactly  $t^*$  jobs on the fast machine, the second claim follows from Lemma 6 if  $y = s \leq MAXSMALL(t^*)$ . Else, we can use Lemma 17. With more than  $t^*$  jobs on the fast machine, we use additionally that MAXSMALL(t) is decreasing in t (Observation 3).

**Lemma 22** If (5) does not hold, there is an equilibrium instance with  $t^* - 1$  equal-sized jobs on the fast machine and  $y = \min(s, (t^* - 1)/(s - 1))$ . Any equilibrium instance with at most  $t^* - 1$  jobs on the fast machine, where all those jobs have size at most 1, has  $y \le \min(s, (t^* - 1)/(s - 1))$ .

**Proof** To prove the first claim, we use an instance analogous to the one from the proof of Lemma 21. There is a job of size  $y = \min(s, (t^* - 1)/(s - 1))$  on one slow machine, and  $t^* - 1$  jobs of size  $y(s - 1)/(t^* - 1) \le 1$  on the fast machine. Each slow machine has an equal amount of sand  $z = \max(0, y\frac{s-1}{s}(1 - s/(t^* - 1)))$ . If x < z, we redistribute the sand among the fast machine and the slow machines excluding  $M_y$  so that all loads are equal (without changing the total size of all the jobs).

Then as in the previous proof, this is an equilibrium with POA = y. (If we redistributed some sand because x < z, the proof is even easier.) We still need to show that the optimal makespan is 1. Note that  $t^* - 1 \le m'$  by Lemma 20, so that in the optimal schedule, we can assign each job which is on the fast machine to its own slow machine, and  $y \le s$  to the fast machine as before. It remains to be shown that the total size of all the jobs is at most m' + s. If z > 0, this follows since  $y \le MAXSMALL(t^* - 1)$  by Lemma 19 so that y satisfies (8). If z = 0, this follows because  $y \le s$  and there are  $t^* - 1 \le m'$  jobs of size at most 1.

For the second claim, note that if all jobs on the fast machine have size at most 1, the total work there, xs, is at most  $t^* - 1$  in this case. The claim then follows from Lemma 6, Lemma 10 and Lemma 19.

Given Lemmas 21 and 22, the only option that we did not yet consider for  $t^* \le m'$  is to have at most  $t^* - 1$  jobs on the fast machine, where one of the jobs is larger than 1. We will consider the case where  $t^* = m' + 1$  separately later.

**Lemma 23** If there is a job which is larger than 1 on the fast machine, then  $y \le 1 + \frac{t-1}{s}$ .

**Proof** If  $y \le 1$ , then this is clear. Else, the total size of the t jobs on the fast machine and the single job (Lemma 8) on  $M_y$  is at most s + t - 1, since the job larger than 1 on the fast machine, together with the job of size y > 1 on  $M_y$  have a total size of at most s, and each other job on the fast machine has size of at most 1. In other words,  $xs + y \le s + t - 1$ , implying that  $xs \le s + t - 1 - y$ , and with the help of Lemma 10 we then find that  $y \le \frac{s+t-1-y}{s-1}$ , or  $y(1 + \frac{1}{s-1}) \le \frac{s+t-1}{s-1}$ , and so  $y \le \frac{s+t-1}{s} = 1 + \frac{t-1}{s}$ .  $\Box$ 

**Definition 3** Let  $t_3$  be the highest value of  $t \in \mathbb{N}$ ,  $t < t^*$  such that

$$\frac{t^*-1}{s-1} < 1 + \frac{t-1}{s} \le \text{GlobMax}.$$

If there is no such value t, let  $t_3 = 1$ .

**Lemma 24** If  $t_3 > 1$  and (5) does not hold, then  $t_3 = t^* - 1$ .

**Proof** We first note that if the condition  $\frac{t^*-1}{s-1} < 1 + \frac{t-1}{s}$  holds for some value of t, then it holds for any t' such that  $t < t' < t^*$ .

Suppose that (5) does not hold and  $1 < t_3 \le t^* - 2$ . Thus the condition  $1 + \frac{t-1}{s} \le \text{GLOBMAX}$ does not hold for  $t = t_3 + 1$ , so  $1 + \frac{t_3}{s} > \text{GLOBMAX}$ . Let  $y = 1 + t_3/s > \text{GLOBMAX}$ . Then  $\lceil s(y-1) \rceil = \lceil t_3 \rceil = t_3$ . Moreover, from the definition of  $t_3$  and the assumptions on  $t_3$  we see that  $\text{GLOBMAX} > \frac{t^*-1}{s-1} \ge \frac{t_3+1}{s-1}$ . Therefore  $y > \frac{t_3+1}{s-1}$ , which implies  $(s-1)y > t_3 + 1$  and therefore  $(s-1)y - \lceil s(y-1) \rceil = (s-1)y - t_3 > 1$ . Now, for any  $y \le 1 + t_3/s$ , in particular for y = GLOBMAX, we clearly have that  $\lceil s(y-1) \rceil \le t_3$ . But since we saw above that  $\text{GLOBMAX} > \frac{t_3+1}{s-1}$ , we find  $(s-1)\text{GLOBMAX} > t_3 + 1$  and therefore (5) holds. This is a contradiction.  $\Box$  **Lemma 25** If y > 1 and  $t_3 > 1$ , there is an equilibrium instance with  $t^* - 1$  jobs on the fast machine where one job is larger than 1 and  $y = 1 + (t^* - 2)/s$ . For any equilibrium instance with at most  $t^* - 1$  jobs on the fast machine,  $y \le 1 + (t^* - 2)/s$  if one of those jobs is larger than 1.

**Proof** Consider the following instance. We have  $t_3 = t^* - 1$  by Lemma 24. There is a job of size  $y = 1 + \frac{t^*-2}{s}$  on the slow machine  $M_y$ . On the fast machine, there are  $k = \lceil s(y-1) \rceil$  jobs of size 1 as well as 1 job of size  $q = (s-1)y - \lceil s(y-1) \rceil \le s - y$ . Thus x = (s-1)y/s. On each empty slow machine, we place  $\max(x-1,0)$  of sand. It then immediately follows that this is an equilibrium, since the condition of Lemma 10 is satisfied, and no job can improve by moving to a slow machine with load  $\max(x-1,0)$ .

We need to show that the optimal makespan is 1. The total size of all the jobs is at most m' + s because  $y \leq GLOBMAX$  by definition of  $t_3$ , so that the loads on the fast machine and on  $M_y$  are at most those from the example from Theorem 4, where the total size was exactly m' + s. This holds because we maintain x = (s - 1)y/s, which is now not larger. If x > 1, the loads on the remaining machines are also smaller in the current example.

Suppose x - 1 < 0. Then y < s/(s - 1). For y = s/(s - 1) < GLOBMAX (Lemma 15), we have x = 1, and the loads on the other machines are zero. It is clear that for smaller y, if we maintain  $x = \frac{s-1}{s}y$ , the total size of the jobs on  $M_y$  and the fast machine is smaller. Thus also in the case that x - 1 < 0 we have that the total size of all the jobs in the current example is not more than m' + s.

We also need that  $q \ge 1$ . For  $y = 1 + (t^* - 2)/s$ , we have  $\lceil s(y - 1) \rceil = \lceil t^* - 2 \rceil = t^* - 2$ . This means that  $q \ge 1$  holds if

$$(s-1)(1+(t^*-2)/s) - (t^*-2) \ge 1,$$

or  $(s-1)(1 + (t^* - 2)/s) \ge t^* - 1$ , which means  $1 + (t^* - 2)/s \ge (t^* - 1)/(s - 1)$ . But this follows from the assumption that  $t_3 > 1$ .

Note also that this immediately implies that  $y \le s - q \le s - 1$ . In addition, we actually find that  $k = t^* - 2 \le m'$ , so each of these t - 1 jobs can be placed on their own machine in the optimal solution, thus the optimal makespan is 1. The second claim follows immediately from Lemma 23.

We are now ready to give a full characterization of the POA in the case that  $t^* \leq m'$  and (5) does not hold.

**Theorem 6** If  $t^* \le m'$  and (5) does not hold, and the POA is achieved on a slow machine, it is given by  $\begin{pmatrix} t^* - 2 & t^* - 1 \end{pmatrix}$ 

$$\min\left(s, \max\left(\mathsf{MAXSMALL}(t^*), 1 + \frac{t^* - 2}{s}, \frac{t^* - 1}{s - 1}\right)\right).$$

**Proof** The first upper bound follows from Lemma 6. There are three cases, depending on where the maximum is achieved. The case numbers indicate the term that achieves the maximum.

Case 1. We use Lemma 21 to get an instance with  $y = \min(s, \text{MAXSMALL}(t^*))$ . The lemma (combined with Lemma 6) states that no higher load can be achieved on a slow machine using at least  $t^*$  jobs on the fast machine. If there are less than  $t^*$  jobs on the fast machine, we have the bounds from Lemma 22 and Lemma 25 which are not larger in this case.

Case 2. We use Lemma 22 to get an instance with  $y = \min(s, (t^* - 1)/(s - 1))$  with  $t^* - 1$  jobs of size at most 1 on the fast machine. Similar to in Case 1, it can be seen that other possibilities for jobs on the fast machine do not give higher values for y.

Case 3. We use Lemma 25 to get an instance with  $y = 1 + (t^* - 2)/s$ . The proof of Lemma 25 shows that if  $1 + (t^* - 2)/s \ge \frac{t^* - 1}{s-1}$ , then  $1 + (t^* - 2)/s \le s - 1$ .

### **3.3.3** The case $t^* = m' + 1$

Suppose that  $t^* = m' + 1$ . This case requires special attention, because if we have m' + 1 jobs on the fast machine, we get an additional condition for the POA-instance: we find that the sum of y and one of the jobs on the fast machine must be at most s.

**Definition 4** Let

ManyJobs = 
$$\frac{(s+1)m'}{s+(m'-1)(2-1/s)}$$
.

**Lemma 26** For  $m' \ge 2$  and  $s \ge 2$ , if MANYJOBS  $\le$  MAXSMALL(m' + 1), then MANYJOBS < 1 + m'/s.

**Proof** First of all, MANYJOBS and 1 + m'/s are both continuous for  $s \ge 2$  and  $m' \ge 2$ . Furthermore, MAXSMALL(m'+1) is also continuous for  $s \ge 2$  and  $m' \ge 2$  (we have  $s^2 > (m'-1)(s-1)(\frac{s}{m'+1}-1)$  because  $s^2 > \frac{m'-1}{m'+1} \cdot (s-1)s$  for  $s \ge 2$  and  $m' \ge 2$ ). Solving for m', we have MANYJOBS = MAXSMALL(m'+1) for

$$m_{1,2} = \frac{s \pm \sqrt{17s^2 - 4s^3 - 20s + 8}}{2s - 4}$$

These values are not real if  $17s^2 - 4s^3 - 20s + 8 < 0$ , that is, if  $s \ge 2.65$ . Taking for instance m = 3 and s = 3, we find MANYJOBS = 1.89 > 1.8 = MAXSMALL(m' + 1). Thus if MANYJOBS  $\le MAXSMALL(m' + 1)$ , we know that  $2 \le s < 2.65$  since both functions are continuous for  $s \ge 2$  and  $m' \ge 2$ . On the other hand, we have MANYJOBS = 1 + m'/s for  $m_3 = s(s-1)^2/(2s-1)$ . The value  $m_3$  is continuously increasing for all  $s \ge 2$ : the derivative is  $(4s^3 - 7s^2 + 4s - 1)/(2s - 1)^2$ , which is positive for all  $s \ge 2$  since the numerator is more than  $4s^3 - 8s^2 + 4s - 1 \ge 4s - 1 > 0$ , and the denominator is positive. Furthermore, for s = 2.65,  $m_3 < 1.678 < 2$ . Thus for  $m' \ge 2$ , we never have MANYJOBS = 1 + m'/s for  $2 < s \le 2.65$ . Since MANYJOBS = 12/7 < 1 + m'/s = 2 for s = 2 and m = 2, and both functions are continuous for  $s \ge 2$  and  $m' \ge 2$ , the lemma is proved.

**Lemma 27** For  $s \ge 2$  and  $m' \ge 2$ , MANYJOBS < s.

**Proof** We have equality for  $s = \frac{1}{2}(-m'+2+\sqrt{(m')^2+4m'})$ . This is less than 2 for all  $m' \ge 2$ , and for s = 2 and m' = 2 we have MANYJOBS = 6/3.5 < 2 = s. Finally, MANYJOBS is continuous in s and m for  $s \ge 2$ ,  $m \ge 2$ . This proves the lemma.

**Lemma 28** Let  $m' \ge 2$  and  $s \ge 2$ . If there are m' + 1 jobs on the fast machine, and y > 1, then

$$y \le \min(MAXSMALL(m'+1), MANYJOBS).$$
 (12)

An instance with this y exists if  $t^* = m' + 1$ .

**Proof** The first upper bound follows from Lemma 17. Denote the size of the smallest job on the fast machine by a. If the optimal makespan is 1, then since we may assume no two jobs on the fast machine have total size less than 1 by Lemma 16, we must have  $a \le s - y$  (and  $y \le s - a < s$ ).

We have  $x \ge \frac{s-1}{s}y$  as usual (Lemma 10), and the condition that  $z + a \ge x$ , because the job of size a may not benefit from moving to a slow machine. This implies

$$z \ge x - a \ge y(s - 1)/s + y - s = y(2 - 1/s) - s.$$

Moreover, the total size of all the jobs must be at most m' + s, leading to the condition that

$$y\left(1+(s-1)+(m'-1)(2-1/s)\right)-(m'-1)s \le m'+s.$$
(13)

For  $m' \ge 2$ ,  $s \ge 2$ , this is equivalent to  $y \le MANYJOBS$ . Note that this bound is also valid in case y(2-1/s) - s < 0. (In this case, it would however be better to use the bound  $z \ge 0$ .) In particular, the denominator of MANYJOBS is positive for all  $s \ge 2$ ,  $m' \ge 2$ .

For the second claim, assume  $t^* = m' + 1$ . Note that MAXSMALL $(t^*) > 0$  by definition, and  $t^* > s$  by Lemma 18. If MAXSMALL $(m'+1) \le$  MANYJOBS, it follows that if we take y = MAXSMALL $(m'+1) \le$  MANYJOBS, it follows that if we take y = MAXSMALL $(m'+1) \le$  MANYJOBS, it follows that if we take y = MAXSMALL $(m'+1) \le$  MANYJOBS, it follows that if we take y = MAXSMALL $(m'+1) \le$  MANYJOBS, it follows that if we take y = MAXSMALL $(m'+1) \le$  MANYJOBS, it follows that if we take y = MAXSMALL $(m'+1) \le$  MANYJOBS, it follows that if we take y = MAXSMALL $(m'+1) \le$  MANYJOBS, it follows that if we take y = MAXSMALL $(m'+1) \le$  MANYJOBS, it follows that if we take y = MAXSMALL $(m'+1) \le$  MANYJOBS, it follows that if we take y = MAXSMALL $(m'+1) \le$  MAXSMALL(m'+

1) > 0, inequality (13) is satisfied, whereas (8) holds with equality. We therefore have

$$\begin{split} y\left(s + (m'-1)(2-1/s)\right) - (m'-1)s &\leq m' + s = y\left(s + (m'-1)\left(1 - \frac{1}{s}\right)\left(1 - \frac{s}{m'+1}\right)\right) \\ &\Rightarrow y(m'-1) - (m'-1)s &\leq y(m'-1)\left(1 - \frac{1}{s}\right)\left(-\frac{s}{m'+1}\right) \\ &\Rightarrow y - s &\leq y\left(-\frac{s-1}{m'+1}\right) \end{split}$$

This implies  $y(s-1)/(m'+1) \le s-y$ . This immediately shows that we can use the instance from Lemma 21 for  $t^* = m'+1$ , and in the optimal solution assign the job of size y = MAXSMALL(m'+1) to the fast machine together with one job of size  $y(s-1)/(m'+1) \le s-y$ . Note that in this case we also have  $y \le s - y(s-1)/(m'+1) < s$ , that is, we do not have to worry about the case MAXSMALL(m'+1) > s.

On the other hand, if MANYJOBS < MAXSMALL(m'+1), it follows that if we take y = MANYJOBS, we find s - y < y(s - 1)/(m' + 1). Since MANYJOBS < s by Lemma 27, we have s - y > 0 also in this case. In this case we place one job of size s - y on the fast machine and m' jobs of total size y(s-1)-(s-y) = s(y-1). In order for the optimal makespan to be 1, we must have  $s(y-1)/m' \le 1$ .

To prove this, we use that  $y = MANYJOBS \le 1 + m'/s$ , which holds by Lemma 26. This implies that  $sy \le m' + s$ , and then  $y(s-1) \le m' + s - y$ . This last value, m' + s - y, would be the total size of the jobs on the fast machine if we placed m' jobs of size 1 there plus a job of size s - y. Thus that last inequality implies that the m' jobs in our instance have size at most 1, since we have xs = y(s-1).

Finally, since we have m' + 1 jobs on the fast machine in this instance, one of them of size s - y < y(s-1)/(m'+1), and the other m' jobs all equal-sized, it follows that those m' jobs all have size more than s - y. Thus the job of size s - y is indeed the smallest on the fast machine, and since xs = y(s-1), this means that (13) is a sufficient condition to have an equilibrium.

**Theorem 7** If (5) does not hold, and the POA is achieved on a slow machine, there are two cases. If  $t^* \leq m'$ , the POA is given by  $\min\left(s, \max\left(\text{MAXSMALL}(t^*), 1 + \frac{t^*-2}{s}, \frac{t^*-1}{s-1}\right)\right)$ . If  $t^* = m' + 1$ , the POA is given by

$$\min\left(s, \max\left(\min\left(\mathsf{MaxSmall}(m'+1), \mathsf{ManyJobs}\right), 1 + \frac{m'-1}{s}, \frac{m'}{s-1}\right)\right).$$

**Proof** Suppose  $t^* = m' + 1$ . If the maximum is achieved in the first term, we use one of the instances from Lemma 28, depending on where the inner minimum is achieved. Else, the bound follows as in the proof of Theorem 6. Note that Lemmas 22 and 25 do not require  $t^* \le m'$ .

It can be verified that the POA is achieved on a slow machine for all  $m' \ge 2$  and  $s \in [2, 4.57]$  (and therefore for all  $s \ge 2$  by Lemma 12 and Theorem 5). For instance, for  $m' \ge 10$ , POA > 2 > FASTMAX in this interval. See Figures 2 and 3 for graphs of the POA as a function of s for several values of m'.

### **3.4** The limit of the POA for $m' \to \infty$

What happens with the POA if m' grows without bound? By Theorem 5, POA = GLOBMAX for  $s \ge 4.562$ . To answer this question for smaller s, we first need to consider the value MAXSMALL(t). By Definition 1, we have

$$\lim_{m' \to \infty} \text{MAXSMALL}(t) = \frac{1}{(s-1)(1-s/t)/s} = \frac{1}{(s-1)/s - (s-1)/t} = \frac{st}{(s-1)(t-s)}$$

From this, we can derive  $\lim_{m'\to\infty} t^*$  using Definition 2. We have

$$t = (s-1) \cdot \frac{st}{(s-1)(t-s)} = \frac{st}{t-s} \Leftrightarrow s = t-s \Leftrightarrow t = 2s.$$



Figure 2: The price of anarchy for m' = 2, 3, 4 as a function of s. The top line in each case is GLOBMAX, a global upper bound on the POA. The bottom line is the actual POA for each s. For m' = 4 and  $s \in [3, 3.7]$ , we have POA = 1 + 3/s < GLOBMAX.



Figure 3: The price of anarchy for m' = 5, 10, 20 as a function of s. The top line in each case is GLOBMAX, a global upper bound on the POA. The bottom line is the actual POA for each s.

Hence, for  $s \le 4.562$  and large enough m', we certainly have  $t^* < m'$ . Moreover, we have

$$\lim_{m' \to \infty} \text{MAXSMALL}(2s) = \frac{2s^2}{(s-1)s} = \frac{2s}{s-1} = \lim_{m \to \infty} \text{GLOBMAX}$$

Since POA  $\leq$  GLOBMAX by Lemma 13, we can conclude the following by Lemma 12 and Lemma 21. **Theorem 8** For  $s \in [2,3]$ ,  $\lim_{m'\to\infty} POA = s$ . For  $s \geq 3$ ,  $\lim_{m'\to\infty} POA = \frac{2s}{(s-1)}$ .

# **4 One fast machine: the SPOA**

In this section we demonstrate the fact that the SPOA is strictly smaller than the POA. We consider the overall bounds (i.e., the supremum bounds over all values of s and m') and compare them. The overall bound on the POA, as implied by the previous sections, is 3.

**Theorem 9** The SPOA is 2 for  $m' \le 5$ . For any m', SPOA  $\le \frac{3+\sqrt{5}}{2} \approx 2.618$ . For  $m' \ge 16$ , SPOA  $\ge \frac{1+\sqrt{13}}{2} \approx 2.3027756$ .

**Proof** We first slow a lower bound of 2 for any value of m'. Consider from the following instance. The fast machine has speed 2. There are m' jobs of size 1, and one job of size 2. An optimal solution is clearly to assign one unit job to each slow machine, and the larger job to the faster machine. This gives OPT = 1. In a schedule S that we consider, two jobs of size 1 are scheduled on the fast machine. One slow machine is empty, one has a job of size 2, and all remaining slow machines have one job of size 1. It can be seen that no coalition can improve from trading places. The two jobs on the fast machine can never obtain smaller load, so they would not move to a slow machine. As long as these two jobs do not move, no other job can benefit from moving.

We next prove an upper bound. Consider a strong equilibrium S. We use the notations x,  $M_y$ , y and z, as before. Let r be the size of the smallest job on the fast machine (slightly abusing the notation, we use r to denote the job as well). Lemma 3 and Lemma 6 both hold for any  $s \ge 1$  and any schedule that is a pure equilibrium, thus we can use them in this proof. If  $x \le 1$  then since any job on  $M_y$  is of size at most s, we get that  $y \le x + 1 \le 2$ . In this case the SPOA is no larger than 2, and therefore, since by Lemma 3, we have x < 2, we only need to consider a case where 1 < x < 2, and the SPOA is achieved on  $M_y$ .

Since x > 1, there must be a machine with load smaller than 1, and therefore z < 1. If  $M_y$  contains a job of size d that OPT assigns to a slow machine, we have  $z + d \ge y$  and therefore  $y \le z + 1 < 2$ . Thus  $M_y$  only contains jobs assigned by OPT to the fast machine (and SPOA  $\le s$ ). We therefore have  $y \le s$  and we can assume that s > 2, otherwise we would again get a SPOA of at most 2.

Since x > 1, in the schedule S, the fast machine must have a job the OPT assigns to a slow machine. Thus  $r \le 1$ . Since the job of size r does not benefit from moving to the least loaded slow machine, we get  $z + r \ge x$ .

We claim that  $xs + y \ge r + sy$  and therefore  $xs \ge (s - 1)y + r$ . Recall that  $M_y$  contains only jobs that belong on the fast machine (otherwise  $y \le 2$ ). Consider the coalition consisting of the jobs scheduled on  $M_y$  and a job of size r, scheduled on the fast machine. Upon a deviation of this coalition, the job r moves to the slow machine  $M_y$  and as a result, has a delay of  $r \le 1$ . Its previous delay was x > 1. Since there exists a job of the coalition which does not reduce it load upon deviation, the jobs of  $M_y$  are those that do not benefit from moving: we find  $(xs - r + y)/s \ge y$ . This proves the claim.

Let W be the total size of all the jobs. We get

$$\begin{split} m's + s^2 &\geq Ws \geq xs^2 + ys + (m'-1)zs \geq rs + s^2y + (m'-1)(x-r)s \\ &= rs + s^2y + xs(m'-1) - rs(m'-1) \\ &\geq r(s - sm' + s) + s^2y + ((s-1)y + r)(m'-1) \\ &= r(2s - sm' + m' - 1) + y(s^2 + sm' - m' - s + 1). \end{split}$$

If  $2s-sm'+m'-1\leq 0,$  then we use  $r\leq 1$  to get  $m's+s^2\geq 2s-sm'+m'-1+y(s^2+sm'-m'-s+1)$  or

$$y \le \frac{s^2 + 2m's - 2s - m' + 1}{s^2 + sm' - s - m' + 1} \tag{14}$$

(note that  $s^2 + sm' - m' - s + 1 = s^2 + (s - 1)(m' - 1) > 0$ ). If  $2s + m' - 1 - sm' \ge 0$  we use  $r \ge 0$  to get,

$$y \le \frac{m's + s^2}{s^2 + m's - s - m' + 1} = 1 + \frac{s + m' - 1}{s^2 + m's - s - m' + 1} \le 2$$

since  $s + m' - 1 \le s^2 + m's - s - m' + 1 \Leftrightarrow s^2 + m's + 2 \ge 2s + 2m'$  which holds for any  $s \ge 2$  (by  $(s - 1)^2 \ge 0$ ).

By (14), if  $m' \leq 5$  then  $y \leq 2$  holds if  $s^2 - m' + 1 \geq 0$ . But  $m' \leq s^2 + 1$  for any  $m' \leq 5$  (since  $s \geq 2$ ). For larger m', we show that  $y \leq \frac{2s-1}{s-1}$ . Since we also know SPOA  $\leq s$ , we get SPOA  $\leq 2.618$ .

To prove  $y \le \frac{2s-1}{s-1}$  or  $y-1 \le \frac{s}{s-1}$  we need to show  $\frac{m's-s}{s^2+sm'-s-m'+1} \le \frac{s}{s-1}$ , i.e.,  $(m'-1)(s-1) \le s^2 + sm'-s - m'+1$  which holds since  $s^2 > 0$ .

For the lower bound, consider a fast machine of speed  $\sigma = \frac{1+\sqrt{13}}{2}$ . In an optimal schedule, the fast machine has a job of size  $\sigma$ , there are 12 slow machines that contain two jobs, of sizes  $\frac{1}{4}$  and  $\frac{3}{4}$  and the remaining slow machines have one job of size 1 each. Therefore OPT = 1. In the schedule we consider, the fast machine has four jobs of size 1, 12 slow machines have jobs of size  $\frac{3}{4}$ , three slow machines have four jobs of size  $\frac{1}{4}$  each, one slow machine has a job of size  $\sigma$  and the remaining slow machines have jobs of size 1. The load on the fast machine is  $\frac{2(\sqrt{13}-1)}{3} \approx 1.736865$  and the makespan is achieved on the slow machine which contains the job of size  $\sigma \approx 2.303$ .

Consider the terms on which each type of job would join a coalition. We first discuss the case where the job of size  $\sigma$  does not join. If no job which is assigned to the fast machine joins, then no job which is scheduled to a slow machine would want to move to the fast machine, and jobs that are single on their machine would not join, so no coalition can be created. On the other hand, since the load on the fast machine is strictly less than 1.75, then the jobs on this machine would join a coalition only if they could move to a slow machine with a resulting load of less than 1.74, i.e. due to the structure of the instance, the load excluding the additional job should be at most  $\frac{1}{2}$ . For that, some jobs of size  $\frac{1}{4}, \frac{3}{4}$  or 1 would need to join the coalition. There is clearly no advantage to exchanges between jobs of size 1, thus we need to consider only smaller jobs. A job of size  $\frac{3}{4}$  benefits from moving to the fast machine only if the resulting total size there is no larger than 1.74, i.e., at most 1.5, but this can happen if all jobs of size 1 on the fast machine join the coalition. Jobs of size  $\frac{1}{4}$  would move to the fast machine if the resulting total size there is at most 2.25. For that, at least two jobs from the fast machine need to join the coalition. We consider three cases based on the number of jobs migrating from the fast machine. If two jobs migrate, only a single job of size  $\frac{1}{4}$  can migrate, so the room created for the migrating jobs of size 1 does not suffice. If three jobs migrate, then five jobs of size  $\frac{1}{4}$  can migrate, and there is room only for two migrating jobs of size 1. If four large jobs migrate, in order to make room for the migrating jobs of size 1, six jobs of size  $\frac{1}{4}$  and one job of size  $\frac{3}{4}$  must migrate (if there are more jobs of size  $\frac{3}{4}$  migrating, and less pairs of jobs of size  $\frac{1}{4}$ , then the load on the fast machine would only be larger). This would create a total size of 2.25 on the fast machine, therefore the job of size  $\frac{3}{4}$  would not join the coalition.

If the job of size  $\sigma$  joins the coalition, at least two jobs of size 1 from the fast machine must join the coalition as well, since  $\frac{3+\sigma}{\sigma} = \sigma$ . In order to make it beneficial for these two jobs to migrate, and since moving both of them to the machine that becomes empty would create a load of 2 there, at least two jobs of size  $\frac{1}{4}$  or one job of size  $\frac{3}{4}$  needs to join the coalition. But then the load on the fast machine is already larger than 1 due to the job of size  $\sigma$ , so no such jobs would join the coalition.

# 5 Conclusion

We studied the POA as a function of the number of different speeds. We found a tight overall bound, and completely resolved the case where all machines are identical, except for one faster machine. It can be interesting for find a tighter result for the SPOA as a function of the number of different speeds, p, and find whether it is strictly smaller than p + 1, which is the POA for this case. Another direction is to study the influence of additional factors on the POA, such as the ratio of the largest and smallest speeds, or even as a function of all the machine speeds, possibly as the solution of a mathematical program.

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