

Improved Lower Bound for Online Strip Packing

Rolf Harren · Walter Kern

Published online: 15 August 2013
© Springer Science+Business Media New York 2013

Abstract We study the online strip packing problem and derive an improved lower bound of $\rho \geq 2.589\dots$ for the competitive ratio of this problem. The construction is based on modified “Brown-Baker-Katseff sequences” (Brown et al. in *Acta Inform.* 18:207–225, 1982) using only two types of rectangles. In addition, we present an online algorithm with competitive ratio $(3 + \sqrt{5})/2 = 2.618\dots$ for packing instances of this type.

Keywords Strip packing · Rectangle packing · Online algorithms · Lower bounds

1 Introduction

In the two-dimensional strip packing problem a number of rectangles have to be packed without rotation or overlap into a strip such that the height of the strip used is minimal. The width of the rectangles is bounded by 1 and the strip has width 1 and infinite height. Baker, Coffman and Rivest [2] show that this problem is NP-hard, while Kenyon and Remila [3] present an approximation scheme for solving this problem.

We study the online version of this packing problem. In the online version the rectangles are given to the online algorithm one by one from a list, and the next rectangle is given as soon as the current rectangle is irrevocably placed into the strip. To evaluate the performance of an online algorithm we employ competitive analysis.

R. Harren
Max-Planck-Institut für Informatik (MPII), Campus E1 4, 66123 Saarbrücken, Germany
e-mail: rharren@mpi-inf.mpg.de

W. Kern (✉)
Department of Applied Mathematics, University of Twente, P.O. Box 217, 7500 AE Enschede,
The Netherlands
e-mail: w.kern@utwente.nl

For a list of rectangles L , the height of a strip used by online algorithm ALG and by the optimal solution are denoted by $\text{ALG}(L)$ and $\text{OPT}(L)$, respectively. The optimal solution is not restricted in any way by the ordering of the rectangles in the list. Competitive analysis measures the absolute worst-case performance of online algorithm ALG by its competitive ratio

$$\rho_{\text{ALG}} = \sup_L \left\{ \frac{\text{ALG}(L)}{\text{OPT}(L)} \right\}.$$

Known Results Regarding the upper bound on the competitive ratio for online strip packing, recent advances have been made by Ye, Han and Zhang [4] and Hurink and Paulus [5]. Independently they showed that a modification of the well-known shelf algorithm yields an online algorithm with competitive ratio $7/2 + \sqrt{10} \approx 6.6623$, improving an earlier “shelf type algorithm” by Baker and Schwarz [6]. Another line of research deals with the so-called asymptotic competitive ratio, cf. [6–8].

In the early 80’s, Brown, Baker and Katseff [1] derived a lower bound $\rho \geq 2$ on the competitive ratio of any online algorithm by constructing certain (adversary) sequences in a fairly straightforward way—see Sect. 2. These sequences, that we call BBK sequences in the sequel, were further studied by Johannes [9] and Hurink and Paulus [10], who derived improved lower bounds of 2.25 and 2.43, respectively. (Both results are computer aided and presented in terms of online parallel machine scheduling, a closely related problem.) The paper of Hurink and Paulus [10] also presents an upper bound of $\rho \leq 2.5$ for packing BBK sequences. Kern and Paulus [11] finally settled the question of how well the BBK sequences can be packed by providing matching upper and lower bounds of $\rho_{\text{BBK}} = 3/2 + \sqrt{33}/6 \approx 2.457$.

Our Contribution Using modified BBK sequences we show an improved lower bound of 2.589... on the absolute competitive ratio of this problem. The modified sequences that we use consist solely of two types of items, namely, *thin* items that have negligible width (and thus can all be packed in parallel) and *blocking* items that have width 1. The advantage of these sequences is that the structure of the optimal packing is simple, i.e., the optimal packing height is the sum of the heights of the blocking items plus the maximal height of the thin items. Therefore, we call such sequences *primitive*. We like to stress that all instances used so far to derive lower bounds are primitive.

On the positive side, we present an online algorithm for packing primitive sequences with competitive ratio $(3 + \sqrt{5})/2 = 2.618\dots$. This result shows that our lower bound analysis of modified BBK sequences is fairly tight and, secondly, that in order to derive new lower bounds for strip packing that are larger than 2.618... (and thus to significantly reduce the gap to the general upper bound of 6.6623), instances with a more complex structure (not just thin and blocking items) must be analyzed. In this sense, the upper bound result can thus be taken as a hint to future research directions, possibly leading to improved lower bounds.

The present paper is a journal version of an extended abstract that was earlier published (without proof) in the proceedings of WAOA 2011 [12].

Organization We start our presentation with a description of the Brown-Baker-Katseff sequences and their modification in Sect. 2. In Sect. 3 we present our lower bound based on these modifications and in Sect. 4 we describe our algorithm for packing primitive sequences. A detailed proof of the main result (lower bound) is presented in Sect. 5.

2 Sequence Construction

In this paper we denote the thin items by p_i and the blocking items by q_i (adopting the notation from [11]). As already mentioned in the introduction, we assume that the width of the thin items is negligible and thus all thin items can be packed next to each other. Moreover, the width of the blocking items q_i is always 1, so that no item can be packed next to any blocking item in parallel. Therefore, all items are characterized by their heights and we refer to their heights by p_i and q_i as well. By definition, for any list $L = q_1, q_2, \dots, q_k, p_1, p_1, \dots, p_\ell$ consisting of thin and blocking items we have

$$\text{OPT}(L) = \sum_{i=1}^k q_i + \max_{i=1, \dots, \ell} p_i.$$

To prove the desired lower bound we assume the existence of a ρ -competitive algorithm ALG for some $\rho < 2.589 \dots$ (the exact value of this bound is specified later) and construct an adversary sequence depending on the packing that ALG generates.

To motivate the construction, let us first consider the GREEDY algorithm for on-line strip packing, which packs every item as low as possible—see Fig. 1(a). This algorithm is not competitive (i.e., has unbounded competitive ratio): Indeed, consider the list $L_n = p_0, q_1, p_1, q_2, p_2, \dots, q_n, p_n$ of items with

$$\begin{aligned} p_0 &:= 1, \\ q_i &:= \varepsilon && \text{for } 1 \leq i \leq n, \\ p_i &:= p_{i-1} + \varepsilon && \text{for } 1 \leq i \leq n \end{aligned}$$

for some $\varepsilon > 0$. GREEDY would pack each item on top of the preceding ones and thus generate a packing of height $\text{GREEDY}(L_n) = \sum_{i=0}^n p_i + \sum_{i=1}^n q_i = n + 1 + \Omega(n^2\varepsilon)$, whereas the optimum clearly has height $1 + 2n\varepsilon$.

The GREEDY algorithm illustrates that any competitive online algorithm needs to create gaps in the packing. These gaps work as a buffer to accommodate small blocking items—or, viewed another way, force the adversary to release larger blocking items.

BBK Sequences The idea of Brown, Baker and Katseff [1] was to try to cheat an arbitrary (non-greedy) online packing algorithm ALG in a similar way by constructing an alternating sequence p_0, q_1, p_1, \dots of thin and blocking items. The heights p_i , respectively q_i are determined so as to force the online algorithm ALG to put each item above the previous ones—see Fig. 1(b) for an illustration. To describe the heights of

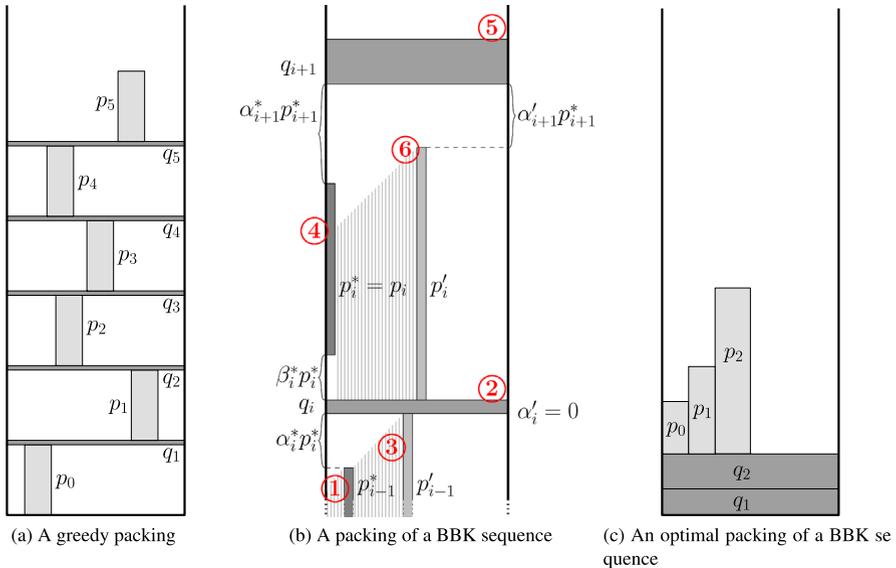


Fig. 1 Online and optimal packings

the items formally, we consider the gaps that ALG creates between the items. We distinguish two types of gaps, namely gaps below and gaps above a blocking item, and refer to these gaps as α - and β -gaps, respectively. These gaps also play an important role in our analysis of the modified BBK sequences. We describe the height of the gaps around the blocking item q_i relative to the thin item p_i . Thus, we denote the height of the α -gap below q_i by $\alpha_i p_i$ and the height of the β -gap above q_i by $\beta_i p_i$. Using this notation, we are ready to formally describe the BBK sequences $L = p_0, q_1, p_1, q_2, \dots$ with

$$\begin{aligned}
 p_0 &:= 1, \\
 q_1 &:= \beta_0 p_0 + \varepsilon, \\
 p_i &:= \beta_{i-1} p_{i-1} + p_{i-1} + \alpha_i p_i + \varepsilon && \text{for } i \geq 1, \\
 q_i &:= \max(\alpha_{i-1} p_{i-1}, \beta_{i-1} p_{i-1}, q_{i-1}) + \varepsilon && \text{for } i \geq 2.
 \end{aligned}$$

In other words, each q_i is chosen such that it is just too high to fit into one of the preceding gaps. This is equivalent to saying that q_i exceeds the preceding α - and β -gaps as well as q_{i-1} (which in turn exceeds all previous gaps). Similarly, each p_i (except the first p_0) is chosen just too large to fit into one of the gaps between two consecutive blocking items. As mentioned in the introduction, Brown, Baker and Katseff [1] used these sequences to derive a lower bound of 2 before Kern and Paulus [11] recently showed that the competitive ratio for packing them is $\rho_{\text{BBK}} = 3/2 + \sqrt{33}/6 \approx 2.457$.

The optimal online algorithm for BBK sequences that Kern and Paulus [11] describe generates packings with striking properties: No α - and β -gaps are created except the first possible gap $\beta_0 = \rho_{\text{BBK}} - 1$ and the second α -gap $\alpha_2 = 1/(\rho_{\text{BBK}} - 1)$,

which are chosen as large as possible while remaining ρ_{BBK} -competitive. Observing this behavior of the optimal algorithm led us to the modification of the BBK sequences.

Modified BBK Sequences By definition, the decisions of the online algorithm, in particular, the gaps it creates, influence the sequence (construction): Creating large α - or β -gaps “forces” the adversary to provide large blocking items in the next step. When packing BBK sequences, a good online algorithm should be eager to “enforce” blocking items of relatively large size (as each blocking item of size q increases the optimal packing by q as well). Thus a good online algorithm should seek to create large gaps.

Modified BBK sequences are designed to counter this strategy: Each time the online algorithm places a blocking item q_i , the adversary, rather than immediately releasing a thin item p_{i+1} (of height defined as in standard BBK sequences) that does not fit in between the last two blocking items, generates a whole sequence of slowly growing thin items, which “continuously” grow from p_i to p_{i+1} . Packing this subsequence causes additional problems for the online algorithm: If the algorithm fits the whole subsequence into the last interval between q_{i-1} and q_i , it would fill out the whole interval and create an α -gap of 0 below q_i . More generally, if the algorithm fits a large part of the subsequence into the last interval between q_{i-1} and q_i , it would create a rather small α -gap below q_i . On the other extreme, if ALG would pack a thin item of height slightly larger than p_i above q_i , then the (relative) β -gap it can generate is much less compared to what it could have achieved with a thin item of larger height p_{i+1} (assuming that the p -items are packed as high as possible, subject to ρ -competitiveness). Thus letting thin items grow continuously from p_i to p_{i+1} forces the online algorithm to either create smaller α - or smaller β -gaps. The next blocking item q_{i+1} will be released as soon as the sequence of thin items has grown from p_i to p_{i+1} .

This general concept of modified BBK sequences applies after the first blocking item q_1 is released. Since subsequences of thin items and single blocking items are released alternately from this point on, we refer to this phase as the *alternating* phase. Before that, we have a *starting* phase in which the algorithm is confronted with a “continuously” increasing sequence of thin items. The starting phase ends with the release of the first blocking item q_1 . The purpose of the starting phase is to prevent the online algorithm from introducing a large gap in the first step (when the first thin item arrives). Indeed, the optimal online algorithm by Kern and Paulus [11] generates an initial gap β_0 of maximal size to enforce a large first blocking item q_1 . Since the first item has height 1, it must be “scheduled” at height $\beta_0 = \rho_{\text{BBK}} - 1$ in order to not exceed the optimal ratio ρ_{BBK} already in the first step. In the starting phase, we seek to prevent the algorithm from creating a large β_0 -gap as described in more detail in the next section.

Summarizing, a *modified BBK sequence* simply consists of a sequence of thin items, continuously growing in height, interleaved with blocking items which (by definition of their height) must be packed above all preceding items, and are released as described above, *i.e.*, when the thin item size has grown up to the largest gap between two blocking items, *cf.* Sects. 3 and 5 for more details.

We will use modified BBK sequences to prove

Theorem 1 *There exists no algorithm for online strip packing with competitive ratio*

$$\rho < \hat{\rho} = \frac{17}{12} + \frac{1}{48} \sqrt[3]{22976 - 768\sqrt{78}} + \frac{1}{12} \sqrt[3]{359 + 12\sqrt{78}} \approx 2.589 \dots$$

3 Lower Bound

For the sake of contradiction, assume there exists an online algorithm ALG that is ρ -competitive for online strip packing with $\rho < \hat{\rho}$. Let $\delta = \hat{\rho} - \rho > 0$. *W.l.o.g.* we assume that δ is sufficiently small. We feed ALG with a sequence r_1, r_2, \dots of thin items, interleaved with blocking items arriving at certain times as described in the following. The initial subsequence of thin items r_1, r_2, \dots, r_i that precedes the first blocking item defines the *starting phase*. The basic idea is to prevent ALG from creating a large β -gap in the first step. So if ALG packs the first thin item r_1 “too high”, we release a slightly larger thin item r_2 . The best ALG can do in this case is to bottom-align r_2 with r_1 , yielding a slight decrease in the (relative!) β -gap. Continuing this way with an increasing sequence r_3, r_4, \dots , ALG will eventually reduce its β -gap to almost 0 or decide to “jump”, *i.e.*, pack some r_j on top of the current packing, leaving a new gap in between r_{j-1} and r_j of reasonable size. In case this is (still) too large compared to the current height of the packing, we continue with r_{j+1} etc.

Thus in the starting phase we seek to decrease the maximal size (relative to the current packing height) of a gap. More specifically, let

$$\frac{h(\text{maxgap}_{\text{ALG}}(r_i))}{\text{ALG}(r_i)}$$

be the *max-gap-to-height* ratio after packing r_i , where $h(\text{maxgap}_{\text{ALG}}(r_i))$ denotes the height of the maximal gap that algorithm ALG created up to item r_i and $\text{ALG}(r_i)$ denotes the height algorithm ALG consumed up to item r_i . We say ALG is (ρ, c) -competitive in the starting phase if ALG is ρ -competitive (*i.e.*, $\text{ALG}(r_i) \leq \rho \text{OPT}(r_i)$) and retains a max-gap-to-height ratio of c (*i.e.*, $h(\text{maxgap}_{\text{ALG}}(r_i))/\text{ALG}(r_i) \geq c$ for $i \geq 1$) for all lists $L = r_1, r_2, \dots$ of thin items.

In the analysis of the starting phase in Sect. 5.1 we show that an increasing sequence of thin items (the starting phase) forces any ρ -competitive algorithm to reach a state with max-gap-to-height ratio less than

$$\hat{c} = \frac{\hat{\rho} - 2\sqrt{\hat{\rho} - 1}}{\hat{\rho} - 1}.$$

Thus there must be a first item r_i that ALG packs, causing a max-gap-to-height ratio of less than \hat{c} . The starting phase ends with the release of the first blocking item q_1 of height $\hat{c} \cdot \text{ALG}(r_i)$ and we enter the second phase which we call the *alternating phase*. (For $\hat{\rho}$ as in Theorem 1 this yields a rather small value of $\hat{c} = 0.04275 \dots$. This means that ALG might equally well pack the first item r_1 of size $r_1 = 1$, say, at height $\hat{c} = 0.04275 \dots$, very close to the bottom of the strip—in which case we would enter the alternating phase immediately.)

In Sect. 5.2 we analyze the alternating phase, more precisely, we investigate, how the competitiveness of ALG in the alternating phase is influenced by its max-gap-to-height ratio in the starting phase. We show that an algorithm with max-gap-to-height ratio of \hat{c} in the starting phase cannot retain ρ -competitiveness in the alternating phase for $\rho < \hat{\rho}$ in case

$$\hat{c} = \frac{1 - \sqrt{4\hat{\rho}^2 - 12\hat{\rho} + 5}}{2(\hat{\rho} - 1)}.$$

Thus our two phases fit together for

$$\hat{c} = \frac{\hat{\rho} - 2\sqrt{\hat{\rho} - 1}}{\hat{\rho} - 1} = \frac{1 - \sqrt{4\hat{\rho}^2 - 12\hat{\rho} + 5}}{2(\hat{\rho} - 1)},$$

which is satisfied for

$$\hat{\rho} = \frac{17}{12} + \frac{1}{48}\sqrt[3]{22976 - 768\sqrt{78}} + \frac{1}{12}\sqrt[3]{359 + 12\sqrt{78}} \approx 2.589 \dots$$

We present the proof of Theorem 1 in Sect. 5. We end this section by observing that modified BBK sequences show a completely different behavior as compared to “ordinary” BBK sequences. The optimal online algorithm dealing with ordinary BBK sequences is such that the sequence becomes stationary after a few steps (*cf.* [11]), whereas modified BBK sequences continuously grow to infinity.

4 Upper Bound

In this section we present the online algorithm ONL for packing instances that consist solely of thin and blocking items. We prove that the competitive ratio of ONL is $\rho = (3 + \sqrt{5})/2 \approx 2.618$. We distinguish two kinds of packings according to the item on top: If the item on top of the packing is a blocking item, we have a *blocked* packing, otherwise we have an *open* packing. Initially, we have a blocked packing (consider the bottom of the strip as a blocking item of height 0 “on top of” the initial empty packing).

The general idea of the algorithm ONL is pretty straightforward: First note that we might assume that the thin items are increasing in height (a thin item that has smaller size than a previous one can always be packed in parallel to the larger one). If a thin item arrives at a blocked packing and the item does not fit into one of the gaps between two blocking items—we say that a “jump is unavoidable” in this case—then we pack it on top of the current closed packing, leaving a β -gap of relative height $\rho - 2$ (*i.e.*, $(\rho - 2)$ times the height of the thin item) between the newly placed thin item and the preceding blocking item. Note that placing a thin item on top of a blocked packing results in an open packing. The relative size ($\beta = \rho - 2$) of the gap induced by this new “top” item is determined so as to ensure a competitive ratio of ρ in the long run (*cf.* the proof of Theorem 2 below. Subsequent thin items are placed bottom-aligned with the thin “top” item causing the jump, so as to not deliberately diminish the current β -gap. Any arriving blocking item is packed as low as possible,

i.e., in case it fits into one of the gaps, we pack it there, otherwise it is put on top of the current packing, resulting in a closed packing. The fact that we pack blocking items as low as possible amounts to saying that we work with α -gaps equal to 0. Summarizing, we apply the following algorithm.

Online Algorithm ONL for primitive sequences

Initially the packing is considered to be blocked.

WHILE a rectangle r_j is released

IF r_j is a blocking item, pack r_j at the lowest possible height

ELSIF r_j is a thin item

IF the packing is open, pack r_j bottom-aligned with the top thin item

ELSIF the packing is blocked, try to pack r_j below the top item.

If this is not possible, pack r_j at distance $(\rho - 2)r_j$ above the packing.

ENDWHILE

The above algorithm does not even try to cope with a “starting phase” in any respect. Nonetheless, it turns out to yield a rather good competitive ratio:

Theorem 2 *ONL is a ρ -competitive algorithm for packing primitive sequences for*

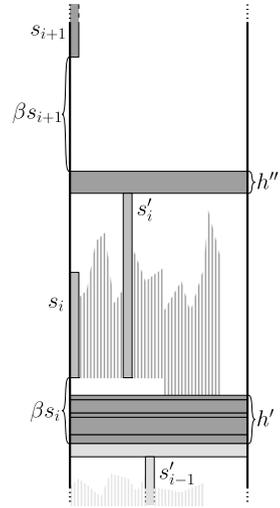
$$\rho = \frac{3 + \sqrt{5}}{2} \approx 2.618.$$

Proof We show that ONL is ρ -competitive for $\rho = (3 + \sqrt{5})/2$ by induction on the number of items. As to the inductive step, observe that whenever we pack a blocking item of height, say q , then the current height of the ONL-packing increases by at most q , whereas the optimum packing height increases by exactly q . Further, whenever we pack a thin item into one of the gaps, the ONL-packing height does not increase at all, so the algorithm stays trivially ρ -competitive in this case. Summarizing, the only critical case occurs when we pack a thin item r_j at distance βr_j with $\beta = \rho - 2$ above the current closed packing, *i.e.*, when a new top item r_j is placed due to an “unavoidable jump”.

We denote the thin items that are packed when generating a new gap by s_i for the i -th jump. Let s'_i be the highest thin item that is bottom-aligned with s_i . Note that the blocking item that blocks the packing after the i -th jump is packed directly above s'_i . See Fig. 2 for an illustration. We may assume *w.l.o.g.* that the sequence starts with a thin item (otherwise, a blocking item is put on the bottom of the strip in the first step, increasing the height of both ONL as well as OPT without any further consequences). So s_1 is the first item and this is packed at distance $(\rho - 2)s_1$ from the bottom line, so the competitive ratio after the first step is $\rho - 1 < \rho$.

For the induction step we assume $\text{ONL}(s_i) \leq \rho \text{OPT}(s_i)$. Before a jump can become unavoidable, new blocking items of total height greater than βs_i (where $\beta = \rho - 2$) need to arrive as otherwise the gap below s_i could accommodate all of them. Let h' be the height of the blocking items that are packed into the β -gap below s_i and let h'' be the total height of blocking items that arrive between s_i and s_{i+1} and are packed above s_i . See Fig. 2. We have $h' \leq (\rho - 2)s_i$ and $h' + h'' > (\rho - 2)s_i$ as

Fig. 2 Packing after jump $i + 1$. Blocking items released after s_i shown in *darker shade*. By definition, s_{i+1} is the first thin item that does not fit into a gap. Thus, in particular, $s_{i+1} > s'_i + \beta s_i - h'$



otherwise no blocking item would be packed on top. As further blocking items could be packed even below s'_{i-1} we get

$$\begin{aligned} \text{OPT}(s_{i+1}) &\geq \text{OPT}(s_i) + h' + h'' + s_{i+1} - s_i, \\ \text{ONL}(s_{i+1}) &= \text{ONL}(s_i) + s'_i - s_i + h'' + \beta s_{i+1} + s_{i+1}. \end{aligned}$$

And thus we have

$$\begin{aligned} \text{ONL}(s_{i+1}) &\leq \rho \text{OPT}(s_{i+1}) \\ &\Leftrightarrow \text{ONL}(s_i) + s'_i - s_i + h'' + \beta s_{i+1} + s_{i+1} \\ &\leq \rho(\text{OPT}(s_i) + h' + h'' + s_{i+1} - s_i) \\ &\Leftrightarrow (\rho - 1)s_i + s'_i - \rho h' - (\rho - 1)h'' \leq (\rho - 1 - \beta)s_{i+1}. \end{aligned}$$

As $\rho - 1 - \beta = 1$ and $s_{i+1} > s'_i + (\rho - 2)s_i - h'$ this is satisfied if

$$\begin{aligned} (\rho - 1)s_i + s'_i - \rho h' - (\rho - 1)h'' &\leq s'_i + (\rho - 2)s_i - h' \\ &\Leftrightarrow s_i \leq (\rho - 1)(h' + h'') \\ &\Leftrightarrow s_i \leq (\rho - 1)(\rho - 2)s_i = s_i. \end{aligned}$$

The last equality holds since $\rho = (3 + \sqrt{5})/2$ and thus $(\rho - 1)(\rho - 2) = 1$. □

So the true best possible competitive ratio for packing primitive sequences is somewhere in between the two values specified by Theorems 1 and 2. We have reasons to believe that it is strictly in between these two. But perhaps an even more challenging question is whether or not (or to what extent) primitive sequences provide worst case instances for online packing in general. So far, all lower bounds for

online strip packing are based on primitive sequences. Theorem 2 states that this approach is limited. In order to achieve significant further improvements (towards the upper bound of 6.6623), suitable *non-primitive sequences* have to be designed.

5 Detailed Proof of Theorem 1

5.1 The Starting Phase

In this section we describe the lower bound for the starting phase. As we explained before, the key parameter of this phase is the *max-gap-to-height* ratio. We will show that for $\rho < \hat{\rho}$, any ρ -competitive algorithm can be forced into a state with max-gap-to-height ratio less than \hat{c} . In this section we use the definition

$$\hat{c} = \frac{\hat{\rho} - 2\sqrt{\hat{\rho} - 1}}{\hat{\rho} - 1}.$$

The starting phase ends as soon as a state is reached with a max-gap-to-height ratio less than \hat{c} . To derive a contradiction, we assume that the ρ -competitive algorithm ALG is (ρ, \hat{c}) -competitive, i.e., retains a max-gap-to-height ratio of \hat{c} .

Let $\eta > 0$ be some very small constant and consider the adversary list $L_{\text{start}} = r_1, r_2, \dots$ consisting of thin items

$$\begin{aligned} r_1 &= 1 \quad \text{and} \\ r_i &= r_{i-1} + \eta \quad \text{for } i \geq 2. \end{aligned}$$

Recall that we denote the thin items by r_i instead of p_i here to be able to designate certain items that correspond to the thin items p_i from the BBK sequence in the analysis of the alternating phase

The sole function of the positive term η is to gradually increase the height of items (we substituted ε from the BBK sequences by η because we use ε later in our analysis). To simplify the calculations, however, we assume that η is chosen small enough such that single instances of η can be omitted from the analysis. (The careful reader might want to check that the bounds we derive for the competitive ratio are actually continuous functions of η and therefore we are well allowed to take the limit ($\eta \rightarrow 0$.)

In the following analysis we consider the phases between the creation of new gaps. See Fig. 3(a) for an illustration of the following notations. We refer to the first items in each phase as the *jump* items s_1, s_2, \dots and we denote the last item in each phase by s'_1, s'_2, \dots . As we argued above, we assume $s_{i+1} = s'_i$. Furthermore, we denote the gaps that ALG creates by g_1, g_2, \dots and refer to the maximal gap after ALG packs an item r_i by $\text{maxgap}_{\text{ALG}}(r_i)$. Note that the height of the gaps might change when further items are packed (in case ALG packs them such that they reach into the gap from above or below). We denote the initial height of gap g_i by $\lambda_i s_i$ and the gap height directly before the next jump, i.e., in the moment s'_i is packed, by $\lambda'_i s_i$. Note that the height of gap g_i is always given relative to the corresponding jump item s_i . Finally, we denote the packing height up to gap g_i by $\mu_i s_i$, again relative to s_i . We

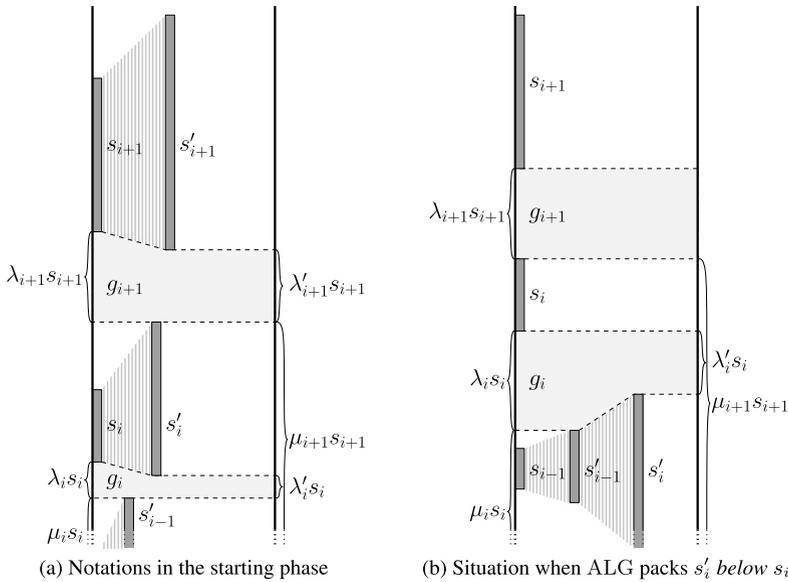


Fig. 3 Starting phase. Lemma 1 shows that the gap sizes are increasing with each jump and Lemma 2 shows that ALG needs to pack s'_i next to s_i

have $ALG(s_i) = \mu_i s_i + \lambda_i s_i + s_i$ and $\mu_i s_i \geq ALG(s'_{i-1})$ as s'_{i-1} is packed below g_i but other items might even reach higher than s'_{i-1} .

Since $OPT(s_i) = s_i$ and $ALG(s_i) = (\mu_i + \lambda_i + 1)s_i$ we directly have

$$\begin{aligned}
 (\mu_i + \lambda_i + 1)s_i &\leq \rho s_i && \text{for all } i \geq 1 \text{ and thus} \\
 \mu_i + \lambda_i &\leq \rho - 1 && \text{for all } i \geq 1.
 \end{aligned}
 \tag{1}$$

Before we are ready to prove that ALG is forced to reach a state with max-gap-to-height ratio less than \hat{c} , we have to show some assumptions that we can make on the algorithm ALG. First, we show that we can assume that ALG generates a packing where the gap preceding s_i is the maximal gap until s_{i+1} is packed for all $i \geq 1$. Or, in other words, ALG generates a packing with increasing gap sizes.

Lemma 1 *We can assume that ALG generates a packing that satisfies*

$$\text{maxgap}_{ALG}(r_j) = g_i \quad \text{for } r_j \in \{s_i, \dots, s'_i\}.$$

Proof The intuition of this proof is simple: A new gap g_i that is not maximal (as long as it is the current gap) is unnecessary and can therefore be omitted. We do this by bottom-aligning all items from s_i to r_j with the top of the previous gap.

More formally, let $\text{maxgap}_{ALG}(r_j) = g_k \neq g_i$ be the first violation of the condition for $r_j \in \{s_i, \dots, s'_i\}$. The modified algorithm ALG' simulates ALG with the exception that it bottom-aligns those items from $\{s_i, \dots, r_j\}$ that were previously packed above g_k with the top of g_k .

As items have only been moved downwards, ALG' remains ρ -competitive. Moreover, for the altered algorithm we have

$$\begin{aligned} \maxgap_{ALG'}(r_\ell) &= \maxgap_{ALG}(r_j) = g_k \quad \text{for } r_\ell \in \{s_i, \dots, r_j\} \quad \text{and} \\ h_{r_\ell}(g_k) &\geq h_{r_j}(g_k) \geq \hat{c}ALG(r_j) \geq \hat{c}ALG'(r_\ell) \quad \text{for } r_\ell \in \{s_i, \dots, r_j\}, \end{aligned}$$

where $h_{r_\ell}(g_k)$ denotes the height of gap g_k in the moment r_ℓ is packed. The second inequality is due to our assumption that ALG is (ρ, \hat{c}) -competitive, so the height of the max-gap at any time is at least \hat{c} -times the current packing height. The last inequality shows that also the modified algorithm retains a max-gap-to-height ratio of \hat{c} . So (ρ, \hat{c}) -competitiveness is not violated.

In total, the altered algorithm ALG' potentially even saves packing height in comparison with the original algorithm ALG . We can apply this method to all violations of $\maxgap_{ALG}(r_j) = g_i$ by induction. □

Now we show that the space *below* a jump item s_i is not large enough to accommodate s'_i before ALG makes the next jump. The implication of this statement is that any (ρ, \hat{c}) -competitive algorithm needs to place new items next to the current jump item.

Lemma 2 *ALG cannot generate a gap with an item s_{i+1} when the last item s'_i is packed completely below the previous jump item s_i .*

Proof For the sake of contradiction assume that ALG generates such a gap with item s_{i+1} while the last item s'_i was packed completely below the previous jump item s_i —see Fig. 3(b). As we will see, the proof of this lemma does not require to consider that ALG retains a max-gap-to-height ratio of \hat{c} .

By inequality (1) we have $s'_i \leq (\mu_i + \lambda_i)s_i \leq (\rho - 1)s_i$ as s'_i is packed below s_i . Thus $s_i \geq s'_i/(\rho - 1)$. With our assumption $s'_i = s_{i+1}$ we have

$$ALG(s_{i+1}) \geq s'_i + s_i + s_{i+1} \geq \left(2 + \frac{1}{\rho - 1}\right)s_{i+1}.$$

The contradiction follows with $\rho < 2.618\dots$ as

$$\begin{aligned} 1 &> (\rho - 2)(\rho - 1) \\ &\Leftrightarrow \left(2 + \frac{1}{\rho - 1}\right)s_{i+1} > \rho s_{i+1} \\ &\Rightarrow ALG(s_{i+1}) > \rho OPT(s_{i+1}). \end{aligned} \quad \square$$

Lemmas 1 and 2 state that each jump is larger than the previous jump and that *w.l.o.g.* ALG bottom-aligns the items next to the current jump item until a subsequent jump is carried out. This gives us sufficient information about the structure of the online packing to derive a contradiction. More specifically, the next two lemmas show that the relative gap height λ_i is decreasing by a constant in every step, which contradicts the trivial lower bound of $\lambda_i \geq \hat{c}/(1 - \hat{c}) \cdot (\mu_i + 1) \geq \hat{c}/(1 - \hat{c})$ as $\lambda_i s_i \geq \hat{c}(\mu_i s_i + \lambda_i s_i + s_i)$.

Lemma 3 $\lambda_1 \leq \rho - 1$ and for any $i \geq 1$

$$\lambda_{i+1} \leq \rho - 2 - \frac{\hat{c}(\rho - 1)}{\lambda_i - \hat{c}(\rho - 1)}.$$

Proof The first part, $\lambda_1 \leq \rho - 1$, follows directly from the ρ -competitiveness.

By Lemma 1 we know that $\text{maxgap}_{\text{ALG}}(s'_i) = g_i$. Since ALG preserves a max-gap-to-height ratio of at least \hat{c} , we have $\lambda'_i s_i \geq \hat{c} \text{ALG}(s'_i)$. Moreover, by Lemma 2 we have $\text{ALG}(s'_i) \geq \mu_i s_i + \lambda'_i s'_i + s'_i$ and thus

$$\begin{aligned} \lambda'_i s_i &\geq \hat{c} \text{ALG}(s'_i) \geq \hat{c}((\mu_i + \lambda'_i) s_i + s'_i) \\ \Rightarrow s_{i+1} = s'_i &\leq \frac{\lambda'_i s_i - \hat{c}(\mu_i + \lambda'_i) s_i}{\hat{c}}. \end{aligned} \tag{2}$$

Now we consider the packing height $\mu_{i+1} s_{i+1}$. We have $\mu_{i+1} s_{i+1} \geq \text{ALG}(s'_i) \geq (\mu_i + \lambda'_i) s_i + s'_i$ and thus

$$\begin{aligned} \mu_{i+1} &\geq (\mu_i + \lambda'_i) \frac{s_i}{s_{i+1}} + \frac{s'_i}{s_{i+1}} \\ &\geq \frac{\hat{c}(\mu_i + \lambda'_i)}{\lambda'_i - \hat{c}(\mu_i + \lambda'_i)} + 1 \quad \text{by inequality (2)} \\ &\geq \frac{\hat{c}(\rho - 1)}{\lambda_i - \hat{c}(\rho - 1)} + 1. \end{aligned}$$

The last step holds since

$$\begin{aligned} \frac{\partial}{\partial \lambda'_i} \left(\frac{\hat{c}(\mu_i + \lambda'_i)}{\lambda'_i - \hat{c}(\mu_i + \lambda'_i)} \right) &= \frac{\hat{c}(\lambda'_i - \hat{c}(\mu_i + \lambda'_i)) - \hat{c}(\mu_i + \lambda'_i)(1 - \hat{c})}{(\lambda'_i - \hat{c}(\mu_i + \lambda'_i))^2} \\ &= \frac{-\hat{c} \mu_i}{(\lambda'_i - \hat{c}(\mu_i + \lambda'_i))^2} < 0 \quad \text{as } \mu_i > 0 \end{aligned}$$

and thus $\frac{\hat{c}(\mu_i + \lambda'_i)}{\lambda'_i - \hat{c}(\mu_i + \lambda'_i)}$ is minimal for λ'_i maximal, which is $\lambda'_i = \lambda_i = \rho - 1 - \mu_i$ by inequality (1).

Using this lower bound for μ_{i+1} we get

$$\begin{aligned} \lambda_{i+1} &\leq \rho - 1 - \mu_{i+1} \quad \text{by inequality (1) for } i + 1 \\ &\leq \rho - 2 - \frac{\hat{c}(\rho - 1)}{\lambda_i - \hat{c}(\rho - 1)}. \end{aligned} \quad \square$$

Using this upper bound for the relative gap height λ_{i+1} we will show that no (ρ, \hat{c}) -competitive algorithm exists. We already gave the lower bound of $\lambda_i \geq \hat{c}/(1 - \hat{c})$. On the other hand, the following lemma shows that the relative gap heights are gradually decreasing over time. This gives a contradiction to the assumption that ALG can retain a max-gap-to-height ratio of \hat{c} . Thus ALG is either not ρ -competitive or we

reach a state with a max-gap-to-height ratio of less than \hat{c} , which ends the starting phase.

Lemma 4 $\lambda_{i+1} \leq \lambda_i - \varepsilon$ for some fixed $\varepsilon > 0$.

Proof Let $\varepsilon = \varepsilon(\rho) = 2\sqrt{\hat{c}(\rho - 1)} - \rho + 2 + \hat{c}(\rho - 1)$. By Lemma 3 we have $\lambda_{i+1} \leq \lambda_i - \varepsilon$ since

$$\begin{aligned} \rho - 2 - \frac{\hat{c}(\rho - 1)}{\lambda_i - \hat{c}(\rho - 1)} &\leq \lambda_i - 2\sqrt{\hat{c}(\rho - 1)} + \rho - 2 - \hat{c}(\rho - 1) \\ \Leftrightarrow -\frac{\hat{c}(\rho - 1)}{\lambda_i - \hat{c}(\rho - 1)} &\leq \lambda_i - 2\sqrt{\hat{c}(\rho - 1)} - \hat{c}(\rho - 1) \\ \Leftrightarrow \lambda_i^2 - (2\sqrt{\hat{c}(\rho - 1)} + \hat{c}(\rho - 1))\lambda_i &\geq -\hat{c}(\rho - 1) - 2\sqrt{\hat{c}(\rho - 1)}\hat{c}(\rho - 1) - \hat{c}^2(\rho - 1)^2 \\ \Leftrightarrow \lambda_i^2 - (2\sqrt{\hat{c}(\rho - 1)} + 2\hat{c}(\rho - 1))\lambda_i &\geq -\hat{c}(\rho - 1) - 2\sqrt{\hat{c}(\rho - 1)}\hat{c}(\rho - 1) - \hat{c}^2(\rho - 1)^2 \\ \Leftrightarrow (\lambda_i - (\sqrt{\hat{c}(\rho - 1)} + \hat{c}(\rho - 1)))^2 &\geq (\sqrt{\hat{c}(\rho - 1)} + \hat{c}(\rho - 1))^2 - \hat{c}(\rho - 1) \\ &\quad - 2\sqrt{\hat{c}(\rho - 1)}\hat{c}(\rho - 1) - \hat{c}^2(\rho - 1)^2 \\ &= 0. \end{aligned}$$

Thus it only remains to show that $\varepsilon(\rho) > 0$.

With $\hat{c} = \frac{\hat{\rho} - 2\sqrt{\hat{\rho} - 1}}{\hat{\rho} - 1}$ we have $\varepsilon(\hat{\rho}) = 0$ since

$$\begin{aligned} \varepsilon(\hat{\rho}) &= 2\sqrt{\hat{c}(\hat{\rho} - 1)} - \hat{\rho} + 2 + \hat{c}(\hat{\rho} - 1) \\ &= 2\sqrt{\hat{\rho} - 2\sqrt{\hat{\rho} - 1}} - \hat{\rho} + 2 + \hat{\rho} - 2\sqrt{\hat{\rho} - 1} \end{aligned}$$

and

$$\begin{aligned} 2\sqrt{\hat{\rho} - 2\sqrt{\hat{\rho} - 1}} - \hat{\rho} + 2 + \hat{\rho} - 2\sqrt{\hat{\rho} - 1} &= 0 \\ \Leftrightarrow 2\sqrt{\hat{\rho} - 2\sqrt{\hat{\rho} - 1}} &= 2\sqrt{\hat{\rho} - 1} - 2 \\ \Leftrightarrow 4(\hat{\rho} - 2\sqrt{\hat{\rho} - 1}) &= 4(\hat{\rho} - 1) - 8\sqrt{\hat{\rho} - 1} + 4. \end{aligned}$$

Note that this calculation actually defines the lower bound of $\frac{\hat{\rho} - 2\sqrt{\hat{\rho} - 1}}{\hat{\rho} - 1}$ for \hat{c} . Now observe that $\hat{c} = \frac{\hat{\rho} - 2\sqrt{\hat{\rho} - 1}}{\hat{\rho} - 1}$ does not depend on ρ and thus we have

$$\frac{\partial}{\partial \rho} (2\sqrt{\hat{c}(\rho - 1)} - \rho + 2 + \hat{c}(\rho - 1)) = \frac{\hat{c}}{\sqrt{\hat{c}(\rho - 1)}} - 1 + \hat{c}.$$

This derivative is negative as

$$\begin{aligned} \frac{\hat{c}}{\sqrt{\hat{c}(\rho - 1)}} &< 1 - \hat{c} \\ \Leftrightarrow \frac{\hat{c}}{\rho - 1} &< (1 - \hat{c})^2 \\ \Leftrightarrow \frac{\hat{\rho} - 2\sqrt{\hat{\rho} - 1}}{(\rho - 1)(\hat{\rho} - 1)} &< \frac{(\hat{\rho} - 1 - \hat{\rho} + 2\sqrt{\hat{\rho} - 1})^2}{(\hat{\rho} - 1)^2} \quad \text{by definition of } \hat{c} \\ \Leftrightarrow \hat{\rho} - 2\sqrt{\hat{\rho} - 1} &< (2\sqrt{\hat{\rho} - 1} - 1)^2 \cdot \frac{\rho - 1}{\hat{\rho} - 1} \\ \Leftrightarrow \hat{\rho} - 2\sqrt{\hat{\rho} - 1} &< \frac{4(\hat{\rho} - 1) - 4\sqrt{\hat{\rho} - 1} + 1}{2} \quad \text{as } \frac{\rho - 1}{\hat{\rho} - 1} > \frac{1}{2} \text{ for } \rho \geq 2 \\ \Leftrightarrow \frac{3}{2} &< \hat{\rho}. \end{aligned}$$

Thus $\varepsilon(\rho)$ is strictly decreasing with respect to ρ and $\varepsilon(\hat{\rho}) = 0$. Hence $\varepsilon(\rho) > 0$ for $\rho < \hat{\rho}$ and the lemma follows. □

Thus the λ_i decrease by a fixed amount in each step, contradicting the lower bound $\lambda_i \geq \hat{c}/(1 - \hat{c})$ for any $i \geq 1$. Thus our original assumption that /ALG/ cannot be true. In other words, we have proved

Lemma 5 Any ρ -competitive algorithm ALG can be forced to reach a state where the max-gap-to-height ratio is less than

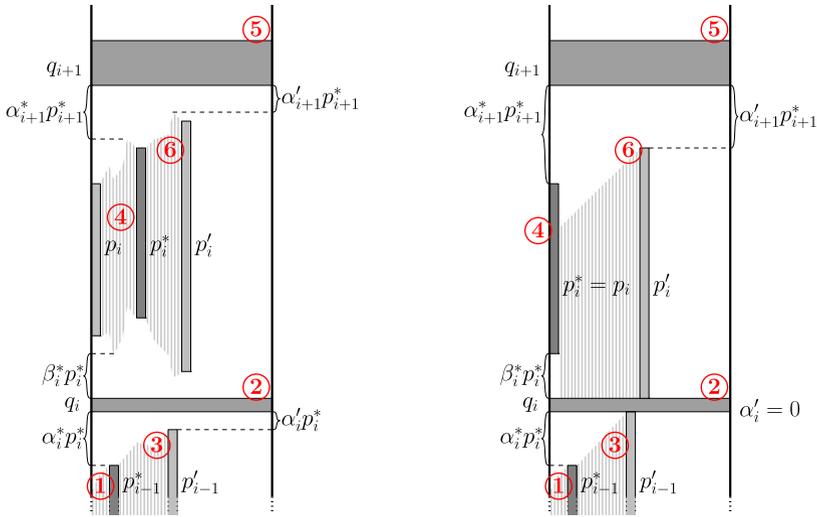
$$\hat{c} = \frac{\hat{\rho} - 2\sqrt{\hat{\rho} - 1}}{\hat{\rho} - 1}.$$

5.2 The Alternating Phase

In this section we describe the lower bound in the alternating phase. In this phase we use that by Lemma 5, any ρ -competitive algorithm ALG is forced to reach a state where the max-gap-to-height ratio is less than $\hat{c} = \frac{\hat{\rho} - 2\sqrt{\hat{\rho} - 1}}{\hat{\rho} - 1}$. As explained earlier in Sect. 3, we seek to analyze how \hat{c} and $\hat{\rho}$ must be related so that an algorithm that finishes the starting phase with a max-gap-to-height-ratio of \hat{c} cannot stay ρ -competitive for $\rho < \hat{\rho}$ in the alternating phase. The outcome will be that the two values must be related by the equation

$$\hat{c} = \frac{1 - \sqrt{4\hat{\rho}^2 - 12\hat{\rho} + 5}}{2(\hat{\rho} - 1)}.$$

Thus, if \hat{c} satisfies both equations above, then no ρ -competitive algorithm can exist for $\rho < \hat{\rho}$. Solving the two equations above yields $\hat{\rho} \approx 2.589\dots$ and $\hat{c} \approx 0.04275\dots$



(a) Notations. α -gaps below blocking items with height $\alpha_i^* p_i^*$ in the moment q_i is packed and final height $\alpha'_i p_i^*$. β -gaps above blocking items with height $\beta_i^* p_i^*$ (b) Structured packing according to the assumptions by Lemmas 6, 7. Here we illustrate the case $\beta_i^* > 0$ and thus $\alpha'_i = 0$ and $p_i^* = p_i$

Fig. 4 Order of the released items. (1) Thin items up to p'_{i-1} ; (2) blocking item q_i ; (3) and (4) thin items up to p_i^* (including the jump item p_i); (5) blocking item q_{i+1} and (6) further thin items up to p'_i

Our adversary sequence in the alternating phase starts with the first blocking item q_1 and then continues with the list of thin items of gradually increasing height from the starting phase interleaved with further blocking items. Let $\eta > 0$ be some very small constant and let r_k be the last item that was released in the starting phase. Then we continue with the list $L_{\text{alternating}} = q_1, r_{k+1}, r_{k+2}, \dots$ where

$$q_1 = \hat{c} \cdot \text{ALG}(r_k) \quad \text{and}$$

$$r_i = r_{i-1} + \eta \quad \text{for } i \geq k + 1.$$

To understand when the blocking items are inserted, let us first introduce the notations in this phase—see Fig. 4(a).

Similar to the starting phase, we consider the jump items, i.e., the thin items that are the first to be packed above a blocking item q_i , and denote them by p_i . The thin item directly before the jump item is denoted by p'_{i-1} (we will later see that we can actually assume that p'_{i-1} is the last item that is packed below q_i). We denote the interval between the blocking items q_{i-1} and q_i by I_i . As in the standard BBK sequences, the thin item whose height exceeds the height of the previous interval plays an important role. We denote the first item that exceeds the height of I_{i-1} by p_i^* and describe all further heights relative to these designated items.

As described in the introduction, we distinguish α -gaps (directly below blocking items) and β -gaps (directly above blocking items). As the gap heights can change during the packing (as further thin items are packed into the same interval) we have to be specific about the moment in which we consider these heights. Let $\alpha_i^* p_i^*$ be the

height of the α -gap below q_i in the moment q_i is packed and let $\alpha'_i p_i^*$ be the final height of the α -gap below q_i , i.e., the height in the moment p_i^* is packed (as afterwards no further item can be packed into I_{i-1}). The notation is due to our assumption that p'_{i-1} is the last item that is packed into I_{i-1} (which we show later). Regarding the β -gap we get along with a single definition: Let $\beta_i^* p_i^*$ be the height of the β -gap above q_i in the moment p_i^* is packed.

The blocking item q_{i+1} is released directly after p_i^* . This ensures that the online algorithm jumps before a new blocking item is released (as the height of p_i^* exceeds the height of the previous interval). We set the height of the blocking items to

$$q_1 := \hat{c} \cdot \text{ALG}(r_k) \quad \text{as already mentioned above and}$$

$$q_i := \max(\alpha'_{i-1} p_{i-1}^*, \beta_{i-1}^* p_{i-1}^*, q_{i-1}) + \eta \quad \text{for } i \geq 2.$$

Note that we use the *final* height $\alpha'_i p_i^*$ of the α -gap in this definition. This definition ensures that the blocking items are always packed above all previous items.

Again the function of the positive term η is to gradually increase the height of thin items and to ensure that the blocking items are always packed above all previous items. As before, we make the assumption that η is chosen small enough to be omitted from the analysis. Thus we assume that $q_i = \max(\alpha'_{i-1} p_{i-1}^*, \beta_{i-1}^* p_{i-1}^*, q_{i-1})$ and that the height of p_i^* equals the height of the previous interval I_{i-1} throughout this section. (Again, this is justified by taking the limit ($\eta \rightarrow 0$)).

We use $\text{succ}(r_i)$ and $\text{prec}(r_i)$ to denote the thin item that succeeds and that precedes r_i , respectively. Using this notations we can rephrase the input list including the blocking items to

$$L_{\text{alternating}} = q_1, r_{k+1}, \dots, p_1^*, q_1, \text{succ}(p_1^*), \dots, p_2^*, q_2, \text{succ}(p_2^*), \dots$$

We also refer to Fig. 4 for an illustration of the order in which the items are released.

5.2.1 Overview

We prove by contradiction that no ρ -competitive algorithm exists for $\rho < \hat{\rho}$. Thus we assume to the contrary that a ρ -competitive algorithm ALG exists. By the analysis of the starting phase we already know that we can force ALG to reach a state with a max-gap-to-height ratio less than \hat{c} . In accordance with the notation given above we introduce the parameter γ_i^* to measure how much ALG improves upon the ρ -competitiveness. Let γ_i^* be defined through

$$\text{ALG}(p_i^*) + \gamma_i^* p_i^* = \rho \text{OPT}(p_i^*).$$

Using this value, we introduce the potential function

$$\Phi_i = \gamma_i^* + \beta_i^*.$$

Note that moving p_i^* (together with all other thin items packed on top of q_i) up or down will increase *resp.* decrease the value of β_i^* and, at the same time, decrease *resp.* increase the value of γ_i^* by the same amount. Thus, such a move would not

affect the potential Φ_i . (This phenomenon seems to be a characteristic feature of suitable potentials, *cf.* also, *e.g.*, [11] or [13].) Yet, of course, the online algorithm's decision on where to pack the thin items above q_i determines the current β -gap and influences subsequent items and potential values.

Obviously, any ρ -competitive algorithm needs to keep Φ_i non-negative over time. We aim at deriving a contradiction by showing that Φ_i decreases by a constant amount in every step. Unfortunately, there is one possible exception to this rule, making the proof substantially more involved: Φ_i might increase exactly once. We will show that even in this case, Φ_i is properly bounded from above and cannot increase a second time.

We start our analysis with some preliminary results (Lemmas 6–11) on the structure of a packing generated by ALG. The general theme is that if a ρ -competitive algorithm exists, then there also exists a ρ -competitive algorithm that generates packings with the assumed structure. In other words: If ALG does not generate such a packing, we can alter the packing (or rather the algorithm) such that the conditions are satisfied and ρ -competitiveness is not violated at any point. In addition, a few helpful estimates are derived, essentially lower and upper bounds for α -gaps: Lemma 7 states that for the online algorithm it is not wise to generate both a nonzero α and a nonzero β -gap, as one large β -gap is more promising. Lemma 8 states that the online algorithm should better not work with small nonzero α -gaps, as these could better be replaced by β -gaps. Thus the size of nonzero α -gaps can be bounded from below and, of course, also from above (Lemmas 9 and 10), as the online algorithm is assumed to be ρ -competitive. The “preliminaries section” then concludes with Lemma 11, saying how exactly Φ_{i+1} depends on Φ_i and the parameters (gap- and item sizes) in step i .

Then follows the subsection “induction” (Lemmas 12–16) where we prove that the potential decreases by some fixed amount in each step (except possibly once). The proof will be by induction as we need to upper-bound the potential—as well as current q/p -values—in each step. Intuitively, a small potential indicates that the online algorithm is in a bad position, as the potential upper bounds both the (relative) distance from the allowed packing height (ρ -times OPT) and the current β -gap. Lemma 12 states that the initial potential is small, due to the fact that the online algorithm enters the alternating phase with a small max-gap-to-height-ratio. In Lemmas 13 to 16 we investigate the change in the potential depending on how q_{i+1}^* is defined (via the α - or β -gap or q_i). In each case we conclude that the potential decreases, assuming that it was low already. The only exception is when $q_{i+1} = q_i$. This has also been the critical case for standard BBK-sequences, where the optimal online algorithm's strategy would create a sequence that becomes stationary after a few steps and items are packed without any gaps. This works only if (in the stationary part) the blocking items q are sufficiently large compared to the thin items p . (In the stationary part, OPT increases by q in each step, while the online height increases with $p + q$, so that $q + p \leq \rho q$, *i.e.*, $\frac{q}{p} \geq \frac{1}{\rho-1}$ is required.)

In the case of modified BBK-sequences that we consider here, we shall see that such relatively large blocking items cannot be generated and therefore the optimum online algorithm will never induce a stationary sequence. Indeed, the blocking items stay small in size (relative to the corresponding thin items, *cf.* Lemmas 12–16 below).

Even in the critical case where $q_{i+1} = q_i$, the potential decreases in case the current q/p ratio is small enough (Lemma 15) or—in case the q/p ratio is slightly larger—it will get small enough in the next step (Lemma 16), so that, eventually, the potential is shown to decrease in each except possibly one single step.

5.2.2 Preliminaries

The following lemmas (6–11) provide some simplifications, *i.e.*, “*w.l.o.g.* assumptions” on the structure of the packing that ALG generates in this phase—see Fig. 4(b) for an illustration.

Lemma 6 *We can assume that ALG generates a packing such that*

1. *the items $p_i, \dots, p_i^*, \dots, p'_i$ lie in interval I_i ,*
2. *the items p_i, \dots, p_i^* are bottom-aligned,*
3. *the items $\text{succ}(p_i^*), \dots, p'_i$ are bottom-aligned at the top of q_i .*

Proof By definition the items p_i^*, \dots, p'_i are taller than the previous interval I_{i-1} and thus all lie in interval I_i . Assume that an item from $p_i, \dots, \text{prec}(p_i^*)$ does not lie in interval I_i and let r_j be the tallest such item. Then we can move down the items $p_i, \dots, \text{prec}(r_j)$ and bottom-align them with r_j . This redefines p_i to $\text{succ}(r_j)$ and hereby satisfies condition 1. Observe that moving down the items $p_i, \dots, \text{prec}(r_j)$ does not violate ρ -competitiveness and as α'_i and β_i^* are not changed, the further packing remains unchanged.

If the items p_i, \dots, p_i^* are not packed bottom-aligned, we move them downwards until they are aligned with the lowest item of this list in order to satisfy condition 2. And to satisfy condition 3 we move the items $\text{succ}(p_i^*), \dots, p'_i$ down until they are aligned with the top of q_i if these items are not bottom-aligned at the top of q_i . In both cases the alteration is possible as the height of the interval I_i and thus the height of p_{i+1}^* remains unchanged. Moreover, the height of q_i does not change (as β_i^* is not changed). The values of α'_{i+1} and α'_{i+1} can actually change, but only become larger. But as the heights of I_i and p_{i+1}^* remain unchanged, the parameter α'_{i+1} only affects q_{i+1} and the value of q_{i+1} contributes to the packing height of OPT and ALG to the same extent. Thus increased values of α'_{i+1} and α'_{i+1} cannot cause a violation of the ρ -competitiveness. \square

Recall that $q_{i+1} = \max(\alpha'_i p_i^*, \beta_i^* p_i^*, q_i)$. Depending on the way in which q_{i+1} is actually defined, we can assume that the other value(s) are zero as the following lemma shows. (Intuitively, as a good online algorithm should seek to create large gaps, it does not make sense to create *both* an α - and a β -gap.)

Lemma 7 *We can assume that ALG generates a packing such that*

1. *if $q_{i+1} = \max(\beta_i^* p_i^*, q_i)$, then we have $\alpha'_i = 0$,*
2. *if $q_{i+1} = \max(\alpha'_i p_i^*, q_i)$, then we have $\beta_i^* = 0$.*

Proof First, assume that $q_{i+1} = \max(\beta_i^* p_i^*, q_i)$ and $\alpha'_i > 0$. By construction of the adversary sequence, the height of p_i^* does not depend on α'_i and is predetermined at the moment q_i is packed. Thus a reduction of α'_i , which corresponds to packing further thin items into the previous interval, does not change q_{i+1} and p_i^* . So we can alter ALG such that all items from $\text{succ}(p'_{i-1}), \dots, \text{pre}(p_i^*)$, are packed into I_{i-1} . This reduces α'_i to 0 and thus satisfies condition 1 without implying any change to the packing after p_i^* .

Now assume that $q_{i+1} = \max(\alpha'_i p_i^*, q_i)$ and $\beta_i^* > 0$. In this case a reduction of β_i^* does not change q_{i+1} and p_i^* . So we can alter ALG to set β_i^* to 0, i.e., bottom-align the items p_i, \dots, p_i^* with the top of q_i , without implying any change to the packing after p_i^* and hereby satisfy condition 2. This alteration increases α_{i+1}^* and might increase α'_{i+1} as well—as we saw in Lemma 6, this does not violate ρ -competitiveness. \square

Observe that by Lemma 6 we have $p_{i+1}^* = \beta_i^* p_i^* + p_i^* + \alpha_{i+1}^* p_{i+1}^*$ and thus

$$p_{i+1}^* = \frac{1 + \beta_i^*}{1 - \alpha_{i+1}^*} p_i^*. \tag{3}$$

Using this equation, we are ready to show the following assumption.

Lemma 8 *We can assume that ALG generates a packing such that if $\alpha'_i > 0$, then we have*

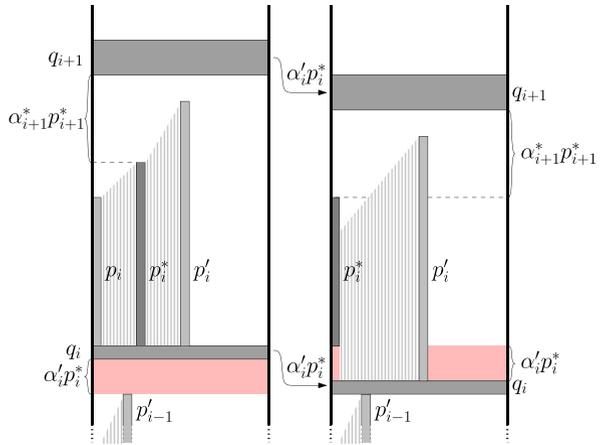
$$\alpha_{i+1}^* > \frac{(\rho - 1)\alpha'_i}{1 + (\rho - 1)\alpha'_i}.$$

Proof We assume that $\alpha'_i > 0$ and $\alpha_{i+1}^* \leq (\rho - 1)\alpha'_i / (1 + (\rho + 1)\alpha'_i)$. By Lemma 7 condition 2 we have $\beta_i^* = 0$ and thus $p_{i+1}^* = p_i^* / (1 - \alpha_{i+1}^*)$. We can alter ALG to save a packing height of $\alpha'_i p_i^*$ without violating ρ -competitiveness by changing the α -gap to a β -gap. To do that, we move down q_i and all items that are released after p'_{i-1} with the exception of p_i by $\alpha'_i p_i^*$. In other words, we close the $\alpha'_i p_i^*$ gap between p'_{i-1} and q_i by moving down q_i and all items above q_i . The only exception is the item p_i that we keep at its position to retain a β -gap at the moment this item is packed. Hereby, we keep a gap of the original size $\alpha'_i p_i^*$ above q_i . See Fig. 5 for an illustration of the altered packing.

Note that this alteration changes the adversary sequence: As there does not remain any α'_i -gap, the item q_{i+1} is released directly after p_i is packed—also redefining p_i^* to p_i . This is the only change in the adversary sequence since the size of q_{i+1} is not changed and also the height of interval I_i stays constant. Since the optimal value changed as q_{i+1} is released earlier than before, we have to check whether the altered packing is actually feasible.

We denote the optimal algorithm for the altered instance by OPT' and the altered algorithm by ALG' . With $\alpha'_i p_i^*$ we refer to the height before the alteration. The height $\alpha_{i+1}^* p_{i+1}^*$ remains unchanged. We have $\text{OPT}'(q_{i+1}) = \text{OPT}(p_i) + q_{i+1} = \text{OPT}(p_i) + \alpha'_i p_i^*$ and $\text{ALG}'(q_{i+1}) = \text{ALG}(p_i) + \alpha_{i+1}^* p_{i+1}^* + q_{i+1} = \text{ALG}(p_i) +$

Fig. 5 If $\alpha_{i+1}^* \leq (\rho - 1)\alpha'_i / (1 + (\rho - 1)\alpha'_i)$, then we can move down q_i and all items that are released after p'_{i-1} , with the exception of p_i , by $\alpha'_i p_i^*$. Hereby the α -gap becomes a β -gap and p_i becomes the new p_i^* as the interval I_{i-1} shrinks



$\alpha_{i+1}^* p_{i+1}^* + \alpha'_i p_i^*$. Thus

$$\begin{aligned}
 \text{ALG}'(q_{i+1}) &\leq \rho \text{OPT}'(q_{i+1}) \\
 &\Leftrightarrow \alpha_{i+1}^* p_{i+1}^* \leq \underbrace{\rho \text{OPT}(p_i) - \text{ALG}(p_i)}_{\geq 0} + (\rho - 1)\alpha'_i p_i^* \\
 &\Leftrightarrow \alpha_{i+1}^* p_{i+1}^* \leq (\rho - 1)\alpha'_i p_i^* \\
 &\Leftrightarrow \alpha_{i+1}^* p_{i+1}^* \leq (\rho - 1)\alpha'_i (1 - \alpha_{i+1}^*) p_{i+1}^* \quad \text{by (3) with } \beta_i^* = 0 \\
 &\Leftrightarrow \alpha_{i+1}^* \leq \frac{(\rho - 1)\alpha'_i}{1 + (\rho - 1)\alpha'_i},
 \end{aligned}$$

which we assumed to be true. Thus q_{i+1} can actually be packed by the altered algorithm. The feasibility for all other items in the altered packing is obvious. \square

On the other hand, it is not possible for ALG to create an arbitrarily large gap when packing a blocking item q_{i+1} . We capture this fact in the following lemma.

Lemma 9 We have

$$\alpha_{i+1}^* \leq \frac{\gamma_i^* + (\rho - 1)\frac{q_{i+1}}{p_i^*}}{1 + \beta_i^* + \gamma_i^* + (\rho - 1)\frac{q_{i+1}}{p_i^*}}.$$

Proof The value of α_{i+1}^* can be bounded by observing the moment when q_{i+1} is packed. We have

$$\begin{aligned}
 \text{OPT}(q_{i+1}) &= \text{OPT}(p_i^*) + q_{i+1}, \\
 \text{ALG}(q_{i+1}) &= \text{ALG}(p_i^*) + \alpha_{i+1}^* p_{i+1}^* + q_{i+1}.
 \end{aligned}$$

And since q_{i+1} needs to be packed ρ -competitively by ALG we get

$$\begin{aligned}
 \text{ALG}(q_{i+1}) &\leq \rho \text{OPT}(q_{i+1}) \\
 \Leftrightarrow \alpha_{i+1}^* p_{i+1}^* &\leq \gamma_i^* p_i^* + (\rho - 1)q_{i+1} \\
 \Leftrightarrow \frac{\alpha_{i+1}^*}{1 - \alpha_{i+1}^*} &\leq \frac{\gamma_i^* + (\rho - 1)\frac{q_{i+1}}{p_i^*}}{1 + \beta_i^*} \quad \text{by (3)} \\
 \Leftrightarrow \alpha_{i+1}^* &\leq \frac{\gamma_i^* + (\rho - 1)\frac{q_{i+1}}{p_i^*}}{1 + \beta_i^* + \gamma_i^* + (\rho - 1)\frac{q_{i+1}}{p_i^*}}. \quad \square
 \end{aligned}$$

The parameter α'_i plays an important role in the analysis as the height of the preceding blocking item depends on it. With the next lemma we get an upper bound for this parameter.

Lemma 10 *We have $(\rho - 1)\alpha'_i \leq \gamma_i^*$.*

Proof The idea of the bound is that if ALG jumps early, i.e., with an $\alpha'_i > 0$, then it generates a packing where $p_i^* = p'_i + \alpha'_i p_i^*$. This additional height directly contributes to the value of γ_i^* with a factor of $\rho - 1$ (as ALG and OPT increase by the same amount).

Formally, we have $\text{ALG}(p_i^*) = \text{ALG}(p_i) + \alpha'_i p_i^*$, $\text{OPT}(p_i^*) = \text{OPT}(p_i) + \alpha'_i p_i^*$ and $\text{ALG}(p_i^*) + \gamma_i^* p_i^* = \rho \text{OPT}(p_i^*)$. And since p_i was feasible we have $\text{ALG}(p_i) \leq \rho \text{OPT}(p_i)$ and get $(\rho - 1)\alpha_i p_i^* \leq \gamma_i^* p_i^*$. \square

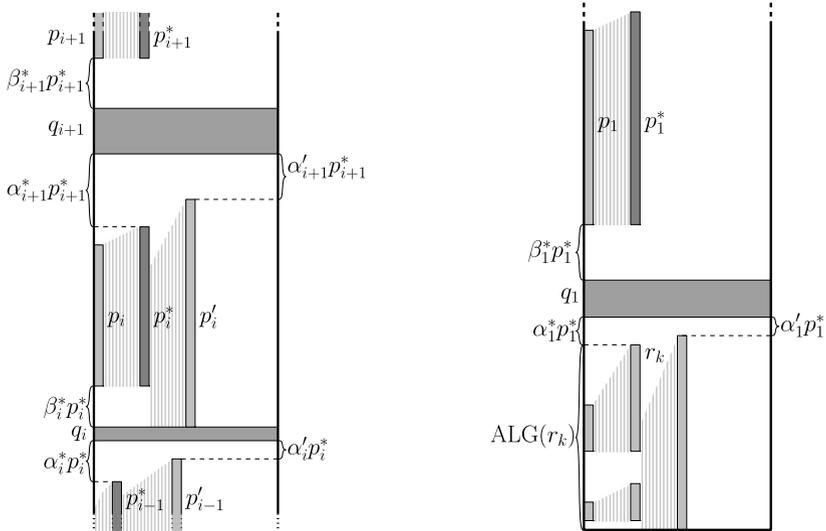
Similar to Kern and Paulus [11] we get the following lemma that bounds the potential function in terms of the parameters of the previous interval.

Lemma 11 *We have*

$$\begin{aligned}
 \Phi_{i+1} = \gamma_{i+1}^* + \beta_{i+1}^* &= \frac{\gamma_i^* + (\rho - 1)\frac{q_{i+1}}{p_i^*} + (\rho - 1)\beta_i^* - 1}{1 + \beta_i^*} (1 - \alpha_{i+1}^*) \\
 &\quad + (\rho - 2)\alpha_{i+1}^*.
 \end{aligned}$$

Proof See Fig. 6(a) for an illustration of the packing. We consider the change between p_i^* and p_{i+1}^* and with $p_{i+1}^* = \beta_i^* p_i^* + p_i^* + \alpha_{i+1}^* p_{i+1}^*$ from (3) we have

$$\begin{aligned}
 \text{OPT}(p_{i+1}^*) &= \text{OPT}(p_i^*) + q_{i+1} + p_{i+1}^* - p_i^* \\
 &= \text{OPT}(p_i^*) + q_{i+1} + \beta_i^* p_i^* + \alpha_{i+1}^* p_{i+1}^*, \\
 \text{ALG}(p_{i+1}^*) &= \text{ALG}(p_i^*) + \alpha_{i+1}^* p_{i+1}^* + q_{i+1} + \beta_{i+1}^* p_{i+1}^* + p_{i+1}^* \\
 &= \text{ALG}(p_i^*) + \alpha_{i+1}^* p_{i+1}^* + q_{i+1} + \beta_{i+1}^* p_{i+1}^* + \beta_i^* p_i^* + p_i^* + \alpha_{i+1}^* p_{i+1}^*.
 \end{aligned}$$



(a) In this illustration we disregard Lemma 7 (which would give $\alpha'_i = 0$ or $\beta_i^* = 0$) to show the general case

(b) Here we show that ALG can pack further items aligned with the bottom of the strip before generating the gap $\beta_1^* p_1^*$

Fig. 6 Illustrations for Lemmas 11 and 12

Thus with $\gamma_i^* p_i^* = \rho \text{OPT}(p_i^*) - \text{ALG}(p_i^*)$ we get

$$\begin{aligned} \text{ALG}(p_{i+1}^*) + \gamma_{i+1}^* p_{i+1}^* &= \rho \text{OPT}(p_{i+1}^*) \\ \Leftrightarrow \gamma_{i+1}^* p_{i+1}^* + \beta_{i+1}^* p_{i+1}^* - (\rho - 2)\alpha_{i+1}^* p_{i+1}^* \\ &= \gamma_i^* p_i^* + (\rho - 1)q_{i+1} + (\rho - 1)\beta_i^* p_i^* - p_i^*. \end{aligned}$$

By (3) we have $(1 - \alpha_i^*)p_{i+1}^* = (\beta_i^* + 1)p_i^*$ and finally get

$$\frac{\gamma_{i+1}^* + \beta_{i+1}^* - (\rho - 2)\alpha_{i+1}^*}{1 - \alpha_{i+1}^*} = \frac{\gamma_i^* + (\rho - 1)\frac{q_{i+1}}{p_i^*} + (\rho - 1)\beta_i^* - 1}{1 + \beta_i^*}. \quad \square$$

This completes our preparations for the induction that we show next.

5.2.3 The Induction

In this section we show the intended contradiction: Any ρ -competitive algorithm needs to satisfy $\Phi_i \geq 0$, however the potential Φ_i decreases indefinitely.

We start the induction with the next lemma, giving a maximal initial value of $\rho - 2 + (\rho - 1)\hat{c}$ for the potential. Afterwards, we distinguish three cases according to the definition of q_{i+1} . If $q_{i+1} = \beta_i^* p_i^*$ or $q_{i+1} = \alpha'_i p_i^*$ we show $\Phi_{i+1} \leq \Phi_i - \varepsilon$ for some $\varepsilon > 0$. The case $q_{i+1} = q_i$ is more involved. Either we also get a decreasing potential or the potential might actually rise, but is still lower than the initial value.

Therefore, this rise can only happen once, as we finally show when we bring together all parts.

In the following calculations (which are very technical) we basically derive a series of upper bounds on the potential Φ_{i+1} . In detail, we get

$$\begin{aligned} \Phi_{i+1} &\leq \frac{2(\rho-1)\Phi_i-1}{1+\Phi_i} && \text{in case } q_{i+1} = \beta_i^* p_i^* \quad (4) \\ \text{and } \Phi_{i+1} &\leq \frac{\rho(\rho-1)\Phi_i-1}{1+\rho\Phi_i} && \text{in case } q_{i+1} = \beta_i^* p_i^* \\ \text{and } \Phi_{i+1} &\leq \frac{2(\rho-1)\Phi_i-1}{1+2\Phi_i} && \text{in case } q_{i+1} = \alpha_i' p_i^* \\ \text{and } \Phi_{i+1} &< \frac{\rho(\rho-1)\Phi_i+\rho^2-3\rho+1}{\rho\Phi_i+2\rho-1} && \text{in case } q_{i+1} = q_i \\ \text{and } \Phi_{i+1} &\leq \frac{(\Phi_i+(\rho-1)\hat{c}-1)(\rho-1)}{\rho(\rho-2+(\rho-1)\hat{c})} + \rho - 2 && \text{in case } q_{i+1} = q_i. \end{aligned}$$

All these conditions eventually imply $\Phi_{i+1} \leq \Phi_i - \varepsilon$ for some $\varepsilon > 0$ and $\rho < \hat{\rho}$. Just for condition (4) we additionally require the induction hypothesis $\Phi_i \leq \rho - 2 + (\rho - 1)\hat{c}$. This is actually exactly the condition that gives us the value of $\hat{c} = \frac{1 - \sqrt{4\hat{\rho}^2 - 12\hat{\rho} + 5}}{2(\hat{\rho} - 1)}$.

We now start with the induction hypothesis. Not only do we give an upper bound for the initial potential Φ_1 , but also for the ratio q_1/p_1^* . This is needed later when we bring the different parts together.

Lemma 12 *We have*

$$\begin{aligned} \Phi_1 &\leq \rho - 2 + (\rho - 1)\hat{c} \quad \text{and} \\ \frac{q_1}{p_1^*} &< \frac{1}{\rho}. \end{aligned}$$

Proof Consider the packing of ALG and the optimal packing after p_1^* is released—see Fig. 6(b). Recall that $\text{ALG}(r_k)$ is the packing height at the end of the starting phase and that $q_1 = \hat{c} \text{ALG}(r_k)$. As p_1^* equals the height of the interval below q_1 we have

$$\begin{aligned} \text{OPT}(p_1^*) &= p_1^* + q_1 = p_1^* + \hat{c} \text{ALG}(r_k) \quad \text{and} \\ \text{ALG}(p_1^*) &= p_1^* + q_1 + \beta_1^* p_1^* + p_1^* = 2p_1^* + \hat{c} \text{ALG}(r_k) + \beta_1^* p_1^*. \end{aligned}$$

Moreover, we have $p_1^* = \text{ALG}(r_k) + \alpha_1^* p_1^*$ and thus $p_1^* = \frac{\text{ALG}(r_k)}{1 - \alpha_1^*}$. We get

$$\begin{aligned} \gamma_1^* p_1^* &= \rho \text{OPT}(p_1^*) - \text{ALG}(p_1^*) \\ &= (\rho - 2)p_1^* + (\rho - 1)\hat{c} \text{ALG}(r_k) - \beta_1^* p_1^* \end{aligned}$$

$$\begin{aligned} \Rightarrow \Phi_i &= \gamma_1^* + \beta_1^* = \rho - 2 + (\rho - 1)\hat{c}(1 - \alpha_1^*) \quad \text{since } p_1^* = \frac{\text{ALG}(r_k)}{1 - \alpha_1^*} \\ &\leq \rho - 2 + (\rho - 1)\hat{c}. \end{aligned}$$

Finally, observe that

$$\frac{q_1}{p_1^*} \leq \frac{\hat{c} \text{ALG}(r_k)}{\text{ALG}(r_k)} = \hat{c} < \frac{1}{\rho}. \quad \square$$

With the next two lemmas we show that the potential decreases if $q_{i+1} = \beta_i^* p_i^*$ or $q_{i+1} = \alpha_i' p_i^*$. At the same time we show that q_{i+1}/p_{i+1}^* is bounded, which we need in the last case $q_{i+1} = q_i$.

Lemma 13 *If $\Phi_i \leq \rho - 2 + (\rho - 1)\hat{c}$ and $q_{i+1} = \beta_i^* p_i^*$, then*

$$\begin{aligned} \Phi_{i+1} &\leq \Phi_i - \varepsilon \quad \text{for some } \varepsilon > 0 \quad \text{and} \\ \frac{q_{i+1}}{p_{i+1}^*} &\leq \frac{\Phi_{i+1}}{\rho - 1} \quad \text{or} \quad \frac{q_{i+1}}{p_{i+1}^*} < \frac{1}{\rho}. \end{aligned}$$

Proof By Lemma 7, condition 1 we can assume $\alpha_i' = 0$. Thus Lemma 11 yields

$$\Phi_{i+1} = \frac{\gamma_i^* + 2(\rho - 1)\beta_i^* - 1}{1 + \beta_i^*} (1 - \alpha_{i+1}^*) + (\rho - 2)\alpha_{i+1}^*.$$

Note that this function is linear in α_{i+1}^* . Thus Φ_{i+1} attains its maximum for maximal or minimal α_{i+1}^* . We show for both cases that $\Phi_{i+1} \leq \Phi_i - \varepsilon$ for some $\varepsilon > 0$.

If Φ_{i+1} is non-increasing in α_{i+1}^* we have

$$\begin{aligned} \Phi_{i+1} &\leq \frac{\gamma_i^* + 2(\rho - 1)\beta_i^* - 1}{1 + \beta_i^*} \quad \text{as } \alpha_{i+1}^* \geq 0 \\ &= \frac{\Phi_i + (2\rho - 3)\beta_i^* - 1}{1 + \beta_i^*} \quad \text{as } \gamma_i^* + \beta_i^* = \Phi_i \\ &\leq \frac{2(\rho - 1)\Phi_i - 1}{1 + \Phi_i}. \end{aligned}$$

The last step holds as $\beta_i^* \leq \Phi_i$ and the function is increasing with respect to β_i^* . With $\varepsilon = \varepsilon(\rho) = \frac{\hat{c}^2(\rho-1)^2 - \hat{c}(\rho-1) - (\rho-1)(\rho-2) + 1}{(\rho-1)(1+\hat{c})}$ we have $\Phi_{i+1} \leq \Phi_i - \varepsilon$ since

$$\begin{aligned} \frac{2(\rho - 1)\Phi_i - 1}{1 + \Phi_i} &\leq \Phi_i - \varepsilon \\ \Leftrightarrow \Phi_i^2 - (2\rho - 3 + \varepsilon)\Phi_i &\geq \varepsilon - 1 \\ \Leftrightarrow (\rho - 2 + (\rho - 1)\hat{c})^2 - (2\rho - 3 + \varepsilon)(\rho - 2 + (\rho - 1)\hat{c}) &\geq \varepsilon - 1 \\ \text{as } \Phi_i &\leq \rho - 2 + (\rho - 1)\hat{c} \text{ and } 2\Phi_i - 2\rho + 3 - \varepsilon \leq 2(\rho - 1)\hat{c} - 1 < 0 \end{aligned}$$

$$\Leftrightarrow \frac{(\rho - 2 + (\rho - 1)\hat{c})^2 - (2\rho - 3)(\rho - 2 + (\rho - 1)\hat{c}) + 1}{1 + \rho - 2 + (\rho - 1)\hat{c}} \geq \varepsilon$$

$$\Leftarrow \frac{\hat{c}^2(\rho - 1)^2 - \hat{c}(\rho - 1) - (\rho - 1)(\rho - 2) + 1}{(\rho - 1)(1 + \hat{c})} = \varepsilon.$$

It remains to show $\varepsilon = \varepsilon(\rho) > 0$. We have $\varepsilon(\hat{\rho}) = 0$ since

$$\hat{c}^2(\hat{\rho} - 1)^2 - \hat{c}(\hat{\rho} - 1) - (\hat{\rho} - 1)(\hat{\rho} - 2) + 1 = 0$$

for $\hat{c} = \frac{1 - \sqrt{4\hat{\rho}^2 - 12\hat{\rho} + 5}}{2(\hat{\rho} - 1)}$. Now observe that ε is strictly decreasing with ρ since

$$\frac{\partial}{\partial \rho}(\varepsilon(\rho)) = \frac{(\hat{c}^2 - 1)(\rho - 1)^2 - (\rho^2 - 2\rho + 2)}{(\rho - 1)^2(1 + \hat{c})} < 0$$

as $\hat{c}^2 - 1 < 0$ and $\rho^2 - 2\rho + 2 > 0$. Thus we have $\varepsilon = \varepsilon(\rho) > \varepsilon(\hat{\rho}) = 0$ in this case.

Now, if Φ_{i+1} is increasing in α_{i+1}^* , we use Lemma 9 to get

$$\begin{aligned} \Phi_{i+1} &\leq \frac{\gamma_i^* + 2(\rho - 1)\beta_i^* - 1}{1 + \beta_i^*} \cdot \left(1 - \frac{\gamma_i^* + (\rho - 1)\beta_i^*}{1 + \gamma_i^* + \rho\beta_i^*}\right) + (\rho - 2) \cdot \frac{\gamma_i^* + (\rho - 1)\beta_i^*}{1 + \gamma_i^* + \rho\beta_i^*} \\ &\leq \frac{\gamma_i^* + 2(\rho - 1)\beta_i^* - 1}{1 + \beta_i^*} \cdot \frac{1 + \beta_i^*}{1 + \gamma_i^* + \rho\beta_i^*} + (\rho - 2) \cdot \frac{\gamma_i^* + (\rho - 1)\beta_i^*}{1 + \gamma_i^* + \rho\beta_i^*} \\ &= \frac{(\rho - 1)\gamma_i^* + \rho(\rho - 1)\beta_i^* - 1}{1 + \gamma_i^* + \rho\beta_i^*} \\ &= \frac{(\rho - 1)\Phi_i + (\rho - 1)^2\beta_i^* - 1}{1 + \Phi_i + (\rho - 1)\beta_i^*} \quad \text{as } \gamma_i^* + \beta_i^* = \Phi_i \\ &\leq \frac{\rho(\rho - 1)\Phi_i - 1}{1 + \rho\Phi_i}. \end{aligned}$$

Again, the last step holds as $\beta_i^* \leq \Phi_i$ and the function is increasing with respect to β_i^* .

With $\varepsilon = 3 - \rho - 1/\rho > 0$ (for $\rho < \hat{\rho}$) we have $\Phi_{i+1} \leq \Phi_i - \varepsilon$ since

$$\begin{aligned} \frac{\rho(\rho - 1)\Phi_i - 1}{1 + \rho\Phi_i} &\leq \Phi_i - 3 + \rho + \frac{1}{\rho} \\ \Leftrightarrow \rho(\rho - 1)\Phi_i - 1 &\leq \Phi_i - 3 + \rho + \frac{1}{\rho} + \rho\Phi_i^2 - 3\rho\Phi_i + \rho^2\Phi_i + \Phi_i \\ \Leftrightarrow \Phi_i^2 + \left(\frac{2 - 2\rho}{\rho}\right)\Phi_i &\geq \frac{2 - \rho - \frac{1}{\rho}}{\rho} \\ \Leftrightarrow \left(\Phi_i + \frac{1 - \rho}{\rho}\right)^2 &\geq \left(\frac{1 - \rho}{\rho}\right)^2 + \frac{2 - \rho - \frac{1}{\rho}}{\rho} = 0. \end{aligned}$$

Thus in both cases we have $\Phi_{i+1} \leq \Phi_i - \varepsilon$ for some $\varepsilon > 0$.

It remains to show

$$\frac{q_{i+1}}{p_{i+1}^*} \leq \frac{\Phi_{i+1}}{\rho - 1} \quad \text{or} \quad \frac{q_{i+1}}{p_{i+1}^*} < \frac{1}{\rho}.$$

We have $\frac{q_{i+1}}{p_{i+1}^*} = \frac{\beta_i^*}{1 + \beta_i^*}(1 - \alpha_{i+1}^*)$ by (3). If $\beta_i^* < \frac{1}{\rho - 1}$ we have

$$\frac{q_{i+1}}{p_{i+1}^*} = \frac{\beta_i^*}{1 + \beta_i^*}(1 - \alpha_{i+1}^*) \leq \frac{\beta_i^*}{1 + \beta_i^*} < \frac{1}{\rho - 1} \cdot \frac{1}{1 + \frac{1}{\rho - 1}} = \frac{1}{\rho - 1} \cdot \frac{1}{\frac{\rho}{\rho - 1}} = \frac{1}{\rho}.$$

Otherwise, we have $\beta_i^* \geq \frac{1}{\rho - 1}$ and thus

$$\begin{aligned} \frac{q_{i+1}}{p_{i+1}^*} &= \frac{\beta_i^*}{1 + \beta_i^*}(1 - \alpha_{i+1}^*) \leq \frac{\Phi_{i+1}}{\rho - 1} \\ \Leftrightarrow (\rho - 1)\beta_i^* &\leq \gamma_i^* + 2(\rho - 1)\beta_i^* - 1 + \underbrace{(\rho - 2)(1 + \beta_i^*) \frac{\alpha_{i+1}^*}{1 - \alpha_{i+1}^*}}_{\geq 0} \\ \Leftrightarrow 1 &\leq \gamma_i^* + (\rho - 1)\beta_i^* \\ \Leftrightarrow \beta_i^* &\geq \frac{1}{\rho - 1}. \quad \square \end{aligned}$$

Lemma 14 *If $\Phi_i \leq \rho - 2 + (\rho - 1)\hat{c}$ and $q_{i+1} = \alpha'_i p_i^*$ then*

$$\begin{aligned} \Phi_{i+1} &\leq \Phi_i - \varepsilon \quad \text{for some } \varepsilon > 0, \text{ and} \\ \frac{q_{i+1}}{p_{i+1}^*} &< \frac{1}{\rho}. \end{aligned}$$

Proof In this case we can assume $\beta_i^* = 0$ (by Lemma 7, condition 2) and hereby have $\Phi_i = \gamma_i^*$. Thus by Lemma 11 and with $(\rho - 1)\alpha'_i \leq \gamma_i^* = \Phi_i$ by Lemma 10 we have

$$\begin{aligned} \Phi_{i+1} &= (\Phi_i + (\rho - 1)\alpha'_i - 1)(1 - \alpha_{i+1}^*) + (\rho - 2)\alpha_{i+1}^* \\ &\leq (2\Phi_i - 1)(1 - \alpha_{i+1}^*) + (\rho - 2)\alpha_{i+1}^*. \end{aligned}$$

We consider the derivative with respect to α_{i+1}^* and with $\Phi_i \leq \rho - 2 + (\rho - 1)\hat{c}$ we get

$$\begin{aligned} \frac{\partial}{\partial \alpha_{i+1}^*} ((2\Phi_i - 1)(1 - \alpha_{i+1}^*) + (\rho - 2)\alpha_{i+1}^*) &= \rho - 1 - 2\Phi_i \\ &\geq 3 - \rho - 2(\rho - 1)\hat{c} > 0. \end{aligned}$$

Thus Φ_{i+1} increases with α_{i+1}^* and since by Lemma 9

$$\alpha_{i+1}^* \leq \frac{\gamma_i^* + (\rho - 1)\alpha'_i}{1 + \gamma_i^* + (\rho - 1)\alpha'_i} \leq \frac{2\gamma_i^*}{1 + 2\gamma_i^*} = \frac{2\Phi_i}{1 + 2\Phi_i}$$

we get

$$\begin{aligned} \Phi_{i+1} &\leq (2\Phi_i - 1) \left(1 - \frac{2\Phi_i}{1 + 2\Phi_i} \right) + (\rho - 2) \frac{2\Phi_i}{1 + 2\Phi_i} \\ &= \frac{2(\rho - 1)\Phi_i - 1}{1 + 2\Phi_i}. \end{aligned}$$

With $\varepsilon = \sqrt{2\rho} - \rho + 1/2 > 0$ (for $\rho < \hat{\rho}$) we have $\Phi_{i+1} \leq \Phi_i - \varepsilon$ since

$$\begin{aligned} \frac{2(\rho - 1)\Phi_i - 1}{1 + 2\Phi_i} &\leq \Phi_i - \sqrt{2\rho} + \rho - \frac{1}{2} \\ \Leftrightarrow 2\Phi_i^2 + (2 - 2\sqrt{2\rho})\Phi_i &\geq \sqrt{2\rho} - \rho - \frac{1}{2} \\ \Leftrightarrow \left(\Phi_i + \frac{1 - \sqrt{2\rho}}{2} \right)^2 &\geq \left(\frac{1 - \sqrt{2\rho}}{2} \right)^2 + \frac{\sqrt{2\rho} - \rho - \frac{1}{2}}{2} = 0. \end{aligned}$$

Thus we proved the first part of the lemma.

It remains to show

$$\frac{q_{i+1}}{p_{i+1}^*} < \frac{1}{\rho}.$$

With $\beta_i^* = 0$ we have $p_{i+1}^* = p_i^*/(1 - \alpha_{i+1}^*)$ by (3). Using $\alpha_{i+1}^* \geq \frac{(\rho-1)\alpha'_i}{1+(\rho-1)\alpha'_i}$ by Lemma 8 and $\alpha'_i \leq \alpha_i^* < 1$ (by definition of α_i^* as a fraction of p_i^*) we get

$$\frac{q_{i+1}}{p_{i+1}^*} = \frac{\alpha'_i p_i^*}{p_{i+1}^*} = \alpha'_i (1 - \alpha_{i+1}^*) \leq \frac{\alpha'_i}{1 + (\rho - 1)\alpha'_i} < \frac{1}{\rho}.$$

This finishes the proof of this lemma. □

Finally, we consider the case $q_{i+1} = q_i$. First, we show that the potential definitely decreases if $q_i/p_i^* < 1/\rho$.

Lemma 15 *If $\Phi_i \leq \rho - 2 + (\rho - 1)\hat{c}$ and $\frac{q_i}{p_i^*} < \frac{1}{\rho}$ and $q_{i+1} = q_i$ then*

$$\Phi_{i+1} \leq \Phi_i - \varepsilon \quad \text{for some } \varepsilon > 0, \text{ and}$$

$$\frac{q_{i+1}}{p_{i+1}^*} < \frac{1}{\rho}.$$

Proof The second part is trivial since $q_{i+1} = q_i$, $p_{i+1}^* \geq p_i^*$ and $q_i/p_i^* < 1/\rho$.

To show the first part we assume $\alpha'_i = 0$ and $\beta_i^* = 0$ according to conditions 1 and 2 of Lemma 7. Thus we have $\Phi_i = \gamma_i^*$ and with Lemma 11 we get

$$\Phi_{i+1} = \left(\Phi_i + (\rho - 1) \frac{q_i}{p_i^*} - 1 \right) (1 - \alpha_{i+1}^*) + (\rho - 2)\alpha_{i+1}^*.$$

We consider the derivative with respect to α_{i+1}^* and get

$$\frac{\partial}{\partial \alpha_{i+1}^*}(\Phi_{i+1}) = \rho - 1 - \Phi_i - (\rho - 1)\frac{q_i}{p_i^*} > 1 - (\rho - 1)\hat{c} - \frac{\rho - 1}{\rho} > 0.$$

Thus Φ_{i+1} increases with α_{i+1}^* and since by Lemma 9

$$\alpha_{i+1}^* \leq \frac{\Phi_i + (\rho - 1)\frac{q_i}{p_i^*}}{1 + \Phi_i + (\rho - 1)\frac{q_i}{p_i^*}}$$

we get

$$\begin{aligned} \Phi_{i+1} &\leq \frac{(\rho - 1)\Phi_i + (\rho - 1)^2\frac{q_i}{p_i^*} - 1}{1 + \Phi_i + (\rho - 1)\frac{q_i}{p_i^*}} \\ &< \frac{(\rho - 1)\Phi_i + \frac{(\rho - 1)^2}{\rho} - 1}{\Phi_i + \frac{2\rho - 1}{\rho}} \quad \text{as } \frac{q_i}{p_i^*} < \frac{1}{\rho} \\ &= \frac{\rho(\rho - 1)\Phi_i + \rho^2 - 3\rho + 1}{\rho\Phi_i + 2\rho - 1}. \end{aligned}$$

With

$$\varepsilon = \frac{3\rho - \rho^2 - 1}{\rho\Phi_i + 2\rho - 1} \geq \frac{3\rho - \rho^2 - 1}{\rho(\rho + (\rho - 1)\hat{c}) - 1} > 0,$$

as $\Phi_i \leq \rho - 2 + (\rho - 1)\hat{c}$ and $3\rho - \rho^2 - 1 > 0$ for $\rho < \hat{\rho}$, we have $\Phi_{i+1} \leq \Phi_i - \varepsilon$ since

$$\begin{aligned} \frac{\rho(\rho - 1)\Phi_i + \rho^2 - 3\rho + 1}{\rho\Phi_i + 2\rho - 1} &\leq \Phi_i - \frac{3\rho - \rho^2 - 1}{\rho\Phi_i + 2\rho - 1} \\ \Leftrightarrow \rho(\rho - 1)\Phi_i + \rho^2 - 3\rho + 1 &\leq \rho\Phi_i^2 + (2\rho - 1)\Phi_i - 3\rho + \rho^2 + 1 \\ \Leftrightarrow 0 &\leq \rho\Phi_i^2 + (3\rho - \rho^2 - 1)\Phi_i \end{aligned}$$

which is satisfied as $3\rho - \rho^2 - 1 > 0$ for $\rho < \hat{\rho}$. □

If we do not have $q_i/p_i^* < 1/\rho$ we can still assume that $q_i/p_i^* \leq (\rho - 2 + (\rho - 1)\hat{c})/(\rho - 1)$ by Lemmas 13 and 14 (as $\Phi_i \leq \rho - 2 + (\rho - 1)\hat{c}$ and this ratio does not increase in case $q_{i+1} = q_i$).

We use this bound to show that either the potential still decreases or we can bound the potential by $\rho - 2$ and the q_{i+1}/p_{i+1}^* ratio is less than $1/\rho$. So from this point on we remain in the case of the previous lemma and the potential decreases by a constant in every step.

Lemma 16 If $\Phi_i \leq \rho - 2 + (\rho - 1)\hat{c}$ and $\frac{q_i}{p_i^*} \leq \frac{\rho - 2 + (\rho - 1)\hat{c}}{\rho - 1}$ and $q_{i+1} = q_i$ then either

$$\Phi_{i+1} < \rho - 2 \quad \text{and}$$

$$\frac{q_{i+1}}{p_{i+1}^*} < \frac{1}{\rho},$$

or

$$\Phi_{i+1} \leq \Phi_i - \varepsilon \quad \text{for some } \varepsilon > 0, \text{ and}$$

$$\frac{q_{i+1}}{p_{i+1}^*} \leq \frac{\rho - 2 + (\rho - 1)\hat{c}}{\rho - 1}.$$

Proof As in Lemma 15 we have $\alpha'_i = 0$, $\beta_i^* = 0$, $\Phi_i = \gamma_i^*$ and

$$\Phi_{i+1} = \left(\Phi_i + (\rho - 1) \frac{q_i}{p_i^*} - 1 \right) (1 - \alpha_{i+1}^*) + (\rho - 2) \alpha_{i+1}^*.$$

Again, Φ_{i+1} increases with α_{i+1}^* . We distinguish two cases according to the value of α_{i+1}^* .

If $\alpha_{i+1}^* > 1 - \frac{\rho - 1}{\rho(\rho - 2 + (\rho - 1)\hat{c})}$ then

$$\frac{q_{i+1}}{p_{i+1}^*} = \frac{q_i}{p_i^*} (1 - \alpha_{i+1}^*) < \frac{\rho - 2 + (\rho - 1)\hat{c}}{\rho - 1} \cdot \frac{\rho - 1}{\rho(\rho - 2 + (\rho - 1)\hat{c})} = \frac{1}{\rho}.$$

As Φ_{i+1} is increasing with α_{i+1}^* we use Lemma 9 to get

$$\begin{aligned} \Phi_{i+1} &\leq \frac{(\rho - 1)\Phi_i + (\rho - 1)^2 \frac{q_i}{p_i^*} - 1}{1 + \Phi_i + (\rho - 1) \frac{q_i}{p_i^*}} \\ &\leq \frac{2(\rho - 1)(\rho - 2 + (\rho - 1)\hat{c}) - 1}{1 + 2(\rho - 2 + (\rho - 1)\hat{c})} \end{aligned}$$

as $\Phi_i \leq \rho - 2 + (\rho - 1)\hat{c}$ and $\frac{q_i}{p_i^*} \leq \frac{\rho - 2 + (\rho - 1)\hat{c}}{\rho - 1}$. We have

$$\begin{aligned} \frac{2(\rho - 1)(\rho - 2 + (\rho - 1)\hat{c}) - 1}{1 + 2(\rho - 2 + (\rho - 1)\hat{c})} &< \rho - 2 \\ \Leftrightarrow 2(\rho - 1)(\rho - 2 + (\rho - 1)\hat{c}) - 1 &< \rho - 2 + 2(\rho - 2)(\rho - 2 + (\rho - 1)\hat{c}) \\ \Leftrightarrow 2(\rho - 2 + (\rho - 1)\hat{c}) &< \rho - 1 \\ \Leftrightarrow 2(\rho - 1)\hat{c} &< 3 - \rho, \end{aligned}$$

which holds for $\rho < \hat{\rho}$. We showed that if $\alpha_{i+1}^* > 1 - \frac{\rho - 1}{\rho(\rho - 2 + (\rho - 1)\hat{c})}$, then we have $\Phi_{i+1} < \rho - 2$ and $q_{i+1}/p_{i+1}^* < 1/\rho$.

Otherwise, we have $\alpha_{i+1}^* \leq 1 - \frac{\rho-1}{\rho(\rho-2+(\rho-1)\hat{c})}$ and get

$$\begin{aligned} \Phi_{i+1} &\leq \frac{(\Phi_i + (\rho - 1)\frac{q_i}{p_i^*} - 1)(\rho - 1)}{\rho(\rho - 2 + (\rho - 1)\hat{c})} + \rho - 2 - \frac{(\rho - 2)(\rho - 1)}{\rho(\rho - 2 + (\rho - 1)\hat{c})} \\ &\leq \frac{(\Phi_i + (\rho - 1)\hat{c} - 1)(\rho - 1)}{\rho(\rho - 2 + (\rho - 1)\hat{c})} + \rho - 2 \quad \text{as } \frac{q_i}{p_i^*} \leq \frac{\rho - 2 + (\rho - 1)\hat{c}}{\rho - 1}. \end{aligned}$$

With $\varepsilon = \Phi_i + \frac{(1-(\rho-1)\hat{c}-\Phi_i)(\rho-1)}{\rho(\rho-2+(\rho-1)\hat{c})} - \rho + 2$ we have $\Phi_{i+1} \leq \Phi_i - \varepsilon$ since

$$\begin{aligned} &\frac{(\Phi_i + (\rho - 1)\hat{c} - 1)(\rho - 1)}{\rho(\rho - 2 + (\rho - 1)\hat{c})} + \rho - 2 \\ &\leq \Phi_i - \Phi_i - \frac{(1 - (\rho - 1)\hat{c} - \Phi_i)(\rho - 1)}{\rho(\rho - 2 + (\rho - 1)\hat{c})} + \rho - 2. \end{aligned}$$

It remains to show that $\varepsilon > 0$. To see this, observe that ε is increasing with respect to Φ_i as

$$\frac{\partial}{\partial \Phi_i} \left(\Phi_i + \frac{(1 - (\rho - 1)\hat{c} - \Phi_i)(\rho - 1)}{\rho(\rho - 2 + (\rho - 1)\hat{c})} - \rho + 2 \right) = 1 - \frac{\rho - 1}{\rho(\rho - 2 + (\rho - 1)\hat{c})} > 0$$

for $2.55 \leq \rho < \hat{\rho}$ (here we assume $\delta = \hat{\rho} - \rho$ is sufficiently small). As $\Phi_i \geq 0$ we have

$$\varepsilon \geq \frac{(1 - (\rho - 1)\hat{c})(\rho - 1)}{\rho(\rho - 2 + (\rho - 1)\hat{c})} - \rho + 2 > 0.$$

Of course,

$$\frac{q_{i+1}}{p_{i+1}^*} \leq \frac{q_i}{p_i^*} \leq \frac{\rho - 2 + (\rho - 1)\hat{c}}{\rho - 1}$$

holds trivially. We showed that if $\alpha_{i+1}^* \leq 1 - \frac{\rho-1}{\rho(\rho-2+(\rho-1)\hat{c})}$, then we have $\Phi_{i+1} \leq \Phi_i - \varepsilon$ for $\varepsilon > 0$ and $q_{i+1}/p_{i+1}^* \leq (\rho - 2 + (\rho - 1)\hat{c})/(\rho - 1)$. This finishes the proof of this lemma. \square

This ends this extensive induction. Let us summarize the complete induction and show that it actually gives the desired contradiction.

Recall that our induction hypothesis in Lemma 12 states that $\Phi_1 \leq \rho - 2 + (\rho - 1)\hat{c}$ and $q_1/p_1^* < 1/\rho$.

First assume that whenever we need to apply Lemma 16, then the second condition holds, i.e., $\Phi_{i+1} \leq \Phi_i - \varepsilon$ and $q_{i+1}/p_{i+1}^* \leq (\rho - 2 + (\rho - 1)\hat{c})/(\rho - 1)$. Then Lemma 13 (for $q_{i+1} = \beta_i^* p_i^*$), Lemma 14 (for $q_{i+1} = \alpha'_i p_i^*$) and Lemmas 15 and 16 (for $q_{i+1} = q_i$) show that the potential decreases by a constant in every step.

Now if Lemma 16 is applied and the second condition does not hold, then $\Phi_{i+1} < \rho - 2$ and $q_{i+1}/p_{i+1}^* < 1/\rho$. Thus Lemma 13 (for $q_{i+1} = \beta_i^* p_i^*$), Lemma 14 (for $q_{i+1} = \alpha'_i p_i^*$) and Lemmas 15 (for $q_{i+1} = q_i$) show that the q_i/p_i^* ratio remains less

than $1/\rho$. So the precondition for Lemma 15 is always satisfied if $q_{i+1} = q_i$ and we do not need to apply Lemma 16 anymore. Thus from this point on, the potential decreases by a constant in every further step.

Summarizing, we derived a contradiction to $\Phi_i \geq 0$ for all $i \geq 1$, thereby proving Theorem 1.

References

1. Brown, D.J., Baker, B.S., Katseff, H.P.: Lower bounds for online two-dimensional packing algorithms. *Acta Inform.* **18**, 207–225 (1982)
2. Baker, B.S., Coffman, E.G., Rivest, R.L.: Orthogonal packings in two dimensions. *SIAM J. Comput.* **9**(4), 846–855 (1980)
3. Kenyon, C., Remila, E.: Approximate strip packing. In: Proc. of 37th FOCS, pp. 31–36 (1996)
4. Ye, D., Han, X., Zhang, G.: A note on online strip packing. *J. Comb. Optim.* **17**(4), 417–423 (2009)
5. Hurink, J., Paulus, J.: Online algorithm for parallel job scheduling and strip packing. In: WAOA: Proc. of 5th Workshop on Approximation and Online Algorithms, pp. 67–74 (2007)
6. Baker, B.S., Schwarz, J.S.: Shelf algorithms for two-dimensional packing problems. *SIAM J. Comput.* **12**(3), 508–525 (1983)
7. Csirik, J., Woeginger, G.J.: Shelf algorithm for online strip packing. *Inf. Process. Lett.* **63**, 171–175 (1997)
8. Han, X., Iwama, K., Ye, D., Zhang, G.: Strip packing vs. bin packing. In: Proc. 3rd International Conference on Algorithmic Aspects in Information and Management (AAIM). Lecture Notes in Computer Science, vol. 4508, pp. 358–367. Springer, Berlin (2007)
9. Johannes, B.: Scheduling parallel jobs to minimize the makespan. *J. Sched.* **9**(5), 433–452 (2006)
10. Hurink, J., Paulus, J.: Online scheduling of parallel jobs on two machines is 2-competitive. *Oper. Res. Lett.* **36**(1), 51–56 (2008)
11. Kern, W., Paulus, J.: A tight analysis of Brown-Baker-Katseff sequences for online strip packing. *J. Comb. Optim.* (2012). doi:[10.1007/s10878-012-9463-1](https://doi.org/10.1007/s10878-012-9463-1)
12. Harren, R., Kern, W.: Improved lower bound for online strip packing. In: WAOA 2011 Proceedings (2011)
13. Fuchs, B., Hochstaettler, W., Kern, W.: Online matching on a line. *Theor. Comput. Sci.* **332**(1–3), 251–264 (2005)