Keep Your Distance: Land Division With Separation

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Abstract
This paper is part of an ongoing endeavor to bring the theory of fair division closer to practice by handling requirements from real-life applications. We focus on two requirements originating from the division of land estates: (1) each agent should receive a plot of a usable geometric shape, and (2) plots of different agents must be physically separated. With these requirements, the classic fairness notion of proportionality is impractical, since it may be impossible to attain any multiplicative approximation of it. In contrast, the ordinal maximin share approximation, introduced by Budish in 2011, provides meaningful fairness guarantees. We prove upper and lower bounds on achievable maximin share guarantees when the usable shapes are squares, fat rectangles, or arbitrary axes-aligned rectangles, and explore the algorithmic and query complexity of finding fair partitions in this setting.

1 Introduction
The problem of fairly allocating a divisible resource has a long history, dating back to the seminal article of Polish mathematician Hugo Steinhaus [1948]. In its basic formulation, the resource, which is metaphorically viewed as a cake, comes in the form of an interval. The aim is to find a division satisfying some fairness criteria, e.g., proportionality, which means that if there are $n$ agents, the value that each of them receives should be at least $1/n$ of the entire cake. Not only does a proportional allocation always exist, but it can also be found efficiently [Dubins and Spanier, 1961].

While the interval cake is simple and consequently useful as a starting point, it is often insufficient for modeling real-world applications, especially when combined with the common requirement that each agent should receive a connected piece of the cake.¹ In particular, when allocating real estate, geometric considerations play a crucial role: it is hard to build a house or raise cattle on a thin or highly zigzagged piece of land even if its total area is large. Such considerations have motivated researchers to study fairness in land division, which also serves to model the allocation of other two-dimensional objects such as advertising spaces [Berliant et al., 1992; Ichishi and Idzik, 1999; Berliant and Dunz, 2004; Dall’Aglio and Maccheroni, 2009; Iyer and Huhns, 2009; Devulapalli, 2014; Segal-Halevi et al., 2017; Segal-Halevi et al., 2020]. These studies have uncovered important differences between land division and interval division: for instance, when agents must be allocated square pieces, Segal-Halevi et al. [2017] show that we cannot guarantee the agents more than $1/(2n)$ of their entire value in the worst case, even when the agents have identical valuations over the land.

A related issue, which frequently arises in practice, is that agents’ pieces may have to be separated from one another: we may need to leave a space between adjacent pieces of land, e.g., to prevent dispute between owners, provide access to the plots, avoid cross-fertilization of different crops, or ensure safe social distancing among vendors in a market. The formal study of fair division with separation constraints was initiated by Elkind et al. [2021c], who focus on the one-dimensional setting. The goal of our work is to extend this analysis to two dimensions, i.e., to analyze fair division of land under separation constraints.

1.1 Our Contribution
We assume that each agent must obtain a contiguous piece of land, and the shares that any two agents receive must be separated by a distance of at least $s$, where $s$ is a given parameter that is independent of the land value. In the presence of separation constraints, no multiplicative approximation of proportionality can be guaranteed, even in one dimension: when all of the agents’ values are concentrated within distance $s$, only one agent can receive a positive utility. Elkind et al. [2021c] therefore consider the well-known criterion of maximin share fairness [Budish, 2011; Kurokawa et al., 2018]—the value that each agent receives must be at least her $1$-out-of-$n$ maximin share, i.e., the best share that she can guarantee for herself by dividing the resource into $n$ bundles and accepting the worst one. Elkind et al. [2021c] show that this criterion can be satisfied for an interval cake, while an ordinal approximation of it can be attained for a one-dimensional circular cake.

We establish that maximin share fairness and relaxations

¹As Stromquist [1980] memorably wrote, without such connectivity requirements, there is a danger that agents will receive a “countable union of crumbs”.

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thereof can also provide worst-case guarantees in land allocation with separation. Moreover, since full proportionality cannot always be attained in this setting even in the case of no separation \((s = 0)\), our results have interesting implications for that case as well.

Our first result is negative: we prove that, when \(s > 0\), it is impossible to guarantee to each agent a positive fraction of her 1-out-of-\(n\) maximin share. Therefore, in the rest of the paper, we focus on an ordinal notion of approximation. Specifically, we ask for the smallest value of \(k \geq n\) such that we can guarantee each agent her 1-out-of-\(k\) maximin share.

We assume that both the land to be divided and each agent’s piece are axes-aligned rectangles. If we additionally require that all rectangles (both in agents’ maximin partitions and in the final allocation) are \(r\)-fat, i.e., the ratio of the length of the longer side to the length of the shorter side is bounded by \(\frac{3}{2}\); in particular, if all land pieces are required to be squares \((r = 1)\), we obtain \(k = 4n - 5\). Without the fatness assumption, the problem is more difficult, and the technique we use for fat rectangles no longer works. However, we still obtain a finite approximation: we show that it suffices to set \(k = 2n + 2\), and provide stronger bounds for small values of \(n\). In particular, for \(n = 2\) we can set \(k = 3\), which is optimal.

Our positive results are constructive, in the sense that, given each agent’s 1-out-of-\(k\) maximin partition (i.e., a partition into \(k\) pieces where the value of each piece is at least the agent’s maximin share), we can divide the land among the agents so that each agent gets her 1-out-of-\(k\) share, using a natural adaptation of the standard Robertson–Webb model [Robertson and Webb, 1998]. However, it is not clear how a 1-out-of-\(k\) maximin partition can be efficiently computed or even approximated. To circumvent this difficulty, we focus on a special class of land partitions known as guillotine partitions; intuitively, these are partitions that can be obtained by a sequence of edge-to-edge cuts. We show that we can efficiently compute an approximately optimal guillotine partition, and that the loss caused by using guillotine partitions can be bounded. Combining these results with our ordinal approximation algorithms, we obtain approximation algorithms for computing a maximin allocation.

1.2 Related Work

In considering fair division with separation, we build on the work of Elkind et al. [2021c], who investigate the one-dimensional variant of this problem. Fair land division with constraints on the shape of usable pieces has been previously studied [Segal-Halevi et al., 2017; Segal-Halevi et al., 2020]. We follow these works in considering fat rectangles and guillotine cuts; however, the fairness notions considered in these papers are (partial) proportionality and envy-freeness, whereas our work concerns maximin fairness.

Our analysis is also somewhat similar in spirit to several recent works on dividing a cake represented by a general graph, which generalizes both the interval and the cycle (a.k.a. pie) setting. Several fairness notions have been studied in this setting: partial proportionality [Bei and Suksompong, 2021], envy-freeness [Igarashi and Zwicker, 2021], and maximin share fairness [Elkind et al., 2021a]. In all of these works, the cake is still one-dimensional—it is a union of a finite number of intervals. As we show in this work, a two-dimensional cake is fundamentally different.

2 Preliminaries

The land is given by a closed, bounded, and connected subset \(L\) of the two-dimensional Euclidean plane \(\mathbb{R}^2\). The land is to be divided among a set of agents \(N = [n]\), where \([k] := \{1, 2, \ldots, k\}\) for any positive integer \(k\). There is a prespecified family \(U\) of usable pieces. Each agent has an integrable density function \(f_i : L \to \mathbb{R}_{\geq 0}\); agent \(i\)’s value for a piece of land \(Z\) is given by \(v_i(Z) := \int_Z f_i(x, y) \, dx \, dy\).

Let \(s \geq 0\) be the separation parameter. An allocation of the land is given by a vector \(\mathbf{A} = (A_1, \ldots, A_n)\), where each \(A_i\) is a single (connected) piece of land allocated to agent \(i\). We require allocations to be \(s\)-separated, i.e., any two pieces \(A_i\) and \(A_j\) are separated by distance at least \(s\), where distance is measured according to the \(\ell_\infty\) norm:

\[
d(A_i, A_j) = \inf_{(x, y) \in A_i, (x', y') \in A_j} \max\{|x - x'|, |y - y'|\}.
\]

Partitions and \(s\)-separated partitions are defined similarly, except that instead of a vector \(\mathbf{A} = (A_1, \ldots, A_n)\), we have a set \(P = \{P_1, \ldots, P_n\}\). Denote by \(\Gamma_n(s)\) the set of all \(s\)-separated partitions. An instance consists of the land, agents, density functions, and the separation parameter.

**Definition 2.1.** The 1-out-of-\(k\) maximin share of agent \(i\) is defined as \(\text{MMS}^{i,k}_n := \sup_{P \in \Gamma_n(s)} \min_{j \in [k]} v_i(P_j)\). We omit \(s\) if it is clear from the context, and write \(\text{MMS}^n_k\) instead of \(\text{MMS}^{i,k}_n\). We refer to \(\text{MMS}^n_k\) as \(i\)’s maximin share.

As with cake cutting [Elkind et al., 2021c], the supremum in Definition 2.1 can be replaced by a maximum. This requires defining a metric on the usable pieces and showing that \(U\) is compact in that metric space—see Appendix C in the work of Segal-Halevi et al. [2017]. An \(s\)-separated partition for which this maximum is attained is called a maximin partition of agent \(i\).

All omitted proofs can be found in the full version of our paper [Elkind et al., 2021b].

3 A General Impossibility Result

We first show that, in contrast to one-dimensional cake cutting with separation [Elkind et al., 2021c], for land division there may be no allocation that guarantees to all agents their maximin share or even any multiplicative approximation of it. This negative result does not depend on the geometric shape of the land or the pieces.\(^2\)

**Proposition 3.1.** For every family \(U\) of usable pieces, integer \(n \geq 2\), separation parameter \(s > 0\), and real number \(r > 0\), there exists a land division instance with \(n\) agents in which no \(s\)-separated allocation gives every agent \(i\) a value of at least \(r \cdot \text{MMS}_i^n\).

**Proof sketch.** We construct \(n\) sets \(S_1, \ldots, S_n\) consisting of \(n\) points each such that the distance between any two points in

\(^2\)We are grateful to Alex Ravsky for the proof idea.
the same set is greater than 1, but if we pick one representative point from each set, then some two representatives are at distance less than 1 apart. We then construct agents’ density functions so that each agent only values a small ‘pool’ of land around each of the points in their set, assigning a value of 1 to each such pool. By scaling this construction by approximately a factor of s, we can ensure that each agent’s maximin share is 1, but there are at most n − 1 agents who can obtain a positive utility in an s-separated partition.

4 Ordinal Approximation

Since no multiplicative approximation of the maximin share can be guaranteed, we instead consider an ordinal notion of approximation. That is, we ask if each agent can be guaranteed her 1-out-of-k maximin share for some k > n.

While the negative result of Section 3 does not depend on geometric assumptions, our positive results concern pieces that have a ‘nice’ geometric shape. Specifically, we first consider the scenario where U consists of ‘fat’ axes-aligned rectangles, i.e., rectangles whose length-to-width ratio is bounded by a constant (e.g., if this constant is 1, the set U consists of axes-aligned squares). We then consider the more general setting where U consists of all axes-aligned rectangles.

Placing such constraints on the shape of each piece is useful in land allocation settings, particularly in urban regions. Note that when we restrict the shape of the usable pieces to be a (fat) rectangle, in our definition of the maximin share we also only consider s-separated partitions in which each piece is a (fat) rectangle.

4.1 Squares and Fat Rectangles

For a real number r ≥ 1, a rectangle is called r-fat if the ratio of its longer side to its shorter side is at most r [Agarwal et al., 1995; Katz, 1997]. In particular, a 1-fat rectangle is a square.

Given a rectangle R, we denote the lengths of its long and short side by long(R) and short(R), respectively; we refer to short(R) as the width of R. In what follows, we say that two pieces of land overlap if their intersection has a positive area, and that they are disjoint if their intersection has a zero area.

In order to obtain maximin share guarantees, the high-level idea is to find a sufficiently large k such that if we consider the agents’ 1-out-of-k maximin partitions, then it is possible to select a representative piece from each partition in such a way that these representatives are s-separated. This will ensure that, by allocating to each agent her representative, we obtain an allocation that is s-separated and in which agent i receives value at least MMS_{i,s}^k. The following theorem shows that k = (2[\frac{r}{r}] + 2)n − (3[\frac{r}{r}] + 2) suffices; for a square (r = 1), this yields k = 4n − 5. Note that our result does not place any assumptions on the shape of the land.

**Theorem 4.1.** Let r ≥ 1 be a real number. For every land division instance with n agents and separation parameter s, where U is the set of axes-aligned r-fat rectangles, there exists an allocation in which every agent i receives value at least MMS_{i,s}^k, where k := (2[\frac{r}{r}] + 2)n − (3[\frac{r}{r}] + 2).

**Proof.** We first consider the following geometric problem:

There are n ≥ 2 sets, each of which contains N ≥ n axes-aligned r-fat rectangles. The rectangles in each set are pairwise disjoint. We want to choose a single representative rectangle from each set so that the representatives are pairwise disjoint. What is the smallest integer N = NFAT(r, n) for which this is always possible?

We first prove that NFAT(r, 2) ≤ \lfloor r \rfloor + 2. 3 Given \lfloor r \rfloor + 2 red and \lfloor r \rfloor + 2 blue r-fat rectangles, we have to show that at least one red and one blue rectangle do not overlap each other. The proof of this statement is left to the full version of our paper [Elkind et al., 2021b].

Next, we prove that NFAT(r, n + 1) ≤ NFAT(r, n) + 2\lfloor r \rfloor + 2. We claim that in any arrangement of axes-aligned r-fat rectangles, there exists a rectangle that overlaps at most 2\lfloor r \rfloor + 2 pairwise disjoint rectangles. To prove this claim, let Q_{min} be a minimum-width rectangle, and denote its width by w. Mark all four corners of Q_{min}. Moreover, on both of its longer sides, mark \lfloor r \rfloor − 1 additional points so that the distance between any two consecutive marks is at most w. The total number of marks is 2\lfloor r \rfloor + 2. Any rectangle Q that overlaps Q_{min} must contain at least one of its marks. Hence, if there are more than 2\lfloor r \rfloor + 2 rectangles overlapping Q_{min}, some two of them will overlap. This establishes the claim.

Combining this with the base case n = 2, we conclude that, for all n ≥ 2 and r ≥ 1:

\[
NFAT(r, n) \leq (2\lfloor r \rfloor + 2)(n - 2) + (\lfloor r \rfloor + 2) = (2\lfloor r \rfloor + 2)n - (3\lfloor r \rfloor + 2).
\]

For the special case of a square we get N(1, n) ≤ 4n − 5.

![Figure 1: The dark rectangles are s-separated if and only if the light rectangles wrapping them with a rectangle ring of width s/2 are disjoint.](image-url)

We can now return to our original problem. Given the separation parameter s, for every rectangle Q with side lengths u and v, define WRAP(Q, s) to be the rectangle with side lengths u + s and v + s and the same center as Q (i.e., wrap Q with a ‘rectangle ring’ of width s/2); note that WRAP(Q, s)
is $r$-fat whenever $Q$ is. Then, two rectangles $Q_1$ and $Q_2$ are
$s$-separated if and only if $\text{WRAP}(Q_1, s)$ and $\text{WRAP}(Q_2, s)$ do
not overlap (see Figure 1).

Now, given an $n$-agent instance, we ask each agent to pro-
duce a 1-out-of-$k$ maximin partition: this is a set of $k$ axes-
aligned rectangles that are $s$-separated. Then, we replace each
rectangle $Q$ with $\text{WRAP}(Q, s)$, so each agent now has a set
of $k$ non-overlapping rectangles. Since $k \geq \text{NFAT}(r, n)$,
there is a set of representative rectangles, one per agent,
that are pairwise disjoint. Suppose that these rectangles are
$\text{WRAP}(Q_1, s), \text{WRAP}(Q_2, s), \ldots, \text{WRAP}(Q_n, s)$, where
$\text{WRAP}(Q_i, s)$ belongs to agent $i$’s set. We allocate the re-
cangle $Q_1$ to agent $i$. The rectangles $Q_i$ are $s$-separated, and
every agent $i$ receives value at least $\text{MMS}^{k,s}_i$, as desired.

Constructions similar to those in Proposition 3.1 show that
$\text{NFAT}(r, n) \geq n + 1$ for all $r$. Thus the bound $4n - 5$ for
squares is optimal for $n = 2$, but may be suboptimal for $n \geq
3$. Closing the gap between the lower bound $n + 1$ and the
upper bound $4n - 5$ seems to require new geometric insights.

4.2 Arbitrary Rectangles

Next, we allow the pieces to be arbitrary axes-aligned rect-
angles, and assume that the land itself is also an axes-aligned
rectangle. Without loss of generality, we suppose further that
the land is a square (otherwise, for positive results, a rectan-
gular land can be completed to a square by attaching to it a
rectangle that all agents value at 0). We scale the axes so that
the land is the unit square $[0, 1] \times [0, 1]$.

The arbitrary rectangle case differs from the fat rectangle
case in two respects. First, without the separation require-
ment, the arbitrary rectangle case is much easier: the land
can be projected onto a one-dimensional interval, for which
full proportionality, and hence $\text{MMS}^{k}_i$, can be achieved [Du-
bins and Spanier, 1961]. In contrast, with the separation re-
quirement, the arbitrary rectangle case is much harder: the
representative-selection technique of Theorem 4.1 (which has
also been used implicitly, in a simpler form, for cake and pie
division by Elkind et al. [2021c]), does not yield a meaning-
fad bound for arbitrary rectangles. An example similar to the
one in Footnote 3 shows that, even for an arbitrarily large $k$,
there exist two size-$k$ sets of pairwise-disjoint rectangles such
that no two representatives are disjoint.

A priori, for $n \geq 2$, it is not clear that there is a finite
$\text{NRECT}(n)$ such that an $\text{MMS}^{\text{NRECT}(n)}$ allocation among n
agents always exists. Below we prove that $\text{NRECT}(n)$ is in-
deed finite for any $n \geq 2$, and derive improved upper bounds
on $\text{NRECT}(n)$ for small values of $n$. Towards this goal, we
develop some new tools.

In what follows, for each agent we fix a 1-out-of-$k$ max-
imin partition—see Figure 2 for some examples of such par-
titions. For all $i \in N$, we assume without loss of generality
that $\text{MMS}_i = 1$, and that $i$’s value is 0 outside the $k$ rect-
cangles in her maximin partition (the latter value being positive
can only make it easier to satisfy the agent). Hence each agent
has a value of $k$ for the land and should get an axes-aligned
rectangle worth at least 1.

We refer to the $k$ rectangles in the agent’s fixed maximin
partition as $\text{MMS}$-rectangles; every rectangular piece of land
that is worth at least 1 to the agent is called a value-$1$ rect-
angle. Due to our normalization, every $\text{MMS}$-rectangle is a
value-$1$ rectangle, but the converse is not necessarily true.

**Definition 4.2.** Consider an agent with a fixed 1-out-of-$k$
maximin partition, and integers $p, q \geq 1$. A vertical $p$:
rectangle cut is a rectangular strip of height 1 and width $s$
that has at least $p$ whole $\text{MMS}$-rectangles on its left and at
least $q$ whole $\text{MMS}$-rectangles on its right. A vertical $p$-
rectangle stack is a sequence of $p$ rectangles of value 1 such
that each consecutive pair is separated by a vertical distance
of at least $s$.

Horizontal rectangle cuts and stacks are defined similarly.

In Figure 2(a), the left vertical cut (the thick red line) is a
1:2-rectangle cut and the right one is a 2:1-rectangle cut.
In Figure 2(c), there is a vertical 3-rectangle stack. In Fig-
ure 2(d), the vertical cut is a 2:1-rectangle cut, and there is a
vertical 2-rectangle stack.

The following lemma shows the existence of either a rect-
gle cut or a rectangle stack with appropriate parameters.

**Lemma 4.3.** Fix an agent and a 1-out-of-$k$ maximin partition
of this agent. For any integers $1 \leq p, q \leq k$ with $p + q \leq
k + 1$, the agent has a vertical $p$:$q$-rectangle cut or a vertical
$(k - p - q + 2)$-rectangle stack.

**Proof.** Starting from the left end of the cake, move a vertical
knife of width $s$ to the right. Stop the knife at the first point
where there are at least $p$ whole $\text{MMS}$-rectangles to its left—
the knife may need to move outside the cake in order for this
to happen, as in Figure 2(c) for any $p$. Consider two cases.

**Case 1:** There are at least $q$ whole $\text{MMS}$-rectangles to the
right of the knife. Then, the knife indicates a vertical $p$:$q$-
rectangle cut. This is the case when $p = q = 1$ in Figures
2(a), (b), and (d).

**Case 2:** There are at most $q - 1$ whole $\text{MMS}$-rectangles
to the right of the knife. Then, by moving the knife slightly
to the left, we obtain a cut for which there are at most $p - 1$
$\text{MMS}$-rectangles entirely to its left, and at most $q - 1$ $\text{MMS}$-
rectangles entirely to its right. Therefore, at least $k - p - q + 2$
$\text{MMS}$-rectangles must intersect the knife itself. Since
the knife width is $s$, these rectangles must lie in order vertically,
with a vertical distance of at least $s$ between consecutive rect-
cangles. Hence, they form a vertical $(k - p - q + 2)$-rectangle
stack. This is the case when $p = q = 1$ in Figure 2(c).

In the remainder of this section, given $y, y' \in [0, 1]$ with
$y \leq y'$, we write $R(y, y') := [0, 1] \times [y, y']$.

We now prove a positive result for two agents matching the
lower bound implied by Proposition 3.1.

**Theorem 4.4.** For any land division instance with a rect-
angular land and $n = 2$ agents, there exists an allocation
in which each agent $i$ receives an axes-aligned rectangle of
value at least $\text{MMS}^3_i$.

**Proof.** Call the agents Alice and Bob. Take a 1-out-of-3 max-
imin partition of each agent, and consider two cases.

**Case 1:** Both agents have a vertical 1:1-rectangle cut. Assu-
me without loss of generality that Alice’s cut lies further to
the left; give the rectangle to its left to Alice and the one to its
right to Bob. Then each agent receives an $\text{MMS}$-rectangle.
Case 2: At least one agent, say Alice, has no vertical 1:1-rectangle cut. By Lemma 4.3, she has a vertical 3-rectangle stack, as in Figure 2(c). For the $i$-th rectangle in this stack (counting from the bottom), denote the $y$-coordinates of its top and bottom sides by $t_i$ and $b_i$, respectively. Note that $t_i + s \leq b_2$ and $t_2 + s \leq b_3$.

If Bob’s value for $R(0, t_2)$ is at least 1, then give $R(0, t_2)$ to Bob and $R(b_3, 1)$ to Alice. Otherwise, Bob values $R(0, t_2)$ less than 1, so his value for $R(b_2, 1)$ is more than 2. Give $R(b_2, 1)$ to Bob and $R(0, t_1)$ to Alice. In both cases Alice’s value is 1 and the pieces are $s$-separated.

For $n \geq 3$ agents, the analysis becomes more complicated. As in classic cake-cutting algorithms (e.g., [Dubins and Spanier, 1961]), we would like to proceed recursively: give one agent a rectangle worth at least 1, and divide the rest of the land among the remaining $n - 1$ agents. In particular, for $n = 3$, after allocating a piece to one agent, we would need to show that, for each of the remaining two agents, the rest of the land is worth at least 3, so that we can apply Theorem 4.4. In fact, to apply Theorem 4.4, we need an even stronger condition: each agent should have three $s$-separated rectangles of value 1. However, the recursion step might yield a remainder land made of many pieces of such rectangles, each of which is worth less than 1. We therefore need to adapt our definitions and lemmas accordingly.

Definition 4.5. A vertical $p$-$q$-value cut of an agent is a rectangular strip of width $s$ such that the agent values the land on its left at least $p$ and the land on its right at least $q$.

For any integers $p$, $q$, every $p$-$q$-rectangle cut is also a $p$-$q$-value cut, but the converse is not necessarily true.

For the following lemma, it is important that the agent’s value function is normalized as explained earlier, i.e., the value of each MMS-rectangle is 1 and the value outside the MMS-rectangles is 0. A land-subset is a subset of the land after some pieces have possibly been allocated to other agents.

Lemma 4.6. Consider an agent with a fixed 1-out-of-$k$ maximin partition of the land, which takes part in a division of a rectangular land-subset. Let $V \leq k$ be the agent’s value for the land-subset. For any integers $p, q \geq 1$ with $p + q \leq V$, the agent has either a vertical $p$-$q$-value cut or a vertical $[([V] - p - q)/2]$-rectangle stack.

The following lemma establishes a weaker bound than Theorem 4.4 does; however, it applies to an arbitrary land-subset, and hence (unlike Theorem 4.4) can be used as part of a recursive argument. Its proof is essentially identical to the proof of Theorem 4.4: the only difference is that we first look for a vertical 1:1-value cut, and if we fail to find one, we invoke Lemma 4.6 to establish the existence of a vertical 3-rectangle stack.

Lemma 4.7. Consider a rectangular land-subset and $n = 2$ agents who value it at least 7 each. There is an allocation in which each agent receives an axes-aligned rectangle of value at least 1.

Proof. We consider two cases.

Case 1: Both agents have a vertical 1:1-value cut. Take the cut to the left, give the rectangle to its left to the cutter, and the rectangle to its right to the other agent.

Case 2: At least one agent, say Alice, has no vertical 1:1-value cut. By Lemma 4.6, she has a vertical 3-rectangle stack, so we can proceed as in Case 2 of Theorem 4.4.

Let $V_{req}(n)$ be the smallest value of $V$ such that if each of $n$ agents values the land-subset at $V$ or higher, then there is an allocation of this land-subset in which each agent’s value for her share is at least 1. Obviously $V_{req}(1) = 1$, and by Lemma 4.7 we know that $V_{req}(2) \leq 7$. We can now provide a finite (exponential) MMS approximation for every positive $n$.

Theorem 4.8. For any $n \geq 1$, given any land division instance with a rectangular land and $n$ agents, there exists an allocation in which each agent $i$ receives an axes-aligned rectangle with value at least $MMS_i^{k}$, where $k = 2n + 2$.

By adjusting the argument in the proof of Theorem 4.8, we can obtain stronger bounds for $n = 3$ and $n = 4$; in particular, we can guarantee each agent $i$ a piece of value at least $MMS_i^{14}$ and $MMS_i^{22}$, respectively. The details can be found in the full version of our paper [Elkind et al., 2021b].

5 Computing Maximin Allocations

The results in Section 4 are stated in terms of approximation guarantees. To convert them into algorithms, we need to formally define our computational model. To do so, we propose a natural modification of the classic Robertson–Webb model [Robertson and Webb, 1998] for the two-dimensional setting.

Consider an axes-aligned rectangle $L = [a_0, a_1] \times [b_0, b_1]$, which may be part of a larger land-subset. We adapt the $\text{CUT}$ and $\text{EVAL}$ queries of the Robertson–Webb model to allow for horizontal and vertical cuts as follows. The $\text{CUT}((L, \delta))$ query returns a value $a$ such that agent $i$ values the rectangle
an optimal

Our algorithm proceeds by discretizing the land and finding a value \( b \) such that agent \( i \) values the rectangle \([a_0, a_1] \times [b_0, b_1]\) at \( b \); we assume that this query returns \( a_1 \) (respectively, \( b_1 \)) if the agent values the entire rectangle less than \( b \). Similarly, the \( \text{EVAL}_i([-L, L, \delta]) \) query with \( a_0 \leq a \leq a_1 \) returns the value that \( i \) assigns to the rectangle \([a_0, a_1] \times [b_0, b_1]\), whereas \( \text{EVAL}_i([-L, b, \delta]) \) query with \( b_0 \leq b \leq b_1 \) returns the value that \( i \) assigns to the rectangle \([a_0, a_1] \times [b_0, b] \).

We can now revisit the proofs of Theorems 4.4 and 4.8 and check if they can be converted into algorithms that use \( \text{CUT} \) and \( \text{EVAL} \) queries. One can see that these proofs are constructive and their basic steps can be expressed in terms of these queries: a \( pq \)-value cut can be implemented by two \( \text{CUT} \) queries, and agents’ values for rectangles of the form \( R(x, y) \) can be determined using \( \text{EVAL} \) queries.

However, these algorithms use the agents’ 1-out-of-\( k \) maximin partitions as their starting points, and it is not clear if such partitions are efficiently computable. Indeed, even in the 1-dimensional case, there is no algorithm that always computes a maximin partition of an agent using finitely many queries, and the best known solution is a \((1 - \varepsilon)\) approximation in time \( O(n \log(1/\varepsilon)) \) [Elkind et al., 2021c]. For the 2-dimensional case, even a \((1 - \varepsilon)\) approximation seems challenging.

To circumvent this difficulty, we focus on maximin partitions with a special structure, namely, guillotine partitions [Gonzalez et al., 1994; Ackerman et al., 2006; Messaoud et al., 2008; Horev et al., 2009; Asinowski et al., 2014; Russo et al., 2020]. This class of partitions is defined recursively, as follows.

Definition 5.1. Consider a land-subset \( L = [a_0, a_1] \times [b_0, b_1] \), a set of rectangles \( P = \{P_1, \ldots, P_t\} \), where \( P_i \subseteq L \) for each \( i \in [t] \), and a separation parameter \( s \). We say that \( P \) forms an \( s \)-separated guillotine partition of \( L \) if one of the following three conditions holds:

- \( t = 1 \) and \( P_1 \subseteq L \);
- there exists an \( a \) with \( a_0 < a < a_1 - s \) and a partition of \( P \) into two disjoint collections of rectangles \( P_1 \) and \( P_2 \) such that \( P_1 \) forms an \( s \)-separated guillotine partition of \([a, a_1] \times [b_0, b_1]\) and \( P_2 \) forms an \( s \)-separated guillotine partition of \([a + s, a_1] \times [b_0, b_1]\);
- there exists a \( b \) with \( b_0 < b < b_1 - s \) and a partition of \( P \) into two disjoint collections of rectangles \( P_1 \) and \( P_2 \) such that \( P_1 \) forms an \( s \)-separated guillotine partition of \([a_0, a_1] \times [b_0, b]\) and \( P_2 \) forms an \( s \)-separated guillotine partition of \([a_0, a_1] \times [b + s, b_1]\).

Intuitively, an \( s \)-separated guillotine partition is obtained by a sequence of cuts, where each cut splits a rectangle into two \( s \)-separated rectangles. All partitions in Figure 2 are guillotine partitions, while Figure 3 provides an example of an \( s \)-separated partition that is not a guillotine partition.

The following theorem shows that we can compute a nearly optimal \( s \)-separated guillotine maximin partition efficiently. Our algorithm proceeds by discretizing the land and finding an optimal \( s \)-separated guillotine partition that is consistent with this discretization; such a partition can be computed by dynamic programming.

Theorem 5.2. Consider a rectangular land-subset \( L \), an agent \( i \) who values \( L \) at 1, and a separation parameter \( s > 0 \). Suppose that there exists an \( s \)-separated guillotine partition of \( L \) into \( k \) parts such that \( i \)’s value for each part is at least \( V \). Then, given \( \varepsilon > 0, s \), and \( k \), we can compute an \( s \)-separated guillotine partition of \( L \) into \( k \) parts such that \( i \)’s value for each part is at least \( V - \varepsilon \), in time polynomial in \( k \) and \( 1/\varepsilon \).

How much value do we lose by considering guillotine partitions instead of general ones? Question 3 illustrates that this loss is non-trivial. The following theorem provides a crude (but positive) lower bound on the approximation ratio.

Theorem 5.3. Let \( \Xi \text{-MMS}^{k,s}i \) denote the maximin share of agent \( i \) with respect to \( s \)-separated guillotine partitions into \( k \) parts. Then, it holds that \( \Xi \text{-MMS}^{k,s}i \geq \text{MMS}^{4k^2,s} \).

Combining Theorems 5.2 and 5.3 with 4.8 gives Corollary 5.4. For each \( \varepsilon > 0 \), we can compute in time polynomial in \( k \) and \( 1/\varepsilon \) an allocation in which each agent \( i \) receives a rectangular piece of land with value at least \( \text{MMS}^{k,s}i - \varepsilon \), where \( k \leq 4 \cdot (2^{2n+2})^2 = 2^{4n+6} \).

6 Conclusion and Future Work

This paper continues the quest of bringing the theory of fair division closer to practice by investigating fair land allocation under separation constraints. Even though the classic fairness notion of proportionality is unsuitable for this setting, we establish meaningful bounds on achievable maximin share guarantees for a variety of shapes and develop a number of new techniques in the process. In particular, for \( r \)-fat rectangles we derive a polynomial bound on the achievable approximation for any number of agents, while for arbitrary rectangular pieces we obtain a finite but exponential bound. Improving the latter bound to polynomial is a challenging question which likely requires novel geometric insights. Other avenues for future work include testing our algorithms on real land division data [Shtechman et al., 2020] and exploring the possibilities of efficient computation with non-guillotine or other types of cuts.

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