

# Fair Allocation of Indivisible Goods to Asymmetric Agents

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## ABSTRACT

We study fair allocation of indivisible goods to agents with *unequal entitlements*. Fair allocation has been the subject of many studies in both divisible and indivisible settings. Our emphasis is on the case where the goods are indivisible and agents have unequal entitlements. This problem is a generalization of the work by Procaccia and Wang [17] wherein the agents are assumed to be symmetric with respect to their entitlements. Although Procaccia and Wang show an almost fair (constant approximation) allocation exists in their setting, our main result is in sharp contrast to their observation. We show that, in some cases with  $n$  agents, no allocation can guarantee better than  $1/n$  approximation of a fair allocation when the entitlements are not necessarily equal. Furthermore, we devise a simple algorithm that ensures a  $1/n$  approximation guarantee.

Our second result is for a restricted version of the problem where the valuation of every agent for each good is bounded by the total value he wishes to receive in a fair allocation. Although this assumption might seem w.l.o.g, we show it enables us to find a  $1/2$  approximation fair allocation via a greedy algorithm. Finally, we run some experiments on real-world data and show that, in practice, a fair allocation is likely to exist. We also support our experiments by showing positive results for two stochastic variants of the problem, namely *stochastic agents* and *stochastic items*.

## CCS Concepts

•Computing methodologies → Multi-agent systems;

## Keywords

fairness, indivisible, entitlements, proportionality, approximation, stochastic

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## 1. INTRODUCTION

In this work, we conduct a study of *fairly* allocating *indivisible goods* among  $n$  agents with unequal claims on the goods. Fair allocation is a very fundamental problem that has received attention in both Computer Science and Economics. This problem dates back to 1948 when Steinhaus [20] introduced the *cake cutting* problem as follows: given  $n$  agents with different valuation functions for a cake, is it possible to divide the cake between them in such a way that every agent receives a piece whose value to him is at least  $1/n$  of the whole cake? Steinhaus answered this question in the affirmative by proposing a simple and elegant algorithm which is called *moving knife*. Although this problem admits a straightforward solution, several ramifications of the cake cutting problem have been studied since then, many of which have not been settled after decades [5, 18, 9, 16, 12, 10, 21, 8, 1]. For instance, a natural generalization of the problem in which we discriminate the agents based on their entitlements is still open. In this problem, every agent claims an entitlement  $e_i$  to the cake such that  $\sum e_i = 1$ , and the goal is to cut the cake into disproportional pieces and allocate them to the agents such that every agent  $a_i$ 's valuation for his piece is at least  $e_i$  fraction of his valuation for the entire cake. For two agents, Brams *et al.* [4] showed that at least two cuts are necessary to divide the cake between the agents. Furthermore, Robertson *et al.* [19] proposed a modified version of cut and choose method to divide the cake between two agents with portions  $e_1, e_2$ , where  $e_1$  and  $e_2$  are real numbers. McAvaney, Robertson, and Web [15] considered the case when the entitlements are rational numbers. They used Ramsey partitions to show that when the entitlements are rational, one can make a proper division via  $O(n^3)$  cuts.

Recently, a new line of research is focused on the fair allocation of indivisible goods. In contrast to the conventional cake cutting problem, in this problem instead of a heterogeneous cake, we have a set  $\mathcal{M}$  of indivisible goods and we wish to distribute them among  $n$  agents. Indeed, due to trivial counterexamples in this setting<sup>1</sup>, the previous guarantee, that is

<sup>1</sup>For instance if there is only one item, at most one agent has a non-zero profit in any allocation.

every agent should obtain  $1/n$  of his valuation for all items from his allocated set, is impossible to deliver. To alleviate this problem, Budish [6] proposed a concept of fairness for the allocation of indivisible goods namely *the maxmin share*. Suppose we ask an agent  $a_i$  to divide the items between the agents in a way that *he thinks* is fair to everybody. Of course, agent  $a_i$  does not take into account other agents' valuations and only incorporates his valuation function in the allocation. Based on this, we define  $MMS_i$  equal to the minimum profit that any agent receives in this allocation, according to agent  $a_i$ 's valuation function. Obviously, in order to maximize  $MMS_i$ , agent  $a_i$  chooses an allocation that maximizes the minimum profit of the agents. We call an allocation fair (approximately fair), if every agent  $a_i$  receives a set of items that is worth at least  $MMS_i$  (a fraction of  $MMS_i$ ) to him.

It is easy to see that  $MMS_i$  is the best possible guarantee that one can hope to obtain in this setting. If all agents have the same valuation function, then at least one of the agents receives a collection of items that are worth no more than  $MMS_i$  to him. A natural question that emerges here is whether a fair allocation with respect to  $MMS_i$ 's is always possible? Although the experiments are in favor of this conjecture, Procaccia and Wang [17] (EC'14) refuted this by an elegant and delicate counterexample. They show such a fair allocation is impossible in some cases, even when the number of agents is limited to 3. On the positive side however, they show an approximately fair allocation can be guaranteed. More precisely, they show that there always exists an allocation in which every agent's profit is at least  $2/3MMS_i$ . Such an allocation is called a  $2/3$ -MMS allocation. Amanatidis, Markakis, Nikzad, and Saberi [2] later provided a proof for the existence of an MMS allocation for the case, when there are large enough items and the value of each agent for every items is drawn independently from a uniform distribution. A generalized form of this result was later proposed by Kurokawa *et al.* [14] for arbitrary distributions. In a recent work, Caragiannis *et al.* [7] proved that the maximum Nash welfare (MNW) solution, which selects an allocation that maximizes the product of utilities, for each agent guarantees a  $2/(1 + \sqrt{4n - 3})$  fraction of her MMS.

Although it is natural to assume the agents have equal entitlements on the items, in most real-world applications, agents have unequal entitlements on the goods. For instance, in various religions, cultures, and regulations, the distribution of the inherited wealth is often unequal. Furthermore, the division of mineral resources of a land or international waters between the neighboring countries is often made unequally based on the geographic, economic, and political status of the countries.

For fairly allocating indivisible items to agents with different entitlements, two procedures are proposed in [5]. The first one is based on *Knaster's procedure of sealed bids*. In this method, we have an auction for selling each item. Therefore, for using it all the agents should have an adequate reserve of money which is the main issue of the procedure. The second procedure mentioned in [5] is based on *method of markers* developed by William F. Lucas which is spiritually similar to the moving knife procedure. In this method, first we line up the items, and then the agents place some markers for dividing the items. This method suffers from high dependency of its final allocation to the order of the items in the line.

Agent duplication is another idea to deal with unequal entitlements. More precisely, when all of the entitlements

are fractional numbers, we can duplicate each agent  $a_i$  to some agents with similar valuation functions to  $a_i$ . The goal of this duplication is to reduce the problem to the case of equal entitlements. After the allocation, every agent  $a_i$  owns all of the allocated items to her duplicated agents. For instance, assume that we have three agents with entitlements  $1/2$ ,  $2/5$ , and  $1/10$ , respectively. In this case, we duplicate the first agent to five agents and the second agent to four agents each having an entitlement of  $1/10$ . This way, we can reduce our problem to the case of equal entitlements. Although agent duplication may be practical when the items are divisible, in the indivisible case, this method does not apply to the indivisible setting. For instance, if the number of the agents is higher than the number of available items, we cannot allocate anything to some agents. Another issue with this method is that it works only for fractional entitlements.

In this paper, we study fair allocation of indivisible items with different entitlements using a model which resolves the mentioned issues. Our fairness criterion mimics the general idea of Budish for defining maxmin shares. Similar to Budish's proposal, in order to define a maxmin share for an agent  $a_i$ , we ask the following question: how much benefit does agent  $a_i$  expect to receive from a fair allocation, if we were to divide the goods *only based on his valuation function*? If agent  $a_i$  expects to receive a profit of  $p$  from the allocation, then he should also recognize a minimum profit of  $p \cdot e_j/e_i$  for any other agent  $a_j$ , so that his own profit per entitlement is a lower bound for all agents. Therefore, a fair answer to this question is the maximum value of  $p$  for which there exists an allocation such that agent  $a_i$ 's profit-per-entitlement can be guaranteed to all other agents (according to his own valuation function). We define the maxmin shares of the agents based on this intuition.

Recall that we denote the number of agents with  $n$  and the entitlement of every agent  $a_i$  with  $e_i$ . We assume the entitlements always add up to 1. For every agent  $a_i$ , we define the weighted maxmin share denote by  $WMMS_i$ , to be the highest value of  $p$  for which there exists an allocation of the goods to the agents in which every agent  $a_j$  receives a profit of at least  $p \cdot e_j/e_i$  based on agent  $a_i$ 's valuation function. Similarly, we call an allocation  $\alpha$ -WMMS, if every agent  $a_i$  obtains an  $\alpha$  fraction of  $WMMS_i$  from his allocated goods. Notice that in case  $e_i = 1/n$  for all agents, this definition is identical to Budish's definition. Since our model is a generalization of the Budish's model, it is known that a fair allocation is not guaranteed to exist for every scenario. However, whether a  $2/3$  approximation or in general a constant approximation WMMS allocation exists remains an open question.

Our main result is in contrast to that of Procaccia and Wang. We settle the above question by giving a  $1/n$  hardness result for this problem. In other words, we show no algorithm can guarantee any allocation which is better than  $1/n$ -WMMS in general. We further complement this result by providing a simple algorithm that guarantees a  $1/n$ -WMMS allocation to all agents. As we show in Section 2, this hardness is a direct consequence of unreasonably high valuation of agents with low entitlements for some items. Moreover, in Section 3 we discuss that not only are such valuation functions unrealistic, but also an agent with such a valuation function has an incentive to misrepresent his valuations (Observation 3.1). Therefore, a natural limitation that one can add to the setting is to assume no item is worth more than  $WMMS_i$  for any agent  $a_i$ . We also study the problem in this mildly restricted

setting and show in this case a 1/2-WMMS guarantee can be delivered via a greedy algorithm.

In contrast to our theoretical results, we show in practice a fair allocation is likely to exist by providing experimental results on real-world data. The source of our experiments is a publicly available collection of bids for eBay goods and services<sup>2</sup>. Note that since those auctions are *truthful*<sup>3</sup>, it is the users' best interest to bid their actual valuations for the items and thus the market is transparent. More details about the experiments can be found in Section 4. We also support our claim by presenting theoretical analysis for the stochastic variants of the problem in which the valuation of every agent for a good is drawn from a given distribution.

## 1.1 Our Model

Let  $\mathcal{N}$  be a set of  $n$  agents, and  $\mathcal{M}$  be a set of  $m$  items. Each agent  $a_i$  has an *additive* valuation function  $V_i$  for the items. In addition, every agent  $a_i$  has an entitlement to the items, namely  $e_i$ . The entitlements add up to 1, i.e.,  $\sum e_i = 1$ .

Since our model is a generalization of maxmin share, we begin with a formal definition of the maxmin shares for equal entitlements, proposed by Budish [6]. In this case, we assume all of the entitlements are equal to  $1/n$ . Let  $\Pi(\mathcal{M})$  be the set of  $n$ -partitionings of the items. Define the maxmin share of agent  $a_i$  ( $MMS_i$ ) of player  $i$  as

$$MMS_i = \max_{\langle A_1, A_2, \dots, A_n \rangle \in \Pi(\mathcal{M})} \min_{j \in [n]} V_i(A_j). \quad (1)$$

One can interpret the maxmin share of an agent as his outcome as a divider in a divide-and-choose procedure against adversaries [6]. Consider a situation that a cautious agent knows his own valuation on the items, but the valuations of other agents are unknown to him. If we ask the agent to run a divide-and-choose procedure, he tries to split the items in a way that the least valuable bundle is as attractive as possible.

When the agents have different entitlements, the above interpretation is no longer valid. The problem is that the agents have different entitlements and this discrepancy must somehow be considered in the divide-and-choose procedure. Thus, we need an interpretation of the maxmin share that takes the entitlements into account.

Let us get back to the case with the equal entitlements. Another way to interpret maxmin share is this: suppose that we ask agent  $a_i$  to fairly distribute the items in  $\mathcal{M}$  between  $n$  agents of  $\mathcal{N}$ , based on his own valuation function. In an ideal situation (e.g., if the goods are completely divisible), we expect  $a_i$  to allocate a share with value  $V_i(\mathcal{M})/n$  to every agent. However, since the goods are indivisible, some sort of unfairness is inevitable. For this case, we wish that  $a_i$  does his best to retain fairness.  $MMS_i$  is in fact, a parameter that reveals how much fairness  $a_i$  can guarantee, regarding his valuation function.

Formally, to measure the fairness of an allocation by  $a_i$ , define a value  $F_A^i$  for any allocation  $A = \langle A_1, A_2, \dots, A_n \rangle$  as

$$F_A^i = \frac{\min_j V_i(A_j)}{V_i(\mathcal{M})/n}.$$

In fact, we wish to make sure  $a_i$  reports an allocation  $A^*$

<sup>2</sup><http://cims.nyu.edu/~munoz/data/>

<sup>3</sup>An action is called truthful, if no bidder has any incentive to misrepresent his valuation

such that  $F_{A^*}^i$  is as close to 1 as possible. The maxmin share of  $a_i$  is therefore defined as

$$MMS_i = F_{A^*}^i (V_i(\mathcal{M})/n). \quad (2)$$

It is easy to observe that Equations (1) and (2) are equivalent, since the fairest allocation in the absence of different entitlements is an allocation that maximizes value of the minimum bundle:

$$\begin{aligned} MMS_i &= F_{A^*}^i (V_i(\mathcal{M})/n) \\ &= \frac{\min_j V_i(A_j^*)}{V_i(\mathcal{M})/n} (V_i(\mathcal{M})/n) = \min_j V_i(A_j^*) \end{aligned}$$

Now, consider the case with different entitlements. Let  $e_i$  be the entitlement of agent  $a_i$ . Similar to the second interpretation for  $MMS_i$ , ask agent  $a_i$  to fairly distribute the items between the agents, but this time, considers the entitlements. In an ideal situation (e.g., a completely divisible resource), we expect the allocation to be proportional to the entitlements, i.e.  $a_i$  allocates a share to agent  $a_j$  with value exactly  $V_i(\mathcal{M})e_j$  (note that when the entitlements are equal, this value equals to  $V_i(\mathcal{M})/n$  for every agent). But again, such an ideal situation is very rare to happen and thus we allow some unfairness. In the same way, define the fairness of an allocation  $A = \langle A_1, A_2, \dots, A_n \rangle$  as

$$F_A^i = \min_j \frac{V_i(A_j)}{V_i(\mathcal{M})e_j} \quad (3)$$

Let  $A^* = \langle A_1^*, A_2^*, \dots, A_n^* \rangle$  be an allocation by  $a_i$  that maximizes  $F_{A^*}^i$ . The weighted maxmin share of agent  $a_i$  is defined in the same way as  $MMS_i$ , that is:

$$WMMS_i = F_{A^*}^i V_i(\mathcal{M})e_i = e_i \min_j \frac{V_i(A_j^*)}{e_j}$$

In summary, the value  $WMMS_i$  for every agent  $a_i$  is defined as follows:

$$WMMS_i = \max_{\langle A_1, A_2, \dots, A_n \rangle \in \Pi(\mathcal{M})} \min_{j \in [m]} V_i(A_j) \frac{e_i}{e_j}.$$

For more intuition, consider the following example:

**EXAMPLE 1.** Assume that we have two agents  $a_1, a_2$  with  $e_1 = 1/3$  and  $e_2 = 2/3$ . Furthermore, suppose that there are 5 items  $b_1, b_2, b_3, b_4, b_5$  with the following valuations for  $a_1$ :  $V_1(\{b_1\}) = V_1(\{b_2\}) = V_1(\{b_3\}) = 4$ ,  $V_1(\{b_4\}) = 3$  and  $V_1(\{b_5\}) = 9$ . For the allocation  $A = \langle \{b_5\}, \{b_1, b_2, b_3, b_4\} \rangle$ , we have  $F_A = \min(\frac{9}{24 \cdot (1/3)}, \frac{15}{24 \cdot (2/3)})$  which means  $F_A = 15/16$ . Moreover, for allocation  $A' = \langle \{b_1, b_2\}, \{b_3, b_4, b_5\} \rangle$ , we have  $F_{A'} = \min(\frac{8}{24 \cdot (1/3)}, \frac{16}{24 \cdot (2/3)})$  which means  $F_{A'} = 1$ . Thus,  $A'$  is a fairer allocation than  $A$ . In addition,  $A'$  is the fairest possible allocation and hence,  $WMMS_1 = 1 \cdot 24 \cdot 1/3 = 8$ .

Example 1 also gives an insight about why agent duplication (as introduced in the Introduction) is not a good idea. For this example, if we duplicate agent  $a_2$ , we have three agents with the same entitlements. But any partitioning of the items into three bundles, results in a bundle with value at most 7 to  $a_1$ .

Finally, an allocation of the items in  $\mathcal{M}$  to the agents in  $\mathcal{N}$  is said to be  $\alpha$ -WMMS, if the total value of the share allocated to each agent  $a_i$  is worth at least  $\alpha WMMS_i$  to him.

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