

Approximately Envy-free and Equitable Allocations of Indivisible Items for Non-monotone Valuations

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Abstract

We revisit the setting of fair allocation of indivisible items among agents with heterogeneous, non-monotone valuations. We explore the existence and efficient computation of allocations that approximately satisfy either envy-freeness or equity constraints. Approximate envy-freeness ensures that each agent values her bundle at least as much as those given to the others, after some (or any) item removal, while approximate equity guarantees roughly equal valuations among agents, under similar adjustments. As a key technical contribution of this work, by leveraging fixed-point theorems (such as Sperner’s Lemma and its variants), we establish the existence of *envy-free-up-to-one-good-and-one-chore* (EF1_g^c) and *equitable-up-to-one-good-and-one-chore* (EQ1_g^c) allocations, for non-monotone valuations that are always either non-negative or non-positive. These notions represent slight relaxations of the well-studied *envy-free-up-to-one-item* (EF1) and *equitable-up-to-one-item* (EQ1) guarantees, respectively. Our existential results hold even when items are arranged in a path and bundles must form connected sub-paths. The case of non-positive valuations, in particular, has been solved by proving a novel multi-coloring variant of Sperner’s Lemma that constitutes a combinatorial result of independent interest. In addition, we also design a polynomial-time dynamic programming algorithm that computes an EQ1_g^c allocation. For monotone non-increasing valuations and path-connected bundles, all the above results can be extended to EF1 and EQ1 guarantees as well. Finally, we provide existential and computational results for certain stronger *up-to-any-item equity* notions under objective valuations, where items are partitioned into goods and chores.

1 Introduction

Fair division (Steinhaus 1948), the field that studies how to fairly allocate resources among a set of agents, has numerous applications across a variety of real-life scenarios, such as divorce settlements, credit assignment, and rent and land division, to name a few. Although fair division has been studied for decades in mathematics and economics, the field has attracted increasing attention from the computer science and AI community in recent years, driven by the flourishing of new fairness concepts and the demand for computation-

ally efficient solutions overcoming the inherent impossibility of achieving optimal fairness guarantees.

Two prominently investigated notions of fairness are *envy-freeness* (Foley 1966) and *equitability* (Dubins and Spanier 1961). An allocation (of items to agents) is *envy-free* (EF) if the value that every agent gives to her assigned bundle (of items) is not less than the value she gives to the bundle assigned to any other agent; it is *equitable* (EQ) if the value that every agent gives to her assigned bundle is not less than the value that the other agents assign to their respective bundles. So, the two notions coincide when agents have identical valuations.

The nature of valuation functions tremendously impacts the solution of a fair division problem. When an agent’s valuation is monotone non-decreasing (resp., non-increasing), items are said to be *goods* (resp., *chores*) for the agent; when it is non-monotone, items are said to be *mixed*. Notable special cases of non-monotone valuations include *non-negative* (resp. *non-positive*) valuations, where every bundle yields a non-negative (resp. non-positive) value, and *objective* valuations, in which items can be partitioned into goods and chores. Valuations, either monotone or non-monotone, are *additive* when the value of a bundle is defined by the sum of the values of its items. Finally, we implicitly assume that, for every valuation function considered, the empty bundle has value zero. While objective valuations have been widely studied (Aziz et al. 2022; Barman et al. 2024), less work has been done for non-negative or non-positive ones, despite their potential applicability in numerous settings. These valuations, for instance, arise when items correspond to nodes in an edge-weighted graph with exclusively positive or negative weights, allocations are interpreted as clusterings, and the value of each bundle/cluster is then determined by factors such as the total weight of internal or cut edges, or other graph-connectivity properties. This class of settings has interesting connections with clustering problems where further fairness guarantees are required (see, e.g., (Dinitz et al. 2022; Chierichetti et al. 2017; Schwartz and Zats 2022)).

Either envy-free or equitable allocations are guaranteed to exist under non-negative or non-positive valuations in the setting of *divisible items*, where items can be arbitrarily split among subsets of agents (Dubins and Spanier 1961; Stromquist 1980). In contrast, in the presence of *indivisible items* which have to be integrally assigned to any of

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the agents, existence cannot be guaranteed even for two agents with additive monotone valuations. To overcome this limitation, a number of relaxations have been proposed in the literature. They allow the removal of one item from a bundle when agents perform bundle comparisons. The removal strategy clearly depends on the nature of the considered items. When an agent compares her bundle A against another bundle B , she can choose between removing a chore from A or removing a good from B . These relaxations have given rise to the notions of *envy-freeness-up-to-any-good* (EFX) (Caragiannis et al. 2019), *envy-freeness-up-to-one-good* (EF1) (Lipton et al. 2004; Budish 2011), *equitability-up-to-any-good* (EQX) (Gourvès, Monnot, and Tlilane 2014) and *equitability-up-to-one-good* (EQ1) (Freeman et al. 2019). By *up-to-any-good*, one means that the fairness property holds irrespectively of which item is selected for removal; by *up-to-one-good*, instead, the property must hold for at least one removed item. Clearly, fairness *up-to-any-good* implies fairness *up-to-one-good*.

An interesting and largely studied generalization of fair division assumes the existence of an *item graph* modeling proximity relationships among items. Every bundle has to induce a connected subgraph and an item removal is allowed only if it does not disconnect the induced subgraph (Bouveret et al. 2017; Bilò et al. 2022; Igarashi 2023; Misra et al. 2021; Suksompong 2019). In this respect, the *path constraint* assumes that the item graph is a connected path.

Our Contribution. Given the lack of positive results for non-monotone valuations under standard approximate fairness notions, often due to impossibility barriers (see, e.g., (Amanatidis et al. 2023; Barman et al. 2024)), we study slight relaxations of EF1, EQ1 and EQX, denoted by $EF1_g^c$ (*envy-free-up-to-one-good-and-one-chore*), $EQ1_g^c$ (*equitable-up-to-one-good-and-one-chore*), and EQX_g^c (*equitable-up-to-any-good-or-any-chore*), respectively. While EF1 and EQ1 require the envy-freeness and equitability properties, respectively, to hold upon the removal of at most one item from “some” bundle, $EF1_g^c$ and $EQ1_g^c$ allow the properties to hold when removing at most one item from “each” bundle (i.e., for each agent, at most one chore from her own bundle and at most one good from others’ bundles). Similarly, while EQX requires the equitability property to hold regardless of which item is removed (i.e., whether it is a good or a chore), EQX_g^c allows it to hold for the removal of either goods only or chores only. We obtain positive results on the existence and computation of allocations satisfying the above fairness criteria across several broad classes of non-monotone valuations.

Results for non-negative or non-positive valuations: Our main contribution concerns the existence and computation of allocations which are fair *up-to-one-good-and-one-chore*, under non-negative or non-positive valuations. In particular, we show that an $EQ1_g^c$ allocation always exists and can be computed in polynomial time, for both non-negative (Theorem 3.3 and 3.4) and non-positive valuations (Theorem 4.2). For $EF1_g^c$ allocations, we only show existence (Theorem 3.5 and Theorem 4.3). The existence and computation of approximately envy-free or equitable allocations under non-

monotone valuations is one of the major open problems in fair division (see, e.g., the surveys by Amanatidis et al. (2023); Liu et al. (2024)). Our results represent a significant step forward in this direction, due to the generality of the non-monotone valuations we consider (either non-negative or non-positive) and the fairness guarantees achieved (requiring the removal of at most one good and one chore).

It is worth noting that our results continue to hold even under path constraints, and in this sense, they generalize the findings of Bilò et al. (2022); Igarashi (2023); Misra et al. (2021); Suksompong (2019), which apply only to monotone non-decreasing valuations, and the results of Bouveret et al. (2017); Bouveret, Cechlárová, and Lesca (2019), which address the computational problem of finding EF and EQ allocations (that, in general, may not exist). Our existential results are obtained using Sperner’s Lemma (Sperner 1928) or its variants, and represent a non-trivial generalization of the approaches previously explored in (Bilò et al. 2022; Igarashi 2023), as handling the non-monotonicity of the valuations poses significant technical challenges (such as the derivation and analysis of the cases described in Figure 1). In particular, to address the specific case of non-positive valuations, we introduce and exploit a novel multi-coloring variant of Sperner’s Lemma (Theorem 4.1), that constitutes a combinatorial result of independent interest. The computational results have been obtained by means of the dynamic-programming paradigm. Finally, when valuations are monotone (non-decreasing or non-increasing), our results extend to the stronger EQ1 and EF1 guarantees under path constraints, thereby generalizing the results of Igarashi (2023); Bilò et al. (2022); Misra et al. (2021), which exhibit EF1 and EQ1 allocations under monotone non-decreasing valuations and path constraints.

Results for objective valuations: To complete the picture for non-monotone valuations, we also consider fair allocations under objective valuations, and in most cases the obtained results hold even under the *up-to-any-good-or-any-chore* approximation guarantee. In particular, we show that an EQX_g^c allocation always exists and can be computed in pseudopolynomial time, via a simple variant of the local-search approach adopted by Barman et al. (2024). This result extends to EQX when valuations are monotone non-increasing, thereby generalizing the result of Barman et al. (2024), which holds only for monotone non-decreasing valuations. For valuations that are both objective and additive, we strengthen the above computational result by showing that an EQX_g^c allocation can be found through a simple and more efficient greedy algorithm. This result strengthens the findings of Hosseini and Sethia (2025), who establish existence and polynomial-time computability of EQ1 allocations under additive objective valuations. We also show that a slight generalization of the polynomial-time algorithm proposed by Hosseini and Sethia (2025), continues to produce EQ1 allocations even for non-additive objective valuations¹. We note that, just as EQX_g^c is considered a relaxation of EQX, similar relaxations for EFX have been studied for

¹The EQ1 algorithm of Hosseini and Sethia (2025) and the algorithms employed in this work were developed independently.

objective (Hosseini et al. 2023) or identical non-monotone valuations (Bérczi et al. 2024). However, these works only show non-existence of these relaxed notions.

Due to the lack of space, most technical details on non-negative and non-positive valuations, as well as all missing proofs, are deferred to the full arXiv version of this work (Bilò, Loeb, and Vinci 2025), while all results on objective valuations are moved to the full version. Table 1 summarizes our results, related work and cases that remains unsolved.

Further Related Work. Lipton et al. (2004) show that an EF1 allocation always exists and can be efficiently computed for non-decreasing valuations. Their result has been extended to objective valuations by Aziz et al. (2022); Bhaskar, Sricharan, and Vaish (2021); Bérczi et al. (2024). Again under objective valuations, (Hosseini and Sethia 2025) show existence and polynomial time computation of an EQ1 allocation; conversely, they show that under additive non-objective valuations EQ1 allocations may not exist. Existence and computation of Pareto optimal EF1 or EQ1 allocations have been studied by Caragiannis et al. (2019); Freeman et al. (2019, 2020); Garg and Murhekar (2024). EF1 and EQ1 allocations under non-objective valuations have been determined for restricted cases only, and their general existence and computation is a major open problem (see surveys by Amanatidis et al. (2023); Liu et al. (2024)).

EQX allocations were first proved to exist for additive monotone non-decreasing valuations in (Gourvès, Monnot, and Tlilane 2014). Efficient algorithms computing one have been later designed by Freeman et al. (2019, 2020), also covering the non-increasing case. Existence and pseudopolynomial time computation of an EQX allocation for monotone non-decreasing (and possibly non-additive) valuations has been shown in (Barman et al. 2024). This result is complemented by showing that, if one drops the monotonicity assumption, EQX allocations may not exist even for two agents with additive valuations. So, our results state that, slightly relaxing EQX to EQX_g^c suffices to recover existence under non-monotone objective valuations.

The EFX criterion was introduced by Caragiannis et al. (2019). Its existence under additive non-negative or non-positive valuations has been addressed only in specific cases, and remains a major open problem in fair division (see surveys by Amanatidis et al. (2023) and Liu et al. (2024)). Whereas, for objective additive valuations (Hosseini et al. 2023) or chore valuations (Christoforidis and Santorinaios 2024), an EFX allocation might not exist. Plaut and Roughgarden (2020) show that an EFX allocation exists and can be efficiently computed for non-negative additive valuations when agents assign the same ranking to all items; this result has also been extended to additive objective valuations with equally ranked items (Aziz and Rey 2020). Restricting to identical valuations, EFX allocations are known to exist for additive non-decreasing valuations (Gourvès, Monnot, and Tlilane 2014) and additive non-increasing valuations (Barman, Narayan, and Verma 2023), whereas they may not exist for non-monotone valuations (Bérczi et al. 2024).

Finally, additional relaxed notions of fairness have been investigated (e.g., in (Akrami et al. 2025; Akrami and Rathi

2025; Amanatidis, Markakis, and Ntokos 2020; Aziz et al. 2024; Bilò, Markakis, and Vinci 2024; Caragiannis et al. 2023; Halpern and Shah 2019; Hoefler, Schmalhofer, and Varricchio 2024; Plaut and Roughgarden 2020)).

2 Model and Definitions

Let $N = \{1, 2, \dots, n\}$ be a finite set of n agents and M be a finite set of m items. Each agent $i \in N$ has an integral valuation function $v_i : 2^M \rightarrow \mathbb{Z}$ with $v_i(\emptyset) = 0$ for any $i \in N$. We denote by $I = (N, M, (v_i)_{i \in N})$ an allocation instance. Given an agent $i \in N$, a bundle of items $S \subseteq M$ and an item $x \in S$, we say that x is a *good* (resp., a *chore*) for i w.r.t. S , if $v_i(S) \geq v_i(S \setminus \{x\})$ (resp., $v_i(S) \leq v_i(S \setminus \{x\})$). We observe that an item x for which $v_i(S) = v_i(S \setminus \{x\})$ is both a good and a chore for i w.r.t. S . An item x is a good (resp., a chore) if it is a good (resp., a chore) for any agent $i \in N$ w.r.t. any bundle $S \subseteq M$. An allocation $\mathcal{A} = (A_1, \dots, A_n)$ is a partition of M into n (possibly empty) bundles of items, such that A_i is the bundle assigned to agent $i \in N$. We aim at finding allocations satisfying fairness criteria related to *envy-freeness* and *equity*, as described below.

Envy-freeness. An allocation $\mathcal{A} = (A_1, \dots, A_n)$ is:

- *envy-free* (EF) if, for any $i, j \in N$, $v_i(A_i) \geq v_i(A_j)$;
- *envy-free-up-to-any-item* (EFX) if, for any $i, j \in N$ such that $v_i(A_i) < v_i(A_j)$, all the following conditions hold: (i) $v_i(A_i) \geq v_i(A_j \setminus \{g\})$ for any good g for agent i w.r.t. A_j ; (ii) $v_i(A_i \setminus \{c\}) \geq v_i(A_j)$ for any chore c for i w.r.t. A_i ; (iii) either there exists a good g for i w.r.t. A_j , or there exists a chore c for i w.r.t. A_i ;
- *envy-free-up-to-one-item* (EF1) if, for any $i, j \in N$ such that $v_i(A_i) < v_i(A_j)$, there exists $x \in A_i \cup A_j$ such that $v_i(A_i \setminus \{x\}) \geq v_i(A_j \setminus \{x\})$;
- *envy-free-up-to-one-good-and-one-chore* (EF1_g^c) if, for any $i, j \in N$ such that $v_i(A_i) < v_i(A_j)$, there exists a subset $X \subseteq M$ with $|A_i \cap X| \leq 1$ and $|A_j \cap X| \leq 1$, such that $v_i(A_i \setminus X) \geq v_i(A_j \setminus X)$.

Under EF1_g^c allocations, agent i stops envying agent j after removing at most one chore from A_i and at most one good from A_j , possibly two items in total. In contrast, under EF1 (resp. EFX), agent i stops envying agent j after removing at most one (resp. any) envy-reducing item from $A_i \cup A_j$. We observe that $\text{EF} \Rightarrow \text{EFX} \Rightarrow \text{EF1} \Rightarrow \text{EF1}_g^c$.

Equitability. The equitability notions we consider are analogous to the envy-freeness criteria described above, but the comparison each agent i makes is not against the valuation $v_i(A_j)$ that she assigns to the bundle given to any other agent j , but rather against the valuation $v_j(A_j)$ that agent j assigns to her own bundle. This naturally leads to the definitions of *equitable* (EQ), *equitable-up-to-any-item* (EQX), *equitable-up-to-one-item* (EQ1) and *equitable-up-to-one-good-and-one-chore* (EQ1_g^c) allocations. In addition, we consider *equitable-up-to-any-good-or-any-chore* (EQX_g^c) allocations, a slight relaxation of EQX that, unlike EQX, allows the equitability to hold after the removal of either only goods or only chores. In particular, an allocation \mathcal{A}

	EQ1_g^c	EF1_g^c	EQ1	EF1	EQX_g^c	EQX	EFX
Gen	χ_a [HS25]	? _a	χ_a [HS25]	? _a	χ_a [HS25]	χ_a [HS25]	χ_a [HSVX23]
NNeg	\checkmark^{cp}	\checkmark^c	?	?	?	?	? _a
NPos	\checkmark^{cp}	\checkmark^c	?	?	?	?	χ [CS24], ? _{ao}
NDec	\checkmark^{cp} [MSVV21]	\checkmark^{p} [LMMS04], \checkmark^c [I23]	\checkmark^{cp} [MSVV21]	\checkmark^{p} [LMMS04], \checkmark^c [I23]	$\checkmark^{\text{p-}}$ [BBPP24]	$\checkmark^{\text{p-}}$ [BBPP24]	? _a
NInc	\checkmark^{cp}	\checkmark^{p} [BSV21], \checkmark^c	\checkmark^{cp}	\checkmark^{p} [BSV21], \checkmark^c	$\checkmark^{\text{p-}}$	$\checkmark^{\text{p-}}$	χ [CS24], ? _{ao}
Obj	\checkmark^{p} , $\checkmark^{\text{p-}}$ _{ao} [HS25]	\checkmark^{p} [BSV21]	\checkmark^{p} , $\checkmark^{\text{p-}}$ _{ao} [HS25]	\checkmark^{p} [BSV21]	$\checkmark^{\text{p-}}$, $\checkmark^{\text{p-}}$ _{ao}	? _a	χ_a [HSVX23]

Table 1: Landscape of results for the considered fairness notions. **Gen**, **NNeg**, **NPos**, **NDec**, **NInc**, and **Obj** denote, respectively, General, Non-negative, Non-positive, Non-decreasing, Non-increasing, and Objective valuations. Gray-highlighted results indicate our contributions, whereas the short citations point to some previous works from which the other (non-highlighted) results follow. Symbols \checkmark , χ , and $?$ indicate, respectively, “always exists”, “does not generally exist”, and “open problem”. Subscript **a** (resp., **ao**) means that an χ or $?$ (resp., \checkmark or $?$) result holds even (resp., only) for additive valuations. Superscripts **p** and **p-** denote existence under polynomial and pseudopolynomial algorithms, respectively; **c** indicates that existence holds under path constraints, and **cp** means that existence under path constraints can be achieved in polynomial time. Although positive (resp., negative) results for some approximate notions of fairness are derived from stronger (resp., weaker) notions, we include repeated citations for completeness.

is EQX_g^c if, for any $i \in N$, at least one of the following conditions holds: (i) for any $j \in N$ such that $v_i(A_i) < v_j(A_j)$, there exists at least one good for j w.r.t. A_j , and for any such good g , $v_i(A_i) \geq v_j(A_j \setminus \{g\})$; (ii) if $v_i(A_i) < v_j(A_j)$ for some $j \in N$, there exists a chore for i w.r.t. A_i , and for any such chore c , $v_i(A_i \setminus \{c\}) \geq v_j(A_j)$ holds for any $j \in N$. We observe that $\text{EQ} \Rightarrow \text{EQX} \Rightarrow \text{EQX}_g^c \Rightarrow \text{EQ1} \Rightarrow \text{EQ1}_g^c$.

Classes of Valuations. We consider the following classes of valuations: (i) *Objective*: any item x is either a good or a chore, independently of the agent and the bundle considered; in such a case, we can partition M into a set of goods G and a set of chores C (choosing arbitrarily how to classify dummy items that qualify as both); (ii) *Non-negative* (resp., *Non-positive*): $v_i(S) \geq 0$ (resp. $v_i(S) \leq 0$) for any $i \in N$, $S \subseteq M$; (iii) *Monotone non-decreasing* (resp., *non-increasing*): each item x is a good (resp., a chore), independently of the agents and bundles considered; (iv) *Additive*: $v_i(S) = \sum_{x \in S} v_i(x)$ for any i , $S \subseteq M$. We observe that monotone non-decreasing (resp. non-increasing) valuations are also non-negative (resp. non-positive) and objective. Finally, we assume the existence of an *oracle* that, given $i \in N$ and $S \subseteq M$, returns $v_i(S)$ in constant time.

3 Non-negative Valuations

To address the case of non-negative valuations, we consider a generalization of the fixed-point approach employed in (Bilò et al. 2022; Igarashi 2023), which is extended in a non-trivial manner to handle the peculiarities of these valuations.

Fairness under Path Constraints. We say that an allocation instance is *path-constrained* if the m items of M are numbered from 1 to m and organized as a path $P = (1, \dots, m)$. Given $s, t \in [m] \cup \{0\}$, let $\llbracket s, t \rrbracket$ denote the bundle $\{s, s+1, \dots, t\}$ if $t \geq s$, and the empty bundle otherwise. A bundle S is *connected* if $S = \llbracket s, t \rrbracket$ for some $s \in [m]$ and $t \in [m] \cup \{0\}$. An allocation \mathcal{A} is *connected* if it is made of connected bundles only. Given a connected bundle $S = \llbracket s, t \rrbracket$, let $\partial S = \{s, t\}$ denote the *boundary* of S (note that $\partial S = S$ if $|S| \leq 2$). For a given connected

allocation $\mathcal{A} = (A_1, \dots, A_n)$, we consider the following path-based notions of EF1_g^c and EQ1_g^c : \mathcal{A} is *envy-free-up-to-one-good-and-one-chore-over-paths* (EF1P_g^c) if, for any $i, j \in N$ such that $v_i(A_i) < v_j(A_j)$, there exists a subset $X \subseteq \partial A_i \cup \partial A_j$ with $|\partial A_i \cap X| \leq 1$ and $|\partial A_j \cap X| \leq 1$, such that $v_i(A_i \setminus X) \geq v_i(A_j \setminus X)$; \mathcal{A} is *equitable-up-to-one-good-and-one-chore-over-paths* (EQ1P_g^c) if, for any $i, j \in N$ such that $v_i(A_i) < v_j(A_j)$, there exists a subset $X \subseteq \partial A_i \cup \partial A_j$ with $|\partial A_i \cap X| \leq 1$ and $|\partial A_j \cap X| \leq 1$, such that $v_i(A_i \setminus X) \geq v_j(A_j \setminus X)$. Note that the notions of *EFOP* and EQ1P_g^c respectively strengthen EF1_g^c and EQ1_g^c by ensuring that items are removed only from the boundary, so that bundles remain path-connected after the removal.

Sperner’s Lemma. Before presenting our results, we provide a brief overview of the underlying theoretical framework based on Sperner’s Lemma. For more details, see, for example, (Flegel 1974).

Let $\text{conv}(v_1, v_2, \dots, v_n)$ denote the convex hull of the n vectors v_1, v_2, \dots, v_n . An $(n-1)$ -simplex Δ is an $(n-1)$ -dimensional polytope defined as the convex hull of its n (affinely independent) vertices v_1, v_2, \dots, v_n . Given $k \in [n]$, a $(k-1)$ -face of an $(n-1)$ -simplex is the $(k-1)$ -simplex obtained as convex hull of a subset of $k-1$ of its vertices. A *triangulation* T of a simplex Δ is a collection of sub- $(k-1)$ -simplices (with $k \in [n]$) whose union is Δ , with the property that the intersection of any two sub-simplices in T is either empty or a face shared by both, which also belongs to T . Each sub-simplex $\Delta' \in T$ is referred to as an *elementary simplex*. The set of vertices of T , denoted as $V(T)$, is the union of the vertices of all the elementary simplices in T (i.e., the union of all the elementary 0-simplices).

Now, let T be a fixed triangulation of an $(n-1)$ -simplex $\Delta = \text{conv}(v_1, v_2, \dots, v_n)$. A *coloring function* of T is a mapping $L : V(T) \rightarrow [n]$ that assigns a number, referred to as a *color*, from the set $[n]$ to each vertex of T . A coloring function L is called *special* if, for any vertex $x \in V(T)$ belonging to the $(n-2)$ -face F_i of Δ that does not include v_i (i.e., the face opposite to v_i , obtained as convex hull of all vertices of Δ except for v_i), the condition $L(x) \neq i$ holds.

We observe that, if L is a special coloring function, then $L(\mathbf{v}_i) = i$ holds for any $i \in [n]$. An elementary $(n-1)$ -simplex $\Delta^* = \text{conv}(\mathbf{x}_1^*, \dots, \mathbf{x}_n^*) \in T$ is said to be *fully-colored* under a coloring function L if each of its n vertices is assigned a distinct color by L , that is, $L(\mathbf{x}_{\sigma(i)}^*) = i$ for any $i \in [n]$, for some permutation $\sigma : [n] \rightarrow [n]$.

Theorem 3.1 (Sperner's Lemma (Sperner 1928)). *Let T be a triangulation of an $(n-1)$ -simplex Δ , where $n \geq 2$, and let L be a special coloring function of T . Then, there exists a fully-colored elementary $(n-1)$ -simplex $\Delta^* \in T$ under L ; moreover, the number of such simplices is odd.*

Below, we also consider a generalized version of Sperner's Lemma, as presented by (Bapat 1989). In this generalized form, there are n special coloring functions L_1, \dots, L_n , and we seek an elementary $(n-1)$ -simplex that is fully-colored according to a broader definition, which holds simultaneously for all of the coloring functions. Let T be a triangulation of an $(n-1)$ -simplex, and let L_1, \dots, L_n be the coloring functions on T . An elementary $(n-1)$ -simplex $\Delta^* = \text{conv}(\mathbf{x}_1^*, \mathbf{x}_2^*, \dots, \mathbf{x}_n^*) \in T$ is *jointly fully-colored* under L_1, \dots, L_n if there exist two permutations $\sigma, \tau : [n] \rightarrow [n]$ such that $L_i(\mathbf{x}_{\sigma(i)}^*) = \tau(i)$ for any $i \in [n]$, i.e., each vertex of Δ^* receives a distinct color under a distinct coloring function.

Theorem 3.2 (Generalized Sperner's Lemma (Bapat 1989)). *Let T be a triangulation of an $(n-1)$ -simplex Δ , and let L_1, \dots, L_n be special coloring functions of T . Then, there exists a jointly-fully-colored elementary $(n-1)$ -simplex $\Delta^* \in T$ under L_1, \dots, L_n .*

EQIP_g^c Allocations

Given a path-constrained allocation instance I with non-negative valuations, we construct a suitable triangulation T of an n -simplex Δ and define a special coloring L for T . Each elementary $(n-1)$ -simplex $\Delta^* \in T$ that is fully-colored under L corresponds to an EQIP_g^c allocation for I . By Sperner's Lemma (Theorem 3.1), the existence of such fully-colored simplices is guaranteed, which in turn ensures the existence of an EQIP_g^c allocation. We also design a polynomial-time algorithm, based on dynamic programming, that efficiently computes such an EQIP_g^c allocation.

Triangulation. Consider the $(n-1)$ -simplex $\Delta = \{\mathbf{x} = (x_1, \dots, x_{n-1}) \in \mathbb{R}^{n-1} : 0 \leq x_1 \leq x_2 \leq \dots \leq x_{n-1} \leq m\}$, which is the convex hull $\text{conv}(\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n)$ of the points $\mathbf{v}_1, \dots, \mathbf{v}_n$, with $\mathbf{v}_i := \underbrace{(0, 0, \dots, 0}_{i-1}, \underbrace{m, m, \dots, m}_{n-i})$ for any $i \in [n]$. We observe that each of the $n(n-2)$ -faces of Δ can be defined as $F_i := \{\mathbf{x} = (x_1, \dots, x_{n-1}) \in \Delta : x_{i-1} = x_i\}$, where we set $x_0 := 0$ and $x_n := m$. We construct a triangulation T of Δ whose set of vertices is $V(T) = \{\mathbf{x} \in \Delta : x_i \in \{0, \frac{1}{3}, \frac{2}{3}, 1, \frac{4}{3}, \frac{5}{3}, 2, \frac{7}{3}, \dots, m-1, m-\frac{2}{3}, m-\frac{1}{3}, m\} \forall i \in [n-1]\}$ and whose simplicial structure is defined below. Each coordinate x_i of vertices $\mathbf{x} \in V(T)$ can be either *integral*, or *1-fractional* or *2-fractional*, where integral (resp. 1-fractional, 2-fractional) means $x_i \in \mathbb{Z}$ (resp. $x_i - \frac{1}{3} \in \mathbb{Z}$,

$x_i - \frac{2}{3} \in \mathbb{Z}$); we write $x_i \equiv 0$ (resp. $x_i \equiv 1, x_i \equiv 2$) if x_i is integral (resp. 1-fractional, 2-fractional). By leveraging Kuhn's triangulation (Kuhn 1960; Scarf 1982; Deng, Qi, and Saberi 2012), we construct the triangulation T such that each elementary $(n-1)$ -simplex $\Delta' = \text{conv}(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n) \in T$ can be generated by fixing the first vertex $\mathbf{x}_1 \in V(T)$ and a permutation $\pi : [n-1] \rightarrow [n-1]$, and then iteratively determining the remaining vertices as follows: $\mathbf{x}_{i+1} = \mathbf{x}_i + \frac{1}{3}\mathbf{e}^{\pi(i)}$ for each $i \in [n-1]$, where \mathbf{e}^i is the i -th vector of the canonical basis of \mathbb{R}^{n-1} (1 in the i -th position, and 0 elsewhere). Each vertex $\mathbf{x} \in V(T)$ can be understood as a vector representing the positions of $n-1$ knives that divide the interval $[0, m]$ into n connected segments having endpoints $a, b \in [0, m] \cap \{\frac{x}{3} | x \in \mathbb{Z}\}$. Following this interpretation, the n vertices of any elementary $(n-1)$ -simplex in T are derived by starting with an initial configuration of $n-1$ cuts (i.e., vertex \mathbf{x}_1) and sequentially shifting each knife one position to the right (by a length of $1/3$) according to a specific ordering defined by a permutation π .

Coloring Function. We now construct the coloring function $L : V(T) \rightarrow [n]$. Given a vertex $\mathbf{x} = (x_1, \dots, x_{n-1}) \in V(T)$, let $\tilde{\mathcal{A}}(\mathbf{x}) = (\tilde{A}_1(\mathbf{x}), \dots, \tilde{A}_n(\mathbf{x}))$ be the *fractional connected allocation* obtained from the partition of $[0, m]$ in n *fractional connected bundles*, defined as $\tilde{A}_i(\mathbf{x}) = [x_{i-1}, x_i]$ for any $i \in [n]$, with $x_0 := 0$ and $x_n := m$; furthermore, each bundle $\tilde{A}_i(\mathbf{x})$ is assigned by default to agent i , for any $i \in [n]$ (i.e., bundles are assigned from left to right following the agents order).

Given $a \in \mathbb{R}_{\geq 0}$, let $a^- := \lfloor a \rfloor$ and $a^+ := \min\{\lfloor a \rfloor + 1, m\}$. Let \tilde{v}_i denote the *virtual valuation* of agent i , which applies to fractional connected bundles $[a, b]$ (where $a, b \in [0, m] \cap \{\frac{x}{3} | x \in \mathbb{Z}\}$ and $a \leq b$) and returns an integer value $\tilde{v}_i([a, b])$ that is defined as follows: *left-value (LV)*: if $a \equiv 0$, $\tilde{v}_i([a, b]) := v_i(\llbracket a^-, b^- \rrbracket)$; *borderline-value (BV)*: if $a \equiv 1$, $\tilde{v}_i([a, b])$ is set equal to the middle value among $v_i(\llbracket a^-, b^- \rrbracket)$, $v_i(\llbracket a^+, b^- \rrbracket)$ and $v_i(\llbracket a^+, b^+ \rrbracket)$; *right-value (RV)*: if $a \equiv 2$, $\tilde{v}_i([a, b]) := v_i(\llbracket a^+, b^- \rrbracket)$. Since the original valuations v_i s are non-negative, the resulting virtual valuations \tilde{v}_i s are also non-negative. Let L be the coloring function that assigns each vertex \mathbf{x} the agent/index i that maximizes the virtual valuation $\tilde{v}_i(\tilde{A}_i(\mathbf{x}))$ applied to the fractional connected bundle $\tilde{A}_i(\mathbf{x})$, where ties are broken in favor of agents receiving a non-empty bundle and, in case of further ties, arbitrarily. We observe that L is a special coloring function. Indeed, for any $i \in [n]$, the $(n-2)$ -face F_i of Δ , which does not contain \mathbf{v}_i , is such that the fractional allocations $\tilde{\mathcal{A}}(\mathbf{x})$ corresponding to vertices $\mathbf{x} \in V(T)$ located on F_i have their i -th bundle empty ($\tilde{A}_i(\mathbf{x}) = \emptyset$). Because of the non-negativity of the virtual valuations, any empty bundle always has the lowest virtual value, regardless of the agent or allocation being considered. Therefore, by the construction of L , we have $L(\mathbf{x}) \neq i$ for any $i \in [n]$ and any vertex $\mathbf{x} \in V(T)$ located on the $(n-2)$ -face F_i . Thus, L is a special coloring function.

From the Fully-colored Simplex to the EQIP_g^c Allocation. According to Sperner's Lemma (Theorem 3.1), there exists at least one fully-colored elementary $(n-1)$ -simplex

$\Delta^* = \text{conv}(\mathbf{x}_1^*, \dots, \mathbf{x}_n^*) \in T$ under the coloring L , where $L(\mathbf{x}_{\sigma(i)}^*) = i$ for all $i \in [n]$, for some permutation $\sigma : [n] \rightarrow [n]$. Equivalently, each $i \in [n]$ is among those agents j who maximize the virtual valuation $\tilde{v}_j(\tilde{A}_j(\mathbf{x}_{\sigma(i)}^*))$ in the fractional connected allocation $\tilde{A}(\mathbf{x}_{\sigma(i)}^*)$ associated with the $\sigma(i)$ -th vertex $\mathbf{x}_{\sigma(i)}^*$ of Δ^* , where the sequence of allocations $\tilde{A}(\mathbf{x}_1^*), \dots, \tilde{A}(\mathbf{x}_n^*)$ is obtained by moving each knife one at a time from left to right in a specific order, starting from the position of knives determined by $\tilde{A}(\mathbf{x}_1^*)$.

Denote by \tilde{A} the first allocation $\tilde{A}(\mathbf{x}_1^*)$, and refer to it as the *main allocation* of Δ^* . By appropriately rounding the fractional bundles of \tilde{A} , we will obtain the desired (integral) allocation \mathcal{A} that satisfies the EQIP $_g^c$ guarantee. The rounding procedure processes all fractional bundles \tilde{A}_j s of the main allocation \tilde{A} from $j = n$ down to $j = 1$, and for each bundle \tilde{A}_j , it returns the integral bundle A_j that will form the final (integral) allocation $\mathcal{A} = (A_1, \dots, A_n)$. Specifically, once the bundles A_{j+1}, \dots, A_n have been determined, the bundle A_j is obtained by rounding the fractional bundle $\tilde{A}_j = [a_j, b_j]$ based on the three possible fractionality levels of the two endpoints (9 = 3 × 3 cases) and other properties. This rounding process is formally described in Figure 1, and is carefully designed so that the following two lemmas hold.

Lemma 3.1. *\mathcal{A} is a connected (integral) allocation.*

Given $i \in N$ and an integral bundle $S = [s, t] \subseteq [m]$, let $v_i^+(S) = \max\{v_i([s, t]), v_i([s+1, t]), v_i([s, t-1])\}$ and $v_i^-(S) = \min\{v_i([s, t]), v_i([s+1, t]), v_i([s, t-1])\}$; $v_i^+(S)$ and $v_i^-(S)$ represent, respectively, the maximum and the minimum valuation that agent i can obtain from S , after possibly removing one of its endpoint items.

Lemma 3.2. *For any $h, i, j \in [n]$, the connected (integral) allocation \mathcal{A} satisfies $v_i^-(A_j) \leq \tilde{v}_i(\tilde{A}_j(\mathbf{x}_h^*)) \leq v_i^+(A_j)$.*

These lemmas yield the following theorem.

Theorem 3.3. *\mathcal{A} is an EQIP $_g^c$ allocation, if valuations are non-negative.*

Proof. First, \mathcal{A} is a connected allocation by Lemma 3.1. Next, we show the EQIP $_g^c$ guarantee. As observed above, the full coloring of simplex Δ^* implies that each $i \in [n]$ is one of the indices $j \in [n]$ that maximize $\tilde{v}_j(\tilde{A}_j(\mathbf{x}_{\sigma(i)}^*))$ (i.e., agent i has the highest virtual valuation in allocation $\tilde{A}(\mathbf{x}_{\sigma(i)}^*)$). Thus, for any $i, j \in N$, we have $v_i^+(A_i) \geq \tilde{v}_i(\tilde{A}_i(\mathbf{x}_{\sigma(i)}^*)) \geq \tilde{v}_j(\tilde{A}_j(\mathbf{x}_{\sigma(i)}^*)) \geq v_j^-(A_j)$, where the second inequality follows from the above observation, and the first and last inequalities follow from Lemma 3.2. Since $v_i^+(A_i) \geq v_j^-(A_j)$ for any $i, j \in N$, we conclude that \mathcal{A} satisfies the EQIP $_g^c$ guarantee (i.e., equitability is obtained by removing at most one chore from the boundary of A_i and one good from the boundary of A_j). \square

Efficient Computation. The EQIP $_g^c$ allocation guaranteed by Theorem 3.3 can be computed by a polynomial-time algorithm. The algorithm first computes the set C_v of

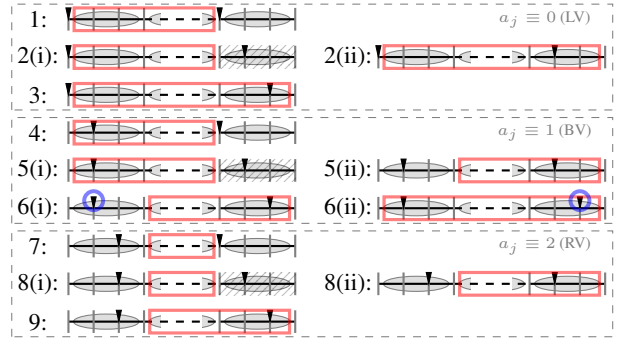


Figure 1: Given $j \in [n]$, we describe how the fractional bundle $\tilde{A}_j = [a_j, b_j]$ of the main allocation of Δ^* can be rounded to obtain the integral bundle A_j in each of the following nine cases, assuming that A_{j+1}, \dots, A_n have already been determined:

- 1: $a_j \equiv 0, b_j \equiv 0$: $A_j \leftarrow \llbracket a_j^-, b_j^- \rrbracket$;
- 2: $a_j \equiv 0, b_j \equiv 1$: (i) $A_j \leftarrow \llbracket a_j^-, b_j^- \rrbracket$ if $b_j^+ \in A_{j+1}$, and (ii) $A_j \leftarrow \llbracket a_j^-, b_j^+ \rrbracket$ if $b_j^+ \notin A_{j+1}$;
- 3: $a_j \equiv 0, b_j \equiv 2$: $A_j \leftarrow \llbracket a_j^-, b_j^+ \rrbracket$;
- 4: $a_j \equiv 1, b_j \equiv 0$: $A_j \leftarrow \llbracket a_j^-, b_j^- \rrbracket$;
- 5: $a_j \equiv 1, b_j \equiv 1$: (i) $A_j \leftarrow \llbracket a_j^-, b_j^- \rrbracket$ if $b_j^+ \in A_{j+1}$, and (ii) $A_j \leftarrow \llbracket a_j^+, b_j^+ \rrbracket$ if $b_j^+ \notin A_{j+1}$;
- 6: $a_j \equiv 1, b_j \equiv 2$: (i) $A_j \leftarrow \llbracket a_j^+, b_j^+ \rrbracket$ if Δ^* is *left-first*, and (ii) $A_j \leftarrow \llbracket a_j^-, b_j^+ \rrbracket$ if Δ^* is *right-first*, where left-first (resp. right-first) means that, in the sequence $\tilde{A}(\mathbf{x}_1^*), \dots, \tilde{A}(\mathbf{x}_n^*)$, the left (resp. right) knife delimiting the j -th bundle is the first to move rightward;
- 7: $a_j \equiv 2, b_j \equiv 0$: $A_j \leftarrow \llbracket a_j^+, b_j^- \rrbracket$;
- 8: $a_j \equiv 2, b_j \equiv 1$: (i) $A_j \leftarrow \llbracket a_j^+, b_j^- \rrbracket$ if $b_j^+ \in A_{j+1}$, (ii) $A_j \leftarrow \llbracket a_j^+, b_j^+ \rrbracket$ if $b_j^+ \notin A_{j+1}$;
- 9: $a_j \equiv 2, b_j \equiv 2$: $A_j \leftarrow \llbracket a_j^+, b_j^+ \rrbracket$.

The figure illustrates each of the nine cases as follows: each item and its three associated fractionality levels are represented by an ellipse divided into three parts; the two black triangles in each case represent the positions (a_j and b_j) of the two knives that determine the j -th fractional bundle $\tilde{A}_j = [a_j, b_j]$ (by possibly cutting the two boundary items of the considered bundle); the red rectangle encloses all items that are fully included in the j -th bundle A_j of the rounded (integral) allocation \mathcal{A} ; the right-hand item, when marked with crossed lines, indicates that it was included in bundle A_{j+1} during the previous step of the rounding procedure; the blue circle on the left (resp. right) knife indicates that bundle $[a_j, b_j]$ is left-first (resp. right-first) in Δ^* .

the valuations $v_i(S)$ that each agent i has for any bundle S (in $O(nm^2)$ time) and then, by dynamic programming, determines for each $c \in C_v$ if there exists an allocation \mathcal{A} such that $v_i^+(A_i) \geq c \geq v_i^-(A_i)$ for any $i \in [n]$ (in $O(nm^2)$ time), where $v_i^+(S)$ and $v_i^-(S)$ denote the maximum and the minimum valuation that i can obtain from a bundle S by deleting at most one item from its boundary;

again, we restrict ourselves to allocations where the i -th left-most bundle is assigned to agent i . We show that finding a value $c \in C_v$ satisfying the above condition is equivalent to finding an EQIP $_g^c$ allocation, whose existence is ensured by Theorem 3.3. Hence, we obtain the following theorem:

Theorem 3.4. *If valuations are non-negative, an EQIP $_g^c$ allocation can be found in time $O(n^2m^4)$.*

EFIP $_g^c$ Allocations

To show the existence of EFIP $_g^c$ allocations, we employ the same framework as in the equitability case, with minor modifications. We use the same triangulation T as in the previous case but equip it with n distinct coloring functions L_1, \dots, L_n , instead of the single coloring function L used earlier. Here, each L_i colors any vertex in $V(T)$ with the index j of the bundle that agent i prefers under virtual valuation \tilde{v}_i (defined as in the previous case); we note that each L_i is special, as the empty bundle is the least valuable.

By applying the Generalized Sperner's Lemma (Theorem 3.2), we show the existence of a jointly fully-colored elementary $(n-1)$ -simplex Δ^* . As in the previous case, this simplex corresponds to a sequence of n connected fractional partitions, but now the bundles are initially unallocated, and there exist two permutations σ and τ such that, in the $\sigma(i)$ -th allocation, agent $i \in [n]$ does not envy any other agent if i receives the $\tau(i)$ -th bundle. Then, by applying the same rounding procedure used for the EQIP $_g^c$ case, the first fractional partition of Δ^* is converted into an integral EFIP $_g^c$ allocation, where each agent i receives the $\tau(i)$ -th bundle. This leads to the following theorem:

Theorem 3.5. *Under non-negative valuations, an EFIP $_g^c$ allocation always exists.*

We conjecture that computing an EFIP $_g^c$ allocation is PPAD-complete, similarly to the result shown in (Deng, Qi, and Saberi 2012); furthermore, even without path constraints, the complexity of finding EF1 or EF1 $_g^c$ remains open. In the subclass of monotone non-decreasing valuations (i.e., goods only), Theorems 3.3–3.5 extend to (the stronger) EF1 and EQ1 under path constraints, thus recovering the results of Bilò et al. (2022); Igarashi (2023); Misra et al. (2021); Suksompong (2019).

4 Non-positive Valuations

To address the case of non-positive valuations under path-connectivity constraints, we resort to a novel multi-coloring variant of Sperner's Lemma, where the underlying coloring functions assign, to each vertex $x \in V(T)$, a set of colors (rather than a single color), including the indices i of the $(n-2)$ -dimensional faces F_i to which x belongs.

Multi-coloring Sperner's Lemma. Let T be a fixed triangulation of an $(n-1)$ -simplex $\Delta = \text{conv}(v_1, \dots, v_n)$. A multi-coloring function of T is a mapping $\mathcal{L} : V(T) \rightarrow 2^{[n]} \setminus \{\emptyset\}$ that assigns a non-empty subset of colors $\mathcal{L}(x) \subseteq [n]$ to each vertex of $x \in V(T)$. We recall that F_i is the $(n-2)$ -dimensional face of Δ opposite to vertex v_i . A multi-coloring function \mathcal{L} is called *special* if, for any vertex $x \in V(T)$, $\mathcal{L}(x) \supseteq \{i \in [n] : x \in F_i\}$ holds (i.e., if

x is a boundary vertex, the set of colors $\mathcal{L}(x)$ contains the indices associated with all $(n-2)$ -faces of Δ on which x is located). We observe that, if \mathcal{L} is a special multi-coloring function and F is a $(k-1)$ -face of Δ spanned by vertices v_{i_1}, \dots, v_{i_k} , it holds that $\mathcal{L}(x) \supseteq [n] \setminus \{i_1, \dots, i_k\}$ for any vertex $x \in V(T)$ that lies on F . An elementary $(n-1)$ -simplex $\Delta^* = \text{conv}(x_1^*, \dots, x_n^*) \in T$ is said to be *fully-colored* under a multi-coloring function \mathcal{L} if there exists a permutation $\sigma : [n] \rightarrow [n]$ such that $i \in \mathcal{L}(x_{\sigma(i)}^*)$ for any $i \in [n]$ (that is, a distinct color i appears in the set $\mathcal{L}(x_{\sigma(i)}^*)$ associated with a distinct vertex $x_{\sigma(i)}^*$).

Theorem 4.1 (Multi-coloring Sperner's Lemma). *Let T be a triangulation of an $(n-1)$ -simplex Δ , where $n \geq 2$, and let \mathcal{L} be a special multi-coloring function of T . Then, there exists a fully-colored elementary $(n-1)$ -simplex $\Delta^* \in T$ under multi-coloring function \mathcal{L} .*

The Multi-coloring Sperner's Lemma can be viewed as the dual of the standard one: whereas the classical version forbids color i on face F_i , the multi-coloring version requires it to appear in $\mathcal{L}(x)$ for every $x \in F_i$.

EQIP $_g^c$ and EFIP $_g^c$ Allocations

To show the existence of an EQIP $_g^c$ allocation for non-positive valuations, we use the same triangulation T as in the case of non-negative one, but we equip T with the multi-coloring function \mathcal{L} that assigns to each vertex $x \in V(T)$ the set $\mathcal{L}(x)$ of indices j which maximize $\tilde{v}_j(A_j(x))$, where \tilde{v}_j is the virtual valuation defined as in Section 3. Unlike the case of non-negative valuations, in this case the empty bundle is always the best for each agent. Thus, \mathcal{L} is a special multi-coloring function and, by the Multi-coloring Sperner's Lemma (Theorem 4.1), there exists an elementary $(n-1)$ -simplex $\Delta^* = \text{conv}(x_1^*, \dots, x_n^*)$ that is fully-colored under the multi-coloring function \mathcal{L} . As in the case of non-negative valuations, this means that there exists a permutation σ such that each agent $i \in [n]$ attains the highest valuation among all agents in allocation $\mathcal{A}(x_{\sigma(i)}^*)$ (where each agent j receives the j -th bundle). From this point onward, we can apply the same approach used for non-negative valuations to transform Δ^* into an EQIP $_g^c$ allocation.

Theorem 4.2. *Under non-positive valuations, an EQIP $_g^c$ allocation always exists and is polynomial-time computable.*

As in the non-negative case, we generalize the multi-coloring Sperner's Lemma to handle n distinct multi-coloring functions, each representing the virtual valuation of an agent i . This yields a jointly fully-colored simplex Δ^* , where each vertex corresponds to an allocation in which a distinct agent prefers a distinct bundle. Applying the rounding procedure then provides the desired EFIP $_g^c$ allocation.

Theorem 4.3. *Under non-positive valuations, an EFIP $_g^c$ allocation always exists.*

The following corollary holds since, in the case of monotone non-increasing valuations, there are chores only.

Corollary 4.1. *Under non-increasing valuations, EQ1 and EF1 allocations always exist, even under path constraints, with the former being computable in polynomial time.*

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AI Use Declaration

ChatGPT (OpenAI) was used exclusively for minor language polishing and figure formatting. All scientific content, analyses, and data were fully conceived and prepared by the authors.

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