

Processing Reserves Simultaneously*

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Abstract

Reserves allow combining competing objectives to allocate scarce resources based on priority. For example, schools may reserve some seats for students from underprivileged backgrounds or hospitals may reserve some ventilators for frontline health workers. An important determinant of the outcome is the order in which reserve categories are processed: categories processed later generally matter more than those processed earlier. The reason is that an agent who qualifies for multiple categories counts towards the quota of whichever category is processed first. I propose a new solution that processes reserve categories simultaneously so that if an agent qualifies for n categories, she takes $1/n$ units from each of them. That approach treats categories symmetrically and offers greater transparency: the relative importance of categories is entirely captured by the size of their quotas.

Keywords: rationing problem, reserve system, simultaneous processing, equal sharing

JEL Classifications: C62, C78, D47, D61, D63.

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1 Introduction

Reserve systems have been developed to solve a variety of real-world rationing problems, including the allocation of medical resources (Pathak, Sönmez, Ünver, and Yenmez, 2020), school seats (Dur, Kominers, Pathak, and Sönmez, 2018; Dur, Pathak, and Sönmez, 2020), and immigration visas (Pathak, Rees-Jones, and Sönmez, 2020a). What these problems have in common is that there are a certain number of identical and indivisible *units* (e.g., medical resources, school seats, immigration visas) and a certain number of *agents* (e.g., patients, students, applicants), each of whom desires one unit. If there are fewer units than agents, some rationing rule must be used to determine which agents are allocated a unit. A simple solution consists in ranking agents in order of priority and allocating units to the highest-priority agents. Patients are for example often prioritized based on their medical situation (Pathak, Sönmez, Ünver, and Yenmez, 2020) and schools often prioritize students based on where they live or whether they have a sibling attending the school (Abdulkadiroglu and Sönmez, 2003). In a reserve system, the priority may vary from one unit to another: units are split into (*reserve*) *categories* and each category has its own priority order over agents. The additional flexibility provided can help achieving a range of objectives, including promoting diversity (Dur, Pathak, and Sönmez, 2020), implementing affirmative action (Sönmez and Yenmez, 2019), or reconciling competing ethical value (Pathak, Sönmez, Ünver, and Yenmez, 2020).

In rationing problems with reserves, an allocation is typically obtained by processing reserve categories sequentially. Categories are processed one at a time following a *precedence order* and allocate their *quotas* (i.e., the number of units reserved) to the agents highest on their respective priority order, among those who have not yet been allocated a unit. With sequential processing, the order in which reserve categories are processed can impact the allocation. To see this, suppose that an agent has the highest priority for multiple categories. Whichever one of them is processed first will allocate one unit of its quota to that agent; hence the other categories no longer have to allocate a unit to that agent by the time they are processed and can instead allocate a unit to their next priority agent. As it turns out, the impact of the precedence order is far from negligible and can be of similar order of magnitude as the size of the quotas (Dur, Kominers, Pathak, and Sönmez, 2018; Dur, Pathak, and Sönmez, 2020). In fact, the reserve system in place for Boston’s public schools was abandoned in 2013, in large part due to concerns over the lack of transparency associated with the processing order (Dur, Kominers, Pathak, and Sönmez, 2018). In the H1-B visa allocation program in the United States, procedure changes made for logistical reasons in 2005 and 2009 had unintended consequences on the outcome (Pathak, Rees-Jones,

and Sönmez, 2020a). In an experimental paper, Pathak, Rees-Jones, and Sönmez (2020b) found that participants tend to correctly account for all decision-relevant parameters but make mistakes as a result of not understanding that reversing the processing order affects the outcome.

I propose a new solution to rationing problem with reserves that processes reserve categories simultaneously. Categories simultaneously allocate units to their highest-priority agents, up to their quotas. If an agent is allocated a unit from, say, n categories, she will only take $1/n$ units of capacity from each of them, allowing these categories to allocate more capacity to agents further down their respective priority orders. The solution I develop has at least two advantages over sequential processing. First, it is more transparent as the outcome depends on the categories' priorities and quotas but not on any precedence order. Second, while sequential processing yields an extreme solution in which an agent qualified for multiple categories counts entirely towards the quota of one category, simultaneous processing treats categories symmetrically and yields a solution in which an agent qualified for multiple categories counts equally towards each of their quotas.

I introduce the *Simultaneous Reserves (SR) algorithm* (Algorithm 1). In each round, categories allocate their quotas to their respective higher-priority agents. If an agent is allocated more than one unit in aggregate (over all categories), then the amount she receives from each category is reduced until she is allocated exactly one unit in aggregate. As a result, some categories have additional capacity, which they can allocate in the next round to agents further down their respective priority orders. Once no category has any additional capacity to allocate, the algorithm has found an allocation. While the SR algorithm may run for infinitely many rounds without finding an allocation, I show that it always converges to one (Theorem 1). I call the allocation to which the SR algorithm converges the *Simultaneous Reserves (SR) allocation*. The SR allocation specifies how much capacity each category allocates to each agent. In contrast to most of the literature, these numbers do not have to be binary so an agent can be allocated one unit in aggregate but receive parts of that unit from different categories. Some agents may be allocated in aggregate an amount of capacity strictly between zero and one. Thanks to the Birkhoff-von Neuman Theorem (Birkhoff, 1946; Von Neumann, 1953), that number can be interpreted as the probability that the agent will be allocated a unit. Moreover, the number of agents in that situation cannot exceed the number of categories (which in practice tends to be small relative to the number of agents).

I analyze the properties of the SR allocation. I show that it satisfies three standard axioms introduced by Pathak, Sönmez, Ünver, and Yenmez (2020)—*compliance with eligibility requirements*, *non-wastefulness*, and the *respect of priorities*—as well as a fourth one that I call *respect of equal sharing* (Theorem 2). An allocation respects equal sharing if every agent

who qualifies for multiple categories receives the same amount of capacity from all of them. I show that the SR allocation is not the only allocation to satisfy all four axioms; however, any other allocation that does generates the same *aggregate allocation*. That is, under any two allocations that satisfy the four axioms, every agent is allocated the same amount of capacity (Theorem 3). Therefore, differences among allocations that satisfy all four axioms amount to a matter of accounting but do not have any tangible impact for agents. Among the allocations that satisfy all four axioms, I characterize the SR allocation to be the one in which agents require the least capacity from any given category (Theorem 4).

The fact that the SR algorithm may run for infinitely many rounds constitutes a clear impediment to practical application. Following a similar approach to that of Kesten and Ünver (2015), I use linear programming to speed up the SR algorithm. The resulting *Simultaneous Reserves algorithm with Linear Programming (SRLP algorithm)* produces the SR allocation in finitely many rounds and polynomial time (Theorem 5).

Related Literature

Pioneered by Kominers and Sönmez (2016), the literature on reserves has flourished in the past few years and spanned a wide range of applications, including health care rationing (Pathak, Sönmez, Ünver, and Yenmez, 2020), school choice (Dur, Kominers, Pathak, and Sönmez, 2018; Dur, Pathak, and Sönmez, 2020), visa applications (Pathak, Rees-Jones, and Sönmez, 2020a), and university admission in India (Sönmez, Yenmez, et al., 2019; Sönmez and Yenmez, 2019; Aygün and Turhan, 2020; Aygun and Turhan, 2020), Brazil (Aygün and Bó, 2020), and Germany (Westkamp, 2013). As a whole, that literature has offered a large number of theoretical and practical insights into rationing problems with reserves. In particular, it has shed light on central role that the precedence order plays and its possible unintended consequences. This paper proposes a new approach for rationing problems with reserves, one that does not use any precedence order. Yilmaz (2020) considers a rationing problem with reserves and develops a solutions concept independent of any precedence order that satisfies basic axioms and is as egalitarian as possible in terms of the probability that each agent has of being allocated a unit. The SR allocation satisfies Yilmaz’s (2020) axioms but pursues a different goal of treating categories symmetrically. In fact, most agents (all but at most the number of categories) receive either zero or one unit.

The present paper also relates to the literature on random assignment and the probabilistic serial assignment, initiated by Hylland and Zeckhauser (1979) and Bogomolnaia and Moulin (2001), and generalized by Budish, Che, Kojima, and Milgrom (2013). The SR allocation resembles a random assignment in the sense that each category allocates to each agent a capacity between zero and one. However, agents do not “eat” categories in the hope to be

assigned one of them. Rather, categories allocate capacity to agents based on their priority orders and, when an agent is allocated one unit in aggregate, that unit may be shared among different categories. Kesten and Ünver (2015) consider a school choice model with priority ties. As a result, a school may assign to a student an amount of capacity strictly between zero and one and a student may be allocated parts of a seat by different school. An important difference is that students have preferences over schools and must ultimately be assigned to one of them while, in this paper, agents do not have preferences over categories and can be assigned parts of a unit from different categories.

The remainder of the paper is organized as follows. Section 2 presents the setup and the four axioms. Section 3 introduces the SR algorithm and analyzes the properties of the SR allocation. Section 4 presents the SRLP algorithm and shows that it produces the SR allocation in polynomial time. Section 5 concludes and all proofs are in the appendix.

2 Preliminaries

2.1 Setup

There are a set of **agents** a , a set of **(reserve) categories** c , and $q \in \mathbb{Z}_{>0}$ identical and indivisible **units**. Each category c has a **quota** $q_c \in \mathbb{R}_{\geq 0}$ with $\sum_{c \in C} q_c = q$. For each category c , there is a linear **priority order** π_c over the set of agents and an eligibility threshold \emptyset . Agent a is **eligible** for category c if $a\pi_c\emptyset$. For every agent a and every category c , I denote by $\hat{A}_{a,c} = \{a' \in A : a'\pi_c a\}$ the set of agents who have a higher priority than a for c and by $\check{A}_{a,c} = \{a' \in A : a\pi_c a'\}$ the set of agents who have a lower priority than a for c . A **rationing problem** is a tuple $R = (A, C, (\pi_c)_{c \in C}, (q_c)_{c \in C})$ specifying a set of agents, a set of categories, and for each category a priority order and a quota. I fix an arbitrary rationing problem R throughout the paper. I say that the rationing problem R has **soft reserves** if every agent is eligible for every category, i.e., if $a\pi_c\emptyset$ for every $a \in A$ and every $c \in C$; otherwise, I say that R has **hard reserves**.

A **(random) allocation** is a $|A| \times |C|$ matrix $\xi = (\xi_{a,c})_{a \in A, c \in C}$ such that, for every agent a and every category c , (i) $\xi_{a,c} \in [0, 1]$, (ii) $\sum_{a \in A} \xi_{a,c} \leq q_c$, and (iii) $\sum_{c \in C} \xi_{a,c} \leq 1$. In words, each element $\xi_{a,c}$ specifies the amount of capacity (between zero and one) that category c allocates to agent a , each category allocates an amount of capacity no larger than its quota, and each agent is allocated at most one unit overall. For every agent a , I denote by $\xi_a = \sum_{c \in C} \xi_{a,c}$ the total amount of capacity allocated to a at the allocation ξ and denote by $\rho(\xi) = (\xi_a)_{a \in A}$ the **aggregate allocation generated by the allocation** ξ . As units are

indivisible, every agent must ultimately be allocated either zero or one unit; hence ξ_a can be interpreted as the probability that a will be allocated a unit.¹

2.2 Rationing Problems in Practice

The rationing problem described in this paper can model a variety of real world situations, such as the allocation of a medical resource, school seats, and immigration visa. In the medical rationing problem of Pathak, Sönmez, Ünver, and Yenmez (2020), each unit is a unit of a medical resource, for example a ventilator or a dose of a vaccine, while each agent is a patient who requires that medical resource. There is a general category, in which patients are prioritized based on their medical situation (typically, based on expected health outcomes and survival probabilities) Pathak, Sönmez, Ünver, and Yenmez (2020) also propose three possible categories, each of which would prioritize a group of patients. An *essential personnel* category would prioritize those patients whose activity is essential during a health emergency, for example frontline health workers, a *disadvantaged* category would prioritize patients in groups that are particularly affected by the crisis, and a *Good Samaritan reciprocity category* would prioritize agents whose selfless acts have saved lives in the past, for example by donating a kidney to a stranger, donating a large amount of blood, or participated in clinical trials. In each of those categories, patients would first be ranked based on whether or not they are part of the target group and then based on their medical situation. Last, Pathak, Sönmez, Ünver, and Yenmez (2020) argue that prioritizing patients based on their medical situation can be discriminatory to certain groups who might not have access at all to the medical resource as a result. They propose to create a *disabled* category that does not take the general medical situation into account but rather prioritizes patients who have a disability and breaks ties with a lottery. It is natural to think of a medical rationing problem as one with soft resource because it is typically better to allocate a unit to a low-priority patient than not at all. In contrast to the rest of the literature, I do not restrict quotas to being integers (even though the total number of units available is an integer).

School choice constitutes another application: each unit is a seat at a given school and each agent is a student who would like to attend that school. Until 2013, Boston had a reserve system with two categories, each of which had a quota equal to half of the total number of seats (Dur, Kominers, Pathak, and Sönmez, 2018). The general category prioritized students who had a sibling attending the schools and then broke ties with a lottery while the *walk*

¹The Birkhoff-von Neumann Theorem (Birkhoff, 1946; Von Neumann, 1953) guarantees the existence of a lottery such that every agent a has a probability ξ_a of being allocated a unit. Moreover, I show that the number of agents who are allocated a total capacity strictly between zero and one is at most equal to the number of categories.

zone category prioritized students who lived within walking distance of the school and broke ties based on the general category priority. In Chicago (Dur, Pathak, and Sönmez, 2020), the city’s neighborhoods are split into four *tiers* based on socioeconomic factors. There is a general category that prioritizes students based on merit and and four tier-specific categories (one per tier), which prioritize students who live in one of the tier’s neighborhoods. The general category’s quota is equal to 30% of the school’s seats and each of the tier-specific categories has a quota equal to 17.5% of the school’s seats. School choice constitutes another example of soft reserves: all seats are allocated as long as there are at least as many students as there are seats.

Pathak, Rees-Jones, and Sönmez (2020a) document a reserve system for the allocation of H-1B visas in the United States. In that application, each unit is a H-1B visa and each agent is a visa applicant and there are two categories: a general category with a quota of 65,000 visas and an *advanced-degree* category with a quota of 20,000 visas. There are two separate lotteries, each of which determines the priority order of one of the categories. Only applicants with an advanced degree are eligible for the advanced-degree category; therefore, the allocation of H-1B visas is an example of a rationing problem with hard reserves: even if the advanced degree category allocates fewer than 20,000 visas, the remaining ones cannot be allocated to applicants who do not have an advanced degree.

Last, a practically relevant aspect of the model is worth a mention. While the total number of units is an integer, there is no such restriction on the categories’ quotas. To the best of my knowledge, this paper is the first to provide that feature in the context of reserves. The additional flexibility may prove useful in practice, particularly when the number of units is small; for instance, if the policy is to allocate 30% of ventilators to essential personnel and 70% to the general population, a hospital with 5 ventilators can set the quotas to 1.5 and 3.5.

2.3 Desirable properties for an allocation

Pathak, Sönmez, Ünver, and Yenmez (2020) introduce three axioms that an allocation should satisfy: compliance with eligibility requirements, non-wastefulness, and respect of priorities. An important difference between my setting and theirs is that Pathak, Sönmez, Ünver, and Yenmez (2020) consider allocations (or *matching* in their terminology) in which each element is either zero or one; that is, each agent allocated either zero units or one unit from exactly one category. I generalize the three properties of Pathak, Sönmez, Ünver, and Yenmez (2020) to my setting and introduce a fourth one—respect of equal sharing—that lies at the heart of the solution I propose.

Axiom 1. *An allocation ξ is **complies with eligibility requirements** if, for every agent a and every category c such that a is not eligible for c , $\xi_{a,c} = 0$.*

The first axiom requires that agents be only allocated capacity by categories for which they are eligible. In a rationing problem with soft reserves (e.g., medical resource rationing or school choice), every allocation trivially complies with eligibility requirements; however Axiom 1 matters in the presence of hard reserves. For instance, in the H1-B visa program, Axiom 1 precludes applicants who do not have an advanced degree from being allocated one of the 20,000 visas reserved for advanced-degree applicants.

Axiom 2. *An allocation ξ is **non-wasteful** if, for every category c such that $\sum_{a \in A} \xi_{a,c} < q_c$ and every agent a such that $\xi_a < 1$, a is not eligible for c .*

The second axiom states that whenever a category has not allocated its full quota, then none of the remaining capacity can benefit an agent who is eligible for that category, as that capacity would then be wasted. In a rationing problem with soft reserves, Axiom 2 requires that either all agents be allocated a unit or all units be allocated, that is $\sum_{a \in A} \xi_a = \min\{|A|, q\}$. With hard reserves, some categories may allocate an amount of capacity smaller than their quotas as long as every agent who could has not been allocated one unit is ineligible for those categories. In the H1-B visa program, an allocation in which all advanced-degree applicants have received a visa is non-wasteful, even if fewer than the 20,000 visas reserved for them have been allocated and some applicants without an advanced degree

Axiom 3. *An allocation ξ **respects priorities** if for every agent a such that $\xi_a < 1$, every category c and every lower-priority agent $a' \in \check{A}_{a,c}$, $\xi_{a',c} = 0$.*

The third axiom ensures that each category allocates its capacity based on priority, that is an agent can only be allocated capacity from a category if all higher-priority agents have been allocated a unit. As Pathak, Sönmez, Ünver, and Yenmez (2020, p.13) note, Axioms 1-3 are widely accepted as properties that a good allocation should possess:

“As far as we know, in every real-life application of a reserve system each of these three axioms are either explicitly or implicitly required. Hence, we see these three axioms as a minimal requirement for reserve systems.”

While Axioms 1-3 do narrow down the set of allocations that should be considered, they leave many possible candidates. In particular, these axioms are silent on a key question of reserve allocation: if an agent qualifies to be allocated a unit from multiple categories, from which one(s) will she receive it? The common solution both in practice and in the literature is to use a *sequential reserve algorithm* in which categories are processed one at a time and

allocate, up to their quota, one unit of capacity to the highest-priority eligible agents who have not yet been allocated a unit (see Pathak, Sönmez, Ünver, and Yenmez (2020, p.17) for a full description of that procedure). The implication is that if an agent qualifies for multiple categories, she will receive a unit from whichever is processed first; hence the categories processed early will tend to allocate units to agents who also qualify for other categories. At the heart of my proposed solution is the idea that, while units are ultimately indivisible, how much capacity categories allocate to agents is merely an accounting exercise; therefore, an agent allocated one unit overall can receive parts of that unit from multiple categories. The fourth axiom, which is new to this paper, stipulates that the unit that an agent receives should be shared equally among the categories for which she qualifies.

Axiom 4. *An allocation ξ **respects equal sharing** if, for every agent a and every category c such that a is eligible for c and $\xi_{a,c} < \max_{c' \in C} \{\xi_{a,c'}\}$, $\xi_{a,c} + \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c} = q_c$.*

Axiom 4 ensures that each agent receives the same amount of capacity from every category that has the capacity to do so.

3 The Simultaneous Reserves (SR) allocation

In this section, I propose an algorithm that processes reserve categories simultaneously (as opposed to sequentially) and produces an allocation that satisfies Axioms 1-4. I call that allocation the Simultaneous Reserves (SR) allocation and show that any other allocation satisfying these axioms generates the same aggregate allocation.

3.1 The Simultaneous Reserves (SR) algorithm

The Simultaneous Reserves (SR) algorithm is formally described in Algorithm 1. In order to describe the algorithm and analyze its properties, it is useful to define the concept of a **pre-allocation**, which is identical to an allocation but allows agents to be allocated more than one unit overall. Formally, a pre-allocation is a $|A| \times |C|$ matrix $x = (x_{a,c})_{a \in A, c \in C}$ such that, for every agent a and every category c , (i) $x_{a,c} \in [0, 1]$, (ii) $\sum_{a \in A} x_{a,c} \leq q_c$. I denote by $x_a = \sum_{c \in C} x_{a,c}$ the total amount of capacity allocated to agent a at the pre-allocation x . Axioms 1-4 are defined analogously over pre-allocations.

At the start of the SR algorithm, each agent has a *demand* of 1. This can be interpreted as the largest amount that an agent could require from any category; as every agent requires one unit overall, the demand starts at 1 but may fall throughout the algorithm as agents are allocated capacity.

Algorithm 1: SIMULTANEOUS RESERVES (SR)

Initialization Set every agent's demand to one: $d_a^0 = 1$ for every agent a .

Round $i \geq 1$:

Capacity Allocation For every agent a and every category c , if a is eligible for c then $x_{a,c}^i = \min\{d_a^{i-1}, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}\}$ and otherwise $x_{a,c}^i = 0$.

Demand Adjustment For every agent a such that $x_a^i < 1$, set $d_a^i = 1$. For every agent a such that $x_a^i = 1$, set $d_a^i = \max_{c \in C} \{x_{a,c}^i\}$. For every agent a such that $x_a^i > 1$, set d_a^i such that $\sum_{c \in C} \min\{d_a^i, x_{a,c}^i\} = 1$.

The first round starts with the Capacity Allocation stage: each category allocates one unit of capacity to one agent at a time in decreasing order of priority until it has less than one unit of capacity left or has allocated one unit of capacity to every eligible agent, whichever comes first. The next agent receives the remaining capacity (which could be 0 or any number smaller than 1) and the remaining agents are not allocated any capacity.

The Capacity Allocation stage generates a pre-allocation x^1 , where for any agent-category pair (a, c) , $x_{a,c}^1$ is the amount of capacity that c has allocated to a . An agent may be allocated more than one unit overall so x^1 is a pre-allocation but not necessarily an allocation. The Demand Adjustment stage reduces the amount of capacity that agents demand in order to turn x^1 into an allocation. The Demand Adjustment stage does not affect agents who have not yet been allocated a unit in aggregate (i.e., a such that $x_a < 1$), those agents continue to demand one unit. The demand of an agent who has been allocated exactly one unit in aggregate (i.e., a such that $x_a = 1$) falls to the maximum capacity she is allocated from any category (i.e., $\max_{c \in C} \{x_{a,c}^i\}$). The rationale is that she does not require more from any category in order to be allocated one unit in aggregate so any additional capacity for which she qualifies can be allocated to the next agent on the priority order. The demand of an agent who has been allocated more than one unit in aggregate (i.e., a such that $x_a > 1$) falls even further so that she abandons any capacity she does not require and is only allocated one unit in aggregate (as $\sum_{c \in C} \min\{d_a^i, x_{a,c}^i\} = 1$).

Every subsequent Round i starts with a demand vector d^{i-1} and, in the Capacity Allocation stage, calculates a pre-allocation x^i . The highest-priority agents are allocated their demand until there is not enough capacity for the next agent. That agent receives whatever capacity remains and the lower-priority agents are not allocated any capacity. For any agent a , the expression $q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}$ represents the amount of capacity left for a once all

higher-priority agents have been allocated their demand. If it is zero or negative, then there is no capacity left for agent a so she is not allocated any capacity. If it is equal to or larger than d_a^{i-1} , then agent a is allocated her demand. If it is anything in between, then agent a is allocated that remaining capacity. In the Demand adjustment stage, d^i is calculated based on x^i and the algorithm continues in Round $i + 1$, in which x^{i+1} and d^{i+1} are calculated.

I next illustrate how the SR algorithm can generate an allocation.

Example 1. There are five agents and four categories, each with a quota of 1. Every agent is eligible for every category and the priorities are

$$\pi_{c_1} : a_1, a_2, a_3, \dots \quad \pi_{c_2} : a_1, a_2, a_4, \dots \quad \pi_{c_3} : a_1, a_3, \dots \quad \pi_{c_4} : a_4, a_5, \dots$$

The pre-allocation calculated in each of the first four rounds of the SR algorithm is displayed in Table 1. In Round 1, each category allocates one unit of capacity to its highest-priority agent. As agent a_1 is allocated a unit from three different category, her demand drops to $1/3$. In Round 2, categories c_1 , c_2 , and c_3 only allocate $1/3$ to a_1 , which leaves $2/3$ to allocate to their second highest-priority agents. As a result, a_2 is allocated $4/3$ in aggregate ($2/3$ from each of c_1 and c_2) and so her demand drops to $1/2$. In Round 3, c_1 and c_2 allocate $1/3$ to a_1 and $1/2$ to a_2 ; hence they have $1/6$ left to allocate to their third highest-priority agents, respectively a_3 and a_4 . Agent a_4 is now allocated $7/6$ in aggregate ($1/6$ from c_2 and 1 from c_4) so her demand drops to $5/6$. In Round 4, c_4 only needs to allocate $5/6$ to a_4 and can therefore allocate $1/6$ to its second highest-priority agent a_5 . Every agent is now allocated at most one unit so the Round 4 pre-allocation x^4 is in fact an allocation. It is easy to see that, in any subsequent round the SR algorithm continues to produce the same (pre-)allocation and demand vector. Hence, in Example 1, the SR algorithm produces the allocation

$$x^4 = \begin{matrix} & c_1 & c_2 & c_3 & c_4 \\ \begin{matrix} a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \end{matrix} & \begin{pmatrix} 1/3 & 1/3 & 1/3 & 0 \\ 1/2 & 1/2 & 0 & 0 \\ 1/6 & 0 & 2/3 & 0 \\ 0 & 1/6 & 0 & 5/6 \\ 0 & 0 & 0 & 1/6 \end{pmatrix} \end{matrix},$$

which generates the aggregate allocation

$$\rho(x^4) = \begin{pmatrix} a_1 & a_3 & a_3 & a_4 & a_5 \\ 1 & 1 & 5/6 & 1 & 1/6 \end{pmatrix}.$$

				<u>Round 1</u>								<u>Round 2</u>			
c_1	(1)	c_2	(1)	c_3	(1)	c_4	(1)	c_1	(1)	c_2	(1)	c_3	(1)	c_4	(1)
a_1	1	a_1	1	a_1	1	a_4	1	a_1	1/3	a_1	1/3	a_1	1/3	a_4	1
a_2	0	a_2	0	a_3	0	a_5	0	a_2	2/3	a_2	2/3	a_3	2/3	a_5	0
a_3	0	a_4	0					a_3	0	a_4	0				
				<u>Round 3</u>								<u>Round 4</u>			
c_1	(1)	c_2	(1)	c_3	(1)	c_4	(1)	c_1	(1)	c_2	(1)	c_3	(1)	c_4	(1)
a_1	1/3	a_1	1/3	a_1	1/3	a_4	1	a_1	1/3	a_1	1/3	a_1	1/3	a_4	5/6
a_2	1/2	a_2	1/2	a_3	2/3	a_5	0	a_2	1/2	a_2	1/2	a_3	2/3	a_5	1/6
a_3	1/6	a_4	1/6					a_3	1/6	a_4	1/6				

Table 1: SR Algorithm applied to Example 1

It is easy to verify that the allocation x^4 satisfies Axioms 1-4. At first sight, it might look as if x^4 does not respect equal sharing because a_4 is allocated 1/6 from c_2 and 5/6 from c_4 . However, this does not violate Axiom 4 as c_2 can only allocate 1/6 to a_4 after having allocated 1/3 to a_1 and 1/2 to a_2 ; formally, $x_{a_4, c_2}^4 + \sum_{a \in \hat{A}_{a_4, c_2}} x_{a, c_2}^4 = 1/6 + 1/3 + 1/2 = 1 = q_{c_2}$. The fact that the SR algorithm produces an allocation that satisfies Axiom 1-4 is not a coincidence. In the next two subsections, I formally define the outcome of the SR algorithm and show that it is an allocation satisfying Axioms 1-4.

3.2 Outcome of the SR algorithm

In Example 1, the SR algorithm finds an allocation after four rounds. In general, the SR algorithm may never reach an allocation; however its outcome is nevertheless well-defined.²

Theorem 1. *The SR algorithm converges to an allocation $\xi^{SR} = \lim_{i \rightarrow \infty} x^i$.*

I call ξ^{SR} the **Simultaneous Reserves (SR) allocation** and discuss its properties in Subsection 3.3. The remainder of this section is devoted to proving and providing intuition for Theorem 1. The SR allocation may reach an allocation (as it does in Example 1), in which case that allocation will be that outcome.

In Example 1, the SR algorithm finds an allocation after four rounds and continues to return the same allocation and demand vector in every subsequent round. As the next result shows, this always occurs once the SR algorithm has found an allocation.

Proposition 1. *Suppose that, in some Round i of the SR algorithm, x^i is an allocation. Then, for every $j > i$, $x^j = x^i$ and $d^j = d^i$.*

²In Section 4, I propose an outcome-equivalent algorithm that works in polynomial time and finds an allocation after fewer than $4|A||C|$ rounds.

		<u>Round 1</u>			<u>Round 2</u>						
c_1	(1)	c_2	(1)	c_3	(2)	c_1	(1)	c_2	(1)	c_3	(2)
a_1	1	a_3	1	a_1	1	a_1	1/2	a_3	1/2	a_1	1/2
a_2	0	a_2	0	a_3	1	a_2	1/2	a_2	1/2	a_3	1/2
a_3	0	a_1	0	a_2	0	a_3	0	a_1	0	a_2	1
a_4	0	a_4	0	a_4	0	a_4	0	a_4	0	a_4	0

		<u>Round 3</u>			<u>Round 4</u>						
c_1	(1)	c_2	(1)	c_3	(2)	c_1	(1)	c_2	(1)	c_3	(2)
a_1	1/2	a_3	1/2	a_1	1/2	a_1	5/12	a_3	5/12	a_1	5/12
a_2	1/3	a_2	1/3	a_3	1/2	a_2	1/3	a_2	1/3	a_3	5/12
a_3	1/6	a_1	1/6	a_2	1/3	a_3	3/12	a_1	3/12	a_2	1/3
a_4	0	a_4	0	a_4	2/3	a_4	0	a_4	0	a_4	5/6

	Round 4	Round 5	Round 6	Round 7	...	Round $i \geq 3$
x_{a_3, c_1}^i	3/12	7/24	15/48	31/96	...	$(2^{i-2} - 1)/(3 \cdot 2^{i-2})$
x_{a_1, c_2}^i	3/12	7/24	15/48	31/96	...	$(2^{i-2} - 1)/(3 \cdot 2^{i-2})$
x_{a_4, c_3}^i	5/6	11/12	23/24	47/48	...	$(3 \cdot 2^{i-3} - 1)/(3 \cdot 2^{i-3})$
$d_{a_1}^i$	9/24	17/48	33/96	65/192	...	$(2^{i-1} + 1)/(3 \cdot 2^{i-1})$
$d_{a_3}^i$	9/24	17/48	33/96	65/192	...	$(2^{i-1} + 1)/(3 \cdot 2^{i-1})$

Table 2: SR Algorithm applied to Example 2

The intuition for Proposition 1 is simple. If x^i is an allocation, then no agent is allocated more than one unit so the demand of every agent is at least what she is allocated from any category and so, in the next Round, every category continues to allocate the same capacity to every agent, i.e., $x^{i+1} = x^i$. In each round, the demand vector depends on the current round's pre-allocation and the pre-allocation depends on the previous round's demand vector; hence the SR algorithm continues to return the same (pre-)allocation and demand vector in every subsequent round.

Proposition 1 implies that the SR algorithm can stop once it finds an allocation. Unfortunately, as the next example shows, the SR algorithm may never reach an allocation.

Example 2. There are four agents and three categories. The priorities and quotas are

$$\pi_{c_1} : a_1, a_2, a_3, a_4, \emptyset \quad \pi_{c_2} : a_3, a_2, a_1, a_4, \emptyset \quad \pi_{c_3} : a_1, a_3, a_2, a_4, \emptyset \quad q_{c_1} = 1 \quad q_{c_2} = 1 \quad q_{c_3} = 2.$$

The working of the SR algorithm is displayed in Table 2. In Round 1, c_1 and c_2 each allocate one unit to their highest-priority agent, respectively a_1 and a_3 . Category c_3 has two units and allocates them to its two highest priority agents, a_1 and a_3 . Each of a_1 and a_3 is allocated a unit from two different categories; hence the demand of both agents drops to 1/2.

In Round 2, each of c_1 and c_2 has an extra half unit to allocation, which goes to their second highest-priority agent, a_2 while c_2 has an extra unit to allocate to its third-highest priority student, who is also a_2 . As a result, a_2 's demand drops to $1/3$. In Round 3, c_1 allocates $1/6$ to a_3 , c_2 allocates $1/6$ to a_1 , and c_3 allocates $2/3$ to a_4 .

At this point, the SR algorithm begins to cycle. Agent a_1 is allocated $7/6$ in aggregate (i.e., $x_{a_1}^3 = 7/6$) so her demand drops in Round 3. However, she can only be allocated $1/6$ from c_2 , meaning she needs to be allocated $5/12$ from each for c_1 and c_3 . It follows that a_1 's demand drops to $5/12$. Analogously, a_3 is allocated $7/6$ in aggregate and her demand drops to $5/12$ as well. In Round 4, as the result of the drop in a_1 and a_3 's demand (by $1/12$ each), c_1 allocates an extra $1/12$ to a_3 , c_2 allocates an extra $1/12$ to a_1 , and c_3 allocates an extra $1/6$ to a_4 . The extra $1/6$ of capacity that a_1 has released has benefit for half to a_4 (at c_3) and for half to a_3 (at c_1) while the extra $1/6$ of capacity that a_3 has released has benefit for half to a_4 (at c_3) and for half to a_1 (at c_2). As a result, each of a_1 and a_3 are allocated $13/12$ in aggregate in Round 4 so their demand drops to $9/24$. In Round 5, as in Round 4, half of the capacity released by a_1 and a_3 goes to a_4 (c_3 allocates an extra $1/12$ to a_3) and the other half comes back to a_1 and a_3 (c_1 allocates an extra $1/24$ to a_3 and c_2 allocates an extra $1/24$ to a_1). The SR algorithm continues to cycle forever, with the amount of capacity reallocated halving in each round. However, even though the SR algorithm never reaches an allocation, it converges to one:

$$\xi^{SR} = \begin{matrix} & c_1 & c_2 & c_3 \\ \begin{matrix} a_1 \\ a_2 \\ a_3 \\ a_4 \end{matrix} & \begin{pmatrix} 1/3 & 1/3 & 1/3 \\ 1/3 & 1/3 & 1/3 \\ 1/3 & 1/3 & 1/3 \\ 0 & 0 & 1 \end{pmatrix} \end{matrix}.$$

I now show that the SR algorithm always converges to an allocation even when it never reaches one. For every Round i of the SR algorithm, I construct the allocation $\xi^i = (\xi_{a,c}^i)_{a \in A, c \in C}$ such that, for every agent a and every category c , $\xi_{a,c}^i = \min\{d_a^i, x_{a,c}^i\}$. By construction, $\xi_a^i \leq 1$ for every agent a so ξ^i is indeed an allocation.³ I also define the matrix $z^i = x^i - \xi^i$ to be the **excess supply** in Round i of the SR algorithm. For every agent a , $z_a^i = \sum_{c \in C} z_{a,c}^i$ can be interpreted as the capacity that a is allocated in addition to the one unit she requires. I denote the total excess supply by $|z^i| = \sum_{a \in A} z_a^i = \sum_{a \in A} \sum_{c \in C} z_{a,c}^i$. In Example 2, the total excess supply is $1/3$ in Round 3 (each of a_1 and a_3 is allocated $7/6$) and is halved in every subsequent round; hence it converges to zero.

Proposition 2. *For every Round $i \geq 1$ of the SR algorithm, $|z^{i+1}| \leq |z^i| \leq |A|(|C| - 1)/i$.*

³I prove formally that ξ^i is an allocation in the Appendix (Lemma 3).

There is a relatively simple intuition for why the total excess supply decreases throughout the SR algorithm. If in some round an agent is allocated more than one unit in aggregate, that extra capacity is reallocated in the next round. It may be reallocated to an agent who was already allocated one unit in aggregate, in which case it continues to count towards the total excess supply, or to an agent who was not yet been allocated one unit in aggregate. In the latter case, as agents always keep the capacity they are allocated up to one unit, the extra capacity no longer counts towards the excess supply in any subsequent round. In Example 2 the excess supply is halved in every Round $i \geq 1$ because half of the excess supply is reallocated to a_1 and a_3 , who are already allocated one unit, and the other half is allocated to a_4 , who does not. The intuition for the upper bound is that, as excess supply is reallocated, categories allocate capacity to agents further down their priority order. Eventually, they must reach the bottom and so there is a bound on how much excess supply can be reallocated throughout the algorithm; hence the more rounds have occurred, the smaller the total excess supply must be in each round.

Proposition 2 implies that the total excess supply converges to zero. As every element of z^i is weakly positive, each of them must then also converge to zero. Therefore, x^i and ξ^i must converge to each other and we have the following corollary.

Corollary 1. $\lim_{i \rightarrow \infty} z^i = \mathbf{0}$ and $\xi^{SR} = \lim_{i \rightarrow \infty} \xi^i = \lim_{i \rightarrow \infty} x^i$.

Corollary 1 guarantees that ξ^{SR} is well defined. As ξ^i is an allocation in every Round i of the SR algorithm, it is natural to think its limit ξ^{SR} is an allocation as well. I formally show this in the appendix, which completes the proof of Theorem 1.

3.3 Properties of the SR allocation

Having defined the SR allocation, I now turn to its properties in regard to the axioms defined in Section 2.3.

Theorem 2. *The SR allocation satisfies Axioms 1-4.*

The key driver behind Theorem 2 is the Demand Adjustment stage in the SR algorithm. An agent's demand sets an upper bound on how much categories can allocate to that agent in subsequent rounds; thus, it ensures that all categories that would allocate at least that upper bound allocate the same amount to that agent. As a result, x^i respects equal sharing in each Round i . The Capacity Allocation stage ensures that x^i also satisfies Axioms 1-3. Theorem 2 follows from the fact that these properties continue to hold in the limit.

Theorem 2 makes the SR allocation an appealing solution for a rationing problem with reserves. It satisfies the natural requirements that are Axioms 1-3 and ensures that an agent

who qualifies for multiple categories affects equally the quota of each of those categories. This is in contrast to current solutions that assign each agent to one category. A natural question at this point is whether the SR allocation is the only one to possess. The next example shows that this is not the case; however I will show that any alternative allocation that satisfies Axioms 1-4 generates the same aggregate allocation.

Example 3. There are two agents and two categories, each with a quota of 1. The priorities are $\pi_{c_1} : a_1, a_2, \emptyset$ and $\pi_{c_2} : a_2, a_1, \emptyset$.

In Example 3, the SR algorithm reaches the allocation

$$\xi^{SR} = \begin{matrix} & c_1 & c_2 \\ \begin{matrix} a_1 \\ a_2 \end{matrix} & \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \end{matrix}$$

in Round 1: each category allocates one unit to its highest-priority agent and no demand adjustment is required. However, for any $\lambda \in [0.5, 1]$, the allocation

$$\xi^\lambda = \begin{matrix} & c_1 & c_2 \\ \begin{matrix} a_1 \\ a_2 \end{matrix} & \begin{pmatrix} \lambda & 1 - \lambda \\ 1 - \lambda & \lambda \end{pmatrix} \end{matrix}$$

satisfies Axioms 1-4. To see this, notice that ξ^λ trivially satisfies Axioms 1-3 since every agent is eligible for every category and $\xi_{a_1}^\lambda = \xi_{a_2}^\lambda = 1$. If $\lambda = 0.5$, then ξ^λ also trivially respects equal sharing since all four elements of ξ^λ are equal to 0.5. If $\lambda > 0.5$, then $\xi_{a_1, c_2}^\lambda < \xi_{a_1, c_1}^\lambda$ and $\xi_{a_2, c_1}^\lambda < \xi_{a_2, c_2}^\lambda$; however, Axiom 4 is not violated since $\xi_{a_1, c_2}^\lambda + \xi_{a_2, c_2}^\lambda = 1 = q_{c_2}$ and $\xi_{a_2, c_1}^\lambda + \xi_{a_1, c_1}^\lambda = 1 = q_{c_1}$.⁴

The SR allocation is not the unique allocation that satisfies Axioms 1-4; in fact, there may be infinitely many such allocations. However, one aspect of Example 3 is worth noting: for every $\lambda \in [0.5, 1]$, ξ^λ allocates one unit of capacity to each of a_1 and a_2 . That is, $\rho(\xi^\lambda) = \rho(\xi^{SR})$ for every $\lambda \in [0.5, 1]$. As it turns out, this is not specific to *Example 3*.

In order to formalize that idea, I introduce two pieces of terminology. I call the aggregate allocation $\rho(\xi^{SR})$ generated by the SR allocation the **SR aggregate allocation** and for any allocation ξ , I call ξ **SR equivalent** if it generates the SR aggregate allocation, that is if $\rho(\xi) = \rho(\xi^{SR})$.

Theorem 3. *Every allocation that satisfies Axioms 1-4 is SR equivalent.*

⁴If $\lambda < 0.5$, ξ^λ no longer respects equal sharing since $\xi_{a_1, c_1}^\lambda < \xi_{a_1, c_2}^\lambda$ and $\xi_{a_1, c_1}^\lambda < 1 = q_{c_1}$.

Corollary 2. *An aggregate allocation is generated by an allocation that satisfies Axioms 1-4 if and only if it is the SR aggregate allocation.*

Theorem 3 and Corollary 2 imply that, even though many allocations may satisfy Axioms 1-4, any difference among them is immaterial as every agent is allocated the same capacity in aggregate. The SR aggregate allocation stands apart as the only one to be generated by an allocation satisfying Axioms 1-4. Moreover, Theorem 3 and Corollary 2 are sharp in the sense that each of the four axioms is needed in order to obtain the SR aggregate allocation.

Proposition 3. *For each Axiom 1-4, there exists a rationing problem in which an allocation that is not SR equivalent satisfies the other three axioms.*

Finally, I return to the SR allocation and show that is characterized by Axiom 1-4 and an additional simple property. For any pre-allocation x and every agent a , I define $d_a(x)$ to be the **demand of agent a associated with x** , as defined in the SR algorithm. That is, if $x_a < 1$, then $d_a(x) = 1$; if $x_a = 1$, then $d_a(x) = \max_{c \in C} \{x_{a,c}\}$; and if $x_a > 1$, then $d_a(x)$ is such that $\sum_{c \in C} \min\{d_a(x), x_{a,c}\} = 1$. I denote by $d(x) = (d_a(x))_{a \in A}$ the vector containing all of the agents' demand associated with x . Note that for an allocation ξ , a 's demand simplifies to

$$d_a(\xi) = \begin{cases} 1 & \text{if } \xi_a < 1 \\ \max_{c \in C} \{\xi_{a,c}\} & \text{if } \xi_a = 1. \end{cases}$$

Theorem 4. *For every allocation $\xi^* \neq \xi^{SR}$ that satisfies Axioms 1-4, $d(\xi^*) < d(\xi^{SR})$.*

Theorem 4 allows fully characterizing the SR allocation: it is the allocation satisfying Axioms 1-4 with the largest demand associated with it. The intuition behind that result is relatively simple. The SR algorithm initially sets every agent's demand to one, the largest possible level. In each round, it calculates a pre-allocation that satisfies Axioms 1-4 and reduces the demands in order to eliminate the excess supply. Thus, the SR algorithm finds in each round an upper bound for the demand vector in any allocation that satisfies Axioms 1-4. The algorithm continues until the demand vector has been reduced just enough to find an allocation satisfying Axioms 1-4; hence it identifies the largest demand vector for which such an allocation exists.

4 Simultaneous Reserves with Linear Programming

A practical shortcoming of the SR algorithm is that it may run for infinitely many rounds. In this section, I propose an alternative algorithm that is outcome equivalent but runs in

Algorithm 2: SIMULTANEOUS RESERVES WITH LINEAR PROGRAMMING (SRLP)

Initialization Set every agent’s demand to one: $d_a^0 = 1$ for every agent a .

Round $i \geq 1$:

Linear Programming If either $i = 1$ or $i > 1$ and there exists an agent-category pair (a, c) such that either $x_{a,c}^{i-1} > x_{a,c}^{i-2} = 0$ or $x_{a,c}^{i-1} \geq d_a^{i-1}$ and $x_{a,c}^{i-2} < d_a^{i-2}$, set $\delta_a^i = d_a^{i-1}$ for every agent a . Otherwise, set $\delta_a^i = \delta_a^{LP}(x^{i-1}, d^{i-1})$ (calculated by Algorithm 3) for every agent a .

Capacity Allocation For every agent a and every category c , if a is eligible for c then $x_{a,c}^i = \min\{\delta^i, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} \delta_{a'}^i, 0\}\}$ and otherwise $x_{a,c}^i = 0$.

Termination If $x_a^i \leq 1$ for every agent a , end and output x^i .

Demand Adjustment For every agent a such that $x_a^i < 1$, set $d_a^i = 1$. For every agent a such that $x_a^i = 1$, set $d_a^i = \max_{c \in C} \{x_{a,c}^i\}$. For every agent a such that $x_a^i > 1$, set d_a^i such that $\sum_{c \in C} \min\{d_a^i, x_{a,c}^i\} = 1$.

polynomial time. The Simultaneous Reserves algorithm with Linear Programming (SRLP algorithm) is formally described in Algorithm 2. Its structure is similar to that of the SR algorithm but in some rounds it solves a linear programming problem (described in Algorithm 3) to update the demand of some agents. As a result, it is outcome equivalent but faster than the SR algorithm.

Theorem 5. *The SRLP algorithm produces ξ^{SR} after fewer than $4|A||C|$ rounds.*

In the remainder of this section, I describe the SRLP algorithm and illustrate its working using Example 2. Along the way, I provide some intuition for Theorem 5, whose formal proof can be found in Appendix A.4. First, a note about the running time of the SRLP algorithm is in order. By Theorem 5, the number of rounds required is polynomial in $|A||C|$. For a round in which linear programming is not used $|A|(|C| + 2)$ operations are required (each category allocates some amount of capacity to each agent and then the total capacity allocated and demand of each agent is calculated). As linear programming can be solved in polynomial time (Khachiyan, 1979), it follows that the SR algorithm works in polynomial-time.

At the high level, the idea behind the SRLP algorithm is to identify when the SR algorithm is at risk of slowing down, and speed up the process by using linear programming. For this purpose, in every Round i of the SRLP algorithm, I split the agent-category pairs into three groups. I say that agent a is **qualified** for category c if $x_{a,c}^i \geq d_a^i$. The term captures the idea that a has a high enough priority for c in order to get the amount of capacity she requires

Algorithm 3: LINEAR PROGRAMMING (LP)

Input Take the input x and d from Algorithm 2.

Linear Program Construction For every agent a , construct the set of categories for which a is qualified and marginal: $C_Q(a) = \{c \in C : x_{a,c} \geq d_a\}$ and $C_M(a) = \{c \in C : x_{a,c} \in (0, d_a)\}$. Let $\tilde{A} = \{a \in A : \min\{|C_Q(a)|, |C_M(a)|\} \geq 1\}$ be the set of agents who are qualified and marginal for at least one category.

Let $\tilde{C} = \{c \in C : c \in C_M(a) \text{ for some } a \in \tilde{A}\}$ be the set of categories that have a marginal agent who is qualified for another category. For every such category $c \in \tilde{C}$, let $a(c)$ be that category's marginal agent, construct the set $A_Q(c) = \{a \in A, x_{a,c} \geq d_a\}$ of agents that are qualified for c and the subset $\tilde{A}_Q(c) = \{a \in \tilde{A}, x_{a,c} \geq d_a\}$ of them that are marginal for another category, and adapt the quota to dismiss the agents that are not marginal for another category: $\tilde{q}_c = q_c - \sum_{a \in A_Q(c) \setminus \tilde{A}_Q(c)} d_a$.

Linear Program Solving Solve the following linear programming problem:

$$\begin{aligned}
 & \max_{(\xi_{a(c),c})_{c \in \tilde{C}}} \sum_{c \in \tilde{C}} \xi_{a(c),c} \\
 \text{subject to } & \xi_{a(c),c} \leq \frac{1 - \sum_{c' \in C_M(a(c)) \setminus \{c\}} \xi_{a(c),c'}}{|C_Q(a(c))| + 1} \tag{LP 1} \\
 \text{and } & \xi_{a(c),c} \leq \tilde{q}_c - \sum_{a \in \tilde{A}_Q(c)} \frac{1 - \sum_{c' \in C_M(a)} \xi_{a,c'}}{|C_Q(a)|} \text{ for every } c \in \tilde{C}.
 \end{aligned}$$

Output Let the vector $(\xi_{a(c),c}^*)_{c \in \tilde{C}}$ be the solution to the linear program (LP 1). For every agent a , set

$$\delta_a^{LP}(x, d) = \begin{cases} \frac{1 - \sum_{c' \in C_M(a(c)) \setminus \{c\}} \xi_{a(c),c'}^*}{|C_Q(a(c))| + 1} & \text{if } a \in \tilde{A} \\ d_a & \text{if } a \in A \setminus \tilde{A}. \end{cases}$$

from it. In contrast, if $x_{a,c}^i = 0$, I say that agent a is **unqualified** for category c in Round i : a 's priority for c is not high enough to obtain any capacity. In the intermediate case in which $0 < x_{a,c}^i < d_a^i$, I say that a is **marginal** for c . I refer to agent a 's quality as either qualified, marginal, or unqualified as a 's **status** for c . Using the convention that $x^0 = \mathbf{0}$, initially every agent is unqualified for every category. Throughout the SRLP algorithm, an agent's status for a category may change to marginal or qualified; however, any such change is by construction irreversible; therefore, there can be at most $2|A||C|$ status changes throughout the entire algorithm. Moreover, once $2|A||C|$ status changes have occurred, every agent is qualified for every category so the only possible allocation is one in which each agent is allocated $1/|C|$ from each category and the SRLP algorithm ends.⁵ The SRLP algorithm works identically to the SR algorithm until either it finds an allocation—in which case the SRLP algorithm ends and returns that allocation—or a round occur in which no status changes. In the latter case, the SRLP algorithm uses the linear program described in Algorithm 3 in the following round, which ensures that either an allocation is found or at least one status changes. As a result, at least one status changes every second round so the SRLP algorithm finds an allocation within $4|A||C|$ rounds.⁶

Whenever no change of status occur has occurred in the previous round (i.e., there is no agent-category pair (a, c) such that either $x_{a,c}^{i-1} > x_{a,c}^{i-2}$ or $x_{a,c}^{i-1} > d_a^{i-1}$ and $x_{a,c}^{i-2} < d_a^{i-2}$), the SRLP algorithm uses Algorithm 3 to calculate a new demand vector. Algorithm 3 takes a pre-allocation x and the associated demand vector d (given by the SRLP algorithm) as inputs and returns a demand vector δ^{LP} . I next describe how Algorithm 3 works and illustrate it with Example 2.

What the SR algorithm does is allocate the excess supply to marginal agents in each round until either a marginal agent is allocated her demand—in which case a change of status occurs—or there is no more excess supply—in which case an allocation has been found. As Example 2 shows, this may take infinitely many rounds; however, linear programming allows doing it in just one round. The idea is to maximize the amount of capacity allocated to marginal agents under the constraints that agents cannot be allocated more than their demand from any category and more than one unit overall and that categories may not allocate more than their quotas.

Consider a Round i in which the SRLP uses linear programming. Then, $x = x^{i-1}$ and $d = d^{i-1}$ are the input to Algorithm 3, which will calculate a demand vector $\delta^{LP}(x, d)$. The idea behind Algorithm 3 is to first calculate an allocation ξ that satisfies Axioms 1-4 and

⁵See the proof of Lemma 25 in Appendix A.4 for a formal argument.

⁶As at least two status changes occur in the first round but one more round may be required once all agents are qualified for all categories, an upper bound on the number of rounds after which the SRLP algorithm ends is $4|A||C| - 2$, see the proof of Lemma 25 in Appendix A.4 for details.

satisfies that $\xi_{a,c} = 0$ for every agent-category pair (a,c) such that $x_{a,c} = 0$; that is, an agent who was unqualified for a category in Round $i - 1$ does not receive any capacity from that category under ξ . The output of Algorithm 3 will turn out to be the demand vector associated with that allocation:

$$\delta_a^{LP} = \begin{cases} 1 & \text{if } \xi_a < 1 \\ \max_{c \in C} \{\xi_{a,c}\} & \text{if } \xi_a = 1. \end{cases}$$

Every agent a who was unqualified for every category in Round $i - 1$ is not allocated any capacity in aggregate: $\xi_a = 0$ so his demand will remain one: $\delta_a^{LP} = d_a = 1$. Consider next an agent a who was qualified for some categories in Round $i - 1$ but not marginal for any. Then, equal sharing dictates that $\xi_{a,c} = x_{a,c} = 1/|C_Q(a)|$ for every $c \in C_Q(a)$ (where $C_Q(a)$ is the set of categories for which a is eligible, as calculated in Algorithm 3). It follows that such an agent's demand does not change either: $\delta_a^{LP} = d_a = 1/|C_Q(a)|$. Finally, consider an agent a who was marginal for some categories in Round $i - 1$ but not qualified for any. That agent's demand in Round $i - 1$ was then one, hence $d_a = 1$. For every category for which a was marginal, all lower-priority agents were unqualified (see Claim 5 in Appendix A.4 for a formal statement); therefore, regardless of a 's demand, those agents are not allocated any capacity from c . Then, what a is allocated does not affect any other agent and one can simply set $\xi_{a,c} = 0$ for every $c \in C$ and $\delta_a^{LP} = d_a = 1$.

The agents for which linear programming is required are those in set \tilde{A} (defined in Algorithm 3) who are qualified for at least one category and marginal for at least one category. Those agents receive one unit in aggregate so $\sum_{c \in C_Q(a)} \xi_{a,c} + \sum_{c \in C_M(a)} \xi_{a,c} = 1$ for every $a \in \tilde{A}$. The challenge is to determine how the unit allocated to a is shared among categories. Equal sharing dictates that a receive her demand from each category for which she is qualified so we have that

$$|C_Q(a)|\delta_a^{LP} + \sum_{c \in C_M(a)} \xi_{a,c} = 1 \quad \text{for every } a \in \tilde{A}. \quad (1)$$

Moreover, equal sharing also dictates that every category $c \in C_M(a)$ allocates to a either all of its remaining capacity or a 's demand:

$$\xi_{a,c} = \min\{\delta_a^{LP}, q_c - \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}\} \quad \text{for every } c \in C_M(a).$$

By construction, $\hat{A}_{a,c} = A_Q(c)$ (see Claim 5 in Appendix A.4 for a formal statement); therefore, as $\xi_{a',c} = \delta_{a'}^{LP}$ for every $a' \in A_Q(c)$ and $\delta_a^{LP} = d_a$ for every $a' \in A_Q(c) \setminus \tilde{A}_Q(c)$, we have

that

$$\sum_{a' \in \hat{A}_{a,c}} \xi_{a',c} = \sum_{a' \in A_Q(c)} \xi_{a',c} = \sum_{a' \in \tilde{A}_Q(c)} \xi_{a',c} + \sum_{a' \in A_Q(c) \setminus \tilde{A}_Q(c)} \xi_{a',c} = \sum_{a' \in \tilde{A}_Q(c)} \delta_{a'}^{LP} + \sum_{a' \in A_Q(c) \setminus \tilde{A}_Q(c)} d_{a'}.$$

Then, by definition, we have that

$$q_c - \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c} = q_c - \sum_{a' \in A_Q(c) \setminus \tilde{A}_Q(c)} d_{a'} - \sum_{a' \in \tilde{A}_Q(c)} \delta_{a'}^{LP} = \tilde{q}_c - \sum_{a' \in \tilde{A}_Q(c)} \delta_{a'}^{LP}.$$

It follows that

$$\xi_{a,c} = \min \left\{ d_a^{LP}, \tilde{q}_c - \sum_{a' \in \tilde{A}_Q(c)} \delta_{a'}^{LP} \right\} \quad \text{for every } c \in C_M(a). \quad (2)$$

The objective is then to find an allocation ξ that satisfies (1) and (2) for every agent $a \in \tilde{A}$ and every category $c \in C_M(a)$. The agents' demands can be substituted out by using (1), which yields

$$\xi_{a,c} = \min \left\{ \frac{1 - \sum_{c' \in C_M(a)} \xi_{a,c'}}{|C_Q(a)|}, \tilde{q}_c - \sum_{a' \in \tilde{A}_Q(c)} \delta_{a'}^{LP} \right\} \quad \text{for every } a \in \tilde{A} \text{ and every } c \in C_M(a).$$

Recalling (see Algorithm 3) that \tilde{C} is the set of categories that have a marginal agent who is qualified for another category and that, for every $c \in \tilde{C}$, c 's marginal agent is denoted $a(c)$, it follows that

$$\xi_{a(c),c} = \min \left\{ \frac{1 - \sum_{c' \in C_M(a(c))} \xi_{a(c),c'}}{|C_Q(a(c))|}, \tilde{q}_c - \sum_{a \in \tilde{A}_Q(c)} \frac{1 - \sum_{c' \in C_M(a)} \xi_{a,c'}}{|C_Q(a)|} \right\} \quad \text{for every } c \in \tilde{C}.$$

Finally, by definition $c \in C_M(a(c))$ so the previous equation can be simplified to

$$\xi_{a(c),c} = \min \left\{ \frac{1 - \sum_{c' \in C_M(a(c)) \setminus \{c\}} \xi_{a(c),c'}}{|C_Q(a(c))| + 1}, \tilde{q}_c - \sum_{a \in \tilde{A}_Q(c)} \frac{1 - \sum_{c' \in C_M(a)} \xi_{a,c'}}{|C_Q(a)|} \right\} \quad \text{for every } c \in \tilde{C}.$$

There are $|\tilde{C}|$ variables and $|\tilde{C}|$ equations, one for each category $c \in \tilde{C}$; however those equations are not linear. What the linear program (LP 1) does is, for each $c \in \tilde{C}$, turn that category's equation into two constraints and maximize $\xi_{a(c),c}$ subject to those constraints.

I illustrate Algorithm 3 this works using Example 2. Round 4 is the first round in which

no status changes (see Table 2) so linear programming is used in Round 5. The inputs are

$$x = x^4 = \begin{matrix} & c_1 & c_2 & c_3 \\ a_1 & \left(\begin{matrix} 5/12 & 3/12 & 5/12 \\ 1/3 & 1/3 & 1/3 \\ 3/12 & 5/12 & 5/12 \\ 0 & 0 & 5/6 \end{matrix} \right) & & \\ a_2 & & & \\ a_3 & & & \\ a_4 & & & \end{matrix} \quad \text{and} \quad d = d^4 = \begin{pmatrix} a_1 & a_2 & a_3 & a_4 \\ 5/12 & 1/3 & 5/12 & 1 \end{pmatrix}.$$

In the Linear Program Construction stage, we have that

$$\begin{aligned} C_Q(a_1) &= \{c_1, c_3\} & C_Q(a_2) &= \{c_1, c_2, c_3\} & C_Q(a_3) &= \{c_2, c_3\} & C_Q(a_4) &= \emptyset \\ C_M(a_1) &= \{c_2\} & C_M(a_2) &= \emptyset & C_M(a_3) &= \{c_1\} & C_M(a_4) &= \{c_3\}; \end{aligned}$$

therefore, $\tilde{A} = \{a_1, a_2\}$ and $\tilde{C} = \{c_1, c_2\}$ so we have that $a(c_1) = a_3$ and $a(c_2) = a_1$. It follows that $A_Q(c_1) = \{a_1, a_2\}$, $\tilde{A}_Q(c_1) = \{a_1\}$, $A_Q(c_2) = \{a_2, a_3\}$, and $\tilde{A}_Q(c_2) = \{a_3\}$; hence $\tilde{q}_{c_1} = \tilde{q}_{c_2} = 1 - d_{a_2} = 2/3$.

In the Linear Program Solving stage, the linear program that must be solved is

$$\begin{aligned} & \max_{(\xi_{a_3, c_1}, \xi_{a_1, c_2})} \xi_{a_3, c_1} + \xi_{a_1, c_2} \\ & \text{subject to} \quad \xi_{a_3, c_1} \leq 1/3, \\ & \quad \quad \quad \xi_{a_1, c_2} \leq 1/3, \\ & \quad \quad \quad \xi_{a_3, c_1} \leq 2/3 - (1 - \xi_{a_1, c_2})/2, \\ & \quad \quad \quad \xi_{a_1, c_2} \leq 2/3 - (1 - \xi_{a_3, c_1})/2. \end{aligned}$$

Setting $\xi_{a_3, c_1} = \xi_{a_1, c_2} = 1/3$ makes all four constraints hold with an equality; hence the vector $(\xi_{a_3, c_1}^*, \xi_{a_1, c_2}^* = (1/3, 1/3)$ is the unique solution to the linear program. Then, the output of Algorithm 3 is the demand vector

$$\delta^{LP} = \begin{pmatrix} a_1 & a_2 & a_3 & a_4 \\ 1/3 & 1/3 & 1/3 & 1 \end{pmatrix}.$$

Returning to Round 5 of Algorithm 2, the linear programming stage has produced the demand vector $\delta^5 = \delta^{LP}$. The Capacity Allocation stage produces the SR allocation— $x^5 = \xi^{SR}$ —and so the SRLP algorithm ends in Round 5 and outputs the SR allocation.

5 Conclusion

In this paper, I have proposed a new solution to rationing problems with reserves. In contrast to existing solutions, in which reserve categories are processed sequentially, I propose to process them simultaneously. A key advantage of this approach is transparency: the Simultaneous Reserves (SR) allocation depends solely on the quotas reserved for each category and how each category prioritizes agent, not on any processing order. The key idea is to allow an agent who is allocated one unit in aggregate to receive parts of that unit from different categories. In fact, I show that the SR allocation respects *equal sharing*: if an agent qualifies for multiple categories, she receives the same amount of capacity from each of them. This is in stark contrast to sequential processing, in which an agent who qualifies for multiple categories receives one unit from whichever is processed first. I show that the SR allocation satisfies three standard axioms—compliance with eligibility criteria, non-wastefulness, and respect for priorities—and respects equal sharing. Moreover, any other allocation satisfying those four properties allocates in aggregate the same amount of capacity to every agent. Finally, I present a polynomial-time algorithm to compute the SR allocation.

This paper opens up various opportunities for future research, I conclude by briefly describing some of them. First, it might be possible to tweak the SR algorithm to handle ties in the priority profile (see Kesten and Ünver (2015) for a similar approach without reserves). Priority ties are often present in real-world application and such a solution would avoid having to break them through a lottery. Second, it would be valuable to explore how the SR algorithm can be combined with the deferred acceptance mechanism (or any other mechanism) so it can be used in matching markets. Third, it may be possible to generalize the approach to sharing rules beyond equal sharing. If an agent qualifies for two categories, with sequential processing the category processed first allocates one unit to that agent while with equal sharing, each category allocates half a unit to the agent. One might consider any sharing rule in between, which would convexify the set of solutions provided by sequential allocation. Ultimately, the hope is that the ideas presented in this paper provide a new perspective on rationing problems with reserves and pave the way to finding new solutions for a wide range of applications.

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A Proofs

The Appendix is organized as follows. I first prove a series of properties of the SR algorithm in Appendix A.1, which I use in Appendix A.2 to prove the results from Section 3. Appendix A.3 contains properties of the SRLP algorithm that mirror those of the SR algorithm. In Appendix A.4, I use those properties to prove the results from Section 4.

A.1 Properties of the SR algorithm

I start with a series of regularity conditions.

Lemma 1. *For every agent a and every Round i such that $x_a^i > 1$, there exists a unique d_a^i such that $\sum_{c \in C} \min\{d_a^i, x_{a,c}^i\} = 1$. Moreover, $d_a^i \in (0, \max_{c \in C}\{x_{a,c}^i\})$.*

Proof. If $d_a^i \leq 0$, then $\sum_{c \in C} \min\{d_a^i, x_{a,c}^i\} \leq 0 < 1$. If $d_a^i = \max_{c \in C}\{x_{a,c}^i\}$, then $\sum_{c \in C} \min\{d_a^i, x_{a,c}^i\} = \sum_{c \in C} x_{a,c}^i = x_a^i > 1$. The expression $\sum_{c \in C} \min\{d_a^i, x_{a,c}^i\}$ is continuous at every d_a^i , strictly increasing in d_a^i at every $d_a^i \leq \max_{c \in C}\{x_{a,c}^i\}$, and constant in d_a^i at every $d_a^i \geq \max_{c \in C}\{x_{a,c}^i\}$. Therefore, there exists a unique value of d_a^i such that $\sum_{c \in C} \min\{d_a^i, x_{a,c}^i\} = 1$ and that value is an element of $(0, \max_{c \in C}\{x_{a,c}^i\})$. \square

Lemma 2. *For every Round $i \geq 1$, x^i is a pre-allocation and, for every agent a , $d_a^i \in [1/|C|, 1]$.*

Proof. ($d_a^i \in [0, 1]$) Towards a contradiction, suppose that $d_a^i \notin [0, 1]$. By definition, $x_a^i \geq 1$ as otherwise $d_a^i = 1$. If $x_a^i = 1$, then $d_a^i = \max_{c \in C}\{x_{a,c}^i\}$. If $x_a^i > 1$, then $d_a^i \in (0, \max_{c \in C}\{x_{a,c}^i\})$ by Lemma 1. In both cases, it follows that there exists a category $c \in C$ such that $x_{a,c}^i \notin [0, 1]$. Then, a is eligible for c , as otherwise $x_{a,c}^i = 0$; therefore, $x_{a,c}^i = \min\{d_a^{i-1}, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}\}$, which implies that $d_a^{i-1} \notin [0, 1]$. By induction, it follows that $d_a^0 \notin [0, 1]$, a contradiction since $d_a^0 = 1$.

(x^i is a pre-allocation) What needs to be shown is that $x_{a,c}^i \in [0, 1]$ for all $a \in A$ and all $c \in C$ and that $\sum_{a \in A} x_{a,c}^i \leq q_c$ for all $c \in C$. Consider any agent a and any category c . If $x_{a,c}^i \notin [0, 1]$, it was established in the first part of this proof that $d_a^{i-1} \notin [0, 1]$, a contradiction.

It remains to show that $\sum_{a \in A} x_{a,c}^i \leq q_c$ for all $c \in C$. Consider any category c and suppose towards a contradiction that $\sum_{a \in A} x_{a,c}^i > q_c$. Then, there exists an agent a such that $x_{a,c} > 0$ and $x_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} x_{a',c}^i > q_c$. By definition, $x_{a,c}^i \leq d_{a'}^{i-1}$ for all $a' \in \hat{A}_{a,c}$ so $x_{a',c}^i + \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1} > q_c$ or, equivalently, $x_{a',c}^i > q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}$. Again by definition, $x_{a,c}^i \leq \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}$; therefore it must be that $x_{a',c} = 0$, a contradiction.

($d_a^i \geq 1/|C|$) If $x_a^i < 1$, the statement is trivially satisfied as $d_a^i = 1$ by definition. If $x_a^i = 1$, then $d_a^i = \max\{c \in C\}\{x_{a,c}^i\}$. As $\sum_{c \in C} x_{a,c}^i = 1$ and $x_{a,c}^i \in [0, 1]$, we have that

$\max_{c \in C} \{x_{a,c}^i\} \geq 1/|C|$ so $d_a^i \geq 1/|C|$. If $x_a^i > 1$, then $\sum_{c \in C} \min\{d_a^i, x_{a,c}^i\} = 1$; therefore, we have that $\sum_{c \in C} d_a^i \geq 1$ so $|C|d_a^i \geq 1$, which is equivalent to $d_a^i \geq 1/|C|$. \square

For every Round $i \geq 1$, let ξ^i to be the Round i allocation (as opposed to the Round i pre-allocation x^i) defined as follows. For every agent a and every category c , $\xi_{a,c}^i = \min\{d_a^i, x_{a,c}^i\}$.

Lemma 3. *For every Round $i \geq 1$, ξ^i is an allocation and, for every agent a , $\xi_a^i = \min\{x_a^i, 1\}$.*

Proof. ($\xi_a^i = \min\{x_a^i, 1\}$) *Case 1:* $x_a^i \leq 1$. If $x_a^i < 1$, then by definition $d_a^i = 1$ and $x_{a,c}^i < 1$ for all $c \in C$. If $x_a^i = 1$, then by definition $d_a^i = \max_{c \in C} \{x_{a,c}^i\}$. It follows that $x_{a,c}^i \leq d_a^i$ for all $c \in C$; therefore,

$$\xi_a^i = \sum_{c \in C} \xi_{a,c}^i = \sum_{c \in C} \min\{d_a^i, x_{a,c}^i\} = \sum_{c \in C} x_{a,c}^i = x_a^i = \min\{x_a^i, 1\}.$$

Case 2: $x_a^i > 1$. By definition, d_a^i satisfies $\sum_{c \in C} \min\{d_a^i, x_{a,c}^i\} = 1$. It follows that

$$\xi_a^i = \sum_{c \in C} \xi_{a,c}^i = \sum_{c \in C} \min\{d_a^i, x_{a,c}^i\} = 1 = \min\{x_a^i, 1\}.$$

(ξ^i is an allocation) By definition, for every agent a and every category c , $\xi_{a,c}^i = \min\{d_a^i, x_{a,c}^i\}$. As $d_a^i, x_{a,c}^i \in [0, 1]$, it follows that $\xi_{a,c}^i \in [0, 1]$. Moreover, $\sum_{a \in A} \xi_{a,c}^i \leq \sum_{a \in A} x_{a,c}^i \leq q_c$. Therefore, ξ^i is a pre-allocation and it remains to show that $\xi_a \leq 1$ for all $a \in A$, which follows from the previously established result that $\xi_a^i = \min\{x_a^i, 1\}$ for all $a \in A$. \square

The next lemma states that the total amount that each agent is allocated weakly increases throughout the algorithm while each agent's demand decreases throughout the algorithm. For notational convenience, let $\xi^0 = \mathbf{0}_{|A| \times |C|}$.

Lemma 4. *For every Round i of the SR algorithm and every agent a , $\xi_a^i \geq \xi_a^{i-1}$ and $d_a^i \leq d_a^{i-1}$.*

Proof of Lemma 4

By definition, $\xi_a^0 = 0$ and $d_a^0 = 1$ and, by Lemmas 2 and 3, $\xi_a^i, d_a^i \in [0, 1]$; therefore the statement holds for Round 1: $\xi_a^1 \geq \xi_a^0$ and $d_a^1 \leq d_a^0$.

The remainder of the proof is by induction. For some $i \geq 2$, suppose that $\xi_a^{i-1} \geq \xi_a^{i-2}$ and $d_a^{i-1} \leq d_a^{i-2}$ for all $a \in A$ (induction hypothesis). I show that $\xi_a^i \geq \xi_a^{i-1}$ and $d_a^i \leq d_a^{i-1}$.

($\xi_a^i \geq \xi_a^{i-1}$) Consider any category c . If a is not eligible for c , then by definition $x_{a,c}^{i-1} = x_{a,c}^i = 0$ and $\xi_{a,c}^{i-1} = \min\{d_a^{i-1}, x_{a,c}^{i-1}\} = 0$; hence $x_{a,c}^i = \xi_{a,c}^{i-1} = 0$. If a is eligible for c , then by definition

$$x_{a,c}^{i-1} = \min\{d_a^{i-2}, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-2}, 0\}\};$$

therefore,

$$\xi_{a,c}^{i-1} = \min\{d_a^{i-1}, d_a^{i-2}, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-2}, 0\}\}.$$

By the induction hypothesis, $d_a^{i-1} \leq d_a^{i-2}$; hence

$$\xi_{a,c}^{i-1} = \min\{d_a^{i-1}, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-2}, 0\}\}. \quad (3)$$

By definition,

$$x_{a,c}^i = \min\{d_a^{i-1}, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}\}. \quad (4)$$

The induction hypothesis implies that $d_{a'}^{i-1} \leq d_{a'}^{i-2}$ for all $a' \in \hat{A}_{a,c}$; therefore the right-hand side of (4) is weakly larger than the right-hand side of (3) and $x_{a,c}^i \geq \xi_{a,c}^{i-1}$.

The previous argument has established that $x_{a,c}^i \geq \xi_{a,c}^{i-1}$ for all $c \in C$; hence $x_a^i \geq \xi_a^{i-1}$. Combining Lemma 3 with that result and the fact that $\xi_a^{i-1} \leq 1$ yields

$$\xi_a^i = \min\{x_a^i, 1\} \geq \min\{\xi_a^{i-1}, 1\} = \xi_a^{i-1},$$

which implies that $\xi_a^i \geq \xi_a^{i-1}$.

($d_a^i \leq d_a^{i-1}$) *Case 1:* $x_a^i < 1$. Lemma 3 and the previously established result that $\xi_a^i \geq \xi_a^{i-1}$ imply that

$$\min\{x_a^i, 1\} = \xi_a^i \geq \xi_a^{i-1} = \min\{x_a^{i-1}, 1\}.$$

It follows that $\min\{x_a^i, 1\} \geq \min\{x_a^{i-1}, 1\}$, which combined with the case assumption that $x_a^i < 1$ implies that $x_a^{i-1} < 1$. By definition, it can therefore be concluded that $d_a^i = d_a^{i-1} = 1$.

Case 2: $x_a^i \geq 1$. If $x_a^i = 1$, then by definition $d_a^i = \max_{c \in C} \{x_{a,c}^i\}$. If $x_a^i > 1$, then by definition $\sum_{c \in C} \min\{d_a^i, x_{a,c}^i\} = 1$. Supposing that $d_a^i > \max_{c \in C} \{x_{a,c}^i\}$ yields

$$\sum_{c \in C} \min\{d_a^i, x_{a,c}^i\} = \sum_{c \in C} x_{a,c}^i = x_a^i > 1,$$

a contradiction. Therefore, the case assumption that $x_a^i \geq 1$ implies that $d_a^i \leq \max_{c \in C} \{x_{a,c}^i\}$.

By definition, for every $c \in C$, $x_{a,c}^i = \min\{d_a^{i-1}, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}\} \leq d_a^{i-1}$; therefore $\max_{c \in C} \{x_{a,c}^i\} \leq d_a^{i-1}$, which means that $d_a^i \leq d_a^{i-1}$.

Lemma 5. *For every Round i and every agent a ,*

$$d_a^i = \begin{cases} 1 & \text{if } \xi_a^i < 1 \\ \max_{c \in C} \{\xi_{a,c}^i\} & \text{if } \xi_a^i = 1. \end{cases}$$

Proof. Case 1: $x_a^i < 1$. In that case, by Lemma 3, $\xi_a^i = x_a^i < 1$ and, by definition, $d_a^i = 1$.

Case 2: $x_a^i = 1$. In that case, by Lemma 3, $\xi_a^i = x_a^i = 1$. By definition, $d_a^i = \max_{c \in C} \{x_{a,c}\}$ and, for every $c \in C$, $\xi_{a,c}^i = \min\{d_a^i, x_{a,c}^i\}$. Combining those two results implies that $\xi_{a,c}^i = x_{a,c}^i$ for all $c \in C$, and therefore $d_a^i = \max\{\xi_{a,c}^i\}$.

Case 3: $x_a^i > 1$. In that case, by Lemma 3, $\xi_a^i = 1$ so it remains to show that $d_a^i = \max_{c \in C} \{\xi_{a,c}^i\}$. If $d_a^i < \max_{c \in C} \{\xi_{a,c}^i\}$, then there exists $c \in C$ such that $d_a^i < \xi_{a,c}^i$. However, by definition, $\xi_{a,c}^i = \min\{d_a^i, x_{a,c}^i\} \leq d_a^i$, a contradiction. If $d_a^i > \max_{c \in C} \{\xi_{a,c}^i\}$, then by definition $d_a^i > \max_{c \in C} \{\min\{d_a^i, x_{a,c}^i\}\}$; therefore $d_a^i > x_{a,c}^i$ for all $c \in C$ so $\sum_{c \in C} \min\{d_a^i, x_{a,c}^i\} = \sum_{c \in C} x_{a,c}^i = x_a^i > 1$. However, by definition, $\sum_{c \in C} \min\{d_a^i, x_{a,c}^i\} = 1$, a contradiction. \square

Lemma 6. *For every Round i and every agent a , $\xi_a^i = 1$ if and only if there exists a category c such that $\xi_{a,c}^i = d_a^i$.*

Proof. If $\xi_a^i < 1$, then $\xi_{a,c}^i < 1$ for all $c \in C$ and, by Lemma 5, $d_a^i = 1$; therefore, $\xi_{a,c}^i < d_a^i$ for all $c \in C$. If $\xi_a^i = 1$, then $d_a^i = \max_{c \in C} \{\xi_{a,c}^i\}$ by Lemma 5; hence there exists $c \in C$ such that $\xi_{a,c}^i = d_a^i$. \square

Lemma 7. *For every Round i , every agent a , and every category c , either $x_{a,c}^i < d_a^{i-1}$ or $\xi_{a,c}^i < d_a^i$ implies that $x_{a',c}^i = \xi_{a',c}^i = 0$ for every lower-priority agent $a' \in \check{A}_{a,c}$.*

Proof. By definition, $\xi_{a,c}^i < d_a^i$ implies that $\xi_{a,c}^i = x_{a,c}^i$; hence, as $d_a^i \leq d_a^{i-1}$ by Lemma 4, it follows that $x_{a,c}^i < d_a^{i-1}$. Moreover, again by definition, we have that $\xi_{a',c}^i = 0$ if and only if $x_{a',c}^i = 0$. Therefore, it is sufficient to show that $x_{a,c}^i < d_a^{i-1}$ implies that $x_{a',c}^i = 0$ for every lower-priority agent $a' \in \check{A}_{a,c}$.

Suppose that $x_{a,c}^i < d_a^{i-1}$ and consider an arbitrary lower-priority agent $a' \in \check{A}_{a,c}$. I show that $x_{a',c}^i = 0$. If a' is not eligible for c , the desired result holds trivially since, by definition, $x_{a',c}^i = 0$. For the remainder of the proof, I assume that a' is eligible for c , which implies that a is eligible for c as well.

As $a\pi_c a'$, the assumption that a' is eligible for c implies that a is also eligible for c . Therefore, by definition, we have that

$$x_{a,c}^i = \min\{d_a^{i-1}, \max\{q_c - \sum_{\tilde{a} \in \hat{A}_{a,c}} d_{\tilde{a}}^{i-1}, 0\}\} < d_a^{i-1},$$

which implies that $q_c - \sum_{\tilde{a} \in \hat{A}_{a,c}} d_{\tilde{a}}^{i-1} < d_a^{i-1}$ or, equivalently, $q_c - \sum_{\tilde{a} \in \hat{A}_{a,c}} d_{\tilde{a}}^{i-1} - d_a^{i-1} < 0$. As $a\pi_c a'$, it follows that

$$q_c - \sum_{\tilde{a} \in \hat{A}_{a',c}} d_{\tilde{a}}^{i-1} \leq q_c - \sum_{\tilde{a} \in \hat{A}_{a,c}} d_{\tilde{a}}^{i-1} - d_a^{i-1} < 0. \quad (5)$$

Moreover, as a' is eligible for c , we have that

$$x_{a',c}^i = \min\{d_a^{i-1}, \max\{q_c - \sum_{\tilde{a} \in \hat{A}_{a',c}} d_{\tilde{a}}^{i-1}, 0\}\}. \quad (6)$$

Combining (5) and (6) yields $x_{a',c}^i = 0$. \square

Lemma 8. *For every Round i , every agent a , and every category c , $\xi_{a,c}^i > 0$ implies that $x_{a',c}^i = d_{a'}^{i-1}$ and $\xi_{a',c}^i = d_{a'}^i$ for every higher-priority agent $a' \in \hat{A}_{a,c}$.*

Proof. Suppose that $\xi_{a,c}^i > 0$. By definition, it must be that $x_{a,c}^i > 0$ so a is eligible for c and

$$q_c > \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}. \quad (7)$$

Consider any higher-priority agent $a' \in \hat{A}_{a,c}$. It needs to be shown that $\xi_{a',c}^i = d_{a'}^i$. By (7), $q_c > d_{a'}^{i-1} + \sum_{\tilde{a} \in \hat{A}_{a',c}} d_{\tilde{a}}^{i-1}$, which is equivalent to

$$q_c - \sum_{\tilde{a} \in \hat{A}_{a',c}} d_{\tilde{a}}^{i-1} > d_{a'}^{i-1}. \quad (8)$$

As a is eligible for c , so is a' ; hence by definition $x_{a',c}^i = \min\{d_{a'}^{i-1}, \max\{q_c - \sum_{\tilde{a} \in \hat{A}_{a',c}} d_{\tilde{a}}^{i-1}, 0\}\}$. By (8), it follows that $x_{a',c}^i = d_{a'}^{i-1}$. By Lemma 4, $d_{a'}^i \leq d_{a'}^{i-1}$; hence $x_{a',c}^i \geq d_{a'}^i$. Using that inequality in conjunction with the definition of $\xi_{a',c}^i$ yields $\xi_{a',c}^i = \min\{d_{a'}^i, x_{a',c}^i\} = d_{a'}^i$. \square

Together, Lemmas 7 and 8 allow defining each category's marginal agents in Round i . If all eligible agents are allocated their demand, the marginal agent is \emptyset . Otherwise, the marginal agent is the highest-priority agent who is not allocated her demand.

Lemma 9. *For every agent a and category c such that a is eligible for c , and for every Round i , $x_{a,c}^i < d_a^{i-1}$ implies that $x_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} x_{a',c}^i = q_c$.*

Proof. Let \tilde{a} be the highest-priority agent such that $x_{\tilde{a},c}^i < d_{\tilde{a}}^{i-1}$. That is, $x_{\tilde{a},c}^i < d_{\tilde{a}}^{i-1}$ and, for every $a \in \hat{A}_{\tilde{a},c}$, $x_{a',c}^i = d_{a'}^{i-1}$. The assumption that $x_{a,c}^i < d_a^{i-1}$ ensures that \tilde{a} exists and either $\tilde{a} = a$ or $\tilde{a} \pi_c a$. Then, as a is eligible for c , so is \tilde{a} and we have that $x_{\tilde{a},c}^i = \min\{d_{\tilde{a}}^{i-1}, \max\{q_c - \sum_{a' \in \hat{A}_{\tilde{a},c}} d_{a'}^{i-1}, 0\}\}$. As $x_{\tilde{a},c}^i < d_{\tilde{a}}^{i-1}$ and $x_{a',c}^i = d_{a'}^{i-1}$ for all $a' \in \hat{A}_{\tilde{a},c}$, it follows that $x_{\tilde{a},c}^i = \max\{q_c - \sum_{a' \in \hat{A}_{\tilde{a},c}} d_{a'}^{i-1}, 0\}$. As x^i is a pre-allocation (by Lemma 2), it must be that $\sum_{a' \in \hat{A}_{\tilde{a},c}} x_{a',c}^{i-1} \leq q_c$; therefore we can conclude that $x_{\tilde{a},c}^i = q_c - \sum_{a' \in \hat{A}_{\tilde{a},c}} x_{a',c}^i$ or, equivalently, $x_{\tilde{a},c}^i + \sum_{a' \in \hat{A}_{\tilde{a},c}} x_{a',c}^i = q_c$.

On the one hand, as either $\tilde{a} = a$ or $\tilde{a}\pi_c a$, we have that $x_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} x_{a',c}^i \geq x_{\tilde{a},c}^i + \sum_{a' \in \hat{A}_{\tilde{a},c}} x_{a',c}^i = q_c$. On the other hand, as x^i is an allocation, we have that $x_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} x_{a',c}^i \leq q_c$. Combining the two statements yields $x_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} x_{a',c}^i = q_c$. \square

Lemma 10. *In every Round i , x^i satisfies Axioms 1-4 and ξ^i satisfies Axioms 1 and 3.*

Proof. (x^i and ξ^i comply with eligibility requirements) By definition, if an agent a is not eligible for a category c , then $x_{a,c}^i = \xi_{a,c}^i = 0$.

(x^i is non-wasteful) Consider any category c such that $\sum_{a \in A} x_{a,c}^i < q_c$ and any agent a who is eligible for c . It needs to be shown that $x_a^i \geq 1$.

Case 1: $x_{a,c}^i = d_a^{i-1}$. By the case assumption and Lemma 4, $x_{a,c}^i = d_a^{i-1} \geq d_a^i$; hence, by definition, $\xi_{a,c}^i = \min\{d_a^i, x_{a,c}^i\} = d_a^i$. By Lemma 6, it follows that $\xi_a^i = 1$ so, by definition, $x_a^i \geq \xi_a^i = 1$.

Case 2: $x_{a,c}^i < d_a^{i-1}$. In that case, Lemma 9 applies and yields $x_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} x_{a',c}^i = q_c$. Then, we have that $\sum_{a \in A} x_{a,c}^i \geq x_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} x_{a',c}^i = q_c$, which contradicts the assumption that $\sum_{a \in A} x_{a,c}^i < q_c$.

(x^i and ξ^i respect priorities) Arbitrarily fix an agent a and a category c . By Lemma 3, $x_a^i < 1$ if and only if $\xi_a^i < 1$ and, by definition, $x_{a,c}^i = 0$ if and only if $\xi_{a,c}^i$. It follows that x^i respects priorities if and only if ξ^i respects priorities; hence it is enough to show that ξ^i respects priorities. Suppose that $\xi_a^i < 1$. In order to establish that ξ^i respects priorities, I need to show that, for every lower-priority agent $a' \in \check{A}_{a,c}$, $\xi_{a',c}^i = 0$. By Lemma 6, the assumption that $\xi_a^i < 1$ implies that $\xi_{a,c}^i < d_a^i$ so, by Lemma 7, $\xi_{a',c}^i = 0$.

(x^i respects equal sharing) Consider any agent a and any category c such that a is eligible for c and $x_{a,c}^i < \max_{c' \in C} \{x_{a,c'}^i\}$. It needs to be shown that $x_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} x_{a',c}^i = q_c$. By definition, $\max_{c' \in C} \{x_{a,c'}^i\} \leq d_a^{i-1}$; hence we have that $x_{a,c}^i < d_a^{i-1}$. Then, by Lemma 9, we have that $x_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} x_{a',c}^i = q_c$. \square

For every Round $i \geq 1$, I define the **excess supply in Round i** to be the $|A| \times |C|$ matrix $z^i = (z_{a,c}^i)_{a \in A, c \in C}$ such that, for every agent a and every category c , $z_{a,c}^i = x_{a,c}^i - \xi_{a,c}^i$. I denote by $z_a^i = \sum_{c \in C} z_{a,c}^i$ the **excess supply of agent a in Round i** . By construction and using Lemma 3, $z_a^i = x_a^i - \xi_a^i = \max\{x_a^i - 1, 0\}$ so z_a^i can be interpreted as the capacity that a is allocated in Round i in excess of the one unit she needs. I denote by $|z^i| = \sum_{a \in A} \sum_{c \in C} z_{a,c}^i = \sum_{a \in A} z_a^i$ the **total excess supply in Round i** , which can be interpreted as the amount of capacity allocated to agents who do not need it and that can be reallocated in the next round. The next result introduces basic properties of the excess demand.

Lemma 11. *For every agent a and every category c , $z_{a,c}^i \in [0, 1]$. Moreover, $|z^i| = 0$ if and only if $x^i = \xi^i$.*

Proof. ($z_{a,c}^i \in [0, 1]$) By definition, $z_{a,c}^i = x_{a,c}^i - \xi_{a,c}^i = x_{a,c}^i - \min\{d_a^i, x_{a,c}^i\} = \max\{x_{a,c}^i - d_a^i, 0\}$. As $x_{a,c}^i, d_a^i \in [0, 1]$ (by Lemma 2), it follows that $z_{a,c}^i \in [0, 1]$.

($|z^i| = 0$ if and only if $x^i = \xi^i$) If $x^i = \xi^i$, then, for every $a \in A$ and every $c \in C$, $x_{a,c} = \xi_{a,c}$ so $z_{a,c} = x_{a,c} - \xi_{a,c} = 0$. It follows that $|z^i| = \sum_{a \in A} \sum_{c \in C} z_{a,c}^i = 0$. If $x^i \neq \xi^i$, then there exist $a \in A$ and $c \in C$ such that $x_{a,c}^i \neq \xi_{a,c}^i$ so $z_{a,c} \neq 0$. As $z_{a,c}^i \in [0, 1]$ for all $a \in A$ and all $c \in C$, it follows that $|z^i| = \sum_{a \in A} \sum_{c \in C} z_{a,c}^i > 0$. \square

Lemma 12. *Suppose that, for some Round $i \geq 1$, $|z^i| = 0$. Then, for every Round $j \geq i$, $|z^j| = 0$, $x^j = \xi^j = x^i = \xi^i$, and, for every agent a , $d_a^j = d_a^i$.*

Proof. Fix an agent a and a category c arbitrarily. The main part of the proof consists in showing that $x_{a,c}^{i+1} = x_{a,c}^i$. By definition, if a is not eligible for c , then $x_{a,c}^{i+1} = x_{a,c}^i = 0$; therefore I focus throughout on the case in which a is eligible for c .

($x_{a,c}^{i+1} \geq x_{a,c}^i$) By definition, $x_{a,c}^{i+1} = \min\{d_a^i, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^i, 0\}\}$. By assumption, $|z^i| = 0$ so, by Lemma 11, $x_{a,c}^i = \xi_{a,c}^i$. As, by definition, $\xi_{a,c}^i = \min\{x_{a,c}^i, d_a^i\}$, it follows that $x_{a,c}^i \leq d_a^i$. Therefore, it remains to show that $x_{a,c}^i \leq \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^i, 0\}$.

By definition, $x_{a,c}^i = \min\{d_a^{i-1}, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}\}$; hence $x_{a,c}^i \leq \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}$. By Lemma 4, $\sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1} \geq \sum_{a' \in \hat{A}_{a,c}} d_{a'}^i$; therefore, we have that $x_{a,c}^i \leq \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\} \leq \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^i, 0\}$.

($x_{a,c}^{i+1} \leq x_{a,c}^i$) *Case 1: $a\pi_c\tilde{a}$.* By the case assumption, $x_{a,c}^i = d_a^{i-1}$. By definition, $x_{a,c}^{i+1} = \min\{d_a^i, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^i, 0\}\}$ so $x_{a,c}^{i+1} \leq d_a^i$. By Lemma 4, $d_a^i \leq d_a^{i-1}$; therefore, we have that $x_{a,c}^{i+1} \leq d_a^i \leq d_a^{i-1} = x_{a,c}^i$.

Case 2: $a = \tilde{a}$. By the case assumption, $x_{a,c}^i < d_a^{i-1}$ and $x_{a',c}^i = d_{a'}^{i-1}$ for every $a' \in \hat{A}_{a,c}$. By definition, $x_{a,c}^i = \min\{d_a^{i-1}, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}\}$ so, as $x_{a,c}^i < d_a^{i-1}$, we have that

$$x_{a,c}^i = \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}. \quad (9)$$

Again by definition, we have that

$$x_{a,c}^{i+1} = \min\{d_a^i, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^i, 0\}\}. \quad (10)$$

For every $a' \in \hat{A}_{a,c}$, $x_{a',c}^i = d_{a'}^{i-1}$ by the case assumption, $x_{a',c}^i = x_{a',c}^{i+1}$ by the result shown in Case 1, $x_{a',c}^{i+1} \leq d_{a'}^i$ by definition, and $d_{a'}^i \leq d_{a'}^{i-1}$ by Lemma 4. Therefore, it can be concluded that $d_{a'}^i = d_{a'}^{i-1}$ for all $a' \in \hat{A}_{a,c}$. Combining that result with (9) and (10) yields

$$x_{a,c}^{i+1} \leq \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^i, 0\} = \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\} = x_{a,c}^i.$$

Case 3: $\tilde{a}\pi_c a$. By definition, as $x_{\tilde{a},c}^i < d_{\tilde{a}}^{i-1}$, $x_{\tilde{a},c}^i = \max\{q_c - \sum_{a' \in \hat{A}_{\tilde{a},c}} d_{a'}^{i-1}, 0\}$. As was shown in Case 2, $d_{a'}^i = d_{a'}^{i-1}$ for all $a' \in \hat{A}_{\tilde{a},c}$; therefore we have that $x_{\tilde{a},c}^i = \max\{q_c - \sum_{a' \in \hat{A}_{\tilde{a},c}} d_{a'}^i, 0\}$, which implies that $q_c - x_{\tilde{a},c}^i - \sum_{a' \in \hat{A}_{\tilde{a},c}} d_{a'}^i \leq 0$. As was shown in Case 2, $x_{\tilde{a},c}^i = x_{\tilde{a},c}^{i+1}$ and, by definition, $x_{\tilde{a},c}^{i+1} \leq d_{\tilde{a}}^i$; therefore, we have that $q_c - d_{\tilde{a}}^i - \sum_{a' \in \hat{A}_{\tilde{a},c}} d_{a'}^i \leq 0$. As $\tilde{a}\pi_c a$, $\sum_{a' \in \hat{A}_{\tilde{a},c}} d_{a'}^i \geq d_{\tilde{a}}^i + \sum_{a' \in \hat{A}_{\tilde{a},c}} d_{a'}^i$; hence it follows that $q_c - \sum_{a' \in \hat{A}_{\tilde{a},c}} d_{a'}^i \leq 0$. Combining that last inequality with the definition of x_a^{i+1} yields $x_a^{i+1} = 0$. By Lemma 2, $x_a^i \geq 0$ so it can be concluded that $x_a^{i+1} \leq x_a^i$.

As a and c were chosen arbitrarily, $x_{a,c}^{i+1} = x_{a,c}^i$ holds for every agent and every category; therefore we have that $x^{i+1} = x^i$. Then, by definition, $d_a^{i+1} = d_a^i$ for every agent a , and $\xi^{i+1} = \xi^i$. The result extends to every $j > i + 1$ by induction. \square

Lemma 13. *Suppose that, for some agent a , some category c , and some Round $i \geq 1$, $x_{a,c}^i < d_a^i$. Then, for every $j \leq i$, $x_{a,c}^j \leq x_{a,c}^i < d_a^j$.*

Proof. If a is not eligible for c , then $x_{a,c}^j = 0$ for every $j \geq 1$ and the result holds as, by Lemma 2, $d_a^j > 0$. The remainder of the proof focuses on the case in which a is eligible for c .

By definition, $x_{a,c}^i = \min\{d_a^{i-1}, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}\}$ and, by Lemma 4, the assumption that $x_{a,c}^i < d_a^i$ implies that $x_{a,c}^i < d_a^{i-1}$. It follows that $x_{a,c}^i = \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}$. Again by definition, $x_{a,c}^{i-1} \leq \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-2}, 0\}$ and, by Lemma 4, $\sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-2} \geq \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}$; hence we have that $x_{a,c}^{i-1} \leq x_{a,c}^i < d_a^{i-1}$. By induction, the statement holds for every $j \leq i$. \square

Lemma 14. *Suppose that, for some agent a , some category c and some Round $i \geq 1$, $x_{a,c}^i \geq d_a^i$. Then, for every $j > i$, $x_{a,c}^j = d_a^{j-1} \geq d_a^j$.*

Proof. By definition, $x_{a,c}^i \leq \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}$, which, combined with the assumption that $x_{a,c}^i \geq d_a^i$, implies that $\max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\} \geq d_a^i$. By Lemma 4, it follows that $\max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^i, 0\} \geq d_a^i$. By definition, we have that $x_{a,c}^{i+1} = \min\{d_a^i, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^i, 0\}\} = d_a^i$. Then, by Lemma 4, it follows that $x_{a,c}^{i+1} = d_a^i \geq d_a^{i+1}$ and the statement holds for all $j > i$ by induction. \square

Lemma 15. *For every allocation ξ^* that satisfies Axioms 1-4, every Round i of the SR algorithm, every agent a , and every category c , $\xi_a^* \geq \xi_a^i$ and $\xi_{a,c}^* \leq d_a^i$.*

Proof. By definition, for every agent a and every category c , $d_a^0 = 1$ and ξ^* is an allocation so $\xi_{a,c}^* \leq 1 = d_a^0$. The remainder of the proof is by induction. Arbitrarily fixing a Round $i \geq 1$, I assume that $\xi_{a,c}^* \leq d_a^{i-1}$ for every agent a and every category c (induction hypothesis) and show that $\xi_a^* \geq \xi_a^i$ and $\xi_{a,c}^* \leq d_a^i$ for every agent a and every category c .

($\xi_a^* \geq \xi_a^i$ for every $a \in A$) Arbitrarily fix an agent a . By Lemma 3, $\xi_a^* \leq 1$ and $\xi_a^i \leq 1$ so the desired result holds trivially if $\xi_a^* = 1$, and therefore only the case in which $\xi_a^* < 1$ needs to be considered. Arbitrarily fixing a category c , I show that, in this case, $\xi_{a,c}^* \geq \xi_{a,c}^i$. That result holds trivially if $\xi_{a,c}^i = 0$; hence I assume for the remainder of the argument that $\xi_{a,c}^i > 0$.

As ξ^i complies with eligibility requirements (by Lemma 10), the assumption that $\xi_{a,c}^i > 0$ implies that a is eligible for c ; hence, by definition, we have that

$$\xi_{a,c}^i = \min\{d_a^i, x_{a,c}^i\} = \min\{d_a^i, d_a^{i-1}, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}\}.$$

It follows that $\xi_{a,c}^i \leq \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}$ so the assumption that $\xi_{a,c}^i > 0$ implies that

$$\xi_{a,c}^i \leq q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}. \quad (11)$$

By the induction hypothesis, $\xi_{a',c}^* \leq d_{a'}^{i-1}$ for every $a' \in \hat{A}_{a,c}$; therefore (11) implies that $\xi_{a,c}^i \leq q_c - \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^*$, which is equivalent to

$$\xi_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^* \leq q_c. \quad (12)$$

By assumption, $\xi_{a,c}^*$ is non-wasteful and $\xi_a^* < 1$; hence, as a is eligible for c , we have that

$$\xi_{a,c}^* + \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^* = q_c. \quad (13)$$

Combining (12) and (13) implies that $\xi_{a,c}^* \geq \xi_{a,c}^i$. As c was chosen arbitrarily, that inequality holds for all categories and it can be concluded that $\xi_a^* = \sum_{c \in C} \xi_{a,c}^* \geq \sum_{c \in C} \xi_{a,c}^i = \xi_a^i$. As a was chosen arbitrarily, it follows that $\xi_a^* \geq \xi_a^i$ for every $a \in A$.

($\xi_{a,c}^* \leq d_a^i$ for every $a \in A$ and every $c \in C$) Arbitrarily fix an agent a and a category c . Towards a contradiction, suppose that $\xi_{a,c}^* > d_a^i$. As $\xi_{a,c}^* \leq 1$ (by Lemma 3), it follows that $d_a^i < 1$. Then, Lemma 5 implies that $d_a^i = \max_{c' \in C} \{\xi_{a,c'}^i\}$ and $\xi_a^i = 1$. It follows that

$$\xi_{a,c}^i \leq \max_{c' \in C} \{\xi_{a,c'}^i\} = d_a^i < \xi_{a,c}^* \leq \max_{c' \in C} \{\xi_{a,c'}^*\}. \quad (14)$$

Moreover, as $\xi_a^* \leq 1$ (by Lemma 3) and as it was established in the previous part of the proof

that $\xi_a^* \geq \xi_a^i$, the fact that $\xi_a^i = 1$ implies that

$$\xi_a^i = \sum_{c' \in C} \xi_{a,c'}^i = \sum_{c' \in C} \xi_{a,c'}^* = \xi_a^* = 1. \quad (15)$$

As (14) implies that $\xi_{a,c}^i < \xi_{a,c}^*$, it follows by (15) that there exists a category $\tilde{c} \in C$ such that

$$\xi_{a,\tilde{c}}^* < \xi_{a,\tilde{c}}^i. \quad (16)$$

By definition, $\xi_{a,\tilde{c}}^i \leq \max_{c' \in C} \{\xi_{a,c'}^i\}$ and, by (14), $\max_{c' \in C} \{\xi_{a,c'}^i\} \leq \max_{c' \in C} \{\xi_{a,c'}^*\}$; therefore, (16) implies that $\xi_{a,\tilde{c}}^* < \max_{c' \in C} \{\xi_{a,c'}^*\}$. As ξ^* respects equal sharing (by assumption), it follows that

$$\xi_{a,\tilde{c}} + \sum_{a' \in \hat{A}_{a,\tilde{c}}} \xi_{a',\tilde{c}}^* = q_{\tilde{c}}. \quad (17)$$

Next, (16) implies that $\xi_{a,\tilde{c}}^i > 0$. As ξ^i complies with eligibility requirements (by Lemma 10), it follows that a is eligible for \tilde{c} so, by definition,

$$\xi_{a,\tilde{c}}^i = \min\{d_a^i, x_{a,\tilde{c}}^i\} = \min\{d_a^i, d_a^{i-1}, \max\{q_{\tilde{c}} - \sum_{a' \in \hat{A}_{a',\tilde{c}}} d_{a'}^{i-1}, 0\}\}.$$

It follows that $\xi_{a,\tilde{c}}^i \leq \max\{q_{\tilde{c}} - \sum_{a' \in \hat{A}_{a',\tilde{c}}} d_{a'}^{i-1}, 0\}$; hence $\xi_{a,\tilde{c}}^i > 0$ implies that $\xi_{a,\tilde{c}}^i \leq q_{\tilde{c}} - \sum_{a' \in \hat{A}_{a',\tilde{c}}} d_{a'}^{i-1}$ or, equivalently, $\xi_{a,\tilde{c}}^i + \sum_{a' \in \hat{A}_{a',\tilde{c}}} d_{a'}^{i-1} \leq q_{\tilde{c}}$. By the induction hypothesis, $\xi_{a,\tilde{c}}^* \leq d_{a'}^{i-1}$ for every $a' \in \hat{A}_{a,\tilde{c}}$; hence we have that $\xi_{a,\tilde{c}}^i + \sum_{a' \in \hat{A}_{a',\tilde{c}}} \xi_{a',\tilde{c}}^* \leq q_{\tilde{c}}$. By (17), it follows that $\xi_{a,\tilde{c}}^i \leq \xi_{a,\tilde{c}}^*$, which contradicts (16). \square

Lemma 16. *For every agent a and every allocation ξ^* that satisfies Axioms 1-4, $\max_{c \in C} \{\xi_{a,c}^*\} \leq \max_{c \in C} \{\xi_{a,c}^{SR}\}$.*

Proof. Towards a contradiction, suppose to the contrary that $\max_{c \in C} \{\xi_{a,c}^*\} > \max_{c \in C} \{\xi_{a,c}^{SR}\}$. Then, there exists a category c' such that $\xi_{a,c'}^* > \xi_{a,c'}^{SR}$. By Theorem 3, we have that $\xi_a^* = \xi_a^{SR}$ so, by definition, $\sum_{c \in C} \xi_{a,c}^* = \sum_{c \in C} \xi_{a,c}^{SR}$. Then, the fact that $\xi_{a,c'}^* > \xi_{a,c'}^{SR}$ implies there exists a category \tilde{c} such that

$$\xi_{a,\tilde{c}}^* < \xi_{a,\tilde{c}}^{SR}. \quad (18)$$

It follows that $\xi_{a,\tilde{c}}^* < \max_{c \in C} \{\xi_{a,c}^*\}$ so, as ξ^* respects equal sharing, we have that $\xi_{a,\tilde{c}}^* + \sum_{a' \in \hat{A}_{a,\tilde{c}}} \xi_{a',\tilde{c}}^* = q_{\tilde{c}}$. Moreover, $\xi_{a,\tilde{c}}^{SR} + \sum_{a' \in \hat{A}_{a,\tilde{c}}} \xi_{a',\tilde{c}}^{SR} \leq q_{\tilde{c}}$ by definition since ξ^{SR} is an allocation. It follows that

$$\xi_{a,\tilde{c}}^{SR} + \sum_{a' \in \hat{A}_{a,\tilde{c}}} \xi_{a',\tilde{c}}^{SR} \leq \xi_{a,\tilde{c}}^* + \sum_{a' \in \hat{A}_{a,\tilde{c}}} \xi_{a',\tilde{c}}^*,$$

which combined with (18) implies that

$$\sum_{a' \in \hat{A}_{a,c}} \xi_{a',\bar{c}}^{SR} < \sum_{a' \in \hat{A}_{a,c}} \xi_{a',\bar{c}}^* \quad (19)$$

By (18), we have that $\xi_{a,\bar{c}}^{SR} > 0$; hence, by definition, $\lim_{i \rightarrow \infty} \xi_{a,\bar{c}}^i > 0$. Similarly, (19) implies by definition that $\lim_{i \rightarrow \infty} \sum_{a' \in \hat{A}_{a,c}} \xi_{a',\bar{c}}^i < \sum_{a' \in \hat{A}_{a,c}} \xi_{a',\bar{c}}^*$. Therefore, there exists a Round i such that

$$\xi_{a,\bar{c}}^i > 0 \quad \text{and} \quad (20)$$

$$\sum_{a' \in \hat{A}_{a,\bar{c}}} \xi_{a',\bar{c}}^i < \sum_{a' \in \hat{A}_{a,\bar{c}}} \xi_{a',\bar{c}}^* \quad (21)$$

By Lemma 8, (20) implies that $\xi_{a,\bar{c}}^i = d_{a'}^i$ for all $a' \in \hat{A}_{a,c}$, which combined with (21) implies that $\sum_{a' \in \hat{A}_{a,\bar{c}}} d_{a'}^i < \sum_{a' \in \hat{A}_{a,\bar{c}}} \xi_{a',\bar{c}}^*$. Therefore, there exists an agent $a' \in \hat{A}_{a,\bar{c}}$ such that $d_{a'}^i < \xi_{a',\bar{c}}^*$, which contradicts Lemma 15. \square

A.2 Proof of the results from Section 3

Proof of Theorem 1

I have shown in the main text that ξ^{SR} is well-defined and equal to both $\lim_{i \rightarrow \infty} x^i$ and $\lim_{i \rightarrow \infty} \xi^i$. It remains to show that ξ^{SR} is an allocation. Arbitrarily fix an agent a and a category c . By definition, it needs to be shown that (i) $\xi_{a,c}^{SR} \in [0, 1]$, (ii) $\sum_{a' \in A} \xi_{a',c}^{SR} \leq q_c$, and (iii) $\xi_a^{SR} \leq 1$.

($\xi_{a,c}^{SR} \in [0, 1]$) *Case 1:* $\xi_{a,c}^i < d_a^i$ for every $i \geq 1$. By definition, the case assumption implies that $x_{a,c}^i < d_a^i$ for every $i \geq 1$, which by Lemma 13 implies that $x_{a,c}^j \leq x_{a,c}^i < d_a^j$ for every $i, j \geq 1$ with $j \leq i$. By definition, it follows that the series $\{\xi_{a,c}^i\}_{i=1}^{\infty}$ is weakly increasing. By Lemma 2, that series is bounded. Then, the Monotone Convergence Theorem implies that $\lim_{i \rightarrow \infty} \xi_{a,c}^i$ is equal to the series' supremum. By Lemma 2, every element of the series $\{\xi_{a,c}^i\}_{i=1}^{\infty}$ is an element of $[0, 1]$; hence so is its supremum.

Case 2: $\xi_{a,c}^i = d_a^i$ for some $i \geq 1$. By definition, the case assumption implies that $x_{a,c}^i \geq d_a^i$; hence Lemma 14 implies that $x_{a,c}^j \geq d_a^j$ for every $j \geq i$. Again by definition, it follows that $\xi_{a,c}^j = d_a^j$ for all $j \geq i$, which implies that $\lim_{i \rightarrow \infty} \xi_{a,c}^i = \lim_{i \rightarrow \infty} d_a^i$ so it remains to show that $\lim_{i \rightarrow \infty} d_a^i \in [0, 1]$.

The series $\{d_a^i\}_{i=1}^{\infty}$ is weakly decreasing by Lemma 4 and bounded below by Lemma 2. By the Monotone Convergence Theorem, $\lim_{i \rightarrow \infty} d_a^i$ is then equal to the infimum of the series $\{d_a^i\}_{i=1}^{\infty}$. By Lemma 2, every element of that series is an element of $[1/|C|, 1]$; hence so is its

infimum.

($\sum_{a' \in A} \xi_{a',c}^{SR} \leq q_c$) As $\lim_{i \rightarrow \infty} \xi_{a',c}^i \in [0, 1]$ for every $a' \in A$, $\sum_{a' \in A} \xi_{a',c}^{SR} = \lim_{i \rightarrow \infty} (\sum_{a' \in A} \xi_{a',c}^i) = \sum_{a' \in A} (\lim_{i \rightarrow \infty} \xi_{a',c}^i)$ converges to a real number. By Lemma 3, ξ^i is an allocation for every $i \geq 1$; therefore every element of the series $\{\sum_{a' \in A} \xi_{a',c}^i\}_{i=1}^{\infty}$ is weakly smaller than q_c . Then, the number to which the series converges cannot exceed q_c .

($\xi_a^{SR} \leq 1$) As $\lim_{i \rightarrow \infty} \xi_{a,c}^i \in [0, 1]$ for every $c \in C$, $\xi_a^{SR} = \sum_{c \in C} \xi_{a,c}^{SR} = \lim_{i \rightarrow \infty} (\sum_{c \in C} \xi_{a,c}^i) = \sum_{c \in C} (\lim_{i \rightarrow \infty} \xi_{a,c}^i)$ is equal to a real number. By Lemma 3, ξ^i is an allocation for every $i \geq 1$; therefore every element of the series $\{\xi_a^i\}_{i=1}^{\infty}$ is weakly smaller than 1. Then, the number to which the series converges cannot exceed 1. \square

Proof of Proposition 2

($|z^{i+1}| \leq |z^i|$) By definition, $|z^{i+1}| = |x^{i+1}| - |\xi^{i+1}|$ and $|z^i| = |x^i| - |\xi^i|$ and, by Lemma 4, $|\xi^{i+1}| \geq |\xi^i|$; therefore, it remains to show that $|x^{i+1}| \leq |x^i|$.

Consider first any category c such that $\sum_{a \in A} x_{a,c}^i < q_c$. I show that, for every $a \in A$,

$$x_{a,c}^i = \begin{cases} 0 & \text{if } a \text{ is not eligible for } c \\ d_a^{i-1} & \text{if } a \text{ is eligible for } c. \end{cases} \quad (22)$$

If a is not eligible for c , then $x_{a,c}^i = 0$ by definition; therefore, it remains to show that, if a is eligible for c , then $x_{a,c}^i = d_a^{i-1}$. Towards a contradiction, suppose that a is eligible for c and $x_{a,c}^i \neq d_a^{i-1}$. Let \tilde{a} be the highest-priority agent in that situation; that is, \tilde{a} is eligible for c , $x_{\tilde{a},c}^i \neq d_{\tilde{a}}^{i-1}$, and, for every $a' \in \hat{A}_{\tilde{a},c}$, $x_{a',c}^i = d_{a'}^{i-1}$. By definition, as \tilde{a} is eligible for c , $x_{\tilde{a},c}^i = \min\{d_{\tilde{a}}^{i-1}, \max\{q_c - \sum_{a' \in \hat{A}_{\tilde{a},c}} d_{a'}^{i-1}, 0\}\}$ so the assumption that $x_{\tilde{a},c}^i \neq d_{\tilde{a}}^{i-1}$ implies that $x_{\tilde{a},c}^i < d_{\tilde{a}}^{i-1}$. It follows that $x_{\tilde{a},c}^i = \max\{q_c - \sum_{a' \in \hat{A}_{\tilde{a},c}} d_{a'}^{i-1}, 0\}$; therefore, $q_c - x_{\tilde{a},c}^i - \sum_{a' \in \hat{A}_{\tilde{a},c}} d_{a'}^{i-1} \leq 0$. Combining the last inequality with the assumption that $x_{a',c}^i = d_{a'}^{i-1}$ for every $a' \in \hat{A}_{\tilde{a},c}$ yields $q_c - x_{\tilde{a},c}^i - \sum_{a' \in \hat{A}_{\tilde{a},c}} x_{a',c}^i \leq 0$. It can then be concluded that $\sum_{a \in A} x_{a,c}^i \geq x_{\tilde{a},c}^i + \sum_{a' \in \hat{A}_{\tilde{a},c}} x_{a',c}^i \geq q_c$, a contradiction; hence (22) holds for every agent a . Letting $A_c = \{a \in A : a \pi_c \emptyset\}$ denote the set of agents that are acceptable for c , it follows that $\sum_{a \in A} x_{a,c}^i = \sum_{a \in A_c} d_a^{i-1}$.

Consider now any category c' . By Lemma 2, $\sum_{a \in A} x_{a,c'}^i \leq q_{c'}$ and, by the previous argument, $\sum_{a \in A} x_{a,c'}^i < q_{c'}$ implies that $\sum_{a \in A} x_{a,c'}^i = \sum_{a \in A_{c'}} d_a^{i-1}$. It follows that $\sum_{a \in A} x_{a,c'}^i = \min\{q_{c'}, \sum_{a \in A_{c'}} d_a^{i-1}\}$ for every $c' \in C$. Therefore, we have that $|x^i| = \sum_{c' \in C} \min\{q_{c'}, \sum_{a \in A_{c'}} d_a^{i-1}\}$ and, analogously, $|x^{i+1}| = \sum_{c' \in C} \min\{q_{c'}, \sum_{a \in A_{c'}} d_a^i\}$; hence, Lemma 4 implies that $|x^{i+1}| \leq |x^i|$.

($|z^i| \leq |A|(|C| - 1)/i$) Fix an agent a and a category c . I first show that

$$\sum_{j=1}^i z_{a,c}^j \leq 1 - 1/|C|. \quad (23)$$

By definition,

$$\sum_{j=1}^i z_{a,c}^j = \sum_{j=1}^i (x_{a,c}^j - \xi_{a,c}^j) = \sum_{j=1}^i (x_{a,c}^j - \min\{d_a^j, x_{a,c}^j\}) = \sum_{j=1}^i \max\{x_{a,c}^j - d_a^j, 0\}. \quad (24)$$

If $x_{a,c}^j \leq d_a^j$ for all $j = 1, \dots, i$, then (23) holds trivially as (24) implies that $\sum_{j=1}^i z_{a,c}^j = 0$. Otherwise, let $k = 1, \dots, i$ be the first round in which a receives her demand from c ; that is, $x_{a,c}^k \geq d_a^k$ and $x_{a,c}^j < d_a^j$ for all $j = 1, \dots, k-1$. Then, by Lemma 14, $x_{a,c}^j = d_a^{j-1}$ for all $j = k+1, \dots, i$. By (24), we have that

$$\sum_{j=1}^i z_{a,c}^j = \sum_{j=1}^i \max\{x_{a,c}^j - d_a^j, 0\} = \sum_{j=k}^i (x_{a,c}^j - d_a^j) = x_{a,c}^k - d_a^i + \sum_{j=k+1}^i (x_{a,c}^j - d_a^{j-1}) = x_{a,c}^k - d_a^i.$$

Then, (23) is satisfied as, by Lemma 2, $x_{a,c}^k \leq 1$ and $d_a^i \geq 1/|C|$.

As a and c were chosen arbitrarily, (23) holds for every agent and every category; therefore, we have that $\sum_{j=1}^i |z^j| \leq |A||C|(1 - 1/|C|) = |A|(|C| - 1)$. The first part of the proof has established that $|z^{i+1}| \leq |z^i|$ and that result holds in every round; hence $|z^i| \leq |z^j|$ for all $j \leq i$. It can then be concluded that $i|z^i| \leq \sum_{j=1}^i |z^j| \leq |A|(|C| - 1)$, which implies that $|z^i| \leq |A|(|C| - 1)/i$. \square

Proof of Theorem 2

By Lemma 10, ξ^i complies with eligibility requirements and respects priorities and x^i satisfies all four properties in every Round i , and by Corollary 1, $\xi^{SR} = \lim_{i \rightarrow \infty} \xi^i = \lim_{i \rightarrow \infty} x^i$. I use those two results to show that ξ^{SR} satisfies all four properties.

(ξ^{SR} complies with eligibility requirements) For every agent a . every category c for which a is not eligible, and every Round i , as x^i complies with eligibility requirements we have that $x_{a,c}^i = 0$ for all $i \geq 1$. It follows that $\xi_{a,c}^{SR} = \lim_{i \rightarrow \infty} x_{a,c}^i = 0$.

(ξ^{SR} is non-wasteful) Consider any category c such that $\sum_{a \in A} \xi_{a,c}^{SR} < q_c$ and any agent a who is eligible for c . It needs to be shown that $\xi_a^{SR} \geq 1$. As $\sum_{a \in A} \xi_{a,c}^{SR} < q_c$, it must be that $\lim_{i \rightarrow \infty} \sum_{a \in A} x_{a,c}^i < q_c$; hence there exists a Round j such that, for all $i \geq j$, $\sum_{a \in A} x_{a,c}^i < q_c$. As x^i is non-wasteful and a is eligible for c , we have that $x_a^i \geq 1$. Then, $x_a^{SR} = \lim_{i \rightarrow \infty} x_a^i \geq 1$.

(ξ^{SR} respects priorities) Arbitrarily fix an agent a such that $\xi_a^{SR} < 1$, a category c , and a

lower-priority agent $a' \in \tilde{A}_{a,c}$. It needs to be shown that $\xi_{a',c}^{SR} = 0$. By Lemma 4, for every Round i , $\xi_a^i \leq \xi_a^{SR} < 1$. As ξ^i respects priorities, it follows that $\xi_{a',c}^i = 0$ for all $i \geq 1$; hence we have that $\xi_{a',c}^{SR} = \lim_{i \rightarrow \infty} \xi_{a',c}^i = 0$.

(ξ^{SR} respects equal sharing) Consider an agent a and a category c such that a is eligible for c and $\xi_{a,c}^{SR} < \max_{c' \in C} \{\xi_{a,c'}^{SR}\}$. It needs to be shown that $\xi_{a,c}^{SR} + \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^{SR} = q_c$. By assumption, we have that $\lim_{i \rightarrow \infty} x_{a,c}^i < \lim_{i \rightarrow \infty} \max_{c' \in C} \{x_{a,c'}^i\}$. Then, there exists a Round j such that, for all $i \geq j$, $x_{a,c}^i < \max_{c' \in C} \{x_{a,c'}^i\}$. For every $i \geq j$, x^i is non-wasteful; hence $x_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} x_{a',c}^i = q_c$ for all $i \geq j$. We can then conclude that $\xi_{a,c}^{SR} + \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^{SR} = \lim_{i \rightarrow \infty} (x_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} x_{a',c}^i) = q_c$. \square

Proof of Theorem 3

Let ξ^* be an allocation that satisfies Axioms 1-4; it needs to be shown that $\xi_a^* = \xi_a^{SR}$ for every agent a . Towards a contradiction, suppose to the contrary that $\xi_{\tilde{a}}^* \neq \xi_{\tilde{a}}^{SR}$ for some agent \tilde{a} . Lemma 15 implies that $\xi_a^* \geq \xi_a^i$ for every agent a and every Round i ; therefore, for every agent a , we have that $\xi_a^* \geq \lim_{i \rightarrow \infty} \xi_a^i = \xi_a^{SR}$. It follows that $\xi_{\tilde{a}}^* > \xi_{\tilde{a}}^{SR}$ and, for every agent $a \neq \tilde{a}$, $\xi_a^* \geq \xi_a^{SR}$; hence we have that $|\xi^*| > |\xi^{SR}|$. Consequently, there must exist a category c such that

$$\sum_{a \in A} \xi_{a,c}^* > \sum_{a \in A} \xi_{a,c}^{SR}. \quad (25)$$

By definition (as ξ^* is an allocation), $\sum_{a \in A} \xi_{a,c}^* \leq q_c$; hence (25) implies that $\sum_{a \in A} \xi_{a,c}^{SR} < q_c$. By Corollary 1, $\lim_{i \rightarrow \infty} \sum_{a \in A} x_{a,c}^i < q_c$; therefore, there exists a Round j such that $\sum_{a \in A} x_{a,c}^i < q_c$ for every $i \geq j$. Then, by Lemma 9, for every agent a who is eligible for c , we have that $x_{a,c}^i = d_a^{i-1}$. By definition, $\xi_{a,c}^i = \min\{d_a^i, x_{a,c}^i\} = \min\{d_a^i, d_a^{i-1}\}$ so, by Lemma 4, $\xi_{a,c}^i = d_a^i$. As $\xi_{a,c}^* \leq d_a^i$ by Lemma 15, we have that $\xi_{a,c}^* \leq \xi_{a,c}^i$. Moreover, for any agent a who is not eligible for c , $\xi_{a,c}^i = 0$ by definition and $\xi_{a,c}^* = 0$ as ξ^* complies with eligibility requirements. Therefore, we can conclude that $\xi_{a,c}^* \leq \xi_{a,c}^i$ for every Round $i \geq j$ and every agent a . Then, for every agent a , $\xi_{a,c}^* \leq \lim_{i \rightarrow \infty} \xi_{a,c}^i = \xi_{a,c}^{SR}$, which contradicts (25). \square

Proof of Proposition 3

Let there be three agents a_1 , a_2 , and a_3 as well as two categories c_1 and c_2 . For each of the four axioms, I construct quotas and priorities such that an allocation that is not SR equivalent satisfies the other three axioms.

(Complies with eligibility requirements) Let the quotas and priorities be

$$q_{c_1} = 2 \quad q_{c_2} = 1 \quad \pi_{c_1} : a_1, \emptyset, a_3, a_2 \quad \pi_{c_2} : a_2, \emptyset, a_3, a_1.$$

The SR algorithm finds the SR allocation after just one round: each category allocates one unit of capacity to its highest-priority agent, respectively a_1 and a_2 . Hence, we have that

$$\xi^{SR} = \begin{array}{cc} & c_1 & c_2 \\ \begin{array}{c} a_1 \\ a_2 \\ a_3 \end{array} & \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} & \text{and } \rho^{SR} = \begin{pmatrix} a_1 & a_2 & a_3 \\ 1 & 1 & 0 \end{pmatrix}.$$

Consider the alternative allocation

$$\xi = \begin{array}{cc} & c_1 & c_2 \\ \begin{array}{c} a_1 \\ a_2 \\ a_3 \end{array} & \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} & \text{with } \rho(\xi) = \begin{pmatrix} a_1 & a_2 & a_3 \\ 1 & 1 & 1 \end{pmatrix}.$$

The allocation ξ is not SR equivalent since $\rho(\xi) \neq \rho^{SR}$. However, ξ is non-wasteful since the amount of capacity allocated by each category is equal to its quota, ξ respects priority since every agent is allocated one unit, and ξ respects equal sharing since, for every agent-object pair (a, c) such that a is eligible for c , we have that $\xi_{a,c} = \max_{c' \in C} \{\xi_{a,c'}\}$ (there are two such pairs, (a_1, c_1) and (a_2, c_2) , and $\xi_{a_1, c_1} = \xi_{a_2, c_2} = 1$).

(Non-wasteful) Let the quotas and priorities be

$$q_{c_1} = 2 \quad q_{c_2} = 1 \quad \pi_{c_1} : a_1, a_3, \emptyset, a_2 \quad \pi_{c_2} : a_2, \emptyset, a_3, a_1.$$

The SR algorithm finds the SR allocation after just one round: c_1 allocates its two units to its two highest-priority agents— a_1 and a_3 —and c_2 allocates its unique unit to its highest-priority agent— a_2 . Hence, we have that

$$\xi^{SR} = \begin{array}{cc} & c_1 & c_2 \\ \begin{array}{c} a_1 \\ a_2 \\ a_3 \end{array} & \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} & \text{and } \rho^{SR} = \begin{pmatrix} a_1 & a_2 & a_3 \\ 1 & 1 & 1 \end{pmatrix}.$$

Consider the alternative allocation

$$\xi = \begin{matrix} & c_1 & c_2 \\ \begin{matrix} a_1 \\ a_2 \\ a_3 \end{matrix} & \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \end{matrix} \quad \text{with} \quad \rho(\xi) = \begin{pmatrix} a_1 & a_2 & a_3 \\ 1 & 1 & 0 \end{pmatrix}.$$

The allocation ξ is not SR equivalent since $\rho(\xi) \neq \rho^{SR}$. However, ξ complies with eligibility requirements since agents only receive capacity from categories for which they are eligible, ξ respects priorities as a_3 is the only agent who is not allocated one unit in aggregate and every agent with a lower priority than a_3 at either category is allocated 0 from that category, and a_3 respects equal sharing as a_1 is not eligible for c_2 , a_2 is not eligible for c_1 , and $\xi_{a_3, c_1} = \xi_{a_3, c_2}$.

(Respects priorities) Let the quotas and priorities be

$$q_{c_1} = 1 \quad q_{c_2} = 1 \quad \pi_{c_1} : a_1, a_3, a_2, \emptyset \quad \pi_{c_2} : a_1, a_2, a_3, \emptyset.$$

The SR algorithm finds the SR allocation after two rounds. In Round 1, both categories allocate one unit to a_1 , which has the highest-priority for both categories. Therefore, a_1 's demand decreases to $1/2$ and in Round 2 each category allocates half a unit to its second highest-priority agent, respectively a_3 and a_2 . Hence, we have that

$$\xi^{SR} = \begin{matrix} & c_1 & c_2 \\ \begin{matrix} a_1 \\ a_2 \\ a_3 \end{matrix} & \begin{pmatrix} 1/2 & 1/2 \\ 0 & 1/2 \\ 1/2 & 0 \end{pmatrix} \end{matrix} \quad \text{and} \quad \rho^{SR} = \begin{pmatrix} a_1 & a_2 & a_3 \\ 1 & 1/2 & 1/2 \end{pmatrix}.$$

Consider the alternative allocation

$$\xi = \begin{matrix} & c_1 & c_2 \\ \begin{matrix} a_1 \\ a_2 \\ a_3 \end{matrix} & \begin{pmatrix} 0 & 0 \\ 1/2 & 1/2 \\ 1/2 & 1/2 \end{pmatrix} \end{matrix} \quad \text{with} \quad \rho(\xi) = \begin{pmatrix} a_1 & a_2 & a_3 \\ 0 & 1 & 1 \end{pmatrix}.$$

The allocation ξ is not SR equivalent since $\rho(\xi) \neq \rho^{SR}$. However, ξ complies with eligibility requirements as agents only receive capacity from categories for which they are eligible, ξ is non-wasteful as each category allocates overall an amount of capacity equal to its quota, and ξ respects equal sharing as each agent is allocated the same amount of capacity from both categories.

(Respects equal sharing) Let the quotas and priorities be identical to the previous example: $q_{c_1} = q_{c_2} = 1$, $\pi_{c_1} : a_1, a_3, a_2, \emptyset$ and $\pi_{c_2} : a_1, a_2, a_3, \emptyset$. We have again that

$$\xi^{SR} = \begin{array}{c} c_1 \quad c_2 \\ a_1 \left(\begin{array}{cc} 1/2 & 1/2 \\ 0 & 1/2 \\ 1/2 & 0 \end{array} \right) \end{array} \quad \text{and} \quad \rho^{SR} = \begin{array}{ccc} a_1 & a_2 & a_3 \\ \left(\begin{array}{ccc} 1 & 1/2 & 1/2 \end{array} \right). \end{array}$$

Consider the alternative allocation

$$\xi = \begin{array}{c} c_1 \quad c_2 \\ a_1 \left(\begin{array}{cc} 1 & 0 \\ 0 & 0 \\ 0 & 1 \end{array} \right) \end{array} \quad \text{with} \quad \rho(\xi) = \begin{array}{ccc} a_1 & a_2 & a_3 \\ \left(\begin{array}{ccc} 1 & 0 & 1 \end{array} \right).$$

The allocation ξ is not SR equivalent since $\rho(\xi) \neq \rho^{SR}$. However, ξ complies with eligibility requirements as agents only receive capacity from categories for which they are eligible, ξ is non-wasteful as each category allocates overall an amount of capacity equal to its quota, and ξ respects priorities as a_2 is the only agent not to be allocated one unit in aggregate and has a lower priority than a_1 for c_1 and a lower priority than a_2 for c_2 . \square

Proof of Theorem 4

It needs to be shown that $d_a(\xi^*) \leq d_a(\xi^{SR})$ for every agent a and $d_a(\xi^*) < d_a(\xi^{SR})$ for some agent a . I prove each of the two statements separately.

($d_a(\xi^*) \leq d_a(\xi^{SR})$ for every agent a) Consider any agent a . If $\xi_a^{SR} < 1$, then by definition $d_a(\xi^*) \leq 1 = d_a(\xi^{SR})$. If $\xi_a^{SR} = 1$, then $\xi_a^* = 1$ by Theorem 3. Therefore, by definition we have that $d_a(\xi^*) = \max_{c \in C} \{\xi_{a,c}^*\}$ and $d_a(\xi^{SR}) = \max_{c \in C} \{\xi_{a,c}^{SR}\}$. By Lemma 16, it follows that $d_a(\xi^*) = \max_{c \in C} \{\xi_{a,c}^*\} \leq \max_{c \in C} \{\xi_{a,c}^{SR}\} = d_a(\xi^{SR})$.

($d_a(\xi^*) < d_a(\xi^{SR})$ for some agent a) As $\xi^* \neq \xi^{SR}$, there exists an agent a and a category c such that $\xi_{a,c}^* \neq \xi_{a,c}^{SR}$. By Theorem 3, $\xi_a^* = \xi_a^{SR}$ so $\xi_{a,c}^* > \xi_{a,c}^{SR}$ implies that $\xi_{a,\tilde{c}}^* < \xi_{a,\tilde{c}}^{SR}$ for some category \tilde{c} . Therefore, without loss of generality, I assume that

$$\xi_{a,c}^* < \xi_{a,c}^{SR} \tag{26}$$

and show that $d_a(\xi^*) \leq d_a(\xi^{SR})$.

First, observe that (26) implies that a is eligible for c ; otherwise, as ξ^* and ξ^{SR} comply with eligibility requirements, we would have that $\xi_{a,c}^* = \xi_{a,c}^{SR} = 0$. Second, I show the following

intermediate result:

Claim 1. $\sum_{a' \in \hat{A}_{a,c}} \xi_{a,c}^* \leq \sum_{a' \in \hat{A}_{a,c}} \xi_{a,c}^{SR}$.

Proof. Suppose to the contrary that $\sum_{a' \in \hat{A}_{a,c}} \xi_{a,c}^* > \sum_{a' \in \hat{A}_{a,c}} \xi_{a,c}^{SR}$. Then, there exists an agent $\tilde{a} \in \hat{A}_{a,c}$ such that $\xi_{\tilde{a},c}^* > \xi_{\tilde{a},c}^{SR}$. As ξ^{SR} is an allocation, by definition $\sum_{a' \in A} \xi_{a',c}^{SR} \leq q_c$ and, by (26), $\xi_{a,c}^{SR} > 0$; therefore, as $\tilde{a} \pi_c a$, it can be concluded that $\xi_{\tilde{a},c}^{SR} + \sum_{a' \in \hat{A}_{\tilde{a},c}} \xi_{a',c}^{SR} < q_c$. As ξ^{SR} respects equal sharing and \tilde{a} is eligible for c (since a is eligible for c and \tilde{a} has a higher priority), it follows that $\xi_{\tilde{a},c}^{SR} = \max_{c' \in C} \{\xi_{\tilde{a},c'}^{SR}\}$. Then, the fact that $\xi_{\tilde{a},c}^* > \xi_{\tilde{a},c}^{SR}$ implies that

$$\max_{c' \in C} \{\xi_{\tilde{a},c'}^*\} \geq \xi_{\tilde{a},c}^* > \xi_{\tilde{a},c}^{SR} = \max_{c' \in C} \{\xi_{\tilde{a},c'}^{SR}\},$$

which contradicts Lemma 16. \square

Having established Claim 1, I now use it to show that $d_a(\xi^*) \leq d_a(\xi^{SR})$. I consider separately the cases in which $\xi_a^{SR} < 1$ and $\xi_a^{SR} = 1$.

Case 1: $\xi_a^{SR} < 1$. In that case, by Theorem 3, $\xi_a^* = \xi_a^{SR} < 1$. As ξ^* and ξ^{SR} are non-wasteful, $\xi_a^* = \xi_a^{SR} < 1$, and a is eligible for c , we have that $\sum_{a \in A} \xi_{a,c}^* = \sum_{a \in A} \xi_{a,c}^{SR} = q_c$. Moreover, as ξ^* and ξ^{SR} respect priorities and $\xi_a^* = \xi_a^{SR} < 1$, we have that $\xi_{a',c}^* = \xi_{a',c}^{SR} = 0$ for every lower-priority agent $a' \in \hat{A}_{a,c}$. It follows that

$$\xi_{a,c}^* + \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^* = \xi_{a,c}^{SR} + \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^{SR} = q_c.$$

Then, (26) implies that $\sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^* > \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^{SR}$, which contradicts Claim 1. Therefore, we must be in Case 2.

Case 2: $\xi_a^{SR} = 1$. In that case, by Theorem 3, $\xi_a^* = \xi_a^{SR} = 1$. I consider separately two subcases: $\xi_{a,c}^* < \max_{c' \in C} \{\xi_{a,c'}^*\}$ and $\xi_{a,c}^* = \max_{c' \in C} \{\xi_{a,c'}^*\}$.

Subcase 2.1: $\xi_{a,c}^* < \max_{c' \in C} \{\xi_{a,c'}^*\}$. As ξ^* respects equal sharing and a is eligible for c , we have that $\xi_{a,c}^* + \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^* = q_c$. Moreover, as ξ^{SR} is an allocation, we have that $\xi_{a,c}^{SR} + \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^{SR} \leq q_c$. Then, (26) implies that $\sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^* > \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^{SR}$, which contradicts Claim 1. Therefore, we must be in Subcase 2.2.

Subcase 2.2: $\xi_{a,c}^* = \max_{c' \in C} \{\xi_{a,c'}^*\}$. As $\xi_a^* = \xi_a^{SR} = 1$, by definition we have that $d_a(\xi^*) = \max_{c' \in C} \{d_a(\xi^*)\}$ and $d_a(\xi^{SR}) = \max_{c' \in C} \{d_a(\xi^{SR})\}$. Using those two results in conjunction with the subcase assumption and (26) yields

$$d_a(\xi^*) = \max_{c' \in C} \{\xi_{a,c'}^*\} = \xi_{a,c}^* < \xi_{a,c}^{SR} \leq \max_{c' \in C} \{\xi_{a,c'}^{SR}\} = d_a(\xi^{SR});$$

hence, it can be concluded that $d_a(\xi^*) \leq d_a(\xi^{SR})$, as required. \square

A.3 Properties of the SRLP algorithm

Lemma 17. *For every Round $i \geq 1$, ξ^i is an allocation and, for every agent a , $\xi_a^i = \min\{x_a^i, 1\}$.*

Proof. Lemma 17 is a counterpart to Lemma 3 for the SRLP algorithm and its proof is completely analogous to that of Lemma 3. \square

Lemma 18. *For every Round i of the SRLP algorithm and every agent a , we have that*

$$d_a^i = \begin{cases} 1 & \text{if } \xi_a^i < 1 \\ \max_{c \in C} \{\xi_{a,c}^i\} & \text{if } \xi_a^i = 1. \end{cases}$$

and $\xi_a^i = 1$ if and only if there exists a category c such that $\xi_{a,c}^i = d_a^i$.

Proof. Lemma 18 is a counterpart to Lemmas 5 and 6 for the SRLP algorithm and its proof is completely analogous to those of Lemmas 5 and 6. \square

Lemma 19. *For every Round i of the SRLP algorithm and every agent a , $\xi_a^i \geq \xi_a^{i-1}$ and $d_a^i \leq \delta_a^i \leq d_a^{i-1}$.*

Proof. Lemma 19 is a counterpart to Lemma 4 for the SRLP algorithm. By an analogous reasoning to that of Lemma 4, the statement holds for Round 1: $\xi_a^1 \geq \xi_a^0$ and $d_a^1 \leq d_a^0$.

The remainder of the proof is by induction. For some $i \geq 2$, suppose that $\xi_a^{i-1} \geq \xi_a^{i-2}$ and $d_a^{i-1} \leq d_a^{i-2}$ for all $a \in A$ (induction hypothesis). I show that $\xi_a^i \geq \xi_a^{i-1}$ and $d_a^i \leq d_a^{i-1}$. The part that differs from the proof of Lemma 4 is that one needs to show that $\delta_a^i \leq d_a^{i-1}$. The result is obtained directly if the SRLP algorithm does not use linear programming in Round i since, in that case, $\delta_a^i = d_a^{i-1}$. If the SRLP does use linear programming, then $\delta_a^i = d_a^{i-1}$ for every $a \notin \tilde{A}$ so I focus on the case in which $a \in \tilde{A}$. As shown in the proof of Lemma 24, the vector $(\xi_{a(c),c}^{i-1})_{c \in \tilde{C}}$ satisfies the constraints of the linear program (LP 1); therefore, the vector $(\xi_{a,c}^{i-1})_{c \in C_M(a)}$ satisfies the constraints of the linear program (LP 3). Let $(\xi_{a(c),c}^*)_{c \in \tilde{C}}$ be the solution to the linear program (LP 1); then, the vector $(\xi_{a,c}^*)_{c \in C_M(a)}$ is the solution to the linear program (LP 3). It follows that $\sum_{c \in C_M(a)} \xi_a^{i-1} \leq \sum_{c \in C_M(a)} \xi_a^*$. By construction,

$$|C_Q(a)|d_a^{i-1} + \sum_{c \in C_M(a)} \xi_{a,c}^{i-1} = |C_Q(a)|\delta_a^i + \sum_{c \in C_M(a)} \xi_{a,c}^* = 1;$$

hence it can be concluded that $\delta_a^i \leq d_a^{i-1}$. As the SRLP algorithm constructs ξ^i from δ^i as well as d^i and ξ^i from x^i identically to the SR algorithm, analogous reasoning to that in the proof of Lemma 4 implies that $\xi_a^i \geq \xi_a^{i-1}$ and $d_a^i \leq d_a^{i-1}$. \square

Lemma 20. *For every Round i of the SRLP algorithm, every agent a , and every category c , either $x_{a,c}^i < \delta_a^i$ or $\xi_{a,c}^i < d_a^i$ implies that $x_{a',c}^i = \xi_{a',c}^i = 0$ for every lower-priority agent $a' \in \tilde{A}_{a,c}$.*

Proof. Lemma 20 is a counterpart to Lemma 7 for the SRLP algorithm. The proof is completely analogous to that of Lemma 9, the only difference is that d_a^{i-1} needs to be replaced throughout by δ_a^i . \square

Lemma 21. *For every agent a and category c such that a is eligible for c , and for every Round i , $x_{a,c}^i < \delta_a^i$ implies that $x_{a,c}^i + \sum_{a' \in \tilde{A}_{a,c}} x_{a',c}^i = q_c$.*

Proof. Lemma 21 is a counterpart to Lemma 9 for the SRLP algorithm. The proof is completely analogous to that of Lemma 9, the only difference is that d_a^{i-1} needs to be replaced throughout by δ_a^i . \square

Lemma 22. *Suppose that, for some agent a , some category c , and some Round $i \geq 1$ of the SRLP algorithm, $x_{a,c}^i < d_a^i$. Then, for every $j \leq i$, $x_{a,c}^j \leq x_{a,c}^i < d_a^j$.*

Proof. Lemma 22 is a counterpart to Lemma 13 for the SRLP algorithm. The proof is completely analogous to that of Lemma 13, the only difference is that Lemma 19 needs to be used instead of Lemma 4. \square

Lemma 23. *Suppose that, for some agent a , some category c and some Round $i \geq 1$ of the SRLP algorithm, $x_{a,c}^i \geq d_a^i$. Then, for every $j > i$, $x_{a,c}^j = d_a^{j-1} \geq d_a^j$.*

Proof. Lemma 23 is a counterpart to Lemma 14 for the SRLP algorithm. The proof is completely analogous to that of Lemma 14, the only difference is that Lemma 19 needs to be used instead of Lemma 4. \square

A.4 Proof of Theorem 5

I present an argument to establish Theorem 5 that relies on four lemmas, whose proof can be found immediately after this proof. The first lemma establishes that the output of Algorithm 3 is well-defined.

Lemma 24. *The linear program (LP 1) in Algorithm 3 has a unique solution.*

Lemma 24 ensures that each Round $i \geq 1$ of the SRLP algorithm, δ^i , x^i , and d^i are well-defined. The next step is to show that, unlike the SR algorithm, the SRLP algorithm eventually terminates.

Lemma 25. *The SRLP algorithm ends after fewer than $4|A||C|$ rounds.*

Lemma 25 guarantees that the SRLP algorithm produces an allocation in finitely many rounds. Letting $N < 4|A||C|$ be the number of rounds after which the SRLP algorithm ends, the **outcome of the SRLP algorithm** is then the allocation x^N . (By construction, x^N must be an allocation, otherwise the SRLP algorithm would not end in Round N .) The next result ensures that the outcome of the SRLP algorithm satisfies all four axioms.

Lemma 26. *In every Round i of the SRLP algorithm, x^i satisfies Axioms 1-4.*

Lemmas 24-26 imply that, after $N < 4|A||C|$ rounds, the SRLP algorithm produces an allocation x^N that satisfies Axioms 1-4. Then, by Theorem 3, the outcome of the SRLP algorithm is SR equivalent. That is, the outcomes of the SR and SRLP algorithms yield the same aggregate allocation. However, there may be multiple allocations satisfying Axioms 1-4 so the last step is to show that x^N is indeed the SR allocation.

Lemma 27. $x^N = \xi^{SR}$.

Combining Lemmas 25 and 27 completes the proof. \square

Proof of Lemma 24

I first show that (LP 1) has a solution and then proceed to showing that there cannot be multiple solutions. For the first part of the proof, I show that the previous round allocation ξ^{i-1} satisfies all $2|\tilde{C}|$ constraints, which guarantees that the (LP 1) has a solution. That is, I show that, for every $c \in \tilde{C}$,

$$\xi_{a(c),c}^{i-1} \leq \frac{1 - \sum_{c' \in C_M(a(c)) \setminus \{c\}} \xi_{a(c),c'}^{i-1}}{|C_Q(a(c))| + 1} \quad (27)$$

$$\text{and } \xi_{a(c),c}^{i-1} \leq \tilde{q}_c - \sum_{a \in \tilde{A}_Q(c)} \frac{1 - \sum_{c' \in C_M(a)} \xi_{a,c'}^{i-1}}{|C_Q(a)|}. \quad (28)$$

Arbitrarily fix an agent $a \in \tilde{A}$. By construction, a is qualified for at least one category in Round $i-1$ so, by Lemma 6, $\xi_a^{i-1} = 1$. It follows that $\sum_{c \in C_Q(a)} \xi_{a,c}^{i-1} + \sum_{c \in C_M(a)} \xi_{a,c}^{i-1} = 1$. As $\xi_{a,c}^{i-1} = d_a$ for every $c \in C_Q(a)$ by definition, we have that $d_a = (1 - \sum_{c \in C_M(a)} \xi_{a,c}^{i-1}) / |C_Q(a)|$. Moreover, again by definition, we have that $\xi_{a,c}^{i-1} \leq d_a$ for every $c \in C_M(a)$. As a was chosen arbitrarily, it follows that

$$\xi_{a,c}^{i-1} \leq \frac{1 - \sum_{c \in C_M(a)} \xi_{a,c}^{i-1}}{|C_Q(a)|} \quad \text{for every } a \in \tilde{A} \text{ and every } c \in C_M(a),$$

which by construction is equivalent to

$$\xi_{a(c),c}^{i-1} \leq \frac{1 - \sum_{c' \in C_M(a(c))} \xi_{a(c),c'}^{i-1}}{|C_Q(a(c))|} \quad \text{for every } c \in \tilde{C}.$$

Through simple algebra (as $c \in C_M(a(c))$, $\xi_{a(c),c}^{i-1}$ can be moved out of the sum and to the left-hand side), it follows that the last inequality is equivalent to (27).

To show that (28) holds, arbitrarily fix a category $c \in \tilde{C}$. As ξ^{i-1} is an allocation (by Lemma 3), we have that $\xi_{a(c),c} + \sum_{a \in A_Q(c)} \xi_{a,c}^{i-1} \leq q_c$. By definition, $\xi_{a,c}^{i-1} = d_a$ for every $a \in A_Q(c)$; hence it follows that $\xi_{a(c),c} + \sum_{a \in A_Q(c)} d_a \leq q_c$, which by definition is equivalent to $\xi_{a(c),c} + \sum_{a \in \tilde{A}_Q(c)} d_a \leq \tilde{q}_c$. Moreover, for every $a \in A_Q(c)$, we have that $|C_Q(a)|d_a + \sum_{c' \in C_M(a)} \xi_{a,c'}^{i-1} = 1$ so $d_a = (1 - \sum_{c' \in C_M(a)} \xi_{a,c'}^{i-1})/|C_Q(a)|$. It follows that

$$\xi_{a(c),c}^{i-1} + \sum_{a \in \tilde{A}_Q(c)} \frac{1 - \sum_{c' \in C_M(a)} \xi_{a,c'}^{i-1}}{|C_Q(a)|} \leq \tilde{q}_c,$$

which is equivalent to (28).

Having established that the linear program in Algorithm 3 has a solution, I now show that the solution is unique. I begin by introducing some notation that will be useful throughout the proof. Given a vector $(\xi_{a(c),c})_{c \in \tilde{C}}$, for every agent $a \in \tilde{A}$ let $S_a = \sum_{c \in C_M(a)} \xi_{a,c}$. Arbitrarily fix an agent $a \in \tilde{A}$ and a vector $S_{-a} = (S_{a'})_{a' \in \tilde{A} \setminus \{a\}}$. For every $c \in C_M(a)$, let

$$\theta_{a,c} = \tilde{q}_c - \sum_{a' \in \tilde{A}_Q(c)} \frac{1 - \sum_{c' \in C_M(a')} \xi_{a',c'}}{|C_Q(a)|} = \tilde{q}_c - \sum_{a' \in \tilde{A}_Q(c)} \frac{1 - S_{a'}}{|C_Q(a)|}. \quad (29)$$

Note that $\theta_{a,c}$ is the right-hand side of the second constraint of the linear program (LP 3) and is fixed by S_{-a} . Label the categories for which a is marginal such that $C_M(a) = \{c_1, c_2, \dots, c_{|C_M(a)|}\}$ with $\theta_{a,c_1} \geq \theta_{a,c_2} \geq \dots \geq \theta_{a,c_{|C_M(a)|}}$. For every $i = 1, \dots, |C_M(a)|$, let

$$T_i = \frac{1 - \sum_{j>i} \theta_{a,c_j}}{|C_Q(a)| + i}. \quad (30)$$

Finally, define the number $n = 0, 1, \dots, |C_Q(a)|$ as follows. If $T_i > \theta_{a,c_i}$ for every $i = 1, \dots, |C_M(a)|$, then $n = 0$. Otherwise, n is the largest number $i = 1, \dots, |C_M(a)|$ such that $T_i \leq \theta_{a,c_i}$; that is, $T_n \leq \theta_{a,c_n}$ and $T_i > \theta_{a,c_i}$ for all $i > n$. Having introduced the required notation, I next introduce the first intermediate result.

Claim 2. *For every $i \geq n$, $T_i > T_{i+1}$.*

Proof. It needs to be shown that

$$\frac{1 - \sum_{j>i} \theta_{a,c_j}}{|C_Q(a)| + i} > \frac{1 - \sum_{j>i+1} \theta_{a,c_j}}{|C_Q(a)| + i + 1},$$

which is equivalent to

$$\begin{aligned} |C_Q(a)| + i + 1 - (|C_Q(a)| + i + 1) \sum_{j>i} \theta_{a,c_j} &> |C_Q(a)| + i - (|C_Q(a)| + i) \sum_{j>i+1} \theta_{a,c_j} \\ \Leftrightarrow 1 + (|C_Q(a)| + i) \sum_{j>i+1} \theta_{a,c_j} &> (|C_Q(a)| + i + 1) \sum_{j>i} \theta_{a,c_j} \\ \Leftrightarrow 1 &> (|C_Q(a)| + i + 1) \theta_{a,c_{i+1}} + \sum_{j>i+1} \theta_{a,c_j} \\ \Leftrightarrow \theta_{a,c_{i+1}} &< \frac{1 - \sum_{j>i+1} \theta_{a,c_j}}{(|C_Q(a)| + i + 1)}. \end{aligned}$$

The right-hand side of the last inequality is equal to T_{i+1} ; therefore, $T_i > T_{i+1}$ is equivalent to $\theta_{a,c_{i+1}} < T_{i+1}$, which is satisfied by the definition of n since, by assumption, $i + 1 > n$. \square

Next, arbitrarily fix a vector $(y_i)_{i>n}$ such that $y_i \leq \theta_{a,c_i}$ for every $i > n$ and consider the following linear program:

$$\begin{aligned} &\max_{(\xi_{a,c_i})_{i=1}^{|C_M(a)|}} \sum_{i=1}^{|C_M(a)|} \xi_{a,c_i} \\ \text{subject to (i)} \quad &\xi_{a,c_i} \leq \frac{1 - \sum_{j \neq i} \xi_{a,c_j}}{|C_Q(a)| + 1} \quad \text{for every } i = 1, \dots, |C_M(a)| \\ \text{and (ii)} \quad &\xi_{a,c_i} = y_i \quad \text{for every } i > n. \end{aligned} \tag{LP 2}$$

The linear program (LP 2) can be interpreted as follows. For every $i > n$, ξ_{a,c_i} is set to y_i so only the first n elements (ξ_{a,c_i} for $i \leq n$) have to be chosen in order to maximize the sum, subject to constraint (i).

Claim 3. *For any vector $(y_i)_{i>n} \leq (\theta_{a,c_i})_{i>n}$, the unique solution to the linear program (LP 2) is the vector $(\xi_{a,c_i}^*)_{i=1}^{|C_M(a)|}$ such that*

$$\xi_{a,c_i}^* = \begin{cases} \frac{1 - \sum_{j>n} y_j}{|C_Q(a)| + n} & \text{if } i \leq n \\ y_i & \text{if } i > n. \end{cases}$$

Proof. I first show that $(\xi_{a,c_i}^*)_{i=1}^{|C_M(a)|}$ satisfies all constraints. Constraint (ii) is satisfied for all $i > n$ by definition; hence I focus on constraint (i).

Case 1: $i \leq n$. It needs to be shown that

$$\frac{1 - \sum_{j>n} y_j}{|C_Q(a)| + n} \leq \frac{1 - \sum_{j \neq i} \xi_{a,c_j}^*}{|C_Q(a)| + 1}. \quad (31)$$

By definition, we have that

$$1 - \sum_{j \neq i} \xi_{a,c_j}^* = 1 - (n-1) \frac{1 - \sum_{j>n} y_j}{|C_Q(a)| + n} - \sum_{j>n} y_j = \frac{(|C_Q(a)| + 1)(1 - \sum_{j>n} y_j)}{|C_Q(a)| + n}.$$

Therefore, the right-hand side of (31) is equal to $(1 - \sum_{j>n} y_j)/(|C_Q(a)| + n)$ and (31) holds with an equality.

Case 2: $i > n$. It needs to be shown that

$$\xi_{a,c_i}^* \leq \frac{1 - \sum_{j \neq i} \xi_{a,c_j}^*}{|C_Q(a)| + 1},$$

which is equivalent to

$$|C_Q(a)| \xi_{a,c_i}^* \leq 1 - \sum_{j=1}^{|C_M(a)|} \xi_{a,c_j}^*. \quad (32)$$

By the definition of $(\xi_{a,c_i})_{i=1}^{|C_M(a)|}$, (32) is equivalent to

$$\begin{aligned} |C_Q(a)| y_i &\leq 1 - n \frac{1 - \sum_{j>n} y_j}{|C_Q(a)| + n} - \sum_{j>n} y_j \\ y_i &\leq \frac{1 - \sum_{j>n} y_j}{|C_Q(a)| + n} \end{aligned} \quad (33)$$

As $y_i \leq \theta_{a,c_i}$ by definition, $\theta_{a,c_i} < T_i$ by the definition of n , $T_i < T_n$ by Claim 2, and T_n is equal to the right-hand side of (32), it can be concluded that (32) holds.

Having shown that the vector $(\xi_{a,c_i}^*)_{i=1}^{|C_M(a)|}$ satisfies all constraints, I proceed to show that it maximizes the objective, which makes it a solution to (LP 2). Consider any vector $(\xi_{a,c_i})_{i=1}^{|C_M(a)|}$ that satisfies constraints (i) and (ii); I show that $\sum_{i=1}^{|C_M(A)|} \xi_{a,c_i} \leq \sum_{i=1}^{|C_M(A)|} \xi_{a,c_i}^*$. Constraint (i) implies that, for every $i \leq n$,

$$|C_Q(a)| \xi_{a,c_i} \leq 1 - \sum_{j=1}^{|C_M(a)|} \xi_{a,c_j}.$$

Summing up over all $i \leq n$ yields

$$\begin{aligned}
\sum_{i \leq n} |C_Q(a)| \xi_{a,c_i} &\leq \sum_{i \leq n} \left(1 - \sum_{j=1}^{|C_M(a)|} \xi_{a,c_j}\right) \\
\Leftrightarrow |C_Q(a)| \sum_{i \leq n} \xi_{a,c_i} &\leq n - n \sum_{i=1}^{|C_M(a)|} \xi_{a,c_i} \\
\Leftrightarrow (|C_Q(a)| + n) \sum_{i \leq n} \xi_{a,c_i} &\leq n - n \sum_{i > n} \xi_{a,c_i} \\
\Leftrightarrow \sum_{i \leq n} \xi_{a,c_i} &\leq n \frac{1 - \sum_{i > n} \xi_{a,c_i}}{|C_Q(a)| + n} \\
\Leftrightarrow \sum_{i=1}^{|C_M(a)|} \xi_{a,c_i} &\leq n \frac{1 - \sum_{i > n} \xi_{a,c_i}}{|C_Q(a)| + n} + \sum_{i > n} \xi_{a,c_i}.
\end{aligned}$$

As constraint (ii) holds for every $i > n$, we obtain that

$$\sum_{i=1}^{|C_M(a)|} \xi_{a,c_i} \leq n \frac{1 - \sum_{i > n} y_i}{|C_Q(a)| + n} + \sum_{i > n} y_i. \quad (34)$$

By definition, we have that

$$\sum_{i=1}^{|C_M(a)|} \xi_{a,c_i}^* = n \frac{1 - \sum_{j > n} y_j}{|C_Q(a)| + n} + \sum_{j > n} y_j,$$

which combined with (34) implies that $\sum_{i=1}^{|C_M(a)|} \xi_{a,c_i} \leq \sum_{i=1}^{|C_M(a)|} \xi_{a,c_i}^*$.

Having shown that is a solution to the linear program (LP 2), I finally show that it is the unique solution. Let $(\xi_{a,c_i}^\#)_{i=1}^{|C_M(a)|}$ be a solution to (LP 2), it needs to be shown that $(\xi_{a,c_i}^\#)_{i=1}^{|C_M(a)|} = (\xi_{a,c_i}^*)_{i=1}^{|C_M(a)|}$. As $(\xi_{a,c_i}^\#)_{i=1}^{|C_M(a)|}$ is a solution to (LP 2), it maximizes the objective so

$$\sum_{i=1}^{|C_M(a)|} \xi_{a,c_i}^\# = \sum_{i=1}^{|C_M(a)|} \xi_{a,c_i}^* = n \frac{1 - \sum_{j > n} y_j}{|C_Q(a)| + n} + \sum_{j > n} y_j.$$

Moreover, as $(\xi_{a,c_i}^\#)_{i=1}^{|C_M(a)|}$ satisfies constraint (i), for every $i \leq n$, we have that

$$|C_Q(a)| \xi_{a,c_i}^\# \leq 1 - \sum_{j=1}^{|C_M(a)|} \xi_{a,c_j}^\#.$$

Therefore, it follows that, for every $i \leq n$

$$|C_Q(a)|\xi_{a,c_i}^\# \leq 1 - n \frac{1 - \sum_{j>n} y_j}{|C_Q(a)| + n} - \sum_{j>n} y_j$$

$$\xi_{a,c_i}^\# \leq \frac{1 - \sum_{j>n} y_j}{|C_Q(a)| + n}.$$

Then, by definition, it follows that $\xi_{a,c_i}^\# \leq \xi_{a,c_i}^*$ for every $i \leq n$. As $\xi_{a,c_i}^\# = \xi_{a,c_i}^*$ for all $i > n$ (by constraint (ii)) and $\sum_{i=1}^{|C_M(a)|} \xi_{a,c_i}^\# = \sum_{i=1}^{|C_M(a)|} \xi_{a,c_i}^*$ (as both vectors maximize the objective), we have that $(\xi_{a,c_i}^\#)_{i=1}^{|C_M(a)|} = (\xi_{a,c_i}^*)_{i=1}^{|C_M(a)|}$. \square

I next use Claim 3 to find the solution to the following linear program.

$$\begin{aligned} & \max_{(\xi_{a,c_i})_{i=1}^{|C_M(a)|}} \sum_{i=1}^{|C_M(a)|} \xi_{a,c_i} \\ \text{subject to (i)} \quad & \xi_{a,c_i} \leq \frac{1 - \sum_{j \neq i} \xi_{a,c_j}}{|C_Q(a)| + 1} \\ & \text{and (ii)} \quad \xi_{a,c_i} \leq \theta_{a,c_i} \quad \text{for every } i = 1, \dots, |C_M(a)|. \end{aligned} \tag{LP 3}$$

The linear program (LP 3) can be thought of as the linear program (LP 1) from Algorithm 3 in which S_{-a} has been fixed so only the vector $(\xi_{a,c_i})_{i=1}^{|C_M(a)|}$, that is the elements that involve agent a , remains to be optimized in order to maximize S_a .

Claim 4. *The unique solution to the linear program (LP 3) is the vector $(\xi_{a,c_i}^*)_{i=1}^{|C_M(a)|}$ such that, for every $i = 1, \dots, |C_M(a)|$,*

$$\xi_{a,c_i}^* = \begin{cases} T_n & \text{if } i \leq n \\ \theta_{a,c_i} & \text{if } i > n. \end{cases}$$

Proof. By Claim 3, $(\xi_{a,c_i}^*)_{i=1}^{|C_M(a)|}$ satisfies constraint (i); otherwise the unique solution to (LP 2) would not satisfy that linear program's constraints. By definition, for every $i \leq n$, $\xi_{a,c_i}^* = T_n \leq \theta_{a,c_i}$ and, for every $i > n$, $\xi_{a,c_i}^* = \theta_{a,c_i}$; hence $(\xi_{a,c_i}^*)_{i=1}^{|C_M(a)|}$ satisfies constraint (ii).

Having shown that $(\xi_{a,c_i}^*)_{i=1}^{|C_M(a)|}$ satisfies all constraints (which implies that (LP 3) has a solution), I now show that it is the unique solution to (LP 3). Let $(\xi_{a,c_i}^\#)_{i=1}^{|C_M(a)|}$ be a solution to (LP 3), I show that $(\xi_{a,c_i}^\#)_{i=1}^{|C_M(a)|} = (\xi_{a,c_i}^*)_{i=1}^{|C_M(a)|}$. By constraint (ii), $\xi_{a,c_i}^\# \leq \theta_{a,c_i}$ for every $i > n$; therefore Claim 3 implies that, for every $i \leq n$,

$$\xi_{a,c_i}^\# = \frac{1 - \sum_{j>n} \xi_{a,c_j}^\#}{|C_Q(a)| + n}, \tag{35}$$

as otherwise $(\xi_{a,c_i}^\#)_{i=1}^{|C_M(a)|}$ would not be optimal. Then,

$$\sum_{i=1}^{|C_M(a)|} \xi_{a,c_i}^\# = n \frac{1 - \sum_{i>n} \xi_{a,c_i}^\#}{|C_Q(a)| + n} + \sum_{i>n} \xi_{a,c_j}^\# = \frac{n + |C_Q(a)| \sum_{i>n} \xi_{a,c_i}^\#}{|C_Q(a)| + n}$$

so the objective is increasing in $(\xi_{a,c_i}^\#)_{i>n}$; therefore, the unique maximizer is obtained by setting $\xi_{a,c_i}^\# = \theta_{a,c_i}$ for every $i > n$, which by (35) implies that $\xi_{a,c_i}^\# = T_n$ for every $i \leq n$. It follows that $(\xi_{a,c_i}^\#)_{i=1}^{|C_M(a)|} = (\xi_{a,c_i}^*)_{i=1}^{|C_M(a)|}$. \square

I finally go back to the linear program (LP 1) in Algorithm 3 and use Claim 4 to show that (LP 1) has a unique solution. For any agent $a \in \tilde{A}$ and any S_{-a} , let $S_a(S_{-a})$ be the maximized objective function of the linear program (LP 3); in words, $S_a(S_{-a})$ is the largest sum that can be reached for the elements involving agent a given S_{-a} . If S_{-a} increases, then by (29), so does $\theta_{a,c}$ for every $c \in C_M(a)$. Therefore, constraint (ii) of the linear program (LP 3) is relaxed, meaning that the largest sum that can be reached increases as well. It follows that $S_a(S_{-a})$ is increasing in S_{-a} . Suppose towards a contradiction that (LP 1) has two solutions giving two distinct sum vectors $S^* = (S_a^*)_{a \in \tilde{A}}$ and $S^\# = (S_a^\#)_{a \in \tilde{A}}$. Then, for every $a \in \tilde{A}$, $S_a^* = S_a(S_{-a}^*)$ and $S_a^\# = S_a(S_{-a}^\#)$. Consider the sum vector $\bar{S} = (\bar{S}_a)_{a \in \tilde{A}}$ with $\bar{S}_a = \max\{S_a^*, S_a^\#\}$ for every $a \in \tilde{A}$. As S^* and $S^\#$ are distinct and derive from solutions of (LP 1), it must be that $\sum_{a \in \tilde{A}} \bar{S}_a > \sum_{a \in \tilde{A}} S_a^* = \sum_{a \in \tilde{A}} S_a^\#$; hence the allocation underpinning \bar{S} must violate some constraint of (LP 1). Consequently, there exists an agent $a \in \tilde{A}$ such that $\bar{S}_a > S_a(\bar{S}_{-a})$. By definition, $\bar{S}_{-a} \geq S_{-a}^*$; hence, as $S_a(S_{-a})$ is increasing in S_{-a} , it follows that $S_a(\bar{S}_{-a}) \geq S_a(S_{-a}^*)$. As $S_a^* = S_a(S_{-a}^*)$, it can be concluded that $\bar{S}_a > S_a^*$. Analogous reasoning yields that $\bar{S}_a > S_a^\#$ so $\bar{S}_a > \max\{S_a^*, S_a^\#\}$, a contradiction.

The preceding reasoning implies that every solution to (LP 1) yields the same sum vector, which I denote by S^* . By Claim 4, for every $a \in \tilde{A}$, there exists a unique vector $(\xi_{a,c}^*)_{c \in C_M(a)}$ such that $\sum_{c \in C_M(a)} \xi_{a,c}^* = S_a^* = S_a(S_{-a}^*)$. Therefore, the vector $(\xi_{a,c}^*)_{a \in \tilde{A}, c \in C_M(a)} = (\xi_{a(c),c}^*)_{c \in \tilde{C}}$ is the unique solution to (LP 1). \square

Proof of Lemma 25

Consider any Round i , any agent a and any category c . Suppose that a is qualified for c in Round i , that is $x_{a,c}^i \geq d_a^i$. By Lemma 23, for every $j > i$, $x_{a,c}^j \geq d_a^j$; therefore, a remains qualified for c in every subsequent round. Suppose next that a is marginal for c in Round i , that is $0 < x_{a,c}^i < d_a^i$. By Lemma 22, in any Round $j > i$, either $x_{a,c}^j \geq x_{a,c}^i$ or $x_{a,c}^j = d_a^j$. In both cases, $x_{a,c}^j > 0$; therefore, a is either marginal or qualified for c in every subsequent round. It follows that throughout the SRLP algorithm, for every agent-object pair (a, c) , a 's

status for c changes at most twice: once from unqualified to marginal and once from marginal to qualified.

Consider next a Round i in which the SRLP algorithm uses linear programming, i.e., $\delta^i = \delta^{LP}(x^{i-1}, d^{i-1})$. I show that either the algorithm ends in Round i or there exists an agent-object pair (a, c) such that a 's status for c changes in Round i . If $x_a^i \leq 1$ for every agent a , the algorithm ends in Round i so the remainder of the argument focuses on the case in which $x_a^i > 1$ for some agent a . Suppose first that a was not qualified for any category in Round $i-1$. The assumption that $x_a^i > 1$ implies by definition that $\xi_a^i = 1$ and by Lemma 18 that $\xi_{a,c}^i = d_a^i$ for some category c . It follows that a 's status for c has changed from either unqualified or marginal to qualified in Round i . Suppose next that a was qualified for some categories but not qualified for any. Then, by assumption, $\xi_{a,c}^{i-1} = d_a^{i-1}$ for all $c \in C_Q(a)$ and $\xi_{a,c}^{i-1} = 0$ for all $c \in C \setminus C_Q(a)$. By Lemma 18, $\xi_a^{i-1} = 1$ so $|C_Q^{i-1}(a)|d_a^{i-1} = 1$. As a is not marginal for any category, $a \notin \tilde{A}$ in the LP algorithm; hence $\delta^i = d^{i-1}$. It follows that $|C_Q^{i-1}(a)|\delta_a^i = 1$; moreover, by definition $x_{a,c}^i \leq \delta_a^i$ for every category c so $\sum_{c \in C_Q^{i-1}(a)} x_{a,c}^i \leq 1$. Then, the assumption that $x_a^i > 1$ implies that there is a category c such that $x_{a,c}^i > x_{a,c}^{i-1} = 0$ so a 's status for c has changed in Round i from unqualified to either marginal or qualified. Last, consider the remaining case in which a is qualified for at least one category and marginal for at least one category. In that case, $a \in \tilde{A}$ in the LP algorithm so $\delta^i = \delta^{LP}(x^{i-1}, d^{i-1})$. Let $(\xi_{a(c),c})_{c \in \tilde{C}}$ be the solution to the linear program (LP 1) in the LP algorithm. By construction, for every $c \in C_M(a)$, $\xi_{a,c}^* = \min\{\delta_a^i, \tilde{q}_c - \sum_{a' \in \tilde{A}_Q(c)} \delta_{a'}^i\}$; otherwise, one constraint in (LP 1) would hold with a strict inequality and $(\xi_{a(c),c})_{c \in \tilde{C}}$ would not be the solution to (LP 1). If $\xi_{a,c}^* = \delta_a^i$, then $x_{a,c}^i \leq \xi_{a,c}^*$ since $x_{a,c}^i \leq \delta_a^i$ by definition. If $\xi_{a,c}^* = \tilde{q}_c - \sum_{a' \in \tilde{A}_Q(c)} \delta_{a'}^i$, then, as $\tilde{q}_c = q_c - \sum_{a' \in A_Q(c) \setminus \tilde{A}_Q(c)} d_{a'}^{i-1}$ and $\delta_{a'}^i = d_{a'}^{i-1}$ for every $a' \in A_Q(c) \setminus \tilde{A}_Q(c)$, we have that $\xi_{a,c}^* = q_c - \sum_{a' \in A_Q(c)} \delta_{a'}^i$. Moreover, as a is marginal for c , $A_Q(c) = \hat{A}_{a,c}$ so $\xi_{a,c}^* = q_c - \sum_{a' \in \hat{A}_{a,c}} \delta_{a'}^i$. Then, for every agent $a' \in \hat{A}_{a,c}$, $\delta_{a'}^i + \sum_{\bar{a} \in \hat{A}_{a',c}} \delta_{\bar{a}}^i < q_c$ so by definition $x_{a',c}^i = \delta_{a'}^i$ for every $a' \in \hat{A}_{a,c}$. It follows that $\xi_{a,c}^* = q_c - \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^i$ or, equivalently, $\xi_{a,c}^* + \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^i = q_c$. As $x_{a,c}$ is a pre-allocation, $x_{a,c}^* + \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^i \leq q_c$ so it can be concluded that $\xi_{a,c}^* \leq \xi_{a,c}^*$. By construction, $|C_Q(a)|\delta_a^i + \sum_{c \in C_M(a)} \xi_{a,c}^* = 1$; therefore the fact that $x_{a,c}^i \leq \xi_{a,c}^*$ for every $c \in C_M(a)$ implies that $\sum_{c \in C_Q(a) \cup C_M(a)} x_{a,c}^i \leq 1$. Then, the assumption that $x_a^i > 1$ implies that $x_{a,c}^i > 0$ for some $c \in C_U(a)$ so a 's status for c changes in Round i from unqualified to either marginal or qualified.

Having established that the status of each agent for each category changes at most twice throughout the SRLP algorithm and that in every round in which the SRLP algorithm uses linear programming the status of at least one agent for at least one category change, I am now in a position to prove that the SRLP algorithm ends after fewer than $4|A||C|$ rounds. In fact, I will show that $4|A||C| - 2$ is an upper bound for the number of rounds of the SRLP

algorithm. First, suppose that fewer than two status changes occur in Round 1. In that case, no agent is qualified for any category in Round 1, that is $x_{a,c}^1 < d_a^1$ for every agent-category pair (a, c) . Then, by Lemma 18, $\xi_a^1 < 1$ and $d_a^1 = 1$ for every agent a . By definition, it follows that $x_a^1 < 1$ for every agent a so the SRLP algorithm ends in Round 1. Therefore, if the SRLP algorithm lasts more than one round, then at least two changes of status occur in Round 1. Suppose that, in some subsequent Round $i > 1$, no change of status occurs. Then, by construction, the SRLP algorithm uses linear programming in Round $i + 1$, which guarantees that a change of status occurs in Round $i + 1$. It follows that at least one change of status occurs every second round. Then, by the end of Round $4|A||C| - 3$, $4|A||C|$ changes of status must have occurred: 2 in Round 1 and $2|A||C| - 2$ in the $4|A||C| - 4$ subsequent rounds. As the status of each agent for each category can change at most twice, it follows that, at the end of Round $4|A||C| - 3$, every agent is qualified for every category. Then, $\xi_{a,c}^{4|A||C|-3} = d_a^{4|A||C|-3}$ for every agent-category pair (a, c) so, by Lemma 18, $\xi_a^{4|A||C|-3} = 1$ for every agent a . It follows that $d_a^{4|A||C|-3} = 1/|C|$ for every agent a . In Round $4|A||C| - 2$, by Lemma 23, $x_{a,c}^{4|A||C|-2} = d_a^{4|A||C|-3} = 1/|C|$ for every agent-object pair (a, c) . Then, for every agent a , $x_a^{4|A||C|-2} = \sum_{c \in C} 1/|C| = 1$ so the SRLP algorithm ends. \square

Proof of Lemma 26

While not entirely analogous, the reasoning is very similar to the proof of Lemma 10.

(Complies with eligibility requirements) By definition, if an agent a is not eligible for a category c , then $x_{a,c}^i = 0$.

(Non-wasteful) Consider any category c such that $\sum_{a \in A} x_{a,c}^i < q_c$ and any agent a who is eligible for c . It needs to be shown that $x_a^i \geq 1$.

Case 1: $x_{a,c}^i = \delta_a^i$. By the case assumption and Lemma 19, $x_{a,c}^i = \delta_a^i \geq d_a^i$; hence, by definition, $\xi_{a,c}^i = \min\{d_a^i, x_{a,c}^i\} = d_a^i$. By Lemma 18, it follows that $\xi_a^i = 1$ so, by definition, $x_a^i \geq \xi_a^i = 1$.

Case 2: $x_{a,c}^i < \delta_a^i$. In that case, Lemma 21 applies and yields $x_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} x_{a',c}^i = q_c$. Then, we have that $\sum_{a \in A} x_{a,c}^i \geq x_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} x_{a',c}^i = q_c$, which contradicts the assumption that $\sum_{a \in A} x_{a,c}^i < q_c$.

(Respects priorities) Consider an agent a such that $\xi_a^i < 1$ and arbitrarily fix a category c and a lower-priority agent $a' \in \check{A}_{a,c}$. It needs to be shown that $x_{a',c}^i = 0$. By Lemma 18, the assumption that $\xi_a^i < 1$ implies that $\xi_{a,c}^i < d_a^i$ so, by Lemma 20, $\xi_{a',c}^i = 0$.

(Respects equal sharing) Consider any agent a and any category c such that a is eligible for c and $x_{a,c}^i < \max_{c' \in C} \{x_{a,c'}^i\}$. It needs to be shown that $x_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} x_{a',c}^i = q_c$. By definition, $\max_{c' \in C} \{x_{a,c'}^i\} \leq \delta_a^i$; hence we have that $x_{a,c}^i < \delta_a^i$. Then, by Lemma 21, we have that $x_{a,c}^i + \sum_{a' \in \hat{A}_{a,c}} x_{a',c}^i = q_c$. \square

Proof of Lemma 27

Consider the allocation ξ^{SR} produced by the SR algorithm and the associated demand vector $d(\xi^{SR})$. By definition, for every agent a , $d_a(\xi^{SR}) = 1$ if $\xi_a^{SR} < 1$ and $d_a(\xi^{SR}) = \max_{c \in C} \{\xi_{a,c}\}$ if $\xi_a^{SR} = 1$. For every category c , it is possible to identify the agents who are *qualified*, *marginal*, and *unqualified* for c under the allocation ξ^{SR} . For every category c , I denote by $A_Q^{SR}(c) = \{a \in A : \xi_{a,c}^{SR} = d_a(\xi^{SR})\}$ the set of agents who are qualified for c under ξ^{SR} , by $A_M^{SR}(c) = \{a \in A : \xi_{a,c}^{SR} \in (0, d_a(\xi^{SR}))\}$ the set of agents who are marginal for c under ξ^{SR} , and by $A_U^{SR}(c) = \{a \in A : \xi_{a,c}^{SR} = 0\}$ the set of agents who are unqualified for a under ξ^{SR} . Similarly, for every agent a , I denote by $C_Q^{SR}(a) = \{c \in C : \xi_{a,c}^{SR} = d_a(\xi^{SR})\}$ the set of categories for which a is qualified, by $C_M^{SR}(a) = \{c \in C : \xi_{a,c}^{SR} \in (0, d_a(\xi^{SR}))\}$ the set of categories for which a is marginal, and by $C_U^{SR}(a) = \{c \in C : \xi_{a,c}^{SR} = 0\}$ the set of categories for which a is unqualified. The next result formalizes the properties of the preceding definitions.

Claim 5. *For every category c , $|A_M(c)| \leq 1$ and, for any two agents a and a' , either $a \in A_Q(c)$ and $a' \in A_M(c) \cup A_U(c)$ or $a \in A_M(c)$ and $a' \in A_U(c)$ implies that $a\pi_c a'$.*

Proof. ($|A_M(c)| \leq 1$.) Towards a contradiction, suppose that there exist two distinct agents $a, a' \in A_M(c)$ with $a \neq a'$. By assumption, we have that $0 < \xi_{a,c}^{SR} < d_a(\xi^{SR})$ and $0 < \xi_{a',c}^{SR} < d_{a'}(\xi^{SR})$; as ξ^{SR} complies with eligibility requirements, it follows that both a and a' are eligible for c . If $\xi_a^{SR} < 1$, then, as ξ^{SR} respects priorities, $\xi_{a',c}^{SR} = 0$, a contradiction. If $\xi_a^{SR} = 1$, then by definition $\xi_{a,c}^{SR} < d_a(\xi^{SR}) = \max_{c' \in C} \{\xi_{a,c'}^{SR}\}$. As ξ^{SR} respects equal sharing, it follows that $\xi_{a,c}^{SR} + \sum_{\bar{a} \in \hat{A}_{a,c}} \xi_{\bar{a},c} = q_c$ so $\xi_{a',c} = 0$, a contradiction.

($a \in A_Q(c)$ and $a' \in A_M(c) \cup A_U(c)$ implies that $a\pi_c a'$.) By assumption, $a \neq a'$ and $\xi_{a,c}^{SR} = d_a(\xi^{SR}) > 0$. Towards a contradiction, suppose that $a\pi_c a'$. If a' is not eligible for c , then neither is a since $a'\pi_a$; hence, as ξ^{SR} complies with eligibility requirements, $\xi_{a,c}^{SR} = 0$, a contradiction. For the remainder of the argument, I assume that a' is eligible for c . If $\xi_{a'}^{SR} < 1$, then, as ξ^{SR} respects priorities, the assumption that $a'\pi_c a$ implies that $\xi_{a,c}^{SR} = 0$, a contradiction. If $\xi_{a'}^{SR} = 1$, then by definition $d_{a'}^{SR} = \min_{c' \in C} \{\xi_{a',c'}^{SR}\}$; as $a' \in A_M(c) \cup A_U(c)$, $\xi_{a',c}^{SR} < d_{a'}^{SR}$ so it follows that $\xi_{a',c}^{SR} < \min_{c' \in C} \{\xi_{a',c'}^{SR}\}$. As ξ^{SR} respects equal sharing, we have that $\xi_{a',c}^{SR} + \sum_{\bar{a} \in \hat{A}_{a',c}} \xi_{\bar{a},c} = q_c$ so $\xi_{a,c}^{SR} = 0$, a contradiction.

($a \in A_M(c)$ and $a' \in A_U(c)$ implies that $a\pi_c a'$.) The reasoning is almost analogous to that of the preceding argument. By assumption, we have that $a \neq a'$ and $\xi_{a,c}^{SR} > 0$. Towards a contradiction, suppose that $a\pi_c a'$. If a' is not eligible for c , neither is a so $\xi_{a,c}^{SR} = 0$. Otherwise, if $\xi_{a'}^{SR} < 1$, we have that $\xi_{a,c}^{SR} = 0$ since ξ^{SR} respects priorities and if $\xi_{a'}^{SR} < 1$, we have that $\xi_{a,c}^{SR} = 0$ since ξ^{SR} respects equal sharing. Therefore, in all cases, $\xi_{a,c}^{SR} = 0$, a contradiction. \square

Next, I construct an alternative rationing problem $\bar{R} = (A, C, (\bar{\pi}_c))_{c \in C}, (q_c)_{c \in C}$ that is identical to the original rationing problem R except that every agent a who is unqualified for a category c under the SR allocation is ineligible for c in \bar{R} (whether or not a is eligible for c in R). That is, for every category c , $\bar{\pi}_c$ is constructed as follows: for any two agents a and a' , $a\bar{\pi}_c a'$ if and only if $a\pi_c a'$ and for every agent a , $a\bar{\pi}_c \emptyset$ if and only if $\xi_{a,c}^{SR} > 0$. In the original rationing problem R , the SR algorithm produces the allocation ξ^{SR} and the SRLP algorithm terminates after N rounds and produces the allocation x^N . In the alternative rationing problem \bar{R} , I denote by $\bar{x}_{a,c}^{SR}$ the allocation produced by the SR algorithm, by \bar{N} the number of rounds after which the SRLP algorithm ends, and by $\bar{x}^{\bar{N}}$ the allocation produced by the SRLP algorithm. In order to prove that $\xi^{SR} = x^N$, I show successively that $\xi^{SR} = \bar{\xi}^{SR}$, $\bar{\xi}^{SR} = \bar{x}^{\bar{N}}$, and $\bar{x}^{\bar{N}} = x^N$.

($\xi^{SR} = \bar{\xi}^{SR}$) For the original rationing problem R , I denote by x^i , ξ^i , and d^i the pre-allocation, the allocation, and the demand vector calculated by the SR algorithm in any given Round i . Similarly, for the alternative rationing problem \bar{R} , I denote by \bar{x}^i , $\bar{\xi}^i$, and \bar{d}^i the pre-allocation, the allocation, and the demand vector calculated by the SR algorithm in any given Round i . I also denote by d^0 and \bar{d}^0 the initial demand vectors in R and \bar{R} , respectively.

By definition, we have that $d^0 = \bar{d}^0 = \mathbf{1}$. Consider any Round $i \geq 1$ of the SR algorithm and suppose, towards an inductive argument, that $d^{i-1} = \bar{d}^{i-1}$. I show that $x^i = \bar{x}^i$ and $d^i = \bar{d}^i$. Arbitrarily fix an agent a and a category c . I first show that $x_{a,c}^i = \bar{x}_{a,c}^i$, considering separately the cases in which $\xi_{a,c}^{SR} > 0$ and $\xi_{a,c}^{SR} = 0$.

Case 1: $\xi_{a,c}^{SR} > 0$. As ξ^{SR} complies with eligibility requirements, the case assumption implies that a is eligible for c in the rationing problem R . By definition, the case assumption also implies that a is eligible for c in the alternative rationing problem \bar{R} . Again by definition, it follows that $x_{a,c}^i = \min\{d_a^{i-1}, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}\}$ and $\bar{x}_{a,c}^i = \min\{\bar{d}_a^{i-1}, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} \bar{d}_{a'}^{i-1}, 0\}\}$ so the induction hypothesis that $d^{i-1} = \bar{d}^{i-1}$ implies that $x_{a,c}^i = \bar{x}_{a,c}^i$. (Note that, by definition, the priority among agents is the same in both rationing problems so $\hat{A}_{a,c}$ can be used to calculate both $x_{a,c}^i$ and $\bar{x}_{a,c}^i$.)

Case 2: $\xi_{a,c}^{SR} = 0$. By definition, the case assumption implies that a is not eligible for c in the original rationing problem R ; hence, $\bar{x}_{a,c}^i = 0$ and it remains to show that $x_{a,c}^i = 0$. If a is not eligible either in the alternative rationing problem \bar{R} , it follows by definition that $x_{a,c}^i = 0$; therefore, for the remainder of the argument, I assume that a is eligible for c in R .

If $\xi_a^{SR} < 1$, as a is eligible for c and ξ^{SR} is non-wasteful, we have that $\sum_{a' \in A} \xi_{a',c}^{SR} = q_c$. Moreover, as ξ^{SR} respects priorities, $\xi_{a',c}^{SR} = 0$ for every $a' \in \check{A}_{a,c}$ and, by the case assumption, $\xi_{a,c}^{SR} = 0$. It follows that $\sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^{SR} = q_c$. If $\xi_a^{SR} = 1$, then $\xi_{a,c}^{SR} + \sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^{SR} = q_c$ as ξ^{SR} respects equal sharing so the case assumption implies that $\sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^{SR} = q_c$. Therefore, it

has been established that

$$\sum_{a' \in \hat{A}_{a,c}} \xi_{a',c}^{SR} = q_c. \quad (36)$$

Consider any agent $a' \in \hat{A}_{a,c}$. By Lemma 5, for every Round $j \geq 1$ of the SR algorithm, we have that $d_{a'}^j \geq \max_{c' \in C} \{\xi_{a',c'}\}$, which implies that $d_{a'}^j \geq \xi_{a',c}^j$. By Lemma 4, it follows that, for every $j \geq 1$, $d_{a'}^j \geq \lim_{j \rightarrow \infty} \xi_{a',c}^j$. Therefore, by Corollary 1, we have that $d_{a'}^j \geq \xi_{a',c}^{SR}$ for every $j \geq 1$, which implies that $d_{a'}^{i-1} \geq \xi_{a',c}^{SR}$. As the last inequality holds for every $a' \in \hat{A}_{a,c}$, (36) implies that $\sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1} \geq q_c$. As a is eligible for c , it can then be concluded that $x_{a,c}^i = \min\{d_a^{i-1}, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} d_{a'}^{i-1}, 0\}\} = 0$.

As a and c were chosen arbitrarily, it has been established that $x_{a,c}^i = \bar{x}_{a,c}^i$ for every agent a and every category c ; hence we have that $x^i = \bar{x}^i$. Then, by construction, it follows that $d^i = \bar{d}^i$. By induction, it can then be concluded that $x^i = \bar{x}^i$ for every $i \geq 1$. Therefore, by Corollary 1, we have that $\xi^{SR} = \lim_{i \rightarrow \infty} x^i = \lim_{i \rightarrow \infty} \bar{x}^i = \bar{\xi}^{SR}$.

($\bar{\xi}^{SR} = \bar{x}^N$) As each of $\bar{\xi}^{SR}$ and \bar{x}^N is an allocation of the alternative rationing problem \bar{R} that satisfies Axioms 1-4, it is sufficient to show that $\bar{\xi}^{SR}$ is the unique allocation of \bar{R} that satisfies Axiom 1-4. Let $\bar{\xi}^*$ be an allocation of \bar{R} that satisfies Axioms 1-4. I show that $\bar{\xi}^* = \bar{\xi}^{SR}$.

First, by Theorem 3, $\bar{\xi}^*$ and $\bar{\xi}^{SR}$ generate the same aggregate allocation; moreover, by Theorem 4, the demand vector associated with $\bar{\xi}^*$ is weakly smaller than the one associated with $\bar{\xi}^{SR}$. It follows that

$$d_a(\bar{\xi}^*) \leq d_a(\bar{\xi}^{SR}) \quad \text{and} \quad \bar{\xi}_a^* = \bar{\xi}_a^{SR} \quad \text{for every } a \in A. \quad (37)$$

Consider any agent-object pair (a, c) such that a is not qualified for c under $\bar{\xi}^{SR}$, i.e., $\bar{\xi}_{a,c}^{SR} = 0$. As $\bar{\xi}^{SR} = \xi^{SR}$, a is not qualified for c under ξ^{SR} ; hence, by definition, a is not eligible for c in the alternative rationing problem \bar{R} . As $\bar{\xi}^*$ complies with eligibility requirements, it follows that $\bar{\xi}_{a,c}^* = 0$ so we have that

$$\bar{\xi}_{a,c}^* = 0 \quad \text{for every } (a, c) \in A \times C \text{ such that } \bar{\xi}_{a,c}^{SR} = 0. \quad (38)$$

Consider next any agent-object pair (a, c) such that a is qualified for c under $\bar{\xi}^{SR}$, i.e., $\bar{\xi}_{a,c}^{SR} = d_a(\bar{\xi}^{SR})$. By definition, $\bar{\xi}_{a,c}^* \leq d_a(\bar{\xi}^*)$ and, by (37), $d_a(\bar{\xi}^*) \leq d_a(\bar{\xi}^{SR})$; therefore, we have that

$$\bar{\xi}_{a,c}^* \leq d_a(\bar{\xi}^*) \leq d_a(\bar{\xi}^{SR}) = \bar{\xi}_{a,c}^{SR}. \quad (39)$$

Consider any agent $a' \in \hat{A}_{a,c}$ and suppose, towards a contradiction, that $\bar{\xi}_{a',c}^{SR} < d_{a'}(\xi^{SR})$. If $\bar{\xi}_{a'}^{SR} < 1$, then the fact that $\bar{\xi}_{a,c}^{SR} = d_a(\bar{\xi}^{SR}) > 0$ implies that $\bar{\xi}^{SR}$ does not respect priorities, a

contradiction. If $\bar{\xi}_{a'}^{SR} = 1$, then by definition $d_{a'}(\bar{\xi}^{SR}) = \max_{c' \in C} \{\xi_{a',c'}^{SR}\}$ so we have that $\bar{\xi}_{a',c}^{SR} < \max_{c \in C} \{\bar{\xi}_{a',c'}^{SR}\}$. As $\bar{\xi}^{SR}$ respects equal sharing, it must then be that $\bar{\xi}_{a',c}^{SR} + \sum_{\bar{a} \in \hat{A}_{a',c}} \bar{\xi}_{\bar{a},c}^{SR} = q_c$. However, as $\bar{\xi}_{a,c}^{SR} = d_a(\bar{\xi}^{SR}) > 0$, we have that $\bar{\xi}_{a,c}^{SR} + \sum_{\bar{a} \in \hat{A}_{a,c}} \bar{\xi}_{\bar{a},c}^{SR} > q_c$, a contradiction. It can then be concluded that $\bar{\xi}_{a',c}^{SR} = d_{a'}(\xi^{SR})$; hence (39) applies to a' and we have that $\bar{\xi}_{a',c}^* \leq \bar{\xi}_{a',c}^{SR}$ for all $a' \in \hat{A}_{a,c}$. It follows that

$$\bar{\xi}_{a,c}^* + \sum_{a' \in \hat{A}_{a,c}} \bar{\xi}_{a',c}^* \leq \bar{\xi}_{a,c}^{SR} + \sum_{a' \in \hat{A}_{a,c}} \bar{\xi}_{a',c}^{SR}. \quad (40)$$

Suppose towards a contradiction that $\bar{\xi}_{a,c}^* < d_a(\bar{\xi}^*)$. By (37), we have that $\bar{\xi}_{a,c}^* < \bar{\xi}_{a,c}^{SR}$ so (40) holds with a strict inequality, which implies that $\bar{\xi}_{a,c}^* + \sum_{a' \in \hat{A}_{a,c}} \bar{\xi}_{a',c}^* < q_c$. If $\bar{\xi}_a^* < 1$, then $\bar{\xi}^*$ either is wasteful or does not respect priorities while if $\bar{\xi}_a^* = 1$, then $d_a(\bar{\xi}^*) = \max_{c' \in C} \{\bar{\xi}_{a,c'}^*\}$ so $\bar{\xi}_{a,c}^* < \max_{c' \in C} \{\bar{\xi}_{a,c'}^*\}$ and $\bar{\xi}^*$ does not respect equal sharing. As $\bar{\xi}^*$ satisfies Axioms 1-4 by assumption, both cases yield a contradiction. It follows that $\bar{\xi}_{a,c}^* = d_a(\bar{\xi}^*)$. Then, if $d_a(\bar{\xi}^*) = 1$, $\bar{\xi}_{a,c}^* = 1$ so $\bar{\xi}_a^* = 1$ and if $d_a(\bar{\xi}^*) < 1$, $\bar{\xi}_a^* = 1$ by definition. As a and c were chosen arbitrarily, it can then be concluded that

$$\bar{\xi}_{a,c}^* = d_a(\bar{\xi}^*) \quad \text{and} \quad \bar{\xi}_a^* = 1 \quad \text{for every } (a,c) \in A \times C \text{ such that } \bar{\xi}_{a,c}^{SR} = d_a(\bar{\xi}^{SR}). \quad (41)$$

Let $A_Q^{SR} = \cup_{c \in C} \{A_Q^{SR}(c)\}$ be the set of agents who are qualified for at least one category under $\bar{\xi}^{SR}$. For every agent $a \in A_Q^{SR}$, (41) implies that $\bar{\xi}_a^* = 1$ so $\sum_{a \in C_Q^{SR}(a)} \bar{\xi}_{a,c}^* + \sum_{a \in C_M^{SR}(a)} \bar{\xi}_{a,c}^* + \sum_{a \in C_U^{SR}(a)} \bar{\xi}_{a,c}^* = 1$. By (38) and (41), it follows that $|C_Q^{SR}(a)| d_a(\bar{\xi}^*) + \sum_{a \in C_M^{SR}(a)} \bar{\xi}_{a,c}^* = 1$; hence we have that

$$d_a(\bar{\xi}^*) = \frac{1 - \sum_{c \in C_M^{SR}(a)} \bar{\xi}_{a,c}^*}{|C_Q^{SR}(a)|} \quad \text{for every } a \in A_Q^{SR}. \quad (42)$$

Next, let $C_M^{SR} = \cup_{a \in A} C_M^{SR}(a)$ be the set of categories that have a marginal agent and, for every $c \in C_M^{SR}$, let $a(c)$ be the agent who is marginal for c (by Claim 5, $a(c)$ is unique). Consider any category $c \in C_M^{SR}$. By definition, $\bar{\xi}_{a,c}^{SR} = 0$ for every $a \in A_U^{SR}(c)$ so $\sum_{a \in A} \bar{\xi}_{a,c}^{SR} = \bar{\xi}_{a(c),c}^{SR} + \sum_{a' \in A_Q^{SR}(c)} \bar{\xi}_{a',c}^{SR}$. As Claim 5 implies that $A_Q^{SR}(c) = \hat{A}_{a(c),c}$, we have that $\sum_{a \in A} \bar{\xi}_{a,c}^{SR} = \bar{\xi}_{a(c),c}^{SR} + \sum_{a' \in \hat{A}_{a(c),c}} \bar{\xi}_{a',c}^{SR}$. If $\bar{\xi}_{a(c)}^{SR} < 1$, then, as $\bar{\xi}^{SR}$ is non-wasteful and respects priorities, it must be that $\bar{\xi}_{a(c),c}^{SR} + \sum_{a' \in \hat{A}_{a(c),c}} \bar{\xi}_{a',c}^{SR} = q_c$. If $\bar{\xi}_{a(c)}^{SR} = 1$, then by definition $d_a(\bar{\xi}^{SR}) = \max_{c' \in C} \{\bar{\xi}_{a(c),c'}^{SR}\}$ so $\bar{\xi}_{a(c),c}^{SR} < \max_{c' \in C} \{\bar{\xi}_{a(c),c'}^{SR}\}$. As $\bar{\xi}^{SR}$ respects equal sharing, it must then be

that $\bar{\xi}_{a(c),c}^{SR} + \sum_{a' \in \hat{A}_{a(c),c}} \bar{\xi}_{a',c}^{SR} = q_c$. The preceding argument has established that

$$\sum_{a \in A} \bar{\xi}_{a,c}^{SR} = q_c \quad \text{for every } c \in C_M^{SR}. \quad (43)$$

As $\bar{\xi}^*$ is an allocation, it must be that $\sum_{a \in A} \bar{\xi}_{a,c}^* \leq q_c$, and therefore (43) implies that $\sum_{a \in A} \bar{\xi}_{a,c}^* \leq \sum_{a \in A} \bar{\xi}_{a,c}^{SR}$.

Consider next a category $c \in C \setminus C_M^{SR}$. For every $a \in A_U^{SR}(c)$, $\bar{\xi}_{a,c}^{SR} = 0$ by definition and $\bar{\xi}_{a,c}^* = 0$ by (38). As $A_M^{SR}(c) = \emptyset$ by assumption, it follows that $\sum_{a \in A} \bar{\xi}_{a,c}^{SR} = \sum_{a \in A_Q^{SR}(c)} \bar{\xi}_{a,c}^{SR}$ and $\sum_{a \in A} \bar{\xi}_{a,c}^* = \sum_{a \in A_Q^{SR}(c)} \bar{\xi}_{a,c}^*$. For every $a \in A_Q^{SR}(c)$, $\bar{\xi}_{a,c}^{SR} = d_a(\bar{\xi}^{SR})$ by definition and $\bar{\xi}_{a,c}^* = d_a(\bar{\xi}^*)$ by (41), which implies that $\sum_{a \in A} \bar{\xi}_{a,c}^{SR} = |A_Q^{SR}(c)| d_a(\bar{\xi}^{SR})$ and $\sum_{a \in A} \bar{\xi}_{a,c}^* = |A_Q^{SR}(c)| d_a(\bar{\xi}^*)$. As $d_a(\bar{\xi}^*) \leq d_a(\bar{\xi}^{SR})$, it must then be that $\sum_{a \in A} \bar{\xi}_{a,c}^* \leq \sum_{a \in A} \bar{\xi}_{a,c}^{SR}$. The argument in the last two paragraphs allows concluding that

$$\sum_{a \in A} \bar{\xi}_{a,c}^* \leq \sum_{a \in A} \bar{\xi}_{a,c}^{SR} \quad \text{for every } c \in C. \quad (44)$$

By (37), $|\bar{\xi}^*| = \sum_{a \in A} \bar{\xi}_a^* = \sum_{a \in A} \bar{\xi}_a^{SR} = |\bar{\xi}^{SR}|$; hence, by definition, we have that $\sum_{c \in C} \sum_{a \in A} \bar{\xi}_{a,c}^* = \sum_{c \in C} \sum_{a \in A} \bar{\xi}_{a,c}^{SR}$, which combined with (44) implies that

$$\sum_{a \in A} \bar{\xi}_{a,c}^* = \sum_{a \in A} \bar{\xi}_{a,c}^{SR} \quad \text{for every } c \in C. \quad (45)$$

For every category $c \in C_M^{SR}$, combining (43) and (45) yields $\sum_{a \in A} \bar{\xi}_{a,c}^* = q_c$. As $\bar{\xi}_{a,c}^* = 0$ for every $a \in A_U^{SR}(c)$ by (38) and $\bar{\xi}_{a,c}^* = d_a(\bar{\xi}^*)$ for every $a \in A_Q^{SR}(c)$ by (41), it must be that

$$\bar{\xi}_{a(c),c}^* + \sum_{a' \in A_Q^{SR}(c)} d_{a'}(\bar{\xi}^*) = q_c \quad \text{for every } c \in C_M^{SR}. \quad (46)$$

Combining (42) and (46), it follows that any allocation $\bar{\xi}^*$ that satisfies Axioms 1-4 must satisfy the following linear system of equations:

$$\bar{\xi}_{a(c),c}^* + \sum_{a' \in A_Q^{SR}(c)} \frac{1 - \sum_{c \in C_M^{SR}(a')} \bar{\xi}_{a',c}^*}{|C_Q^{SR}(a')|} = q_c \quad \text{for every } c \in C_M^{SR}. \quad (47)$$

The linear system of equations defined in (47) has $|C_M^{SR}|$ variables and $|C_M^{SR}|$ equations. For any $\bar{\xi}^*$ that satisfies Axioms 1-4, $(\bar{\xi}_{a(c),c}^*)_{c \in C_M^{SR}}$ must be a solution to the linear system of equations defined in (47); moreover, for every agent-category pair (a, c) such that $a \in A_Q^{SR}(c)$,

$\bar{\xi}_{a,c}^*$ must be determined by (41) and (42), and for every agent-category pair (a, c) such that $a \in A_U^{SR}(c)$, it must be that $\bar{\xi}_{a,c}^* = 0$, as per (38). As $\bar{\xi}^{SR}$ satisfies Axioms 1-4, $(\bar{\xi}_{a(c),c}^{SR})_{c \in C_M^{SR}}$ is a solution to the linear system of equations defined in (47). If all $|C_M^{SR}|$ equations in (47) are linearly independent, then $(\bar{\xi}_{a(c),c}^{SR})_{c \in C_M^{SR}}$ is the unique solution so $\bar{\xi}^{SR}$ is the unique allocation in the alternative allocation problem \bar{R} to satisfy Axioms 1-4; hence the proof is complete. In the remainder of the proof, I show that the opposite case leads to a contradiction.

Towards a contradiction, suppose that the linear system of equations defined in (47) has strictly fewer than $|C_M^{SR}|$ linearly independent equations. Then, there is at least one degree of freedom; hence, arbitrarily fixing a category $c \in C_M^{SR}$, for any value of $\bar{\xi}_{(a(c),c)}^*$ there exists a vector $(\bar{\xi}_{a(c'),c'}^*)_{c' \in C_M^{SR} \setminus \{c\}}$ such that $(\bar{\xi}_{a(c'),c'}^*)_{c' \in C_M^{SR}}$ is a solution to the linear system of equations defined in (47).

Given an arbitrarily small positive number $\epsilon > 0$, I construct an allocation $\bar{\xi}^\epsilon$ as follows. Let $\bar{\xi}_{a(c),c}^\epsilon = \bar{\xi}_{a(c),c}^{SR} + \epsilon$ and, for every $c' \in C_M^{SR} \setminus \{c\}$, let $\bar{\xi}_{a(c'),c'}^\epsilon$ be such that $(\bar{\xi}_{a(c'),c'}^\epsilon)_{c' \in C_M^{SR}}$ is a solution to the system of equations defined in (47). Then, for every agent-category pair (a, c) such that $a \in A_Q^{SR}(c)$, let $\bar{\xi}_{a,c}^\epsilon$ be determined by (41) and (42) and, for every agent-category pair (a, c) such that $a \in A_U^{SR}(c)$, let $\bar{\xi}_{a,c}^\epsilon = 0$ (as per (38)). By definition, for every $c' \in C_M^{SR}$, $0 < \bar{\xi}_{a(c'),c'}^{SR} < d_{a(c')}(\bar{\xi}^{SR})$. As all equations in (41) and (47) are linear, there exists a value $\epsilon > 0$ small enough so that $0 < \bar{\xi}_{a(c'),c'}^\epsilon < d_{a(c')}(\bar{\xi}^\epsilon)$ for every $c' \in C_M^{SR}$. Fixing such an ϵ , I next show that $\bar{\xi}^\epsilon$ satisfies Axioms 1-4.

By definition, $\bar{\xi}_{a,c}^\epsilon = 0$ for every agent-category pair (a, c) such that a is not eligible for c in the alternative problem \bar{R} so $\bar{\xi}^\epsilon$ complies with eligibility requirements. I next introduce a small result that is useful to prove that $\bar{\xi}^\epsilon$ satisfies the other three axioms. For every agent-object pair (a, c) such that $a \in A_Q^{SR}(c)$, by the definition $\bar{\xi}^\epsilon$ satisfies the first part of (41): $\bar{\xi}_{a,c}^\epsilon = d_a(\bar{\xi}^\epsilon)$. I show that the second part of (41) also holds. If $d_a(\bar{\xi}^\epsilon) = 1$, then $\bar{\xi}_{a,c}^\epsilon = 1$ so $\bar{\xi}_a^\epsilon = 1$. If $d_a(\bar{\xi}^\epsilon) < 1$, then by definition $\bar{\xi}_a^\epsilon = 1$. Therefore, we have that

$$\bar{\xi}_a^\epsilon = 1 \quad \text{for every } (a, c) \in A \times C \text{ such that } \bar{\xi}_{a,c}^{SR} = d_a(\bar{\xi}^{SR}). \quad (48)$$

Suppose towards a contradiction that $\bar{\xi}^\epsilon$ is wasteful. Then, there exists an agent-category pair (a, c) such that $\sum_{a' \in A} \bar{\xi}_{a',c}^\epsilon < q_c$, $\bar{\xi}_a^\epsilon < 1$, and a is eligible for c in \bar{R} . By (47) the fact that $\sum_{a' \in A} \bar{\xi}_{a',c}^\epsilon < q_c$ implies that $c \notin C_M^{SR}$ so $A_M^{SR}(a) = \emptyset$; hence it must be that either $a \in A_Q^{SR}(c)$ or $a \in A_U^{SR}(c)$. If $a \in A_Q^{SR}(c)$, then by (48), $\bar{\xi}_a^\epsilon = 1$, a contradiction. If $a \in A_U^{SR}(c)$, then by definition a is not eligible for c in \bar{R} , a contradiction. It can then be concluded that $\bar{\xi}^\epsilon$ is non-wasteful. Next, suppose towards a contradiction that $\bar{\xi}^\epsilon$ does not respect priorities. Then, there exists an agent-category pair (a, c) and a lower-priority agent $a' \in \check{A}_{a,c}$ such that $\bar{\xi}_a^\epsilon < 1$ and $\bar{\xi}_{a',c}^\epsilon > 0$. If $a \in A_Q^{SR}(c)$, then (48) implies that $\bar{\xi}_a^\epsilon = 1$, a contradiction. If

$a \notin A_Q^{SR}(c)$, then by Claim 5, $a' \in A_U^{SR}(c)$ so, by (38), $\bar{\xi}_{a',c}^\epsilon = 0$, a contradiction. It follows that $\bar{\xi}^\epsilon$ respects priorities. Finally, suppose towards a contradiction that $\bar{\xi}^\epsilon$ does not respect equal sharing. Then, there exists an agent-category pair (a, c) such that a is eligible for c in \bar{R} , $\bar{\xi}_{a,c}^\epsilon < \max_{c' \in C} \{\bar{\xi}_{a,c'}^\epsilon\}$, and $\bar{\xi}_{a,c}^\epsilon + \sum_{a' \in \hat{A}_{a,c}} \bar{\xi}_{a',c}^\epsilon < q_c$. By definition, $\max_{c' \in C} \{\bar{\xi}_{a,c'}^\epsilon\} \leq d_a(\bar{\xi}^\epsilon)$, which by (41) implies that $a \notin A_Q^{SR}(c)$. Moreover, the assumption that a is eligible for c in \bar{R} implies by definition that $a \notin A_U^{SR}(c)$ so it must be that $a \in A_M^{SR}(c)$. In that case, however, we have that $\bar{\xi}_{a,c}^\epsilon + \sum_{a' \in A_Q^{SR}(c)} \bar{\xi}_{a',c}^\epsilon = q_c$ by (47) and that $A_Q^{SR}(c) = \hat{A}_{a,c}$ by Claim 5. It follows that $\bar{\xi}_{a,c}^\epsilon + \sum_{a' \in \hat{A}_{a,c}} \bar{\xi}_{a',c}^\epsilon = q_c$, a contradiction.

The preceding argument has established that there exists an allocation $\bar{\xi}^\epsilon$ in the alternative rationing problem \bar{R} that satisfies Axioms 1-4. By definition, there exists a category $c \in C_M^{SR}$ such that $\bar{\xi}_{a(c),c}^\epsilon = \bar{\xi}_{a(c),c}^{SR} + \epsilon > \bar{\xi}_{a(c),c}^{SR}$. Moreover, by construction, both $\bar{\xi}^\epsilon$ and $\bar{\xi}^{SR}$ satisfy (46) so we have that

$$\bar{\xi}_{a(c),c}^\epsilon + \sum_{a' \in A_Q^{SR}(c)} d_{a'}(\bar{\xi}^\epsilon) = \bar{\xi}_{a(c),c}^{SR} + \sum_{a' \in A_Q^{SR}(c)} d_{a'}(\bar{\xi}^{SR}) = q_c.$$

Then, as $\bar{\xi}_{a(c),c}^\epsilon > \bar{\xi}_{a(c),c}^{SR}$, there must exist an agent $a' \in A_Q^{SR}(c)$ such that $d_{a'}(\bar{\xi}^\epsilon) < d_{a'}(\bar{\xi}^{SR})$, which contradicts (37).

($\bar{x}^N = x^N$) Though not entirely analogous, the reasoning is this last part of the proof is similar to that of the first part (which shows that $\xi^{SR} = \bar{\xi}^{SR}$). The main difference is that I follow the SRLP algorithm instead of the SR algorithm. For the original rationing problem R , I denote by x^i , δ^i , and d^i the pre-allocation, the LP demand vector, and the demand vector calculated by the SRLP algorithm in any given Round i . Similarly, for the alternative rationing problem \bar{R} , I denote by \bar{x}^i , $\bar{\delta}^i$, and \bar{d}^i the pre-allocation, the LP demand vector, and the demand vector calculated by the SRLP algorithm in any given Round i . I also denote by d^0 and \bar{d}^0 the initial demand vectors in R and \bar{R} , respectively.

By definition, we have that $d^0 = \bar{d}^0 = \delta^1 = \bar{\delta}^1 = \mathbf{1}$. Consider any Round $i = 1, \dots, \min\{N, \bar{N}\}$ of the SRLP algorithm and suppose, towards an inductive argument, that $\delta^i = \bar{\delta}^i$. I show that $x^i = \bar{x}^i$ and, if $i < \min\{N, \bar{N}\}$, $\delta^{i+1} = \bar{\delta}^{i+1}$. Arbitrarily fix an agent a and a category c . I first show that $x_{a,c}^i = \bar{x}_{a,c}^i$, considering separately the cases in which $\xi_{a,c}^{SR} > 0$ and $\xi_{a,c}^{SR} = 0$.

Case 1: $\xi_{a,c}^{SR} > 0$. As ξ^{SR} complies with eligibility requirements, the case assumption implies that a is eligible for c in the rationing problem R . By definition, the case assumption also implies that a is eligible for c in the alternative rationing problem \bar{R} . Again by definition, it follows that $x_{a,c}^i = \min\{\delta_a^i, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} \delta_{a'}^i, 0\}\}$ and $\bar{x}_{a,c}^i = \min\{\bar{\delta}_a^i, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} \bar{\delta}_{a'}^i, 0\}\}$ so the induction hypothesis that $\delta^i = \bar{\delta}^i$ implies that

$$x_{a,c}^i = \bar{x}_{a,c}^i.$$

Case 2: $\xi_{a,c}^{SR} = 0$. By definition, the case assumption implies that a is not eligible for c in the alternative rationing problem \bar{R} ; hence, $\bar{x}_{a,c}^i = 0$ and it remains to show that $x_{a,c}^i = 0$. If a is not eligible either in the alternative rationing problem R , it follows by definition that $x_{a,c}^i = 0$; therefore, for the remainder of the argument, I assume that a is eligible for c in R .

As $\bar{x}^{\bar{N}} = \bar{\xi}^{SR} = \xi^{SR}$, (36) implies that

$$\sum_{a' \in \hat{A}_{a,c}} \bar{x}_{a',c}^{\bar{N}} = q_c. \quad (49)$$

Consider any agent $a' \in \hat{A}_{a,c}$. By construction, $\bar{x}^{\bar{N}}$ is an allocation (otherwise the SRLP algorithm would not end in Round \bar{N}) so by definition $d_{a'}(\bar{x}^{\bar{N}}) = \max_{c' \in C} \{x_{a',c'}^{\bar{N}}\}$ if $\bar{x}_a^{\bar{N}} = 1$ and $d(\bar{x}^{\bar{N}}) = 1$ if $\bar{x}_a^{\bar{N}} < 1$. It follows that $\bar{d}_{a'}^{\bar{N}} = d_{a'}(\bar{x}^{\bar{N}}) \geq \max_{c' \in C} \{x_{a',c'}^{\bar{N}}\} \geq \bar{x}_{a',c}^{\bar{N}}$. As a' was chosen arbitrarily, we have that $\bar{d}_{a'}^{\bar{N}} \geq \bar{x}_{a',c}^{\bar{N}}$ for every $a' \in \hat{A}_{a,c}$; hence (49) implies that $\sum_{a' \in \hat{A}_{a,c}} \bar{d}_{a'}^{\bar{N}} \geq q_c$. Then, by Lemma 19, we have that $\sum_{a' \in \hat{A}_{a,c}} \bar{\delta}_{a'}^i \geq q_c$. As a is eligible for c , it follows by definition that $\bar{x}_{a,c}^i = \min\{\bar{\delta}_a^i, \max\{q_c - \sum_{a' \in \hat{A}_{a,c}} \bar{\delta}_{a'}^i, 0\}\} = 0$.

As a and c were chosen arbitrarily, it has been established that $x_{a,c}^i = \bar{x}_{a,c}^i$ for every agent a and every category c ; hence we have that $x^i = \bar{x}^i$. Then, by construction, it follows that $d^i = \bar{d}^i$. Moreover, if $i < \min\{N, \bar{N}\}$, then by construction the fact that $x^i = \bar{x}^i$ and $d^i = \bar{d}^i$ implies that $\delta^{i+1} = \bar{\delta}^{i+1}$. By induction, it follows that $x^{\min\{N, \bar{N}\}} = \bar{x}^{\min\{N, \bar{N}\}}$ so it must be that, in both rationing problems, the SRLP algorithm ends in the same Round $N = \bar{N}$ and produces the same allocation $x^N = \bar{x}^{\bar{N}}$. \square