



The Algorithmic Landscape of Priority-Respecting Allocations

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Based on joint work with Sid Banerjee and David Kempe

Paper: <https://arxiv.org/abs/2204.13019>

A Motivating Example: Pandemic Response

Supply-chain constraints place limits on available resources

- Ventilators, Vaccines, Anti-viral treatments

Many considerations for who to prioritize

- Healthcare / essential workers
- Individuals with comorbidities
- Residents of high-density housing

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What is a “fair” way to allocate care?

Commonly used priority schemes have issues

Formalizing the Reserve Allocation Setting

Agents : \mathcal{A} , $n = |\mathcal{A}|$

- *Unit demand* for the resource
- *Indifferent* about how they are allocated

Categories : \mathcal{C} , $m = |\mathcal{C}|$

Each category $c \in \mathcal{C}$ has:

Quota : $q_c \in \mathbb{N}$, $q = \sum_{c \in \mathcal{C}} q_c$

Eligibility : $\mathcal{E}_c \subseteq \mathcal{A}$

Priorities : *Total pre-order* \succeq_c over \mathcal{E}_c

- \succeq_c separates agents into ranked *priority tiers*
- $a \succeq_c a' \implies c$ gives priority to a over a'

Visualizing an Instance

α (2)	β (1)	γ (1)
<i>a</i>	<i>b</i>	<i>b</i>
<i>b</i>	<i>c</i> , <i>e</i>	<i>a</i>
<i>c</i>	<i>d</i>	
<i>d</i>		
<i>e</i>		

Feasible Allocations

Goal: Select an allocation map $\varphi : \mathcal{A} \rightarrow \mathcal{C} \cup \{\emptyset\}$

Determines recipient set $\mathcal{A} \setminus \varphi^{-1}(\emptyset)$

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Eligibility Respecting [ER]: Categories only allocate to eligible agents

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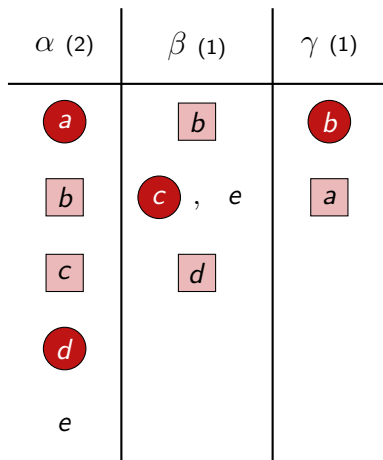
Priority Respecting [PR]: A category allocates to an agent only if all higher-priority agents have been allocated

$$\varphi(a') = c \wedge a \succeq_c a' \implies \varphi(a) \neq \emptyset$$

Visualizing an Allocation

α (2)	β (1)	γ (1)
	b	
b	 , e	a
c	d	
		
e		

Visualizing an Allocation



Pareto Efficient [PE]: No alternate allocation satisfying [ER], [QR], [PR], allocates to a strict superset of agents

$$\neg \exists \psi : \psi^{-1}(\emptyset) \subsetneq \varphi^{-1}(\emptyset)$$

Locating Good Allocations

Pareto Efficient [PE]: No alternate allocation satisfying [ER], [QR], [PR], allocates to a strict superset of agents

$$\neg \exists \psi : \psi^{-1}(\emptyset) \subsetneq \varphi^{-1}(\emptyset)$$

Is there an efficient algorithm to find allocations with these properties?

Existing Approaches

Pathak et al (2021) [1]: Variant of Deferred Acceptance [2]

- Agents have arbitrary preferences over eligible categories
- Run DA with agents proposing to categories
- [QR], [ER], [PR], not necessarily [PE]

Delacrétaz (2021) [3]: Simultaneous Reserves Algorithm

- “Water-filling” down priority lists determines who gets allocated
- [QR], [ER], [PR], not necessarily [PE]

Aziz and Brandl (2021) [4]: Reverse Rejecting Algorithm

- Iteratively certifies whether a maximal allocation can be found without allocating to a particular agent
- All four properties, but requires $O(n)$ max matching problems

Toward an Efficient Algorithm

Decision variables: $\mathbf{x} = \{x_{a,c}\}_{a \in \mathcal{A}, c \in \mathcal{C}}$.

$$x_{a,c} = \mathbb{I}(\varphi(a) = c),$$

(P_0)

$$\max \quad V(\mathbf{x}) := \sum_{a \in \mathcal{A}} \sum_{c \in \mathcal{C}} x_{a,c} \quad [\text{PE}]$$

$$\text{s.t.} \quad \sum_{a \in \mathcal{A}} x_{a,c} \leq q_c \quad \forall c \in \mathcal{C} \quad [\text{QR}]$$

$$\sum_{c \in \mathcal{C}} x_{a,c} \leq 1 \quad \forall a \in \mathcal{A} \quad [\text{UD}]$$










$$x_{a,c} = 0 \quad \forall a, c : a \notin \mathcal{E}_c \quad [\text{ER}]$$

$$x_{a,c} \in \{0, 1\} \quad \forall a \in \mathcal{A}, c \in \mathcal{C}$$

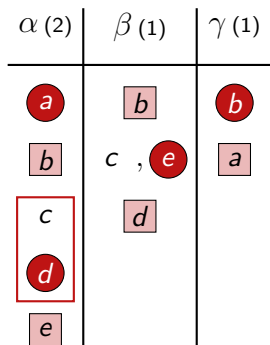
(P_0) encodes a bipartite b -matching problem

LP-relaxation is totally unimodular \implies integer corner points











(P_0) Doesn't Account for Priorities

$\alpha(2)$	$\beta(1)$	$\gamma(1)$
		
	c , 	
c		
		
		

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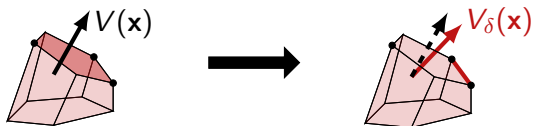
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To incorporate priorities, we'll modify the IP objective.

Adding Priorities

Idea: Tilt the objective so remaining optima respect priorities

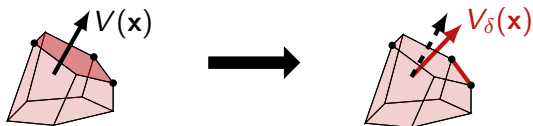


Replace $V(\mathbf{x})$ with $V_\delta(\mathbf{x}) = \sum_{a \in \mathcal{A}} \sum_{c \in \mathcal{C}} (1 - \delta_{a,c}) x_{a,c}$.

Interpreting $\delta_{a,c}$ as the cost of allocating a through c , a *valid* δ satisfies:

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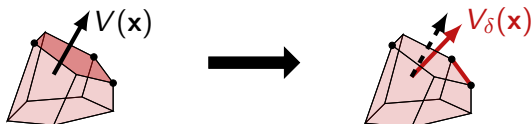
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Small Effect: Costs don't disincentivize allocation

$$\sum_{a \in \mathcal{A}} \sum_{c \in \mathcal{C}} \delta_{a,c} \leq \frac{1}{2}$$

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Consistent: Prioritized agents have lower cost $a \succeq_c a' \iff \delta_{a,c} \leq \delta_{a',c}$

Our Perturbed LP

Given any δ , define the LP

(P_δ)

$$\begin{array}{ll} \max & V_\delta(\mathbf{x}) \\ \text{s.t.} & \sum_{a \in \mathcal{A}} x_{a,c} \leq q_c \quad \forall c \in \mathcal{C} \\ & \sum_{c \in \mathcal{C}} x_{a,c} \leq 1 \quad \forall a \in \mathcal{A} \\ & x_{a,c} = 0 \quad \forall a, c : a \notin \mathcal{E}_c \\ & x_{a,c} \geq 0 \quad \forall a \in \mathcal{A}, c \in \mathcal{C} \end{array}$$

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Theorem

Let \mathbf{x}^* be a solution of (P_δ) for any valid δ . Then, \mathbf{x}^* corresponds to an allocation satisfying [ER], [QR], [PR], [PE].

Realizability of Good Allocations

Converse result:

Theorem (Informal)

Every recipient set determined by a fair allocation can be located by solving (P_δ) for some valid δ .

- Our perturbed matching framework is a standard setting
- The restrictions we've placed on δ are minimal

How far can we extend our techniques to handle related problems?

3 Case Studies: “Computational knife’s edge” of priority-respecting allocation

1. Reasoning about a Particular Agent

Must agent a be allocated?

Remove a and all lower-ranked agents from instance.

Check if max matching size decreases

$\alpha(2)$	$\beta(1)$	$\gamma(1)$
w	x	x
x	a, z	w
a	y	
y		

Can agent a be allocated?

A *serviceable* agent is a recipient in *some* good allocation.

Deciding whether an agent a is serviceable is NP-Hard.

Proof Idea: Reduction from X3C problem.

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


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2. Incorporating Agent Utility

Agent a has a utility function $u_a : \mathcal{C} \rightarrow (0, 1]$ encoding preference for certain categories. ($u_a(\emptyset) = 0$.)

Utility *Pareto-Efficient* Allocation

Run our algorithm twice

First run uses arbitrary δ to determine recipient set.

Second run removes agents outside of recipient set and sets δ according to agent utilities.

Utility *Maximizing* Allocation

NP-Hard via a reduction from serviceable problem.

Proof Idea: One agent has high utility in all categories, others have low utility.

*Hardness reduction can be generalized to other optimization objectives (e.g. Nash Social Welfare)

3. Global Fairness Heuristics

For each eligible agent $a \in \mathcal{E}_c$, let $r_c(a)$ be their priority tier in c (1 = highest priority, 2 = next priority tier, etc.)

Minimizing Maximum Allocated Rank

Run algorithm with “geometric” perturbation

$$\delta_{a,c} \propto n^{r_c(a)}$$

Proof Idea: Cost of highest ranked allocation dominates all others.

Maximizing Minimum Unallocated Rank

NP-Hard via an XC3 reduction similar to serviceable problem.

Proof Idea: Serviceable candidate has low rank in their only category. Other categories fill more ranks.

Conclusion

- Reserve Allocation is a reasonable modeling framework for assignment problems with “competing” objectives
- Can locate good allocations via a weighted matching LP
 - ▶ More efficient than existing approaches
 - ▶ Provides flexibility to many problem extensions

Open Questions:

Natural desiderata that locate a unique (fractional) allocation?





Can this allocation be computed efficiently?

- Our perturbation technique seems useful in other related problems

Thank You!



References

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-  H. Aziz and F. Brandl, “Efficient, fair, and incentive-compatible healthcare rationing,” in *Proceedings of the 22nd ACM Conference on Economics and Computation*, pp. 103–104, 2021.

Proving the Main Theorem

Theorem

Let \mathbf{x}^* be a solution of (P_δ) for any valid δ . Then, \mathbf{x}^* corresponds to an allocation satisfying [ER], [QR], [PR], [PE].

Proof Sketch.

[ER],[QR]: Ensured by (P_δ) constraints.

[PR]: δ is Consistent.

[PE]: Small Effect of δ and integrality:

$$V(\hat{\mathbf{x}}) \geq V(\mathbf{x}^*) \geq V_\delta(\mathbf{x}^*) \geq V_\delta(\hat{\mathbf{x}}) = V(\hat{\mathbf{x}}) - \sum_{a,c} \delta_{a,c} \geq V(\hat{\mathbf{x}}) - \frac{1}{2}.$$

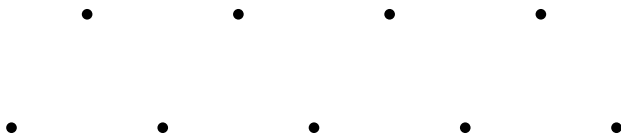
so $V(\hat{\mathbf{x}}) = V(\mathbf{x}^*)$ for a solution $\hat{\mathbf{x}}$ to (P_0) . □

Exact Cover by 3-Sets (X3C)

Input: Ground set $E = \{e_1, e_2, \dots, e_{3n}\}$.

Collection of subsets $\mathcal{S} = \{S_1, \dots, S_m\}$, each $|S_i| = 3$.

Decide: Is there a collection of subsets $\{S_{i_1}, \dots, S_{i_n}\}$ such that
 $E = \bigcup_{j=1}^n S_{i_j}$?



Lemma

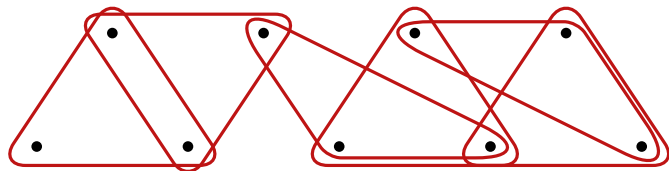
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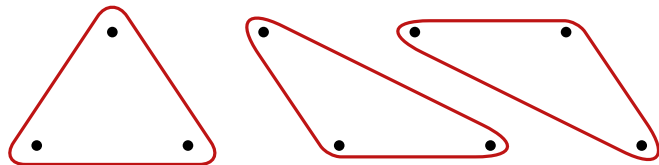
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Lemma

X3C is NP-Complete.

The Reduction

X3C Input: $E = \{e_1, e_2, \dots, e_{3n}\}$, $\mathcal{S} = \{S_1, \dots, S_m\}$

$$S_j = \{e_{j,1}, e_{j,2}, e_{j,3}\}$$

Allocation Instance: $\mathcal{A} = E \cup \{s_1, \dots, s_m\} \cup \{f_1, \dots, f_{4(m-n)}\} \cup a$

set categories			
α_1 (4)	\dots	α_m (4)	β (1)
f_1		f_1	e_1
\vdots		\vdots	\vdots
$f_{4(m-n)}$		$f_{4(m-n)}$	e_{3n}
s_1		s_m	a
$e_{i_1,1}$		$e_{i_m,1}$	
$e_{i_1,2}$		$e_{i_m,2}$	
$e_{i_1,3}$		$e_{i_m,3}$	