

A New Solution To The Random Assignment Problem

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The Assignment Problem

- How to best assign n objects to n agents
- Lotteries
 - Random assignments of objects to agents
- Random Priority mechanism
 - AKA Random Serial Dictatorship
 - Draw a random ordering of agents, then let them pick objects in that order

Properties

- **Random Priority** is fair
- Incentive compatible
 - **Agents** have no reason to lie about their preference
- Inefficient in a certain setting
 - When **agents** have Von Neumann-Morgenstern (VNM) **preferences** over lotteries
 - VNM **preferences** are characterized by VNM **utility** function
 - Simply the expected value over the lotteries

The Assignment Problem

- CEEI
 - View VNM **utility** function as **utility** over **shares**
 - **Shares** are the probability of receiving
- Properties
 - Not strategyproof
 - In fact no such mechanism can be strategyproof
 - Efficient for VNM **utilities**

Different types of Efficiencies

- Ex-Post Efficiency
 - All possible **assignments** are Pareto optimal
- Ex-Ante Efficiency
 - Efficient in terms of the profile of VNM **utilities**
- New! Ordinal Efficiency
 - In terms of distributions over **assignments**
 - Most probable and most valuable in terms of **utilities**
 - Will get into more detail later

Notation

- N is the set of n agents, A is the set of n objects
- Π is some bistochastic matrix of 1s and 0s
 - Deterministic assignment
- \underline{D} is the set of all Π
- P is some bistochastic matrix
 - Random assignment
 - Weighted sum of all $\Pi \in \underline{D}$
- \underline{R} is the set of all P
- $>$ is all agents strict preference orders over A
- \underline{A} is the domain of A

More notation

- A random allocation to an **agent** is a probability distribution over **A**
- $\underline{L}(A)$ is the set of all such allocations
- u_i is a mapping of **A** $\rightarrow \mathbf{R}^n$, the VNM utility
 - u is the profile over all of these
- Compatibility: $>_i$ is compatible with u_i
means that for any **a**, **b** $\in A$,
 - **a** $>$ **b** in $>_i$ iff $u_i(\mathbf{a}) > u_i(\mathbf{b})$

Even more notation

- σ is an ordering of agents
- θ is the set of all such orderings
- $\text{Prio}(\sigma, >)$ is a function mapping the orderings and the set of preferences to a deterministic assignment
- Prio creates an assignment by going through the ordering σ and giving each agent their top-ranked available item by $>$

Efficiencies

- Given some **random assignment** matrix P and a profile of utilities u compatible with a profile of preferences \succ
 - Ex-ante efficiency comes from:
 - Pareto optimality at u
 - Ex-post efficiency
 - If P can be represented as a sum over a distribution of $\text{Prio}(\sigma, \succ)$ from all possible orderings σ with some weights

Random Priority

- In this notation, easy to define **random priority assignment**
- **P** is the average over all $\text{Prio}(\sigma, >)$
 - All weights are $1/n!$
 - That is, average over all serial dictatorships

Stochastic Dominance

- A strict ordering $>_i$ implies a partial ordering on $\underline{L}(A)$
- This is called the stochastic dominance relation, $sd(>_i)$
- Formally, given some P_i and Q_i from $\underline{L}(A)$
 - $P_i \text{ } sd(>_i) \text{ } Q_i$ iff for all t in $[1, n]$, the sum over the row P_i from 1 to t is greater than or equal to Q_i 's sum
 - Example

Stochastic Dominance

- Given some preference \succ_i , $P_i \text{ sd}(\succ_i) Q_i$ is equivalent to $u_i P_i \succeq u_i Q_i$ for all compatible utilities u_i
- Definition: If some random assignment P dominates some other random assignment Q for all agents, then Q is stochastically dominated by P

Ordinal Efficiency (O-efficiency)

- A random assignment P is O-efficient if it is not stochastically dominated by any other random assignment
- Some corollaries
 - If P is ex-ante efficient for u , then it is O-efficient at $>$
 - If P is ex-post efficient for $>$, then it is O-efficient at $>$
 - Extra conditions when $n \leq 4$

Simultaneous Eating Algorithm

- Each **object** is an infinitely divisible commodity
- Each agent has an eating speed function $\omega_i(t)$
 - Each agent is allowed to consume an **object** with speed $\omega_i(t)$ at time t
 - $\omega_i(t)$ is non-negative and integrates to 1 over the interval $[0,1]$

Simultaneous Eating Algorithm

- Simply allow **agents** to 'eat' from their best available **objects** at the specified eating speeds
- Example

Simultaneous Eating Algorithm

- Getting P_ω can be done with an iterative algorithm
- $M(a, A)$ is the set of agents who prefer a to all other objects in A .
- Initialize: $A^0 = A$, $y^0 = 0$, $P^0 = \text{zeros}(n, n)$
- Basically this formalizes having each agent eat from their best available object, and the algorithm finds best times to allow

Simultaneous Eating Algorithm

- Let $y^s(a)$ be the minimum y such that the
 - sum over all agents i in $M(a, A^{s-1})$ of the integral from y^{s-1} to y of $\omega_i(t)$
 - plus the sum over all agents of the probability of that agent getting a in P^{s-1}
 - is equal to 1.
 - With the condition that $y^s(a)$ be ∞ if there are no agents that prefer a to all other objects in A^{s-1}

Simultaneous Eating Algorithm

- At each step s , let
 - y^s be the minimum $y^s(\mathbf{a})$ over all objects in A^{s-1}
 - A^s be A^{s-1} without the object that minimized y^s
 - P^s be the following
 - Update each cell $P^s[i, \mathbf{a}]$ by using the previous if i is not in the set of agents that prefer \mathbf{a} to any other object
 - Otherwise add the eating speed $\omega_i(t)$ integrated from y^{s-1} to y^s to $P^{s-1}[i, \mathbf{a}]$

Simultaneous Eating Algorithm

- Since at each step we remove an object, at A^n there will be no objects, so P^n is the final random assignment
- Theorem:
 - P_ω is ordinally efficient for all profiles of eating functions.
 - Conversely, there exists a profile of eating functions for any ordinally efficient P

Probabilistic Serial Assignment

- Apply Simultaneous Eating Algorithm to profile of uniform eating speeds
 - All $\omega_i(t) = 1$ for all t in $[0,1]$ and all agents i in N
- This makes $y^s(\mathbf{a})$ easy to compute at any step
- Has some nice properties

Probabilistic Serial Assignment

- Anonymous
- Only equitable mechanism
 - In order to construct an anonymous assignment, we will always end up with the Probabilistic Serial assignment

Fairness and Incentives of PS vs RP

- Random Priority may generate envy
- Probabilistic Serial may be manipulated
- Both only happen under limited conditions
- For small n :
 - $n = 2$, trivially RP and PS give the same results
 - $n = 3$, RP may generate envy and PS may be manipulated
 - $n \geq 4$?

For $n = 3$

- **RP**
 - O-efficient
 - Strategy-proof
 - Treats equal utilities with equal random allocations
- **PS**
 - O-efficient
 - No envy
 - Weakly strategy-proof

For $n \geq 3$

- Proposition:
- PS
 - Envy free
 - Weakly strategy-proof
- RP
 - Weakly envy free
 - Strategy-proof

Impossibility Result

- For $n \geq 4$, there is no possible mechanism such that
 - It is O-efficient
 - It is strategyproof
 - Treats equal preferences equally
 - Proof is very long

Further caveats

- Note some assumptions
 - Same number of agents and objects
 - Models can be easily adjusted for either more agents than objects or more objects than agents
 - Objective Indifferences
 - Some pair of objects are the same to all agents
 - Subjective Indifferences
 - Some pair of objects are the same to some agents

n agents and m objects

- Both **RP** and **PS** still work
 - If there are more objects than agents, everything still holds if the bistochastic matrices loosen to allow the columns to sum to less than one
 - If there are more agents than objects, then rows sum to m/n and if the eating functions integrate to m/n instead of 1.
 - Can instead add the remainder of null objects, which are the same to all agents

Objective Indifferences

- The simultaneous eating theorem still holds since the choice is inconsequential
- This provides no issue with the current results

Subjective Indifferences

- Since the difference could be unimportant to some agent but not to others, an agent can't be allowed to choose arbitrarily
- Best option seems to be eliciting more preferences from those agents
- Could be a subject of further research

Discussion Considerations

- Other caveats?
- How computable is
 - Probabilistic Serial
 - Random Priority