

# SCHOOL ALLOCATION PROBLEM WITH OBSERVABLE CHARACTERISTICS

YOUNG WU

## 1. MODEL

**1.1. Allocations.** There are finite number of schools  $l \in L = \{1, 2, \dots, \bar{L}\}$  and finite number of groups of students divided by characteristics  $k \in K = \{1, 2, \dots, \bar{K}\}$ . Let  $c_l$  be the measure of seats in school  $l$  and  $\mu_k$  be the measure of students in group  $k$ .

**Definition 1.** A profile of orderings is a full support distribution  $\mu \in \Delta(K \times \mathcal{P}(L))$ , where  $\mathcal{P}(L)$  is the set of all permutations (strict orderings) of  $L$ , and  $\mu$  satisfies,

$$\sum_{p \in \mathcal{P}(L)} \mu(k, p) = \mu_k.$$

We implicitly assume that all preferences are strict, and use the notation  $\mu(k, p)$  to denote the fraction of students in group  $k$  who have preference ordering  $p$ . We call these students' type  $(k, p)$ .

We also assume full support

$$\mu(k, p) > 0 \quad \forall (k, p) \in (K, \mathcal{P}(L)).$$

**Definition 2.** Let the set of students be  $\mathcal{I}$ . An allocation is a function  $q : \mathcal{I} \rightarrow \Delta(L)$ , where  $q(l; i)$  is the probability that a student  $i \in \mathcal{I}$  is assigned to school  $l$ .

In the case where every student with the same type gets the same allocation, we will use the notation  $q(l; k, p)$  to denote the probability that a student with type  $(k, p)$  is assigned to school  $l$ . We call this condition group symmetric. Then the amount of students with type

$(k, p)$  who are assigned to school  $l$  satisfies

$$\mu(k, p) \cdot q(l; k, p) = \int_{i \in \mathcal{I}: \text{type}(i) = (k, p)} q(l; i) d\mu,$$

where  $\mu(k, p)$  is the mass of students with type  $(k, p)$  and

$$\mu(k, p) = \int_{i \in \mathcal{I}: \text{type}(i) = (k, p)} d\mu.$$

We also impose feasibility assumption that no student will be left unassigned

$$\begin{aligned} \sum_{l \in L} c_l &\geq \sum_{k \in K} \sum_{p \in \mathcal{P}(L)} \mu(k, p) \\ &= \sum_{k \in K} \mu_k \\ &= 1. \end{aligned}$$

**Definition 3.** An allocation  $q$  is feasible given a profile  $\mu$  if

$$\int_{i \in \mathcal{I}} q(l; i) d\mu \leq c_l.$$

In the group symmetric case, the condition becomes

$$\sum_{k \in K} \sum_{p \in \mathcal{P}(L)} \mu(k, p) q(l; k, p) \leq c_l.$$

**Definition 4.** A mechanism is function  $q : \mathcal{I} \rightarrow Q$ , where  $Q$  is the set of all feasible allocations.

**1.2. Ordinal Efficiency.** We first define ordinal efficiency using first order stochastic domination of the allocation distribution:

**Definition 5.** An allocation  $q$  is dominated by  $q'$  for a student with type  $(k, p)$  where  $p = l_1 \succ l_2 \succ l_3 \dots \succ l_L$  if

$$\sum_{s=1}^t q(l_s; k, p) \leq \sum_{s=1}^t q'(l_s; k, p) \quad \forall t \in L.$$

The main desirable properties of a ordinal mechanism in this model are:

**Definition 6.** Group Symmetry:  $q(i) = q(j)$  for any  $i, j \in \mathcal{I}$  such that  $\text{type}(i) = \text{type}(j)$ .

**Definition 7.** Envy-free:  $q(i)$  is not dominated by any  $q(j)$  for any  $i, j \in \mathcal{I}$  such that  $\text{group}(i) = \text{group}(j)$ .

**Definition 8.** Efficiency:  $q(i)$  is not dominated by any  $q'(i)$  for any  $i \in \mathcal{I}$ .

Group symmetry states that students with the same characteristics are assigned the same probabilistic allocation. Given this assumption, we can use the notation  $q(l; k, p)$  to denote the probability of a student with type  $(k, p)$  getting allocated the school  $l$ . Envy-free states that any a student  $(k, p)$  will not prefer the allocation of another student  $(k, p')$  for any  $p' \neq p$ . It is weaker condition of strategy-proof, but it guarantees that no student has incentive to misreport her preference  $p$ . We will show that a mechanism satisfying all three properties Definition 6, Definition 7, Definition 8 must be the modified Probabilistic Serial mechanism (PS) defined in the following section.

## 2. MODIFIED PROBABILISTIC SERIAL

**2.1. Algorithm.** We define the probabilistic serial mechanism as a simultaneous eating mechanism with equal constant eating speed over all groups.

**Definition 9.** subcapacity  $c_l^k$  of school  $l$  for students in group  $k$ , are functions  $\mu \rightarrow [0, c_l]$  satisfying

$$\sum_{k \in K} c_l^k(\mu) \leq c_l \quad \forall l \in L.$$

**Algorithm 1.** A simultaneous eating mechanism (PSKT) with eating speed  $\omega = \{\omega_1, \omega_2, \dots, \omega_{\bar{K}}\}$ , where  $\omega_k : [0, 1] \rightarrow \mathbb{R}_+$  satisfying  $\int_0^1 \omega_k(t) dt = 1$ , and subcapacities  $\{c_l^k\}_{l \in L, k \in K}$ , is given by

Initialize:  $L_k^0 = L, y_k^0 = 0$  for each  $k \in K$  and  $q^0(l; k, p) = 0$  for each  $l \in L, k \in K, p \in \mathcal{P}(L)$ ,

*Iteration:* assume  $L_k^{s-1}, y_k^{s-1}, q_k^{s-1}$  are defined for each  $k$ , set:

$$M(l, A) = \{p \in \mathcal{P}(L) : l \succ l' \forall l' \in A \setminus \{l\}\},$$

$$y_k^s(l) = \arg \min_y \left\{ \sum_{p \in M(l, L_k^{s-1})} \int_{y_k^{s-1}}^y \mu(k, p) \omega_k(t) dt + \sum_{p \in \mathcal{P}(L)} \mu(k, p) q^{s-1}(l; k, p) = c_l^k(\mu) \right\},$$

$$F_k^s = \arg \min_l y_k^s(l),$$

$$y_k^s = \min_l y_k^s(l),$$

$$L_k^s = L_k^{s-1} \setminus F_k^s,$$

$$t_l^k = y_{s_k}, \text{ for } l \in F_k^s,$$

$$q^s(l; k, p) = q^{s-1}(l; k, p) + \mathbb{I}_{p \in M(l, L_k^{s-1})} \int_{y_k^{s-1}}^{y_k^s} \omega_k(t) dt.$$

Here, in step  $s$ ,

$M(l, A)$  is the set of students where will be consuming school  $l$  if the remaining set of available schools is  $A$ ,

$y_k^s(l)$  is the smallest amount of time needed for students in group  $k$  to finish consuming school  $l$ ,

$y_k^s$  is the smallest amount of time needed for students in group  $k$  to finish consuming any school,

$F_k^s$  is the set of schools that are completely consumed by students in group  $k$ ,

$t_l^k$  is the time school  $l$  is completely consumed by students in group  $k$ ,

$L_k^s$  is the remaining set of available schools for students in group  $k$ ,

$q^s(l; k, p)$  is the temporary allocation of school  $l$  for students with type  $(k, p)$ .

**Algorithm 2.** A probabilistic serial mechanism (PS) is a simultaneous eating algorithm with  $\omega_k(t) = \omega = 1 \forall k \in K, t \in [0, 1]$ .

**2.2. Ordinal Properties.** For a full support profile, Bogomolnaia and Moulin showed that an allocation is efficient if and only if it is generated by a simultaneous eating algorithm,

and Liu and Pycia showed that an allocation is efficient and envy-free if and only if it is generated by probabilistic serial. The result can be extended to the problem with multiple groups with a similar proof to Theorem 1 in Liu and Pycia.

**Proposition 1.** *An allocation  $q$  is group symmetric, envy-free and efficient for full support profile  $\mu$  if and only if it is generated by Algorithm 2 (PS) with subcapacities,*

$$c_l^k(\mu) = \sum_{p \in \mathcal{P}(L)} \mu(k, p) q(l; k, p).$$

**2.3. Cardinal Efficiency.** We also define cardinal utility functions that induce the preference relations in the following way.

**Definition 10.** A utility distribution that is consistent with a preference profile  $\mu$  is one density function  $f_k : \mathbb{R}^L \rightarrow \mathbb{R}$  for each  $k$  such that:

$$\int_{p(u)=p} f_k(u) du = \frac{\mu(k, p)}{\mu_k} \quad \forall p \in \mathcal{P}(L).$$

where  $p(u)$  is the preference relation induced by the utility ranking  $u$ .

**Definition 11.** Efficiency: the allocation  $q^*$  is cardinally efficient if it maximizes the following expected welfare:

$$q^* = \arg \max_q \sum_{k=1}^K \int u \cdot q(l; k, p(u)) \cdot f_k(u) du.$$

**2.4. Cardinal Extension.** Since PS is ordinally efficient, the cardinally efficient allocation must be obtained by PS for some subcapacities. We will rewrite the welfare maximization problem Definition 11 as a maximization problem by choosing the subcapacities  $c_l^k$ .

**Definition 12.** The welfare function  $W_k : \mathbb{R}^k \rightarrow \mathbb{R}$  for group  $k$  is defined as the following,

$$W_k(c^k) = \int u \cdot q(l; k, p(u)) \cdot f_k(u) du.$$

The welfare maximization problem becomes:

$$\begin{aligned} & \max_{\{c^k\}_{k \in K}} \sum_{k \in K} W_k(c^k) \\ & \text{such that } \mu_k = \sum_l c_l^k \forall k \in K \\ & \text{and } c_l = \sum_k c_l^k \forall l \in L \end{aligned}$$

**Proposition 2.** *The function  $W_k(c^k)$  is non-decreasing and concave in  $c^k$ .*

**Conjecture 1.** *The function  $W_k(c^k)$  is piecewise linear in each subcapacity  $c_l^k$ .*

The above two properties of the welfare function guarantee that the problem can be solved by linear programming, although formulating the problem into a simple linear program is difficult.

### 3. EQUIVALENCE BETWEEN PS AND DA

**3.1. Two-school example.** Optimal PS and DA are equivalent if there are two characteristics and two schools. Suppose students live in two districts ( $k = 1, 2$ ) and there is one school in each district ( $l = 1, 2$ ).

Fix  $k$ , let,

$$\begin{aligned} \mu_1 &= \int_{\{u_1 > u_2\}} f(u) du \\ \mu_2 &= \int_{\{u_1 < u_2\}} f(u) du \end{aligned}$$

Then,

$$\mu = \mu_1 + \mu_2$$

The PS allocations are:

If  $c_1 \leq c_2 \cdot \frac{\mu_1}{\mu_2}$ ,

$$q(k, 1 \succ 2) = 1^{\frac{c_1}{\mu_1}} 2^{1 - \frac{c_1}{\mu_1}}$$

$$q(k, 2 \succ 1) = 1^0 2^1$$

If  $c_1 \geq c_2 \cdot \frac{\mu_1}{\mu_2}$ ,

$$q(k, 1 \succ 2) = 1^1 2^0$$

$$q(k, 2 \succ 1) = 1^{1 - \frac{c_2}{\mu_2}} 2^{\frac{c_2}{\mu_2}}$$

The value function is:

$$W(c) = \begin{cases} \int_{\{u_1 > u_2\}} \left( \frac{c_1}{\mu_1} \right) \cdot u_1 + \left( 1 - \frac{c_1}{\mu_1} \right) \cdot u_2 du + \int_{\{u_1 < u_2\}} 0 \cdot u_1 + 1 \cdot u_2 du & \text{if } c_1 \leq c_2 \cdot \frac{\mu_1}{\mu_2}, \\ \int_{\{u_1 > u_2\}} 1 \cdot u_1 + 0 \cdot u_2 du + \int_{\{u_1 < u_2\}} \left( 1 - \frac{c_2}{\mu_2} \right) \cdot u_1 + \left( \frac{c_2}{\mu_2} \right) \cdot u_2 du & \text{otherwise.} \end{cases}$$

Or

$$W(c) = \begin{cases} \mu_1 \cdot \left( \frac{c_1}{\mu_1} \cdot \mathbb{E}[u_1 | u_1 > u_2] + \left( 1 - \frac{c_1}{\mu_1} \right) \cdot \mathbb{E}[u_2 | u_1 > u_2] \right) + \mu_2 \cdot \mathbb{E}[u_2 | u_1 < u_2] & \text{if } c_1 \leq c_2 \cdot \frac{\mu_1}{\mu_2} \\ \mu_1 \cdot \mathbb{E}[u_1 | u_1 > u_2] + \mu_2 \cdot \left( \left( 1 - \frac{c_2}{\mu_2} \right) \cdot \mathbb{E}[u_1 | u_1 < u_2] + \left( \frac{c_2}{\mu_2} \right) \cdot \mathbb{E}[u_2 | u_1 < u_2] \right) & \text{otherwise.} \end{cases}$$

The derivative with respect to  $c_1$  for fixed  $c_2$  is

$$\frac{\partial W}{\partial c_1}(c) = \begin{cases} \mathbb{E}[u_1 - u_2 | u_1 > u_2] > 0 & \text{if } c_1 \leq c_2 \cdot \frac{\mu_1}{\mu_2} \\ 0 & \text{otherwise,} \end{cases}$$

and for  $c_2 = \mu - c_1$  is

$$\frac{\partial W}{\partial c_1}(c) = \begin{cases} \mathbb{E}[u_1 - u_2 | u_1 > u_2] > 0 & \text{if } c_1 \leq \mu_1 \\ \mathbb{E}[u_1 - u_2 | u_1 < u_2] < 0 & \text{otherwise.} \end{cases}$$

Therefore resulting allocation is cardinally efficient and it is identical to the one obtained from the differed acceptance algorithm.

Include an example of 3 school case when PS and DA are not equivalent?

#### 4. PROOFS

Proofs still have a lot of inconsistent notations.

##### 4.1. Proof of Proposition 1.

*Proof.* The equivalence can be obtained from the following three lemmas. □

**Lemma 1.** (Modified from Liu and Pycia Theorem 1) *If an allocation  $q$  is group symmetric, envy-free and efficient for profile  $\mu \succ \succ 0$ , then it is generated by PS with constraints  $c_l^k(\mu) = \sum_p \mu(k, p) q(l; k, p)$ .*

*Proof.* Consider any allocation  $q'$  and the allocation obtained by PS  $q^1$ . Let  $q^t$  be the partial allocation at time  $t \in [0, 1]$  for PS.

Need to show that for any  $(k, p), l$  and at any time  $t \in [0, 1]$ :

$$\sum_{l' \succ_{k,p} l} q'(l'; k, p) \geq \sum_{l' \succ_{k,p} l} q^t(l'; k, p)$$

Assume for a contradiction, there is  $\tau \in [0, 1]$  such that:

$$\tau = \inf \left\{ t : \sum_{l' \succ_{k,p} l} q'(l'; k, p) < \sum_{l' \succ_{k,p} l} q^t(l'; k, p) \text{ for some } (k, p) \in (K \times \mathcal{P}(L)) \text{ } l \in L \right\}$$

Note that the original inequality are satisfied for all  $t \in [0, \tau]$  and  $(k, p)$  must be eating  $l$  at  $\tau$ , which implies.

$$\sum_{l' \succ_{k,p} l} q'(l'; k, p) \geq \sum_{l' \succ_{k,p} l} q^\tau(l'; k, p) = \tau$$

By continuity of  $q^t$  in  $t$ :

$$\sum_{l' \succ_{k,p} l} q'(l'; k, p) = \sum_{l' \succ_{k,p} l} q^\tau(l'; k, p) = \tau$$



If assumed full support condition,  $l$  is favorite object of some agent  $(k, p')$ , then:

$$q(l; k, p') \geq q^\tau(l; k, p') = \tau$$

Envy-free assumption implies  $(k, p)$  does not prefer the allocation of  $(k, p')$ :

$$q(l; k, p') \leq \tau$$

Therefore,

$$q(l; k, p') = \tau$$

Since  $l$  is not exhausted at  $\tau$ :

$$q^1(l; k, p') > \tau$$

Therefore,  $(k, p')$  gets less  $l$  in  $q$  than  $q^1$ , efficiency assumption implies that there is another student  $(k, \hat{p})$  who gets:

$$q(l; k, \hat{p}) > q^1(l; k, \hat{p})$$

And there is some  $\hat{l} \neq l$  that student  $(k, \hat{p})$  prefers just more than  $l$ :

$$\begin{aligned} \sum_{l' \succ_{k, \hat{p}} \hat{l}} q'(l'; k, \hat{p}) &\geq \sum_{l' \succ_{k, \hat{p}} \hat{l}} q^\tau(l'; k, \hat{p}) = \tau - q^\tau(l; k, \hat{p}) \geq \tau - q^1(l; k, \hat{p}) \\ \Rightarrow \sum_{l' \succ_{k, \hat{p}} l} q'(l'; k, \hat{p}) &> \tau \end{aligned}$$

Comparing students  $(k, p')$  and  $(k, \hat{p})$ , envy-free assumption leads to a contradiction.  $\square$

**Lemma 2.** (*Bogomolnaia and Moulin Proposition 1*) *PS is envy-free.*

*Proof.* For  $y_k^{s-1} \leq t \leq y_k^s$ , define:

$$\begin{aligned} N(l, t) &= M(l, L^{s-1}) \text{ if } l \in L^{s-1} \text{ and } \emptyset \text{ otherwise} \\ n(l, t) &= \sum_{p, k} \mu(k, p) \mathbb{I}_{(k, p) \in M(l, L^{s-1})} \\ t(l) &= \sup \{t | n(l, t) \geq c_l^k(\mu)\} \end{aligned}$$

Consider a student  $(k, p)$ , let  $q$  be the allocation if she reports  $p$  and  $q'$  be the allocation if she reports  $p'$ .

Let  $p$  be the preference  $l_1 \succ_{k,p} l_2 \succ_{k,p} l_3 \succ_{k,p} \dots$

If  $q(l_1; p, k) \leq q'(l_1; p, k)$ , then  $t(l_1) \leq t'(l_1)$ .

We want to show that  $N(l, t) = N'(l, t) \quad \forall t \in (0, t(l))$ , which implies  $q(l_1; p, k) = q'(l_1; p, k)$

Repeat this process for  $l_2, l_3 \dots$  □

**Lemma 3.** (*Bogomolnaia and Moulin Theorem 1*) *PS is efficient.*

*Proof.* Suppose, for a contradiction that  $q$  is obtained by PS and it is not efficient, and  $q$  is dominated by  $q'$ .

Define binary relation  $\tau : l\tau l' \Leftrightarrow \{\exists (k, p) \in (K \times \mathcal{P}(L)) : l \succ_{k,p} l' \text{ and } q(l; k, p) > 0\}$ .

Let  $(k_1, p_1)$  be the student such that  $q(k_1, p_1) \neq q'(k_1, p_1)$ , then there are  $l_0, l_1$  such that

$$l_1 \succ_{p_1} l_0, q(l_1; k_1, p_1) > q'(l_1; k_1, p_1), q(l_0; k_1, p_1) < q'(l_0; k_1, p_1)$$

Then  $l_0\tau l_1$ . Similarly, there is  $l_1\tau l_2$  and since  $L$  is finite, there  $\exists$  a cycle in the relation  $\tau$ :

$$l_0\tau l_1, \dots, l_R\tau l_0$$

Let  $(k_r, p_r)$  be the student such that  $l_{r-1} \succ_{p_r} l_r$  and  $q(l_r; k_r, p_r) > 0$ .

Define  $s_r = \min_s \{s : p^s(l_r; k_r, p_r)\}$ , and note that  $l_{r-1} \notin A^{s_r-1}$ , meaning  $s_{r-1} < s_r$ .

This implies  $s_0 < s_1 < \dots < s_{R-1} < s_0$ , contradiction. □

4.2. **Proof of.** The concavity of the welfare functions can be obtained using the following lemmas:

**Lemma 4.** *Convex combination of envy free allocations are envy free.*

*Proof.* Consider arbitrary pair of students  $(k, p)$  and  $(k, p')$  under two different allocations  $q_1$  and  $q_2$ .

Let  $p = l_1 \succ l_2 \succ l_3 \dots \succ l_L$  be the preference ranking of the first student, and define the following,

$$\begin{aligned} t_1^i &= \min_t \left\{ \sum_{s=1}^t q_1(l_s; k, p) < \sum_{s=1}^t q_1(l_s; k, p') \right\} \\ t_1^a &= \max_t \left\{ \sum_{s=1}^t q_1(l_s; k, p) \leq \sum_{s=1}^t q_1(l_s; k, p') \right\} \\ t_2^i &= \min_t \left\{ \sum_{s=1}^t q_2(l_s; k, p) < \sum_{s=1}^t q_2(l_s; k, p') \right\} \\ t_2^a &= \max_t \left\{ \sum_{s=1}^t q_2(l_s; k, p) \leq \sum_{s=1}^t q_2(l_s; k, p') \right\} \end{aligned}$$

By envy-freeness,  $t_1^i \neq t_1^a$  and  $t_2^i \neq t_2^a$ ,

Consider a convex combination  $q_0 = (\alpha) q_1 + (1 - \alpha) q_2$  for  $\alpha \in [0, 1]$ ,

For  $t \leq \min \{t_1^i, t_2^i\}$ ,

$$\begin{aligned} \sum_{s=1}^t q_0(l_s; k, p) &= \sum_{s=1}^t (\alpha) q_1(l_s; k, p) + (1 - \alpha) q_2(l_s; k, p) \\ &< \sum_{s=1}^t (\alpha) q_1(l_s; k, p') + (1 - \alpha) q_2(l_s; k, p') \\ &= \sum_{s=1}^t q_0(l_s; k, p') \end{aligned}$$

And for  $t \geq \max \{t_1^a, t_2^a\}$ ,

$$\begin{aligned} \sum_{s=1}^t q_0(l_s; k, p) &= \sum_{s=1}^t (\alpha) q_1(l_s; k, p) + (1 - \alpha) q_2(l_s; k, p) \\ &> \sum_{s=1}^t (\alpha) q_1(l_s; k, p') + (1 - \alpha) q_2(l_s; k, p') \\ &= \sum_{s=1}^t q_0(l_s; k, p') \end{aligned}$$

Therefore, under  $q_0$ , no student strictly prefer the alloaction of another student,  $q_0$  is envy-free.  $\square$

**Lemma 5.** *Any inefficient envy-free allocation has an envy-free Pareto improvement.*

*Proof.* Consider an allocation  $q$  and another allocation  $q'$  that (Pareto) dominates  $q$ .

For each student  $(k, p)$  and pair of schools  $(i, j)$ , define the flow from school  $i$  to  $j$  by  $\Delta(k, p; i, j)$  satisfying:

$$\begin{aligned} \sum_j \Delta(k, p; i, j) &= \max \{0, q(k, p; i) - q'(k, p; i)\}; \\ \sum_i \Delta(k, p; i, j) &= \max \{0, q(k, p; j) - q'(k, p; j)\}; \end{aligned}$$

$$\Delta(k, p; i, j) \geq 0$$

Then define another allocation  $q^*$  by:

$$\begin{aligned} q^*(k, p; i) &= q(k, p; i) - \sum_j (\mathbb{I}_{\Delta(k, p; i, j) > 0 \text{ or } j \succ_p i} \cdot \Delta^*(k, p; i, j)) \\ &\quad + \sum_j \mathbb{I}_{\Delta(k, p; j, i) > 0 \text{ or } j \succ_p i} \cdot \Delta^*(k, p; j, i) \end{aligned}$$

where  $\Delta^*$  is defined as:

$$\Delta^*(k, p; i, j) = \frac{\sum_{k', p'} \Delta(k', p'; j, i) \cdot \mu(k', p')}{\sum_{k', p': i \succ_{p'} j \text{ and } \Delta(k', p'; i, j) = 0} \mu(k', p') + \sum_{k', p'} \Delta(k', p'; j, i) \cdot \mu(k', p')}$$

Note that the flows from  $q$  to  $q'$  and the flows from  $q$  to  $q^*$  are the same (the previous system for  $\Delta$  is still satisfied).

Also,  $q^*$  still dominates  $q$  since:

$$\begin{cases} \Delta^*(k, p; i, j) > 0 & \text{if } i \succ_p j \\ \Delta^*(k, p; i, j) < \Delta(k, p; i, j) & \text{if } j \succ_p i \end{cases}$$

And  $q^*$  is envy-free since:

$$\begin{cases} \Delta^*(k, p; i, j) \geq \Delta^*(k, p'; i, j) \quad \forall p' & \text{if } i \succ_p j \\ \Delta^*(k, p; i, j) \geq 0 & \text{if } j \succ_p i \end{cases}$$

Therefore,  $q^*$  is an envy-free Pareto improvement to  $q$ . □

**Lemma 6.** *The set of envy-free allocations are closed.*

*Proof.* Consider any sequence of allocations  $\{q_i\}_{i=1}^{\infty}$  and the limit  $q^*$ .

Fix any two students  $(k, p)$  and  $(k, p')$ , since  $q_i$  are envy-free for each  $i$ :

$$\sum_{s=1}^t q_i(l_s; k, p') \leq \sum_{s=1}^t q_i(l_s; k, p) \quad \forall t$$

where  $l_s$  is the  $s$ -th school in the preference ranking of student  $(k, p)$ .

Then,

$$\begin{aligned} \lim_{i \rightarrow \infty} \sum_{s=1}^t q_i(l_s; k, p') &\leq \lim_{i \rightarrow \infty} \sum_{s=1}^t q_i(l_s; k, p) \quad \forall t \\ \sum_{s=1}^t q^*(l_s; k, p') &\leq \sum_{s=1}^t q^*(l_s; k, p) \quad \forall t \end{aligned}$$

Therefore,  $q^*$  is envy-free. The set is closed under limits.

Similarly, the set of Pareto improvements of any allocation is closed. □

*Proof.* Let  $c, c'$  be two vector of capacities, and  $q, q'$  be the PS allocation with capacities  $c, c'$  respectively.

Consider allocation  $q_0 = \frac{1}{2}q + \frac{1}{2}q'$  and the welfare of allocation  $q_0$  is  $\frac{1}{2}(W(c) + W(c'))$

If  $q_0$  can be obtained from PS with capacities  $\frac{1}{2}(c + c')$ , then  $\frac{1}{2}(W(c) + W(c')) = W\left(\frac{1}{2}(c + c')\right)$ .

Suppose  $q_0$  cannot be obtained from PS, then  $q_0$  is not both envy-free and efficient.

Since  $q_0$  is envy-free from Lemma 6,  $q_0$  is not efficient.

Let  $V$  be the set of envy-free allocations that is more efficient than  $q_0$ .

$V$  is bounded since the set of allocations is bounded and the set of all envy-free allocations and the set of allocations that are Pareto improvements to  $q_0$  are closed by Lemma 8. Then,  $V$  is an intersection of two compact sets implying that  $V$  is compact.

Therefore, there is an allocation  $q^* \in V$  that maximizes  $W\left(\frac{1}{2}(c + c')\right)$ .

Note that  $q^*$  must be efficient because if not, by Lemma 7, there is a envy-free Pareto improvement of  $q^*$  in  $V$  which contradicts the definition that  $q^*$  maximizes  $W\left(\frac{1}{2}(c + c')\right)$ .

$q^*$  is envy-free and efficient, implying that  $q^*$  is the PS allocation with capacity  $\frac{1}{2}(c + c')$ .

Therefore,  $\frac{1}{2}(W(c) + W(c')) \leq W\left(\frac{1}{2}(c + c')\right)$ ,  $W$  is concave in  $c$ .  $\square$

The function  $W_k(c)$  is strictly increasing in  $c_l^k$  for  $c_l^k \in \left[0, \sum_{p: l \text{ is the most preferred school}} \mu(k, p)\right]$ , and non-decreasing for  $c_l^k$  in  $\left[\sum_{p: l \text{ is the most preferred school}} \mu(k, p), c_l\right]$ .

#### 4.3. Proof of Conjecture 1.

*Proof.* Fix  $ak$ , define the set of capacities for which no two schools are finished being eaten at the same time,  $\mathcal{D}$ :

$$\mathcal{D} = \{c : y^s(l) \neq y^s(l') \ \forall s \ \forall l' \neq l\}$$

Define the set of capacities for which  $l$  and  $l'$  are finished being eaten at the same time,  $E_l$ :

$$E_{l,l'} = \{c_l : y^s(l) = y^s(l') \text{ for some } s\}$$

We first show that the value function on this set is linear.

Take  $\varepsilon < \min_s y^s$ , consider the change from  $c_l$  to  $c_l + \varepsilon$  and  $c_{l'}$  to  $c_{l'} - \varepsilon$ .

Let  $s$  be the iteration with  $y^s(l) = y^s$ , and  $s'$  be the iteration with  $y^{s'}(l') = y^{s'}$

Then  $W(c_l)$  will change by

$$\frac{\varepsilon}{|M(l, L^{s-1})|} \cdot \left( \sum_{p \in M(l, L^{s-1})} u(p, l) \right) - \frac{\varepsilon}{|M(l', L^{s'-1})|} \cdot \left( \sum_{p \in M(l', L^{s'-1})} u(p, l') \right)$$

where the  $u(p, l)$  is the expected utility of students with preference  $p$  getting into school  $l$ .

The students with preferences in  $M(l, L^{s-1})$  will spend  $\frac{\varepsilon}{|M(l, L^{s-1})|}$  extra time on eating  $l$ , and  $L^s$  will stay the same since  $\varepsilon < y^s$ .

Similarly, the students with preferences in  $M(l', L^{s'-1})$  will spend  $\frac{\varepsilon}{|M(l', L^{s'-1})|}$  less time on eating  $l'$ , and  $L^{s'}$  will stay the same since  $\varepsilon < y^{s'}$ .

Then note that  $E_{l,l'}$  contains at most one point, since otherwise,  $y^s(l) \neq y^{s'}(l')$  in one of the points in  $E_{l,l'}$ . Therefore,  $\mathcal{D}^c$  is a finite union of  $E_{l,l'}$ , the value function is linear on all but a finite set of points, i.e. piecewise linear.  $\square$

**Lemma 7.** *The value function is piecewise linear in the capacities  $c_l^k$*

*Proof.* Fix  $k$ , re-index the school according the time it is eaten in the  $\text{PS}_k$  algorithm. From the definition of PS, if school  $s_1$  and  $s_2, s_1 < s_2$  are eaten at the same time,  $y_{s_2} = 0$ , the school with smaller original index is eaten first, then the school with larger index is eaten in 0 units of time.

Then the time school  $l$  is eaten is  $t_l = \sum_{s=1}^l y_s$ .

Define the change in finish time of  $l$ :

$$\Delta t_l = \frac{1}{\sum_{p \in M(l, L^{l-1})} \mu(p)}$$

and the change in total value due to change in capacity  $l$ :

$$\Delta V_l(\varepsilon) = V(c_l + \varepsilon) - V(c_l)$$

Define expected utility from eating  $l$ :

$$\Delta u_l = \sum_{p \in M(l, L^{l-1})} \mathbb{E}[u(l; p) \mu(p)]$$

Then we have

$$\Delta V_l(\varepsilon) = \Delta t_l \cdot \varepsilon \cdot \Delta u_l + \sum_{i=1}^{L-l} \Delta V_{l+i}(\eta_{l+i}) \text{ for some } \eta_s < \varepsilon \forall s.$$

$$\text{where } \eta_s = -\Delta t_l \cdot \varepsilon \cdot \left( \sum_{p \in M(s, L^{l-1})} \mu(p) \right) \cdot \Delta t_s$$

$$\text{and } \Delta V_L(\varepsilon) = \Delta t_L \cdot \varepsilon \cdot \Delta u_L$$

$$\begin{aligned} \frac{dV}{dc_l} &= \Delta t_l \cdot \left( \Delta u_l + \sum_{i=1}^{L-l} \left( \sum_{p \in M(l+i, L^{l-1})} \mu(p) \right) \cdot \Delta t_{l+i} \frac{dV}{dc_{l+i}} \right) \\ &= \sum_{i=0}^{L-l} w_i \Delta t_{l+i} \cdot \Delta u_{l+i} \text{ for some weights } w_i \text{ with } w_0 = 1, w_i < 0 \text{ for } i > 0 \end{aligned}$$

For  $\varepsilon_{l_1}, \varepsilon_{l_2}$  small enough, define the following for  $l_1 < l_2$ :

$$\Delta V(\varepsilon_{l_1}, \varepsilon_{l_2}) = V(c_{l_1} + \varepsilon_{l_1}, c_{l_2} + \varepsilon_{l_2}) - V(c_{l_1}, c_{l_2})$$

If  $t_{l_1} = t_{l_2}$ , meaning  $l_1$  and  $l_2$  are eaten at the same time, then:

(1) If  $\Delta t_{l_1} \varepsilon_{l_1} > \Delta t_{l_2} \varepsilon_{l_2}$ ,

$l_2$  will be eaten before  $l_1$  after the  $\varepsilon$  change in capacity:

$$\begin{aligned} \Delta V(\varepsilon_{l_1}, \varepsilon_{l_2}) &= \Delta t_{l_2} \cdot \varepsilon_{l_2} \cdot \left( \Delta u_{l_1} + \Delta u_{l_2} + \sum_{i=1}^{L-l} \Delta V_{l_2+i}(\eta_{l_2+i}) \right. \\ &\quad \left. + \Delta V_{l_2} \left( \varepsilon_{l_1} - \frac{\Delta t_{l_2} \varepsilon(l_2)}{\Delta t_{l_1}} \right) \right) \end{aligned}$$

(2) If  $\Delta t_{s_1} \varepsilon_1 < \Delta t_{s_2} \varepsilon_2$ ,



$s_1$  will be eaten before  $s_2$  after the  $\varepsilon$  change in capacity:

$$\begin{aligned} \Delta V(\varepsilon_{l_1}, \varepsilon_{l_2}) = & \Delta t_{l_1} \cdot \varepsilon_{l_1} \cdot \left( \Delta u_{l_1} + \Delta u_{l_2} + \sum_{i=1}^{L-l} \Delta V_{l_2+i}(\eta_{l+i}) \right. \\ & \left. + \Delta V_{l_1} \left( \varepsilon_{l_2} - \frac{\Delta t_{l_1} \varepsilon(l_1)}{\Delta t_{l_2}} \right) \right) \end{aligned}$$

(3) If  $\Delta t_{s_1} \varepsilon_1 = \Delta t_{s_2} \varepsilon_2$ ,

$s_1$  and  $s_2$  will remain getting eaten at the same time. Same as the one-dimension case.

The directional derivative in the direction  $v$  is:

$$\begin{aligned} \nabla_v V &= \sum_{l=1}^L w_l \Delta t_l \cdot \Delta u_l \text{ for some weights } w_l \\ &= \sum_{l=1}^L w_l \frac{\sum_{p \in M(l, L^{l-1})} \mathbb{E}[u(p, l) \cdot \mu(p)]}{\sum_{p \in M(l, L^{l-1})} \mu(p)} \text{ for some weights } w_l \end{aligned}$$

Moving in direction  $v$  satisfying the following condition will maintain the ordering of the schools:

$$\begin{aligned} \Delta t_{l_1} v_{l_1} &= \Delta t_{l_2} v_{l_2} \quad \forall l_1, l_2 \text{ such that } t_{l_1} = t_{l_2} \\ \frac{v_{l_1}}{\sum_{p \in M(l_1, L_1^{l_1-1})} \mu(p)} &= \frac{v_{l_2}}{\sum_{p \in M(l_2, L_2^{l_2-1})} \mu(p)} \quad \forall l_1, l_2 \text{ such that } t_{l_1} = t_{l_2} \end{aligned}$$

This can be extended to cases where more than two schools are eaten at the same time.

□