

On scheduling fees to prevent merging, splitting and transferring of jobs

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Abstract

A deterministic server is shared by users with identical linear waiting costs, requesting jobs of arbitrary lengths. Shortest jobs are served first for efficiency. When the server can monitor the length of a job, but not the identity of its user, merging, splitting or partially transferring jobs may offer strategic opportunities to cooperative agents. Can we design cash transfers to neutralize such manipulations?

We prove that merge-proofness and split-proofness are not compatible, and that it is similarly impossible to prevent all transfers of jobs involving three agents or more. On the other hand, robustness against pairwise transfers is feasible, and essentially characterizes a one-dimensional set of scheduling methods. This line is borne by two outstanding methods, the merge-proof S^+ and the split-proof S^1 .

Splitproofness, unlike Mergeproofness, is not compatible with several simple tests of equity. Thus the two properties are far from equally demanding.

Key words: scheduling, queuing, merging, splitting, transferring, linear waiting cost.

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1 The problem and the punch lines

Dividing the burden of joint externalities raises many issues of incentive-compatibility. One of these is the strategic transfer, or merging, or splitting, of certain private characteristics of the participants. This type of manipulation is discussed in the fair division literature (see details in section 2); here we study it in a simple scheduling problem with transferable utility. A single deterministic server/machine is shared by users with linear waiting costs, requesting jobs of arbitrary lengths. A job of length x_i takes x_i units of time to process; an agent's disutility is the waiting time until her job is completed, augmented by a (positive or negative) cash payment selected by the mechanism. The key assumption is that the server can monitor the length of a job, but not the identity of its user. This creates opportunities for manipulation if the agents can costlessly merge two jobs of lengths x_i, x_j into a single job of length $x_i + x_j$, sent under one of their names; or if they can split a job x_i into two smaller jobs x_i^1, x_i^2 with $x_i^1 + x_i^2 = x_i$, and send them under two aliases; or, ...nally, if they can transfer a fraction of job x_i and add it to job x_j .

The key assumption is realistic when the usage of the server/machine is private, and can't be traced to its actual bene...ciary. Think of a tool that agents carry to their private workplace, for instance a software used on a private machine. Or consider single access to a database, when the needs of each user of the link are private, so the server cannot detect if and when the link is used by agent i on behalf of another agent j . In huge networks such as the internet, assuming a false identity is very easy, and an important issue of the network design is to protect the system performance against such moves: Douceur [2002]. On the other hand, merging, splitting or transferring jobs is costless if the job takes the form of an electronic document, or of a physical tool easily transported from one job to the next.

Two very simple scheduling mechanisms illustrate the cooperative manipulations that we wish to prevent. Given identical linear waiting costs and the availability of cash transfers, e¢ciency requires to serve the shortest jobs ...rst (Smith [1956]). Suppose the server does this and performs no monetary transfer (at least when all jobs are of di¢erent length, so the e¢cient scheduling order is unique). This mechanism is highly vulnerable to splitting maneuvers: If the two real jobs are $x_1 = 4, x_2 = 3$, agent 1 splits his job as $x_1^0, x_1^{00} = 2$ and cuts his waiting cost by \$3. Partial transfers may also be pro...table: say the three jobs are $(x_1, x_2, x_3) = (1, 4, 5)$; if agent 3 transfers 2

units of job to agent 1, resulting in $(x_1^0, x_2, x_3^0) = (3, 4, 3)$, she will complete x_3 before agent 2 is served, and the net gain \$4 can be divided between agents 1 and 3. But the merging of jobs is clearly not profitable, as this can only delay the completion of these jobs.

Consider next a mechanism serving the longest jobs first, namely one that maximizes total waiting cost. No matter how it deals with ties, this mechanism is badly vulnerable to merging maneuvers, as well as to partial transfers: simply use the above examples backward. On the other hand, the splitting of a job is never profitable.

Can we design a system of cash transfers to prevent in all problems single agents from splitting their job, and coalitions from merging them under a single identity? And what about partial transfers of jobs?

Despite the simplicity of our scheduling model, some of the answers to these questions are disappointingly negative. If the potential set of users contains at least 4 agents, a mechanism treating equals equally cannot be both merge-proof and split-proof: Theorem 1 in Section 4. Moreover every continuous mechanism (i.e., net waiting costs depend continuously upon the profile of job lengths) is vulnerable to transfers involving three agents or more: Section 8.

On the other hand the family of merge-proof scheduling mechanisms is fairly large, and so is that of split-proof mechanisms. Moreover, each family contains many mechanisms invulnerable to job transfers involving only two agents, as explained below. Yet we find that split-proofness, in contrast to merge-proofness, is incompatible with several compelling fairness requirements. Proposition 1 in Section 5 gives a precise content to this statement. Restrict attention to efficient mechanisms (serving successively jobs of increasing length) treating equals equally, and continuous. Every splitproof mechanism in this class must charge a positive fee to null jobs, who create no externality whatsoever; it must also subsidize some jobs in the sense that their net waiting cost is smaller than x_i ; next, the net cost of agent i is not always weakly increasing in his job length x_i ; and finally, the ordering of net costs must sometime contradict that of job lengths. By contrast, merge-proofness is compatible with all four properties just described.

In Section 6, we construct a large family of efficient scheduling mechanisms, treating equals equally and continuous as above, and for which the role of merge-proofness and split-proofness is especially easy to describe. Pick a continuous function θ from \mathbf{R}_+^2 into \mathbf{R} such that $\theta(a, b) + \theta(b, a) = \min\{a, b\}$ for all a, b . Label the set of users $N = \{1, 2, \dots, n\}$ in such a way that

$x_1 \cdot x_2 \cdot \dots \cdot x_n$. The θ -mechanism serves the job in the efficient order $1, 2, \dots, n$ and performs cash transfers resulting in the net waiting cost $y_i = x_i + \sum_{j \in I_i} \theta(x_i, x_j)$ for all i . By construction of θ , this gives $\sum_i y_i = nx_1 + (n-1)x_2 + \dots + x_n$, so that these transfers are balanced.

We call the above mechanism separable because it divides the externality $\min_{i,j} x_i x_j$ between any two agents i, j without paying attention to other job lengths. Proposition 2 in Section 6 characterizes merge-proof separable methods by a system of inequalities slightly less demanding than the super-additivity of θ in its first variable, and split-proof separable methods by a similar system slightly more demanding than the sub-additivity of θ in its first variable.

Two separable mechanisms stand out. The first one, called S^+ , splits the (i, j) -externality equally, namely $\theta^+(a, b) = \frac{1}{2} \min\{a, b\}$. The second mechanism, called S^i , uses the function $\theta^i(a, b) = b \wedge \frac{1}{2} \max\{a, b\}$. The method S^+ corresponds to the Shapley value of the optimistic stand alone cooperative game (a coalition S standing alone is served before N/S ; the method S^i corresponds to the Shapley value of the pessimistic stand alone cooperative game (a coalition S standing alone is served after N/S).

We find that S^+ is merge-proof, whereas S^i is split proof - hence the latter shares all unpalatable consequences of splitproofness discussed above.

In Section 7 we turn to the strategic transfer of jobs. We restrict attention to job transfers involving only two agents, combined with cash transfers within a coalition of arbitrary size. We show that S^+, S^i as well as their affine combinations $y = a \cdot y^+ + (1-a) \cdot y^i, a \in \mathbb{R}$, are not vulnerable to such manipulations. Our main result, Theorem 2, is a characterization of the line of methods borne by S^+ and S^i based on this property of pairwise transfer-proofness. Then we characterize the S^+ method either by adding the requirement that null jobs should not pay (or receive) anything, or by ruling out subsidies beyond the optimistic stand alone wait $(x_i \cdot y_i)$.

2 Related literature

The earliest discussion in the fair division literature of manipulation by merging, splitting, and transferring, is in the rationing problem: each agent has a claim/liability over an amount of money smaller than the sum of individual claims/liabilities. If the claims take the form of anonymous, transferable bonds, dividing the money in proportion to individual claims is the only

method making transfers - as well as merging or splitting - unprovable: Banker [1984]. Variants and extensions of this result are in Moulin [1987], DeFrutos [1999], and Ju [2003]. Related properties of transfer-proofness appear in the quasi-linear social choice problem (Moulin [1985] Chun [?]), in axiomatic costsharing (Sprumont [2004]) and more: Ju and Miyagawa [2003] offer a unified treatment of most of this literature.

We now review the recent and growing microeconomic literature on scheduling. A familiar variant of our model has linear waiting costs that may vary across participants. A scheduling problem consists of a profile of job lengths x_i and waiting cost δ_i per unit of time. Agent i 's utility is then $\delta_i w_i + t_i$, where w_i is waiting time until completion of job i and t_i is the cash payment. Minimizing total waiting cost requires to serve the jobs in the increasing order of the ratios $\frac{x_i}{\delta_i}$ (Smith [1956]).

The mechanism designer can use the cash transfers to ensure truthful (dominant strategy) elicitation of the privately known waiting costs: utilities are linear in money (and waiting costs) so that Vickrey-Clarke-Groves mechanisms can be readily applied. The first authors to explore this idea are Dolan [1978] and Mendelson and Whang [1990]. In fact, given linear waiting costs, we can construct a budget-balanced (fully efficient) VCG mechanism: Suijs [1996], Mitra and Sen [1998], Mitra [2002]. If we must elicit job lengths instead of waiting costs, a similar construction is possible (Hain and Mitra [2001], Kittsteiner and Moldovanu [2003a,b]), provided the VCG mechanisms are suitably generalized to take into account the more complicated allocative externalities from misreporting the length of one job.

Another way to use cash transfers in the linear scheduling model is to ensure fairness, namely an equitable sharing of the congestion externality. Several authors simply apply off-the-shelf solution concepts like the Shapley value or the core to a relevant cooperative game: Curiel et al. [1989], [1993], [2002], Hamers et al. [1996]. The most natural solution is the Shapley value of a stand alone cooperative game. This solution is axiomatized by Maniquet [2003], in the case of identical job lengths, and also discussed by Curiel et al. [1993] and Klijn and Sanchez [2002]. It plays an important role in the current paper as solution S^+ . Our second solution S^i is similarly axiomatized by Chun [2004].

Our approach is original on two accounts. First we explore a new kind of cooperative manipulation, quite different from the misreport of waiting costs or of job lengths. In our model, individual preferences are known to the server, and job lengths are observable. All the action comes from the inability

of the server to detect the true identity of users: this allows participants to request a job, or part of a job, without revealing its true beneficiary.

Secondly we explore the compatibility of our strategy-proofness properties with four classic equity tests, based on monotonicity and bounds on individual disutilities (see Section 5). These tests are all familiar to the fair division literature, and play a role as well in the work of Maniquet [2003] and Chun [2004]. Here they reveal a fundamental asymmetry between the requirements of merge-proofness and split-proofness (Proposition 1).

In related work in progress, Maniquet and Moulin [2004] explore splitting, merging and transferring maneuvers in the model with variable job length and waiting costs described above. The methods S^+ , S^i generalize, and share similar robustness properties. Finally Moulin [2004] discusses the same strategic maneuvers when the server instead of cash transfers, uses randomization. In that context, the properties of merge-proofness and split-proofness are compatible, yet the latter remains a much more demanding property than the former.

3 The model

The set N contains all potential users of the simple machine. It may be finite or infinite. A scheduling problem involves a finite subset N of N . Agent i 's job is completed in exactly x_i units of machine-time. Given a scheduling problem (N, x) , where $x \in \mathbb{R}_+^N$, the mechanism designer - thereafter "the server" - must choose the ordering σ of N - the schedule - in which the jobs will be served, and a vector $t, t \in \mathbb{R}^N$, of monetary transfers such that $\sum_{i \in N} t_i = 0$.

Each agent incurs a waiting cost of \$1 per unit of time, until completion of his/her job (a partially completed job is useless). The equality of waiting costs is a simplifying assumption. Several of our results are preserved when we allow arbitrary linear waiting costs, known to the server: see Maniquet and Moulin [2004].

We write $\sigma(i) < \sigma(j)$ to mean that agent i precedes agent j in the ordering σ , and $P(i, \sigma) = \{j \in N / \sigma(j) < \sigma(i)\}$ is the set of agents preceding i in σ . Thus the disutility of agent i given σ and t is

$$y_i = x_i + \sum_{j \in P(i, \sigma)} x_j + t_i \quad (1)$$

Notice that t_i is a tax on agent i when $t_i > 0$ and a subsidy when $t_i < 0$.

Because monetary transfers are unrestricted, efficiency amounts to choose an ordering σ minimizing total waiting cost

$$\sum_N (x_i + \sum_{P(i,\sigma)} x_j) = x_N + \sum_{(i,j):\sigma(i)<\sigma(j)} x_i$$

An ordering is efficient if and only if it schedules shortest jobs first. In other words their set $E(N, x)$ is characterized by

$$\sigma \in E(N, x) \iff \text{for all } i, j \in N : x_i < x_j \implies \sigma(i) < \sigma(j)$$

We use the notations $a \wedge b = \min\{a, b\}$ and $N(2)$ for the set of all subsets (non-ordered pairs) (i, j) of distinct agents. Then the minimal total waiting cost $v(N, x)$ can be written as

$$v(N, x) = x_N + \sum_{N(2)} x_i \wedge x_j$$

Definition 1 Given N , a scheduling mechanism μ associates to every problem (N, x) , where $N \subseteq \mathbb{N}$ and $x \in \mathbb{R}_+^N$, a pair $\mu(N, x) = (\sigma, t)$, where σ is an ordering of N and $t \in \mathbb{R}^N$ with $t_N = 0$. A scheduling method m associates to every problem (N, x) a profile of net waiting costs $m(N, x) = y, y \in \mathbb{R}^N$, such that

$$y_N = x_N + \sum_{(i,j):\sigma(i)<\sigma(j)} x_i, \text{ for some ordering } \sigma \text{ of } N.$$

To each mechanism μ , we associate a method $x \mapsto y$ by formula (1). We call the mechanism μ efficient if $\sigma \in E(N, x)$ for all N, x ; we call the method m efficient if $y_N = v(N, x)$ for all N, x . To an efficient method m corresponds essentially a unique efficient mechanism μ : the only qualification is at those problems x where some jobs have equal length, $x_i = x_j$, so that $E(N, x)$ is not a singleton. As this will cause no confusion, we shall state some of our axioms for mechanisms (e.g. Merge-proofness) and some of them for methods.

The next property is the standard requirement of horizontal equity:

$$\begin{aligned} &\text{Equal Treatment of Equals (ETE):} \\ &\text{for all } (N, x), i, j \in N : x_i = x_j \implies y_i = y_j \end{aligned}$$

All methods discussed below meet ETE, yet this property is not necessary to our main characterization result (Theorem 2). By contrast, the following axiom plays a key role in Theorem 2.

Continuity (CONT):

for all N , the mapping $x \mapsto y(N, x)$ is continuous on \mathbb{R}_+^N

Continuity ensures that microscopic variations in the job lengths do not have a macroscopic impact on the profile of net waiting costs. In particular when $x_i = x_j$, a small tremble of x_i - the result of a measurement error, or of a strategic move - is not a matter of concern to agents i, j , or to anyone else.

Our first example is a natural and discontinuous mechanism.

Example 1 Shortest job first

For every (N, x) where $x_i \neq x_j$ for all i, j , the mechanism selects the unique efficient ordering σ and performs no transfers. At other profiles, it performs the minimal transfers required by ETE. If at x we have exactly k agents with $x_i = a$ for some a , it orders them arbitrarily, say $i_1 < i_2 < \dots < i_k$, and performs the transfers

$$t_{i_1} = \frac{k-1}{2}a, t_{i_2} = \frac{k-3}{2}a, \dots, t_{i_k} = \frac{(k-1)}{2}a$$

In other words, the mechanism is defined up to a tie-breaking rule, but the corresponding method is unique:

$$y_{i_1} = y_{i_2} = \dots = y_{i_k} = \frac{k+1}{2}a + \sum_{j: x_j < a} x_j$$

It is easy to define efficient scheduling methods meeting ETE and CONT. For instance, the proportional method:

$$y_i = \frac{x_i}{x_N} v(N, x) \text{ for all } x \neq 0; y = 0 \text{ for } x = 0 \quad (2)$$

and the egalitarian method:

$$y_i = x_i + \frac{1}{n} \left(\sum_{j \in N(2)} x_i \wedge x_j \right) \text{ for all } N, x \quad (3)$$

The latter charges the same net cost to every agent beyond his/her own stand alone cost. Both methods are reasonable in terms of the four criteria discussed in Section 5, yet they are vulnerable to the coalitional maneuvers to which we now turn.

4 Merging and Splitting

The server can recognize the length of the jobs it performs, but not the identity of the beneficiary of those jobs. This allows agents to merge several jobs under a single identity, or to split a given job in several small jobs under multiple identities.

Given $N \in \mathbf{N}$, a coalition S , $S \in \mathbf{N}$, and an agent $i \in S$, we associate to every problem (N, x) the (S, i) - merged problem (N^S, x^S) as follows

$$N^S = (N \setminus S) \cup \{i\}; x_{i^S}^S = x_S \text{ and } x_j^S = x_j \text{ for all } j \in N \setminus S$$

We also use the notation $v(S, x) = x_S + \sum_{i \in S} x_i$ for the stand alone waiting cost of coalition S , namely the efficient total wait of S when it is served before $N \setminus S$. Given a mechanism μ on \mathbf{N} we define:

Merge-proofness (MPF)
for all N, S, i as above and all $x \in \mathbf{R}^N$:

$$\mu(N^S, x^S) = (\sigma^S, t^S) \Rightarrow y_S(N, x) \cdot v(S, x) + \sum_{i \in S^0} x_{P(i^S, \sigma^S)} + t_{i^S}^S \quad (4)$$

where S^0 is the subset of S defined by $x_i > 0$.

In this inequality the left-hand side is the net waiting cost of coalition S before merging, and the right-hand side its net cost after merging. Indeed coalition S uses efficiently the slot of length x_S allocated to agent i^S , and moreover everyone in S with a non null job must wait until completion of all jobs in $P(i^S, \sigma^S)$. Note that for $S = N$, the merge-proofness inequality is just the efficiency property.

Given $N \in \mathbf{N}$, $i \in N$, and a finite set $T \in \mathbf{N}$, $T \setminus N = \emptyset$, we associate to every problem (N, x) , the family of (T, i) -splitted problems (N_π, x_π) as follows

$$N_\pi = N \cup T; (x_\pi)_{T \setminus i} = x_{i_\pi} \text{ and } (x_\pi)_j = x_j \text{ for all } j \in N \setminus i_\pi$$

Given a mechanism μ on \mathbf{N} we define:

Split-proofness (SPF)
for all N, T, i as above, all $x \in \mathbf{R}^N$ and all (T, i) -splitted problem (N_π, x_π)

$$\mu(N_\pi, x_\pi) = (\sigma_\pi, t_\pi) \Rightarrow y_{i_\pi}(N, x) \cdot x_{i_\pi} + x_{P(j_\pi, \sigma_\pi)} + (t_\pi)_{T \setminus i_\pi} \quad (5)$$

where j_π is the last agent in $T \setminus i_\pi$ for σ_π .

Agent i_n^0 net cost before splitting is on the left-hand side; after the split, i_n must wait until all jobs in $P(j_n, \sigma_n)$ are completed¹, therefore his net cost is on the right-hand side.

As discussed in the Introduction, Shortest Job First is not split-proof, but it is merge-proof. Symmetrically, Longest Job First is split-proof, but not merge-proof.

We check now that the egalitarian method (3) is neither merge-proof nor split-proof. In the problem $N = f1, 2, 3g, x = (1, 1, 4)$, consider the split of x_3 into x_{n3} and x_{n4} , with $x_{n3} = x_{n4} = 2$. The actual wait of agent 3 in (N, x) is the same as in $N_n = f1, 2, 3, 4g, x_n = (1, 1, 2, 2)_i$ under the assumed identities - and the monetary transfer is smaller in the latter, hence the split is profitable:

$$y_3 = 6 + t_3 = 4 + \frac{1}{3}(1 + 1 + 1) \quad t_3 = 1$$

$$(y_n)_{34} = 4 + 6 + (t_n)_{34} = 2(2 + \frac{1}{4}(7)) \quad (t_n)_{34} = 2.5$$

In the problem $N = f1^n, 2, 3, 4g, x = (2, 2, 5, 5)$, consider the merging of x_{1^n} and x_2 . The actual total wait of agents 1, 2 in (N, x) is unchanged as they merge into 1^n in $N^n = f1^n, 3, 4g, x^n = (4, 5, 5)$. And the net transfer decreases, making the move profitable:

$$y_{1^n 2} = 6 + t_{12} = 2(2 + \frac{1}{4}(15)) \quad t_{12} = 5.5$$

$$y_{1^n}^n = 4 + t_{1^n}^n = 4 + \frac{1}{3}(13) \quad t_{1^n}^n = 4.33$$

We let the reader check similarly that the proportional method (2) is not split-proof, by considering the split of agent 1 from $(f1, 2g, (5, 4))$ into $(f1, 2, 3g, (5, 2, 2))$. On the other hand, the proportional method is merge-proof. We omit the easy proof.

Theorem 1 Assume $|N| \geq 4$. There is no scheduling mechanism satisfying Merge-proofness, Split-proofness, and either Continuity or Equal Treatment of Equals.

The proof is in the Appendix. Recall that merge-proofness implies in particular efficiency. If we restrict the merge-proofness property by allowing

¹At least if $(x_n)_{j_n} > 0$. If $(x_n)_{j_n} = 0$, we should replace j_n by the last agent in $T \setminus i_n$ with a positive job. But this does not affect the statement of SPF.

only the merging of proper coalitions, then there may exist some (inefficient) mechanisms meeting MPF and SPF. I conjecture that this is not the case.

5 Unpalatable consequences of Split-proofness

The formal similarity between the two maneuvers of merging and splitting suggests that the properties MPF and SPF are comparably demanding. This intuition is not correct. We list below four mild normative requirements that we may want to impose on a scheduling method. Then we show that any "reasonable" split-proof method must violate each one of these four properties. In the following statements, we say a method $(N, x) \rightarrow y$:

- Monotonicity (MON)
for all $N, i \in N, x_i \in \mathbb{R}^{N_i} : x_i \rightarrow y_i(N, x)$ is non-decreasing
- Ranking (RKG)
for all $N, i, j \in N, x \in \mathbb{R}^N : (x_i \leq x_j) \rightarrow (y_i \leq y_j)$
- Stand Alone bound (SAB)
for all $N, i \in N, x \in \mathbb{R}^N : y_i \leq x_i$
- Zero Charge for Null Jobs (ZCNJ)
for all $N, i \in N, x \in \mathbb{R}^N : x_i = 0 \rightarrow y_i = 0$

The first two properties are standard equity tests. The Stand Alone bound sets a minimal net waiting cost, namely my disutility in the most optimistic case where I have absolute priority for service. It rules out the subsidization of any agent beyond this most advantageous situation. Monotonicity says that my net waiting cost weakly increases when my job becomes longer: besides its clear normative meaning, this property also rules out "sabotage" by artificially increasing one's job length. Ranking conveys a related idea by way of interpersonal comparisons: if my job is longer than yours, my responsibility in the total waiting burden is higher.

Finally, ZCNJ frees a "null job" agent of any responsibility: such an agent is served first by efficiency, and causes no additional waiting cost to any one. Under ZCNJ, he is not taxed either, $t_i = 0$. The combination of Continuity and ZCNJ implies that y_i converges to zero with x_i .

Many scheduling methods meet these four properties. Examples include Shortest Job First (example 1), the proportional method (2), their convex combinations and much more. The egalitarian method (3) fails ZCNJ and

meets the other three. We now state a negative result about split-proof scheduling methods.

Proposition 1 Fix N and an efficient and continuous scheduling method treating equals equally. If this method is split-proof and $N \geq 5$ then it fails Monotonicity, Ranking and the Stand Alone bound. If $|N| \geq 1$, it fails Zero Charge for Null Jobs as well.

Note that we construct in the next section an efficient merge-proof method meeting all the other axioms listed here: ETE, CONT, MON, RKG, SAB and ZCNJ.

Proof of Proposition 1

Monotonicity. Let $N = \{1, 2, 3, 4\}$ and $x(\varepsilon) = (1, 1, 1, 2(1 + \varepsilon))$. Consider the split of agent 4 into agents 4,5 and $x_\pi(\varepsilon) = (1, 1, 1, 1 + \varepsilon, 1 + \varepsilon)$. By CONT and ETE:

$$\lim_{\varepsilon \downarrow 0} y(x_\pi(\varepsilon)) = y(x_\pi(0)) = (3, 3, 3, 3, 3)$$

By efficiency and CONT again:

$$y_{45}(x_\pi(\varepsilon)) = (4 + \varepsilon) + (5 + \varepsilon) + t_{45}(x_\pi(\varepsilon)) \leq \lim_{\varepsilon \downarrow 0} t_{45}(x_\pi(\varepsilon)) = 3$$

Because the split is not profitable for agent 4, and her real wait after the split is unchanged, we have

$$y_4(x_\pi(\varepsilon)) \leq 5 + 2\varepsilon + t_{45}(x_\pi(\varepsilon)) \leq y_4(x(0)) \leq 2$$

On the other hand, at $\bar{x} = (1, 1, 1, 1)$ ETE gives $y_4(\bar{x}) = 2.5$, and we see that Monotonicity is violated as x_4 goes from 1 to 2.

Ranking. Let $N = \{1, 2, 3\}$ and $x(\varepsilon) = (1, 1, 2(1 + \varepsilon))$. Consider the split of agent 3 into 3,4 and $x_\pi(\varepsilon) = (1, 1, 1 + \varepsilon, 1 + \varepsilon)$. Mimicking the argument of the proof above we get successively

$$\lim_{\varepsilon \downarrow 0} y(x_\pi(\varepsilon)) = (2.5, 2.5, 2.5, 2.5), \lim_{\varepsilon \downarrow 0} t_{34}(x_\pi(\varepsilon)) = 2, \text{ and } y_3(x(0)) \leq 2$$

Now efficiency and ETE give $y_1(x(0)) = y_2(x(0)) \geq 2.5$, a contradiction of RKG.

Stand Alone bound. Let $N = \{1, 2\}$ and $x(\varepsilon) = (1, 3(1 + \varepsilon))$. Consider the split of agent 2 into 2,3,4 and $x_\pi(\varepsilon) = (1, 1 + \varepsilon, 1 + \varepsilon, 1 + \varepsilon)$. As before we have successively

$$\begin{aligned} \lim_{\varepsilon \downarrow 0} y(x_\pi(\varepsilon)) &= (2.5, 2.5, 2.5, 2.5), \lim_{\varepsilon \downarrow 0} t_{234}(x_\pi(\varepsilon)) = 1.5, \\ &\leq y_2(x(0)) \leq 2.5 < 2 = x_2(0) \end{aligned}$$

This contradicts SAB.

Zero Charge Null Jobs. Let $N = \{1, 2\}$ and let an integer $p, p \geq 2$. Set $x(\varepsilon) = (\frac{1}{p}, 1 + p\varepsilon)$ and consider the split of 2 into 2, 3, ..., $p + 1$ and $x_\pi(\varepsilon) = (\frac{1}{p}, \frac{1}{p} + \varepsilon, \dots, \frac{1}{p} + \varepsilon)$. As before we have $\lim_{\varepsilon \downarrow 0} y(x_\pi(\varepsilon)) = \frac{p+2}{2p}$ implying $\lim_{\varepsilon \downarrow 0} t_1(x_\pi(\varepsilon)) = \frac{1}{2}$. Then split-proofness implies $y_2(x_\pi(\varepsilon)) \leq \frac{1}{p} + 1 + p\varepsilon \leq t_1(x_\pi(\varepsilon))$ hence $y_2(x(0)) \leq \frac{1}{2} + \frac{1}{p}$ and $y_1(x(0)) \leq \frac{1}{2} + \frac{1}{p}$. But CONT and ZCNJ imply $\lim_{p \rightarrow \infty} y_1((\frac{1}{p}, 1)) = 0$, contradiction.

Remark 1 For the statements about Ranking and the Stand Alone bound, the assumption Equal Treatment of Equals is redundant. In other words, any efficient, continuous and split-proof method must violate Ranking and the Stand Alone bound for $|N| \geq 4$. To check this, take a set N_π with four agents. Setting $x_\pi(0) = (1, 1, 1, 1)$, we have $y_{N_\pi}(x_\pi(0)) = 10$ thus there exists a pair i, j in N_π such that $y_{ij}((1, 1, 1, 1)) \leq 5$. Label the agents so that $i = 3, j = 4$ and $N_\pi = \{1, 2, 3, 4\}$. Define as in the proof about Ranking $N = \{1, 2, 3\}$, $x(\varepsilon)$ and $x_\pi(\varepsilon)$. CONT ensures $\lim_{\varepsilon \downarrow 0} y_{34}(x_\pi(\varepsilon)) \leq 5$, then $\lim_{\varepsilon \downarrow 0} t_{34}(x_\pi(\varepsilon)) \leq 2$, and split-proofness gives $y_3(x(0)) \leq 2$. Therefore $y_{12}(x(0)) \leq 5$ so that $y_i(x(0)) \leq 2.5$ for at least one of 1, 2. Thus ranking fails. The similar proof about SAB is omitted for brevity.

Whether or not we can drop ETE from the assumptions on our method in the two remaining statements is an open question.

6 Separable scheduling methods

The total waiting externality in the problem (N, x) is $v(N, x) = x_N + \sum_{i < j} x_i \wedge x_j$, namely the cost of having to share the server. A separable method shares each pairwise externality $x_i \wedge x_j$ independently of the rest of the jobs.

Definition 1 Choose a continuous function θ from \mathbb{R}_+^2 into \mathbb{R} such that $\theta(a, b) + \theta(b, a) = a \wedge b$ for all $a, b \in \mathbb{R}_+$. The θ -separable scheduling method is given by

$$y_i(N, x) = x_i + \sum_{j \in N, j \neq i} \theta(x_i, x_j) \text{ for all } N, i \in N \text{ and } x \in \mathbb{R}_+^N$$

Notice that the θ -separable method places no restriction on the set of agents N , or on N . It is obviously efficient and continuous, and treats equals equally.

The Shortest Job First method is θ -separable, except that the function θ is not continuous:

$$\theta(a, b) = 0 \text{ if } a < b; = \frac{a}{2} \text{ if } a = b; = a \text{ if } a > b$$

Neither the egalitarian nor the proportional method is separable.

We speak of a θ_i separable mechanism for any mechanism generating the method in Definition 1.

Proposition 2 The θ -separable scheduling method meets

- i) Monotonicity , $\theta(a, b)$ is non-decreasing in a .
- ii) Ranking , $\theta(a, b)$ is non-decreasing in a and $[a, b] \Rightarrow \theta(a, b) \cdot \frac{b}{2}$.
- iii) Stand Alone Bound , $\theta(a, b) \geq 0$ for all a, b .
- iv) Zero Charge for Null Job , $\theta(0, b) = 0$ for all b .

A θ -separable mechanism is merge-proof if and only if

$$\begin{aligned} \theta(a_1, b) + \theta(a_2, b) &\leq \theta(a_1 + a_2, b) \text{ for all } b, a_1, a_2 \text{ s.t. } a_1 + a_2 \leq b \quad (6) \\ \theta(a_1, b) + \theta(a_2, b) &\leq \theta(a_1 + a_2, b) + b \text{ for all } b, a_1, a_2 \end{aligned}$$

A θ -separable mechanism is split-proof if and only if

$$\begin{aligned} \theta(a_1 + a_2, b) + b &\leq \theta(a_1, b) + \theta(a_2, b) \text{ for all } b, a_1, a_2 \text{ s.t. } b \leq a_1, a_2 \quad (7) \\ \theta(a_1 + a_2, b) &\leq \theta(a_1, b) + \theta(a_2, b) \text{ for all } b, a_1, a_2 \quad (8) \end{aligned}$$

Proof of Proposition 2

Statement i. Suppose $\theta(a, b) > \theta(a^0, b)$ for some $a < a^0$. Fix n and consider the $(n + 1)_i$ agents profiles $x = (a, b, \dots, b)$ and $x^0 = (a^0, b, \dots, b)$. For n large enough, we have

$$y_1(x) = a + n \cdot \theta(a, b) > a^0 + n \cdot \theta(a^0, b) = y_1(x^0)$$

contradicting MON. The converse statement is obvious.

Statement ii. Suppose $\theta(a, b) > \theta(a^0, b)$ for some $a < a^0$. Consider the $(n + 2)_i$ agents profile $x = (a, a^0, b, \dots, b)$. For n large enough we get $y_1 > y_2$, contradicting Ranking. Thus θ must be monotonic in its first variable. Next we fix $a, b, a < b$, and apply Ranking to $x = (a, b)$:

$$y_1(x) = a + \theta(a, b) \cdot y_2(x) = b + \theta(b, a) = b + a \leq \theta(a, b)$$

establishing the second property in statement ii. The converse property is just as easy.

3). We turn to the characterization of Merge-proofness. Fix N, S, i^π, x as in the premises of (4) and develop this inequality for our θ_i separable method. Compute first $t_{i^\pi}^\pi$:

$$\begin{aligned} y_{i^\pi}(N^\pi, x^\pi) &= x_{i^\pi}^\pi + x_{P(i^\pi, \sigma^\pi)}^\pi + t_{i^\pi}^\pi = x_{i^\pi}^\pi + \sum_{NnS} \theta(x_{i^\pi}^\pi, x_j) \\ \Rightarrow t_{i^\pi}^\pi &= \sum_{NnS} x_{P(i^\pi, \sigma^\pi)} + \theta(x_S, x_j) \end{aligned}$$

Next the definition of θ implies

$$\begin{aligned} y_S(N, x) &= x_S + \sum_{S(2)} f\theta(x_i, x_j) + \theta(x_j, x_i)g + \sum_{i \in S, j \in NnS} \theta(x_i, x_j) \\ &= v(S, x) + \sum_{j \in NnS} \sum_{i \in S} \theta(x_i, x_j) \end{aligned}$$

Therefore inequality (4) amounts to

$$0 \leq (j \in S^0 \mid i \in 1) \sum_{j \in NnS} x_{P(i^\pi, \sigma^\pi)} + \sum_{j \in NnS} f\theta(x_S, x_j) \leq \sum_S \theta(x_i, x_j)g \quad (9)$$

We prove now that if θ satisfies (6), then (8) holds for all N, S, i^π and x . The top inequality in (6) implies $\theta(0, b) \leq 0$. Repeated applications of the bottom one give

$$\begin{aligned} \sum_S \theta(x_i, b) &\leq \theta(x_S, b) + (j \in S^0 \mid i \in 1) \sum_S b, \text{ and} \\ \sum_S \theta(x_i, b) &\leq \theta(x_S, b) \text{ if } x_S \leq b \end{aligned}$$

Applying the top inequality to $b = x_j$ for all $j \in P(i^\pi, \sigma^\pi)$, and the bottom one to x_j for all $j \in Nn(S \cup P(i^\pi, \sigma^\pi))$ gives the desired inequality (8).

Next we prove that (6) must hold if θ meets (8) for all problems and all merging. Consider $N = \{1, 2, 3\}, S = \{1, 2\}, i^\pi = 1$ and $x = (a_1, a_2, b)$ for arbitrary a_i, b in \mathbb{R}_+ . If $a_1 + a_2 < b$, $P(i^\pi, \sigma^\pi)$ is empty and (8) yields the top inequality in (6). Continuity of θ takes care of the case $a_1 + a_2 = b$. If $a_1 + a_2 > b$, $P(i^\pi, \sigma^\pi) = \{3\}$ and (8) gives the bottom inequality in (6).

4) Finally we consider Split-proofness, defined by the inequality (5), that we develop similarly for the θ -separable method. First we compute $(t_{\pi})_{T[i_{\pi}]}$ in (5). Set $T[i_{\pi}] = R$ and $|R| = r$, then label $e_i, i \in R$, in such a way that $e_1 \leq e_2 \leq \dots \leq e_r$. Thus $x_{i_{\pi}} = \sum_{i=1}^r e_i$. In the split problem (N_{π}, x_{π}) , the total wait of coalition R is $v(R, x_{\pi}) + \sum_{k=1}^r k \cdot x_{S_k}$ where S_k contains those agents in $N \setminus i_{\pi}$ ranked before k and after $k+1$ in σ_{π} . In particular for $j \in S_k, e_k \leq x_j \leq e_{k+1}$. We are ready to compute $(t_{\pi})_R$

$$\begin{aligned} y_R(N_{\pi}, x_{\pi}) &= v(R, x_{\pi}) + \sum_{k=1}^r k \cdot x_{S_k} + (t_{\pi})_R \\ &= v(R, x_{\pi}) + \sum_{j \in N \setminus i_{\pi}} \sum_{i \in R} \theta(x_i, x_j) \end{aligned}$$

Therefore the split-proofness inequality (5) writes as

$$\begin{aligned} x_{i_{\pi}} + \sum_{j \in N \setminus i_{\pi}} \theta(x_{i_{\pi}}, x_j) &\leq x_{i_{\pi}} + \sum_{k=1}^r x_{S_k} + (t_{\pi})_R \\ &= \sum_{i \in N \setminus i_{\pi}} \theta(x_{i_{\pi}}, x_i) + \sum_{k=1}^r (k-1) \cdot x_{S_k} + \sum_{i \in N \setminus i_{\pi}} \sum_{s=1}^r \theta(e_s, x_i) \end{aligned} \quad (10)$$

We show finally that (7) is true if and only if (9) holds for all N, i_{π}, T and x . The "if" statement follows easily from applying (9) to $N = \{1, 2\}, i_{\pi} = 2, T = \{3\}, x = (b, a_1 + a_2)$ and $x_{\pi} = (b, a_1, a_2)$. If $b \leq a_1, a_2$ we have $S_2 = \{1\}$ and we get the top inequality in (7). For other values of b , S_2 is empty and we get the rest of (7).

Before proving the "only if" statement, we notice a consequence of (7). Fix $k, a_k, k = 1, \dots, r$, such that $a_1 \leq a_2 \leq \dots \leq a_r$, and b . We have

$$b \leq a_k \Rightarrow \theta(a_1 + \dots + a_r, b) + (k-1) \cdot b \leq \sum_{s=1}^r \theta(a_s, b)$$

We omit the easy proof. Apply this inequality to $a_k = e_k$ and to $b = x_j$ for some agent j in S_k , we get

$$\theta(x_{i_{\pi}}, x_j) + (k-1) \cdot x_j \leq \sum_{s=1}^r \theta(e_s, x_j)$$

Summing up over all $j \in N \setminus i$ gives the desired inequality.

Proposition 2 shows that among separable scheduling methods, it is easy to ensure merge-proofness or split-proofness. The former requires θ to be something less than superadditive in its first variable; the latter requires θ to be something more than subadditive in its first variable. The two requirements are incompatible: this results from Theorem 1, or can be checked directly by comparing systems (6) and (7).

Two separable methods stand out for the simplicity of their definition and their multiple interpretations. Moreover, they are the backbone of the characterization of transferproof methods in the next section.

Definition 2 The S^+ and S^i separable methods are associated with θ^+ and θ^i respectively.

$$\theta^+(a, b) = \frac{1}{2}(a \wedge b); \quad \theta^i(a, b) = b \vee \frac{1}{2}(a - b) \text{ for all } a, b$$

The corresponding net waiting costs and transfers for a problem (N, x) with $|N| = n$ and $x_1 \cdot x_2 \cdot \dots \cdot x_n$ are:

$$y_i^+ = \frac{1}{2}x_{f1,i-1g} + (1 + \frac{n-i}{2})x_i \text{ and } t_i^+ = \frac{1}{2}((n-i)x_i - x_{f1,i-1g})$$

$$y_i^i = x_{f1,i-1g} + (\frac{i-3}{2})x_i + \frac{1}{2}x_{f_{i+1},ng} \text{ and } t_i^i = \frac{1}{2}(x_{f_{i+1},ng} - (i-1)x_i)$$

where we use the notation $x_{f_{i,j}g} = \sum_{k \in N \setminus i} x_k$.

The computation of the net waiting costs and transfers from θ^+ , θ^i and Definition 1 is straightforward.

The S^+ method divides equally the externality $x_i \wedge x_j$ between x_i and x_j . If $x_i < x_j$, agent i is served first and gets a "rebate" $\frac{1}{2}x_i$ from agent j . But with the S^i method, agent i gets a larger rebate $\frac{1}{2}x_j$.

Notice that for $|N| = 2$, S^i simply equalizes net costs $y_1^i = y_2^i = x_1 + \frac{1}{2}x_2$, a fairly reasonable compromise. But for larger sizes of N , the method S^i has several unappealing features.

Proposition 3

- i) The scheduling method S^+ is merge-proof. It also satisfies Monotonicity, Ranking, Stand Alone Bound and Zero Charge for Null Jobs.
- ii) The scheduling method S^i is split-proof. hence it violates MON, RKG, SAB, and ZCNJ.

Proof of Proposition 3

That S^+ meets the four properties MON, RKG, SAB and ZCNU is obvious, either by direct inspection of the formula for y_i^+ , or by invoking Proposition 2. Similarly, Proposition 1 and splitproofness imply that S^i violates all four properties; this fact can also be checked directly on the formula for y_i^i , or by invoking Proposition 2. In particular S^i has the following "anti-ranking" property: $x_i < x_j \Rightarrow y_i > y_j$.

Next one checks easily that the function θ^+ has the subadditivity properties (6), whereas θ^i has the superadditivity properties (7), and the proof is complete.

We conclude this section with a few alternative interpretations of S^+ and S^i .

Lemma 1 The profile of net costs selected by the method S^+ is the Shapley value of the optimistic Stand Alone cooperative game $S^+ : v(S, x)$ for all $S \subseteq N$. The profile selected by the method S^i is the Shapley value of the pessimistic stand alone game $S^i : w(S, x) = \sum_{j \in S} x_{N \setminus S} + v(S, x)$.

In the optimistic (resp. pessimistic) Stand Alone game, the total cost of a coalition S is its efficient cost when it is served before (resp. after) the complement coalition $N \setminus S$.

Proof of Lemma 1. The interpretation of S^+ as the Shapley value of the optimistic game is already in Curiel et al [2002]. For the sake of completeness, we give a proof here. Given $N, S \subseteq N$ and $i \in N \setminus S$, the marginal contribution of agent i to S is

$$v(S \cup \{i\}, x) - v(S, x) = x_i + \sum_{j \in S, j > i} x_i \wedge x_j$$

Therefore the (i, j) -externality $x_i \wedge x_j$ is charged to agent i if and only if j appears before i in the random ordering of N : this happens with probability .5, so the Shapley value awards precisely y_i^+ to agent i .

Next we check that y^i is the Shapley value of the game w . By additivity of the value this amounts to check that $y^i - y^+$ is the value of the game $\alpha = w - v$. Compute:

$$y_i^i - y_i^+ = \frac{1}{2}x_N - \frac{n}{2}x_i \text{ and } \alpha(S \cup \{i\}, x) - \alpha(S, x) = x_{N \setminus S \cup i} - sx_i$$

from which the desired conclusion follows easily.

Remark 2 Yet another interpretation of S^+ is by means of the serial cost sharing formula of Friedman and Moulin [1999]. Consider the scheduling

problem (N, x) as a cost sharing problem with the demand profile x and the cost function $C(x) = v(N, x)$. One checks easily that y^+ is the profile of cost shares under the serial cost sharing formula defined there. Finally, we note that under S^+ , the transfer t_i^+ to agent i does not depend upon the length of jobs longer than x_i ; whereas under S^i , t_i^i is independent of the length of jobs shorter than x_i . In combination with efficiency and equal treatment of equals, these properties are clearly characteristic. In the related scheduling model where all jobs are of equal length but agents differ by their linear waiting cost, Maniquet [2003] and Chun [2004] use similar independence properties to characterize respectively the analog of our S^+ and S^i scheduling methods.

7 Transfer of jobs and the main result

We consider a manipulation related to merging and splitting, yet more subtle because it involves a partial transfer of jobs. The number of agents remains constant during the transfer, therefore in this section we may assume $N = \mathbf{N}$.

Our main result (Theorem 2 below) characterizes the scheduling mechanisms robust against partial transfers of jobs involving only two agents, together with monetary transfers among possibly more agents. This restriction is crucial. In Section 8 we derive an impossibility result when transfers among three agents or more are feasible.

Given $N, x, i, j \in N$ and $\varepsilon > 0$, we call x^0 an ε -shrink of x by i, j if $x_i \geq x_j$ and $(x_i^0, x_j^0) = (x_i + \varepsilon, x_j - \varepsilon)$ or if $x_i \leq x_j$ and $(x_i^0, x_j^0) = (x_i - \varepsilon, x_j + \varepsilon)$. We call x^0 an ε -spread of x by i, j if $x_i \geq x_j$ and $(x_i^0, x_j^0) = (x_i - \varepsilon, x_j + \varepsilon)$ or if $x_i \leq x_j$ and $(x_i^0, x_j^0) = (x_i + \varepsilon, x_j - \varepsilon)$. Finally the notation $\Phi(\sigma; i, j)$ stands for the set of agents in N that ordering σ ranks between i and j .

We are now ready to define the two sides of the transfer-proofness axiom. Throughout these definitions we fix the set N of agents, $|N| \geq 3$.

Pairwise Shrink proofness: for all $S, S' \subseteq N$, $x, x^0 \in \mathbf{R}_+^N$, and ε -shrink x^0 of x by $i, j \in S$

$$y_S(N, x) \leq y_{S'}(N, x^0) + \varepsilon \quad (11)$$

Pairwise Spread-proofness: for all $S, S' \subseteq N$, $x, x^0 \in \mathbf{R}_+^N$, and ε -spread x^0 of x by $i, j \in S$

$$\sigma^0 = \mu(N, x^0) \Rightarrow y_S(N, x) \leq y_{S'}(N, x^0) + \varepsilon + x_{\Phi(\sigma^0; i, j)} \quad (12)$$

Definition 3 We call the mechanism μ pairwise transferproof (PTP) if it is pairwise shrinkproof and spreadproof.

Several comments on this definition are in order. Firstly, the PTP concept applies to scheduling mechanisms because the choice of $\sigma^0 = \mu(N, x^0)$ matters to the spreadproofness property (but not to that of pairwise shrinkproofness).

The second observation is that PTP rules out certain maneuvers by coalitions S of arbitrary size: although the partial transfer of jobs only concerns two agents, other agents in S are involved in a redistribution of money inside S .

Next we comment on the inequality defining shrinkproofness. The left-hand side is the total net cost of coalition S before the (job and cash) transfers. We claim that the right-hand side is its total net cost after the job transfer. Without loss of generality, suppose $i = 1, j = 2$ and $x_1^0 = x_1 + \varepsilon \cdot x_2$; $\varepsilon = x_2^0$. The real job x_1 will be completed whenever x_1^0 is served, and job x_2 when x_2^0 is served. If a reported job x_1^0 or x_2^0 is served after some agent $j, j \in \{1, 2\}$, so does the corresponding real job, and vice-versa. Thus the difference between the waiting time of the real jobs x_1, x_2 , and that of the reported jobs x_1^0, x_2^0 is $2x_1 + x_2 - (2x_1^0 + x_2^0) = -\varepsilon$. Hence inequality (10).

For instance, we check that the proportional mechanism is not pairwise shrinkproof. Let $N = \{1, 2, 3\}, x = (1, 6, 5)$ and $S = \{1, 2\}$ with $x_1^0 = 3, x_2^0 = 4$. Thus x^0 is a 2-shrink of x by 1, 2, involving no other agents. Compute

$$y_{12}(x) = \frac{7}{12} \cdot (10) > \frac{7}{12} \cdot (22) - 2 = y_{12}(x^0) - \varepsilon$$

Recall that this method is in fact merge-proof. We let the reader check that the egalitarian method also fails (10) for the following three-person example:

$$x = (1, 8, 2), x^0 = (4, 5, 2) \quad y_{12}(x) = 11\frac{2}{3}, y_{12}(x^0) = 14\frac{1}{3}$$

Finally we explain inequality (11). Suppose as before $i = 1, j = 2, x_1^0 = x_1 - \varepsilon, x_2^0 = x_2 + 5, x_1 \leq x_2$. After the report, the real job x_1 will not be completed when job x_1^0 is done, but only during the service of job x_2^0 . Thus the difference between the wait of the real jobs and that of the reported jobs is

$$2x_1 + x_2 + x_{\Phi(\sigma^0, 1, 2)} - (2x_1^0 + x_2^0) = \varepsilon + x_{\Phi(\sigma^0, 1, 2)}$$

If the set $\Phi(\sigma^0; 1, 2)$ is not empty, a spread from x to x^0 introduces the additional waiting time $x_{\Phi(\sigma^0; 1, 2)}$ to the reported waiting time of S at x^0 . Thus pairwise spread-proofness ends up being easier to meet than pairwise shrink-proofness. For instance, all three methods Shortest Job First, proportional and egalitarian are spread-proof. For an example where this property is violated, consider the following θ_i separable method:

$$\theta(a, b) = \frac{ab}{a+b} \text{ if } a \leq b; = \frac{b^2}{a+b} \text{ if } b < a$$

Set $N = \{1, 2, 3\}$ and $x = (1, 2, 3)$. Consider the ε_i spread by $\{1, 2\}$ to $x^0 = (1 - \varepsilon, 2 + \varepsilon, 3)$ with $0 < \varepsilon < 1$. Inequality (11) for $S = \{1, 2\}$ reads

$$y_{12}(x) = 4 + \theta(1, 3) + \theta(2, 3) \leq 4 + \theta(1 - \varepsilon, 3) + \theta(2 + \varepsilon, 3) = y_{12}(x^0) + \varepsilon$$

It is violated because $\theta(a, b)$ is strictly concave in a on $[0, b]$.

One last remark about our definition 3. We do not allow pairwise transfers exchanging the ordering of jobs 1 and 2, as when x_1, x_2 with $x_1 < x_2$ becomes x_1^0, x_2^0 with $x_1 > x_2^0$. This restriction is without any real loss of generality, because the deviating agents have every incentive to use efficiently the time slots allocated to their reported jobs. In the configuration above, the slot for x_2^0 will be used to complete job x_1 and start job x_2 . Therefore the shift from x to x^0 is equivalent to a shrink from (x_1, x_2) to (x_2^0, x_1^0) .

We are ready to state our main characterization result.

Theorem 2 Fix N with $|N| \geq 4$.

(i) Choose two continuous functions, $\alpha : \mathbb{R}_+ \rightarrow \mathbb{R}$ and $\gamma : \mathbb{R}_+ \rightarrow \mathbb{R}^N$ such that $\sum_{i \in N} \gamma_i(z) = 0$ for all z . The following equality defines a scheduling method y :

$$y(x) = \alpha(x_N) y^+(x) + (1 - \alpha(x_N)) y^i(x) + \gamma(x_N) \text{ for all } x \in \mathbb{R}_+^N$$

where y^+, y^i correspond to S^+ and S^i as in Definition 2. Any corresponding mechanism is efficient, continuous, and pairwise transfer-proof.

(ii) Conversely, if a mechanism μ is efficient, continuous and pairwise transfer-proof, the associated method y takes the above form.

The PTP axiom, almost single handedly, captures a fairly small family of scheduling methods/mechanisms. This family contains the affine combinations of S^+, S^i to which we can add some "constant" γ , where the coefficients of the affine combination and the "constant" depend only upon x_N .

Corollary 1 to Theorem 2 Consider a mechanism μ defined as in statement *i*) by the functions α and γ :

- i) μ treats equals equally if and only if $\gamma(z) = 0$ for all z .
- ii) μ is scale invariant if and only if α is constant in \mathbb{R} and γ is homogeneous of degree 1.

$$\text{Scale Invariance (SI): } y(\lambda x) = \lambda y(x) \text{ for all } \lambda > 0, x \in \mathbb{R}_+^N$$

If we combine the mild properties CONT, ETE and SI with efficiency and PTP, Corollary 1 tells us that we are left with the one-dimensional line of methods joining S^+ and S^i . These methods are all separable, with corresponding function θ :

$$\theta(a, b) = \frac{1}{2}(a \wedge b) + \frac{(1 - \alpha)}{2}(a \vee b) \text{ for all } a, b \geq 0$$

The parameter α is any real number. The method S^+ obtains for $\alpha = 1$ and S^i for $\alpha = 0$.

Corollary 2 to Theorem 2 Consider a mechanism μ defined as in statement *i*) by the functions α and γ :

- i) μ is merge-proof if and only if $\alpha(z) \leq 1$ for all z .
- ii) μ is split-proof if and only if $\alpha(z) \geq 0$ for all z .

This establishes the polar role of S^+ and S^i within the family described in Theorem 2, or its Corollary 1: they stand out respectively for their robustness to merging and splitting maneuvers.

If Theorem 2 and its Corollaries 1 and 2 give a symmetrical role to S^+ and S^i , this symmetry is destroyed as soon as we introduce the normative requirements of Section 4. Not surprisingly, these properties point toward the method S^+ .

Corollary 3 to Theorem 2 The mechanism S^+ is characterized by the combination of efficiency, Continuity, Pairwise Transfer-proofness and either Zero Change for Null Jobs, or the Stand Alone bound.

Remark 3 Two additional properties can be used to single out the S^+ method. They both place an upper bound on individual net waiting costs, which is the familiar idea of a lower bound on individual welfare. The pessimistic stand alone bound for agent i is simply $y_i + w(\text{fig}, x) = x_N$. Both S^+ and S^i meet this bound. The unanimity bound for agent i is $y_i + \frac{n+1}{2}x_i$. It is this agent's net cost in a hypothetical problem (N, θ) where all jobs are

of length x_i . Clearly S^+ meets this bound, whereas S^i violates it, even for $jNj = 2$.

Now S^+ is characterized by the combination of e¢ciency, CONT, PTP and either {the pessimistic stand alone bound plus the unanimity bound}, or {the pessimistic stand alone bound and merge-proofness}. The proof is in the Appendix.

8 Transfers among three or more agents

The two benchmark methods S^+ and S^i , and their a¢ne combinations, are not vulnerable to bilateral partial transfers of jobs, but trilateral problems can be a problem.

A simple example with $N = \{1, 2, 3, 4\}$ illustrates this important point. In the problem $x = (1, 1, 8, 3)$ coalition $T = \{1, 2, 3\}$ rearranges its three jobs as $x^0 = (2, 4, 4, 3)$. The actual wait of everyone in T is the same at x and at the reported x^0 : in the latter, the slot $x_1^0 = 2$ is used to complete jobs x_1 and x_2 , whereas the slots $x_1^0 = x_3^0 = 4$ are devoted to job x_3 . We check that under both S^+ and S^i , the total tax on T decreases from x to x^0 . Equivalently, the tax on agent 4 increases. By Definition 2

$$\begin{aligned} \text{under } S^+ \text{ at } x : t_4^+ &= \frac{1}{2}; \text{ at } x^0 : (t_4^+)^0 = 2 \\ \text{under } S^i \text{ at } x : t_4^i &= 1; \text{ at } x^0 : (t_4^i)^0 = \frac{5}{2} \end{aligned}$$

Now any mechanism described in statement *i*) of Theorem 2 is vulnerable to the same trilateral transfer: indeed $\alpha(x_N)$ and $\gamma(x_N)$ do not change from x to x^0 . therefore we have proved

Corollary 4 to Theorem 2 If $jNj \geq 4$, any e¢cient and continuous mechanism is vulnerable to job transfers involving three or more agents. For the sake of brevity, we do not give a formal definition of profitable transfers of jobs involving 3 or more agents. The definition is notationally cumbersome, and brings no additional intuition beyond that provided by the numerical example above. Notice that the shift from x to x^0 may be interpreted as the combination of merging jobs x_1, x_2 and splitting job x_3 . This suggests that our first negative result, theorem 1, is closely related to Corollary 4.

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9 Appendix

9.1 Theorem 1

We fix $N, |N| \geq 4$ and a mechanism μ satisfying MPF and SPF, as well as either CONT and ETE, and we derive a contradiction. Recall that MPF implies efficiency.

Step 1 A limited symmetry property

Fix $N, |N| \geq 2$, and two agents $1, 2 \in N$. Fix any $x \in \mathbb{R}^N$ such that $x_1 = x_2 = a > 0$. We write x_{-1}, x_{-2} for its projection on $N \setminus \{1\}$ and $N \setminus \{2\}$ respectively, and define $z^i \in \mathbb{R}^N$ by $z_2^1 = 0, z_{-2}^1 = x_{-2}$, and $z_1^2 = 0, z_{-1}^2 = x_{-1}$. We claim

$$y_{12}(N, z^1) = y_{12}(N, z^2) = y_1(N \setminus \{2\}, x_{-2}) = y_2(N \setminus \{1\}, x_{-1})$$

In the merging of 1 and 2 in z^1 to agent 1 in x_{-2} , merge-proofness (4) gives

$$y_{12}(z^1) \cdot a + x_{P(1, \sigma^x)} + t_1^x(x_{-2}) = y_1(x_{-2})$$

because $S^0 = \{1\}$. In the split of 1 in x_{-2} to agents 1, 2 in z^1 , split-proofness implies

$$y_1(x_{-2}) \cdot a + x_{P(1, \sigma^x)} + (t_x)_{12}(z^1) = y_{12}(z^1)$$

because 1 is the only agent with a positive job in $S = \{1, 2\}$. Thus we get $y_{12}(z^0) = y_1(x_{-2})$. Consider similarly the merging of 1, 2 in z^1 to agent 2 in x_{-1} , and the split of 2 in x_{-1} to agents 1, 2 in z^1 : we get $y_{12}(z^1) = y_2(x_{-1})$. Exchanging the roles of 1, 2 gives the remaining equalities in the claim.

Step 2 The case of two agents problems

Fix a vector $(a, b) \in \mathbb{R}_+^2$ s.t. $0 < a < b$, and an arbitrary triple $N = \{1, 2, 3\}$ in \mathcal{N} . From Step 1 applied to $x = (a, a, b)$ we get

$$y_1(f13g, (a, b)) = y_2(f23g, (a, b))$$

Set $y_i(f_i, jg, (a, b)) = u_i(ij)$ and $y_j(f_i, jg, (a, b)) = v_j(ij)$. We have proven $u_i(ij) = u_k(kj)$ for i, j, k all distinct. By exchanging the role of a and b we get similarly $v_j(ij) = v_k(ik)$. Efficiency implies $u_i(ij) + v_j(ij) = 2a + b$, therefore $v_j(kj) = v_j(ij)$ and $u_i(ij) = u_i(ik)$. We can now set $u_i = u_i(ij)$ and $v_i = v_i(ji)$ for all $j \in N \setminus i$. Efficiency shows that $u_i + v_j$ does not depend on the pair (i, j) in N , therefore u_i and v_i are both independent of $i \in N$. We define a function $f(a, b)$ as follows

$$y_i(f_i, jg, (a, b)) = a + f(a, b); \quad y_j(f_i, jg, (a, b)) = a + b - f(a, b)$$

keeping in mind that the pair (i, j) is arbitrary.

Step 3

We now compute explicitly the vector of transfers for a three-person problem $N = \{1, 2, 3\}$ and $x = (a, b, c)$ with $0 < a < b$ and $a + b < c$. This vector is independent of the choice of a triple in N .

Consider the merging of 2,3 in x to 2 in $x^\pi = (a, b + c)$. As the total physical wait of agents $\{2, 3\}$ is the same before and after merging, MPF implies

$$t_{23}(N, x) + t_2(f_1, 2g, x^\pi) = f(a, b + c)$$

When splitting agent 2 in x^π to $\{2, 3\}$ in x , the physical wait of agent 2 is similarly constant, thus SPF implies the opposite inequality and we get $t_{23}(x) = f(a, b + c)$.

A similar argument about the merging of 1,2 in x to 1 in $x_\pi = (a + b, c)$, and the symmetrical split of 1 in x_π to 1,2 in x gives: $t_{12}(x) = f(a + b, c)$. Our choice of a, b, c guarantees that the actual wait of 1,2 is constant in the merging, and that of 1 is constant in the split. Because $t_{123} = 0$ the vector $t(x)$ is now computed explicitly. It is convenient to change f to the function g defined by $f(\alpha, \beta) = g(\alpha, \alpha + \beta)$ for all $\alpha, \beta \geq 0$. We have:

$$t(N, x) = (g(a, d), g(a + b, d) - g(a, d), g(a, d), g(a + b, d)) \quad (13)$$

for all a, b, d such that $0 < a < b$, and $2(a + b) < d$.

Next we invoke ETE or CONT at a triple (a, b, d) where $a = b$. ETE implies $t_1 = t_2 + a$. On the other hand CONT implies that f is continuous in both variables, and so is g . For a small positive ε , the net waiting cost of agent 1 at $(a - \varepsilon, a, d)$ is $g(a - \varepsilon, d) + a - \varepsilon$, and it is $g(2a + \varepsilon, d) - g(a, d) + 2a + \varepsilon$ at $(a + \varepsilon, a, d)$. By continuity, $t_1 = t_2 + a$ follows, namely

$$g(2a, d) = 2g(a, d) \text{ ; } a \text{ for all } a, 0 < a \leq \frac{d}{2} \quad (14)$$

The last step of the proof extends the above argument to four agents problems like $x = (a, b, c, d \text{ ; } (a+b+c))$ with $0 < a \leq b \leq c$ and $2(a+b+c) \leq d$. Looking to the merging of 3,4 and its reverse split, we deduce $t_{34}(x) = g(a+b, d)$ from (12). From the merging of f2, 3, 4g and the reverse split, we get $t_{234}(x) = g(a, d)$, and ...nally from the merging of 1,2,3 and its reverse split we have $t_{123}(x) = g(a+b+c, d)$. Gathering our results

$$t(x) = (g(a, d), g(a+b, d) \text{ ; } g(a, d), g(a+b+c, d) \text{ ; } g(a+b, d) \text{ ; } g(a+b+c, d))$$

At a pro...le x where $b = c$, ETE or the same continuity argument as above gives

$$g(a+b, d) \text{ ; } g(a, d) = g(a+2b, d) \text{ ; } g(a+b, d) + b \quad (15)$$

We derive ...nally a contradiction between (14) and (13). Taking $a = b$ in (14), and omitting d for simplicity, we get

$$g(3a) = 2g(2a) \text{ ; } g(a) \text{ ; } a = 3g(a) \text{ ; } 3a$$

Taking $b = 2a$ gives similarly $g(5a) = 5g(a) \text{ ; } 8a$. Finally taking $a = 2x, b = 3x$ gives

$$g(8x) = 2g(5x) \text{ ; } g(2x) \text{ ; } 3x = 10g(x) \text{ ; } g(2x) \text{ ; } 19x$$

and a contradiction with (13) follows easily.

9.2 Theorem 2

9.2.1 Proof of Statement i

Consider the method associated with the functions α and γ . As a spread or a shrink leaves the sum x_N , and therefore $\gamma(x_N)$, unchanged, we can simply ignore γ while checking PTP. Recall that y^+ and y^i are separable with associated functions θ^+ and θ^i . Thus $y = \alpha y^+ + (1 - \alpha)y^i$ can be written

$$y_i(N, x) = x_i + \sum_{j \in N, j \neq i} \theta(x_i, x_j; x_N)$$

where $\theta(a, b; z) = \alpha(z)\theta^+(a, b) + (1 - \alpha(z))\theta^-(a, b) = \frac{1}{2}(a \wedge b) + \frac{1}{2}(\alpha(z) - 1)(a - b)$.

For fixed b and z , the function $a \mapsto \theta(a, b; z)$ is linear before b and linear after b , and its slope drops by $\frac{1}{2}$ at b . In particular, this function is concave. Thus all we need to show is that any mechanism coming from method y , meets PTP.

Consider first S, x, x^0 and ε as in the premises of (10). Assume without loss of generality $x_1 < x_2$ and that agent 2 transfers ε of his job to agent 1. As $x_N = x_N^0$, we omit x_N in $\theta(x_i, x_j; x_N)$ and compute the total net cost of S before and after the shrink:

$$\begin{aligned} y_S(x) &= v(S, x) + \sum_{N \setminus S} \theta(x_1, x_j) + \theta(x_2, x_j) \\ y_S(x^0) &= v(S, x^0) + \sum_{N \setminus S} \theta(x_1^0, x_j) + \theta(x_2^0, x_j) \\ v(S, x^0) - v(S, x) &= (2x_1^0 + x_2^0) - (2x_1 + x_2) + \sum_{k \in S \setminus \{1, 2\}} p_k \end{aligned}$$

where the term $p_k = x_1^0 \wedge x_k + x_2^0 \wedge x_k - x_1 \wedge x_k - x_2 \wedge x_k$ is nonnegative because $a \mapsto a \wedge x_k$ is concave. Therefore $v(S, x^0) - v(S, x) \leq \varepsilon$. The same concavity argument shows $\theta(x_1^0, x_j) + \theta(x_2^0, x_j) \leq \theta(x_1, x_j) + \theta(x_2, x_j)$, and the proof of (10) is complete.

Next we consider a spread, namely S, x, x^0 and ε as in the premises of (11) with $x_1 \leq x_2$ and agent 1 transferring of her job to agent 2. With the same notation p_k as above, we get: $v(S, x^0) - v(S, x) = -\varepsilon + \sum_{S \setminus \{1, 2\}} p_k$. Setting $q_j = \theta(x_1^0, x_j) + \theta(x_2^0, x_j) - \theta(x_1, x_j) - \theta(x_2, x_j)$, we now have

$$y_S(N, x^0) + \varepsilon - y_S(N, x) = \sum_{S \setminus \{1, 2\}} p_k + \sum_{N \setminus S} q_j \quad (16)$$

where the concavity argument shows this time $p_k \leq 0$ and $q_j \geq 0$. Check first that for any agent $i \notin \Phi(\sigma^0; 1, 2)$, we have $p_i = 0$ if $i \in S \setminus \{1, 2\}$ and $q_i = 0$ if $i \in N \setminus S$. This is clear because the functions $a \mapsto a \wedge x_i$ and $a \mapsto \theta(a, x_i)$ are linear on $[0, x_i]$ and on $[x_i, +\infty[$. Next we pick $i \in \Phi(\sigma^0; 1, 2)$ and suppose first $i \in S \setminus \{1, 2\}$. We have

$$x_1^0 \leq x_i \leq x_2^0 \Rightarrow p_i + x_i = x_1^0 \wedge x_i - x_1 \wedge x_i - x_2 \wedge x_i + 2x_i \leq 0$$

Finally consider $i \in \Phi(\sigma^0; 1, 2) \setminus NnS$. If we show $q_i + x_i \geq 0$, the desired inequality (11) will follow from (15). Recall that on the interval $[x_1^0, x_2^0]$, the function $a \mapsto \theta(a, x_i)$ has 2 linear pieces connecting at x_i and such that the slope drops by $\frac{1}{2}$ at x_j . Therefore

$$q_i = (\theta(x_2^0, x_i) - \theta(x_2, x_i)) - (\theta(x_1, x_i) - \theta(x_1^0, x_i)) \cdot \frac{1}{2}(x_1 - x_1^0)$$

The inequality $q_i + x_i \geq 0$ follows if $x_i \geq x_1$. If $x_1^0 \leq x_i \leq x_1$, compute $q_i = \frac{x_1^0 - x_i}{2}$, ensuring $q_i + x_i \geq 0$. This concludes the proof of statement i.

9.2.2 Proof of Statement ii

We fix N , and an efficient mechanism μ meeting CONT and PTP.

Step 1

For all nonempty and proper subset S of N , we write $H(S) = \{x \in \mathbb{R}_+^N \mid x_i < x_j \text{ for } i \in NnS\}$. We prove the existence of a function $g^S(a, b)$ such that

$$g^S(a, b) \text{ is defined for } a, b \geq 0 \text{ such that } \frac{a}{|S|} < \frac{b}{|N|} \quad (17)$$

$$g^S(x_S, x_N) = t_S(x) \text{ for all } x \in H(S)$$

where $t(x)$ is the monetary transfer selected by μ . For any $x \in H(S)$, efficiency of μ implies that $\sigma(x)$ ranks S ahead of NnS , therefore $y_S(x) = v(S, x) + t_S(x)$. Given $x \in H(S)$, we write x^\square , also in $H(x)$, the vector $x_i^\square = \frac{x_S}{|S|}$ if $i \in S$, $x_i^\square = x_i$ if $i \in NnS$. Our first step toward proving (16) is to show $t_S(x) = t_S(x^\square)$.

We call two agents i, j adjacent at x if $\Phi(\sigma(x); i, j) = ?$. Given x and $i, j \in S$, adjacent at x , consider x^0 obtained from x by averaging x_i and x_j : $x_i^0 = x_j^0 = \frac{1}{2}(x_i + x_j)$, $x_k^0 = x_k$ otherwise. Thus x^0 is a shrink of x , and x a spread of x^0 , and PTP implies $y_S(x) = y_S(x^0) \pm \varepsilon$, where $\varepsilon = \frac{1}{2}|x_i - x_j|$.

Now suppose $x \in H(S)$ as well. Recall $y_S(x) = v(S, x) + t_S(x)$. Note that x^0 is in $H(S)$ as well, and that $v(S, x^0) = v(S, x) \pm \varepsilon$, because $[x_i, x_j]$ contains no x_k , $k \notin i, j$. We conclude $t_S(x) = t_S(x^0)$.

For any $x \in H(S)$ such that $x \notin x^\square$, we can find two agents $i, j \in S$, adjacent at x , and average x_i and x_j without changing $t_S(x)$. Thus we construct a sequence $x^\pm = x, x^1, x^2, \dots$, by averaging at each step some pair x_i, x_j where

i, j are adjacent at x . This sequence either stops at x^π or converges to x^π . By construction

$$y_S(x^p) = v(S, x^p) + t_S(x) \text{ for } p = 0, 1, 2, \dots$$

By continuity of y_S and of $v(S, \cdot)$, we deduce $t_S(x) = t_S(x^\pi)$ as announced.

A symmetrical construction, starting from any $x \in H(S)$, and successively averaging x_i, x_j for some $i, j \in N \setminus S$ adjacent at x , delivers $t_{N \setminus S}(x) = t_{N \setminus S}(x^\pi)$ where $(x^\pi)_i = \frac{x_{N \setminus S}}{|N \setminus S|}$ if $i \in N \setminus S$ and $(x^\pi)_i = x_i$ if $i \in S$. Combining this with $t_S(x) = t_S(x^\pi)$, and $t_S + t_{N \setminus S} = 0$, we see that $t_S(x)$ only depends upon x_S and $x_{N \setminus S}$, and can be written as in (16) for some function g^S . Finally $x \in H(S)$ implies $\frac{x_S}{|S|} < \frac{x_N}{|N|}$ and the proof of Step 1 is complete.

Step 2

We observe first that each function g^S is continuous on its domain. For each $a, b \geq 0$ such that $\frac{a}{|S|} < \frac{b}{|N|}$, we define $x(a, b) = z$ by $z_i = \frac{a}{|S|}$ for $i \in S$, and $z_i = \frac{b|S|}{|N \setminus S|}$ for $i \in N \setminus S$. By Step 1

$$g^S(a, b) = t_S(x(a, b)) = y_S(x(a, b)) - v(S, x(a, b))$$

and the claim follows by CONT. Next we apply continuity again at those profiles where two coordinates are equal, and derive a functional equation ((17) below) linking the different functions g^S .

In the rest of the proof we use the simplified notation $S \ll i, j = S \cup \{i, j\}$, etc...

Fix S , nonempty, and two agents $i, j \in N \setminus S$. Choose also any three a, b, c such that $0 < a < b < c$. We construct x and, for ε small enough, $x(\varepsilon)$ as follows:

$$x_k = a \text{ if } k \in S; x_i = x_j = b; x_k = c \text{ if } k \in N \setminus \hat{A}Si j$$

$$x(\varepsilon) = x + \varepsilon(e^i - e^j) \text{ where } e^i \text{ is the } i\text{-th unit vector in } \mathbb{R}^N$$

For ε small enough and positive, any efficient ordering of N ranks S before j, j before i , and i before the rest. For ε small and negative, the order is $S \hat{A} i \hat{A} j \hat{A} N \setminus Si j$. From Step 1, we have

$$y_i(x(\varepsilon)) = x_{Si j} + t_i(x(\varepsilon)) = x_{Si j} + g^{Si j}(x_{Si j}, x_N) - g^{Sj}(x_{Sj} - \varepsilon, x_N) \text{ for } \varepsilon > 0$$

$$y_i(x(\varepsilon)) = x_{Si} - \varepsilon + g^{Si}(x_{Si} - \varepsilon, x_N) - g^S(x_S, x_N) \text{ for } \varepsilon < 0$$

By continuity of y_i and of g^T , for all T , we deduce

$$b + g^{Sij}(sa + 2b, d) - g^{Sj}(sa + b, d) = g^{Si}(sa + b, d) - g^S(sa, d) \quad (18)$$

where $s = jSj$, $n = jNj$ and $d = sa + 2b + (n - s - 2)c$. As our choice of c is only limited by $0 < a < b < c$, if $Sij \notin N$ equation (17) holds for all a, b, d such that $0 < a < b$ and $\frac{sa+2b}{s+2} < \frac{d}{n}$. In the case $Sij = N$, we have $g^N \equiv 0$ and (17) holds for $0 < a < b$ with $d = sa + 2b$.

Finally the continuity argument applies also to the case $S = ?$, $a = 0$. Thus (17) holds in this case as well, provided $0 < b < d$, and with the convention $g^? = 0$.

Step 3

We derive a first consequence of (17)

$$g^S(sb, d) = \sum_S g^i(b, d) - \frac{s(s-1)}{2}b \text{ for all } ? \notin S \notin N, \text{ and all } 0 < b < \frac{d}{n} \quad (19)$$

Equation (17) for $S = ?$, $a = 0$, gives (18) for $S = ij$. Apply (17) next to $S = k$ and $a < b$, $\frac{1}{3}(a + 2b) < \frac{d}{n}$:

$$g^{ijk}(a + 2b, d) = (g^{ki} + g^{kj})(a + b, d) - g^k(a, d) - b \quad (20)$$

Fix d , let a, b converge to $a^0 = \frac{1}{3}(a + 2b)$, and use (18) for $S = ki$ and $S = kj$: we obtain (18) for $S = kij$. An easy induction argument, omitted for brevity, concludes Step 3.

Step 4

We prove that each function $g^i(a, d)$ is affine in a , and its slope is independent of $i \in N$. The assumption $jNj \geq 4$ plays a key role in this step, and in this one only.

Develop (19) using (18) for $S = ki$ and $S = kj$. We get

$$(g^i + g^j + g^k)\left(\frac{a + 2b}{3}, d\right) = 2g^k\left(\frac{a + b}{2}, d\right) + (g^i + g^j)\left(\frac{a + b}{2}, d\right) - g^k(a, d)$$

for $0 < a < b$ and $\frac{a+2b}{3} < \frac{d}{n}$. As the choice of i, j, k in N is arbitrary, the term $g^k(\frac{a+b}{2}, d) \dot{=} g^k(a, d)$ is independent of $k \in N$. Set it equal to $h(a, b, d)$ so the equation above becomes

$$\sum_{\omega=i,j,k} g^\omega(\frac{a+2b}{3}, d) \dot{=} g^\omega(\frac{a+b}{2}, d) = h(a, b, d)$$

As $|N| \geq 4$, and i, j, k are arbitrary, this implies for all i $g^i(\frac{a+2b}{3}, d) \dot{=} g^i(\frac{a+b}{2}, d) = \frac{1}{3}h(a, b, d)$. Thus, for fixed d , every function g^i meets the equation

$$g(\frac{a+b}{2}, d) = \frac{1}{4}g(a, d) + \frac{3}{4}g(\frac{a+2b}{3}, d)$$

Changing variables to $a^0 = \frac{a+2b}{3}$ we get

$$g(\frac{1}{4}a + \frac{3}{4}a^0) = \frac{1}{4}g(a) + \frac{3}{4}g(a^0) \text{ for all } 0 < a < a^0 < \frac{d}{n}$$

where we omit d for simplicity. This is a simple variant of the classic Cauchy equation (see Aczel [1970]). As $g(\cdot, d)$ is continuous on the interval $[0, d]$, it must be affine, namely $g(a, d) = \lambda(d)a + \beta(d)$. Back to the functions g^i , recall that $g^i(\frac{a+b}{2}) \dot{=} g^i(a)$ is independent of i : thus the slope $\lambda(d)$ is the same for all i and we conclude

$$g^i(a, d) = \lambda(d)a + \beta_i(d) \text{ for all } 0 < a < \frac{d}{n} \quad (21)$$

Step 5 End of proof

We obtain one more equation connecting $\lambda, \beta_i, i \in N$, by applying (17) to $S = N \setminus i, a, b, 0 < a < b$ and $d = (n-2)a + 2b$ (as discussed at the end of Step 2). Applying (18) and (20) we get

$$b = (g^{N \setminus i} + g^{N \setminus j})((n-2)a + b, d) \dot{=} g^{N \setminus j}((n-2)a, d)$$

$$\text{where } g^{N \setminus i}((n-2)a + b, d) = (\lambda(d) \dot{=} \frac{n-2}{2})((n-2)a + b) + \beta_{N \setminus i}(d),$$

$$g^{N \setminus j}((n-2)a, d) = (\lambda(d) \dot{=} \frac{n-3}{2})(n-2)a + \beta_{N \setminus j}(d)$$

and we omit the similar formula for $g^{N \setminus j}$. Upon replacing and rearranging:

$$b = \beta_N(d) + (\lambda(d) \dot{=} \frac{n-2}{2})((n-2)a + 2b) \dot{=} \frac{1}{2}(n-2)a \dot{=} \beta_N(d) = \frac{n-1}{2}d \dot{=} \lambda(d)d$$

Now we set

$$\alpha = \frac{2\lambda + 1}{n} \quad \lambda = \frac{n\alpha - 1}{2}; \text{ and } \beta_i = \frac{1}{n} \left(\frac{n-1}{2} - \lambda \right) d + \gamma_i$$

where α and $\gamma_i, i \in N$, depend on d . From the continuity of g^i in a, d follows that of β_i and of λ in d , hence of α and γ_i in d . Moreover $\gamma_N \rightarrow 0$ by construction. We compute now, with the help of (18), g^i and g^S in terms of α and γ_i :

$$g^i(a, d) = \lambda a + \beta_i = \frac{n\alpha - 1}{2} a + \frac{1 - \alpha}{2} d + \gamma_i$$

$$g^S(a, d) = \frac{n\alpha - s}{2} a + \frac{1 - \alpha}{2} sd + \gamma_S = \alpha \frac{(n - s)}{2} a + (1 - \alpha) \frac{s(d - a)}{2} + \gamma_S$$

For our two basic methods y^+ and y^i , it is easy from Definition 2 to compute $t_S^+(x), t_S^i(x)$ whenever $x \in H(S)$:

$$t_S^+(x) = \frac{1}{2}(n - s) \mathbb{I}_{x_S}; \quad t_S^i(x) = \frac{1}{2}s \mathbb{I}_{x_{N \setminus S}}$$

Compare with the sum of transfers to $t_S(x) = g^S(x_S, x_N)$ in our mechanism μ :

$$t_S(x) = \alpha(x_N) \mathbb{I}_{t_S^+(x)} + (1 - \alpha(x_N)) \mathbb{I}_{t_S^i(x)} + \gamma_S(x_N) \text{ for all } x \in H(S)$$

Recall, for any efficient method, any S and any $x \in H(S)$, the equation $y_S(x) = v(S, x) + t_S(x)$. We have just proven that the method y associated with μ , and the method $\tilde{y} = \alpha y^+ + (1 - \alpha)y^i + \gamma$ have $\tilde{y}_S(x) = y_S(x)$ for all S and $x \in H(S)$. Now if all coordinates of x are different, this forces $y(x) = \tilde{y}(x)$. By continuity the equality holds everywhere on \mathbb{R}_+^N . This concludes the proof of Theorem 2.

9.3 Corollaries of Theorem 2

9.3.1 Corollary 1

Statement i is obvious as y^+, y^i treat equals equally. For statement ii , the "if" part is obvious. To prove "only if," consider $x = d \mathbb{I}_{e^i}$, where e^i is, as before, the i -th coordinate vector. Compute

$$y^+(x) = de^i; y^i(x) = \frac{d}{2}(1, \dots, 1) + \frac{(n-i-1)}{2}de^i$$

Scale Invariance implies $y(x) = dy(e^i)$. Taking the j -th coordinate of this equation for $j \neq i$, gives

$$(1 - \alpha(d))\frac{d}{2} + \gamma_j(d) = \frac{1 - \alpha(1)}{2} + d\gamma_j(1)$$

As j varies in N and $\gamma_N \nearrow 0$, this implies $\gamma_j(d) = d\gamma_j(1)$ for all j and $\alpha(d) = \alpha(1)$, as claimed.

9.3.2 Corollary 2

Clearly the component $\gamma(x_N)$ in $y(x)$ plays no role in the properties of merge-proofness and splitproofness, so we can assume $\gamma \nearrow 0$ without loss of generality. Observe that the method $\alpha \sharp y^+ + (1 - \alpha) \sharp y^i$ behaves essentially like a separable method with respect to the function

$$\theta^\alpha(a, b, d) = \frac{1}{2}(a \wedge b) + \frac{(1 - \alpha(d))}{2}(a \vee b)$$

That is to say, the net cost $y_i(N, x)$ is computed as $y_i(N, x) = x_i + \sum_{N \setminus i} \theta^\alpha(x_i, x_j, x_N)$ for all N, i and x . We can then mimick the proof of Proposition 2: any mechanism with method $\alpha(x_N) \sharp y^+ + (1 - \alpha(x_N)) \sharp y^i$ is mergeproof if and only if, for any fixed d , the function $\theta^\alpha(\cdot, \cdot, d)$ meets the system (6); any such mechanism is split-proof if and only if $\theta^\alpha(\cdot, \cdot, d)$ meets system (7).

One consequence of MPF is that $a \mapsto \theta(a, b, d)$ is superadditive on $[0, b]$. In particular

$$2\theta\left(\frac{b}{2}\right) \geq \theta(b, b) \quad \left(\right) \quad \frac{b}{2} + (1 - \alpha)\frac{b}{2} \geq \frac{b}{2} \Rightarrow \alpha \leq 1$$

Conversely, θ^α meets (6) whenever $\alpha \leq 1$. Indeed $a \mapsto \theta^\alpha(a, b, d)$ has slope $\frac{\alpha}{2} \leq \frac{1}{2}$ on $[0, b]$ with $\theta(b) = \frac{b}{2}$, therefore $\theta(0) \geq 0$. On $[0, b]$ we have $\theta(a_1 + a_2) \geq \theta(a_1) + \theta(a_2) - \theta(0)$, and the top inequality in (6) follows. The bottom one is equally easy.

A consequence of SPF is $\theta(2b, b, d) + b \geq 2\theta(b, b, d)$, from the top inequality in (7). This amounts to $\theta(2b, b, d) \geq 0 \Rightarrow \frac{1}{2}b + \frac{(1-\alpha)}{2}b \geq 0 \quad \left(\right) \quad \alpha \leq 0$. Checking that, conversely, θ^α meets (7) whenever $\alpha \leq 0$ is routine and omitted.

9.3.3 Corollary 3

Choose α, γ as in the statement of Theorem 2. When does the corresponding method meet ZCNJ? For all x , all $i \in N$, $x_i = 0$ implies $y_i^+(x) = 0$ and $y_i^-(x) = \frac{1}{2}x_N$. Therefore ZCNJ implies $\frac{1}{2}(1 - \alpha(d))d + \gamma_i(d) = 0$ for all i and all $d \geq 0$. From this, $\gamma \leq 0$ and $\alpha(d) = 1$ for all d follow at once.

Suppose next that the method associated with α, γ meets SAB. Apply this property first for x, i such that $x_i = 0$. We get $\frac{1}{2}(1 - \alpha)d + \gamma_i \leq 0$. Summing over i gives $\alpha(d) \leq 1$ for all d . Apply next SAB to $x = de^i$ and to agent i

$$\begin{aligned} d &\leq \alpha y_i^+(x) + (1 - \alpha) y_i^-(x) = \alpha d + (1 - \alpha) \frac{3 - n}{2} d + \gamma_i \quad (22) \\ () &\quad \frac{n - 1}{2} (\alpha - 1) d + \gamma_i \leq 0 \end{aligned}$$

Summing up over i yields $\alpha(d) \leq 1$. Thus $\alpha \leq 1$ and the inequality above gives $\gamma \leq 0$ as well.

9.3.4 Remark 3

The pessimistic stand alone bound applied to $x = de^i$ gives the inequality opposed to (21) above, hence $\alpha(d) \leq 1$ for all d . Mergeproofness, on the other hand, amounts to $\alpha(d) \geq 1$ for all d .

Next apply the unanimity bound to any x, i such that $x_i = 0$. We get $\frac{1}{2}(1 - \alpha)d + \gamma_i \leq 0$, hence $\alpha(d) \leq 1$ by summing over i . Thus the combination of the bounds $y_i \leq x_N$ and $y_i \leq \frac{n+1}{2} x_i$ captures, again, the method S^+ .

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