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> **Optimal Mechanisms with
Finite Agent Types**

by

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LEADING IN THOUGHT AND ACTION

Optimal Mechanisms with Finite Agent Types

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Abstract

Myerson's seminal paper on mechanism design invokes differential reasoning in deriving his classical results. The arguments used do not readily lend themselves to parallel derivations for finite type spaces. Because of this difference, and the transparency and intellectual dominance of Myerson's work, relatively few papers have looked at the case with finite agent types beyond the two-type case. This paper provides that parallel argument in the regular case, for which Myerson's results are most often invoked. The results here for finite type spaces can be read alongside Myerson's results with no intuitive leaps. A minor error in Myerson's presentation, irrelevant with symmetrical agents, is corrected, allowing the results to handle the symmetrical and asymmetrical cases in one unified way.

The number of papers and presentations on optimal mechanism design with finite agent types is far fewer than those for continuous type spaces. This probably follows from a difficulty in replicating the line of argument in Myerson's (1981) seminal paper, which invokes differential reasoning in deriving his classical results for optimal mechanisms. Because of this difference, and the transparency and intellectual dominance of Myerson's work, relatively few papers have looked at the case with finite agent types beyond the two-type case.

However, the difference between continuous and finite type spaces can be more than cosmetic. For example, in procurement auctions types represent the cost structures of agent firms, and an analysis of optimal pre-auction investments by agents may lead to single points, or several discrete points, as possible investment (hence, type) solutions. As research moves forward to deconstruct the decision processes that lead up to type the difference between finite and continuous type spaces may become more than a mathematical detail.

The articles that do exist on finite agent types do not follow Myerson's line of argument, for the reasons noted. Rather, they either analyze the two-type case and exploit the natural simplifications that attend that assumption (c.f. Fudenberg and Tirole 2002), or they use mathematical programming or optimal control theory logic (references provided below). The alternative derivations end up, as they must, with accurate expressions for the optimization problem to be solved, sometimes in the form of necessary rather than sufficient conditions. However, these expressions do not as readily suggest the form of the optimal mechanism (at least, in the case of "regular auctions," which we will define below) as Myerson's results. We restrict attention to regular auctions in this paper, because these are where the cleanliness and transparency of Myerson's framework gives it an advantage over the alternative approaches mentioned. Finally, all existing derivations for finite agent types known to the author assume symmetrical agents, while Myerson's paper and this one handle the symmetrical and asymmetrical cases in one unified way.

The mathematical tricks needed to overcome the obstacles to a derivation that parallels Myerson's arguments are well-known and documented, but have yet to be used in that fashion. This note provides this parallel derivation and the finite-type counterpart to Myerson's famous lemma 3. Hence, the component parts of this derivation are not new, although their use in concert to generate our eventual expressions is. The advantage of presenting the results in this way is the intuitive transparency that accrues to those who have already internalized Myerson's classical work. Theorem 1 below can be read alongside Myerson's lemma 3 with no intuitive leaps, and the results here converge to Myerson's as the feasible types become dense in a type space.

The critical result that allows us to circumvent the need for differential arguments is the replacement of global incentive compatibility constraints with local counterparts (lemma 1 below) and showing that even greater simplifications must hold at optimality (lemma 2 below). These arguments are well-known. For the intellectual heritage of this replacement in the context of finite agent types see Harris, Kriebel and Raviv (1982); Hart (1983) and Moore (1988). For text-book articulations of the argument see Stole (2001); Li, Malakhov and Vohra (2002) and Fudenberg and Tirole (2002). Stole and Li et al provide general results for finite agent types through a control theory and mathematical programming lens.

Assume that there are n agents, and agent i has a type $\nu_i \in \Omega_i$ where Ω_i is a finite set with $m_i \geq 1$ elements. We will number these elements as follows. Let $a_i = \nu_i^1 < \nu_i^2 < \dots < \nu_i^{m_i} = b_i$, so a_i and b_i (equivalently ν_i^1 and $\nu_i^{m_i}$) are the least and greatest elements of Ω_i and ν_i^k is increasing in k for $1 \leq k \leq m_i$.

The principal's uncertainty about the value each agent places on the contract is captured in a probability distribution $\pi_i(\nu_i)$, which is the probability that agent i has value ν_i . All other agents share this assessment of agent i 's possible values. The values of various agents are assumed to be independent, so that the principal will assess the probability of any value vector $\nu \in R^n$ with the product $\pi(\nu) = \prod_{i=1}^n \pi_i(\nu_i)$. Agent i knows her own value, but she share's the generic understanding of other agent's values, so that she assesses the probability that the other $(n-1)$ agents have values $\nu_{-i} \in R^{n-1}$ as $\pi_{-i}(\nu_{-i}) = \prod_{j \neq i} \pi_j(\nu_j)$. The principal's value for keeping the contract and not giving it to any of the n agents is ν_0 , and is known to all.

The principal wishes to select a mechanism to maximize his utility. By invoking the revelation principle, the principal can without loss of generality restrict himself to direct revelation mechanisms. In these, each agent reports her true type ν_i so that the principal receives the n -dimensional vector ν in total. The principal declares, prior to the agents issuing these reports, the probability law that will map ν into the allocation of the contract and the expected payment by each agent. That is, the principal declares a *mechanism*, which is a pair of functions $p(\nu)$ and $x(\nu)$ both with ranges in R^n . $p_i(\nu)$ is the probability that agent i will receive the contract, as a function of the vector of messages (ν) received, and $x_i(\nu)$ is the expected payment by agent i , also as a function of the messages received. The principal wishes to select a mechanism (p, x) to maximize his utility, given that the agents will decide what message to deliver after they learn what p and x will be.

We assume that all parties are risk-neutral with additive and separable utilities for money, so that the expected utility for agent i who reports her type truthfully is

$$U_i(p, x, \nu_i) = \sum_{\nu_{-i}} \pi_{-i}(\nu_{-i}) [\nu_i p_i(\nu_{-i}, \nu_i) - x_i(\nu_{-i}, \nu_i)].$$

We ignore the linear revision functions e_i in Myerson's model. The expected utility for the principal is

$$U_0(p, x) = \sum_{\nu} \pi(\nu) [\nu_0 (1 - \sum_{i=1}^n p_i(\nu)) + \sum_{i=1}^n x_i(\nu)].$$

As noted above, and consistent with Myerson's development, we restrict ourselves to direct revelation mechanisms without loss of generality. In these, the principal must choose a mechanism that is consistent with several feasibility constraints. Specifically, a mechanism (p, x) is feasible if it satisfies the following three conditions:

1. p must reflect a legitimate probability distribution, so that $p_i(\nu) \geq 0$ for all i and ν , and $\sum_{i=1}^n p_i(\nu) \leq 1$ for all ν . $1 - \sum_{i=1}^n p_i(\nu)$ is the probability that the principal does not award the contract to any agent (in this case the principal gets ν_0).

2. No agent can be coerced into joining the bidding. That is, each agent must perceive at least as high an expected utility from joining the game as staying out. Mathematically this is the constraint that for all agents i and values $\nu_i \in \Omega_i$, $U_i(p, x, \nu_i) \geq 0$.

3. No agent should be able to do better (in expected utility) by lying, relative to telling the truth about her type. Mathematically this is

$$U_i(p, x, \nu_i) \geq \sum_{\nu_{-i}} \pi_{-i}(\nu_{-i}) [\nu_i p(\nu_{-i}, s) - x(\nu_{-i}, s)]$$

for all agents i , $\nu_i \in \Omega_i$, and $s \in \Omega_i$.

Much of the analysis here (and in Myerson) involves finding a more convenient, but equivalent, form for feasibility. In the literature, (2) are called the "individual rationality" or I.R. constraints, and (3) are called the "incentive compatibility" or I.C. constraints. A key result in the continuous type case is Myerson's lemma 2, which provides an equivalent form for the I.C. constraints. As we have noted, his argument relies on differential reasoning, which does not translate directly for finite type spaces. However, the following two lemmas (which catalog results found in the references cited above) lead to essentially the same result.

First, we define the following variables. Let $Q_i(p, \nu_i) = \sum_{\nu_{-i}} \pi_{-i}(\nu_{-i}) p(\nu_{-i}, \nu_i)$. Thus, $Q_i(p, \nu_i)$ is the probability that agent i will be awarded the contract if she reports ν_i . Also, define $\mathcal{U}_i(s|\nu_i) = \sum_{\nu_{-i}} [\nu_i p_i(\nu_{-i}, s) - x_i(\nu_{-i}, s)]$, which is the expected utility for agent i if she reports value s when her true value is ν_i . If $m_i = 1$ (there is only one possible type) for any agent i , then the I.C. constraints for that agent are automatically satisfied. Otherwise we can invoke the following.

Lemma 1: For any agent i with $m_i \geq 2$, the I.C. constraints are equivalent to the following:

- a) $Q_i(p, \nu_i)$ is nondecreasing in ν_i on Ω_i .
- b) $\mathcal{U}_i(\nu_i^k | \nu_i^k) \geq \mathcal{U}_i(\nu_i^{k-1} | \nu_i^k)$ for all $2 \leq k \leq m_i$.
- c) $\mathcal{U}_i(\nu_i^k | \nu_i^k) \geq \mathcal{U}_i(\nu_i^{k+1} | \nu_i^k)$ for all $1 \leq k \leq m_i - 1$.

Proof: The I.C. constraints are equivalent to $\mathcal{U}_i(\nu_i | \nu_i) \geq \mathcal{U}_i(s | \nu_i)$ for all i , $\nu_i \in \Omega_i$ and $s \in \Omega_i$. Hence, the I.C. constraints imply (b) and (c). But, the I.C. constraints also imply (a). To see this, note that $\mathcal{U}_i(s | \nu_i) = \sum_{\nu_{-i}} \pi_{-i}(\nu_{-i}) [sp_i(\nu_{-i}, s) - x_i(\nu_{-i}, s) + p_i(\nu_{-i}, s)(\nu_i - s)]$
 $= U_i(p, x, s) + \sum_{\nu_{-i}} \pi_{-i}(\nu_{-i}) p_i(\nu_{-i}, s)(\nu_i - s)$
 $= U_i(p, x, s) + Q_i(p, s)(\nu_i - s)$

for all i , $\nu_i \in \Omega_i$ and $s \in \Omega_i$. Now, the I.C. constraints imply (b) and (c), as noted, which in turn imply that

$$\mathcal{U}_i(\nu_i^k | \nu_i^k) - \mathcal{U}_i(\nu_i^{k-1} | \nu_i^k) \geq 0 \geq \mathcal{U}_i(\nu_i^k | \nu_i^{k-1}) - \mathcal{U}_i(\nu_i^{k-1} | \nu_i^{k-1})$$

for all i , and all $2 \leq k \leq m_i$. We can rewrite this as

$$U_i(p, x, \nu_i^k) - U_i(p, x, \nu_i^{k-1}) - Q_i(p, \nu_i^{k-1})(\nu_i^k - \nu_i^{k-1}) \geq U_i(p, x, \nu_i^k) + Q_i(p, \nu_i^k)(\nu_i^{k-1} - \nu_i^k) - U_i(p, x, \nu_i^{k-1}),$$

which is equivalent to

$$Q_i(p, \nu_i^{k-1})(\nu_i^{k-1} - \nu_i^k) \geq Q_i(p, \nu_i^k)(\nu_i^{k-1} - \nu_i^k)$$

which is equivalent to

$$Q_i(p, \nu_i^{k-1})(\nu_i^k - \nu_i^{k-1}) \leq Q_i(p, \nu_i^k)(\nu_i^k - \nu_i^{k-1}).$$

Now, since $\nu_i^k > \nu_i^{k-1}$ we have that

$Q_i(p, \nu_i^{k-1}) \leq Q_i(p, \nu_i^k)$ for all i and $2 \leq k \leq m_i$. So, the I.C. constraints imply that $Q_i(p, \nu_i)$ is nondecreasing in ν_i . So, the I.C. constraints imply (a) - (c). The proof is complete if we can show that (a) - (c) imply the I.C. constraints.

To show this, suppose (a) and (b) hold, then

$$\mathcal{U}_i(\nu_i^k | \nu_i^k) = U_i(p, x, \nu_i^k) \geq \mathcal{U}_i(\nu_i^{k-1} | \nu_i^k) = U_i(p, x, \nu_i^{k-1}) + Q_i(p, \nu_i^{k-1})(\nu_i^k - \nu_i^{k-1}).$$

Likewise by the same argument

$$U_i(p, x, \nu_i^{k-1}) \geq U_i(p, x, \nu_i^{k-2}) + Q_i(p, \nu_i^{k-2})(\nu_i^{k-1} - \nu_i^{k-2}).$$

Hence,

$$\begin{aligned} & \mathcal{U}_i(\nu_i^k | \nu_i^k) \\ &= U_i(p, x, \nu_i^k) \geq U_i(p, x, \nu_i^{k-2}) + Q_i(p, \nu_i^{k-2})(\nu_i^{k-1} - \nu_i^{k-2}) + Q_i(p, \nu_i^{k-1})(\nu_i^k - \nu_i^{k-1}) \end{aligned}$$

which since Q_i is monotone is greater than or equal to

$$\begin{aligned} & U_i(p, x, \nu_i^{k-2}) + Q_i(p, \nu_i^{k-2})(\nu_i^k - \nu_i^{k-1} + \nu_i^{k-1} - \nu_i^{k-2}) \\ &= U_i(p, x, \nu_i^{k-2}) + Q_i(p, \nu_i^{k-2})(\nu_i^k - \nu_i^{k-2}) = \mathcal{U}_i(\nu_i^{k-2} | \nu_i^k). \end{aligned}$$

So, by induction if Q_i is monotone we will have

$$\mathcal{U}_i(\nu_i^k | \nu_i^k) \geq \mathcal{U}_i(s | \nu_i^k) \text{ for all } i, \text{ all } 2 \leq k \leq m_i \text{ and all } s \in \Omega_i \text{ with } s \leq k.$$

A symmetrical argument starting with the upward local constraints (c) and monotonicity (a) reveals that

$$\begin{aligned} \mathcal{U}_i(\nu_i^k | \nu_i^k) &= U_i(p, x, \nu_i^k) \\ &\geq U_i(p, x, \nu_i^{k+1}) + Q_i(p, \nu_i^{k+1})(\nu_i^k - \nu_i^{k+1}) \\ &\geq U_i(p, x, \nu_i^{k+2}) + Q_i(p, \nu_i^{k+2})(\nu_i^{k+1} - \nu_i^{k+2}) + Q_i(p, \nu_i^{k+1})(\nu_i^k - \nu_i^{k+1}) \\ &\geq U_i(p, x, \nu_i^{k+2}) + Q_i(p, \nu_i^{k+2})(\nu_i^k - \nu_i^{k+2}) \end{aligned}$$

and by induction

$$\mathcal{U}_i(\nu_i^k | \nu_i^k) \geq U_i(p, x, s) + Q_i(p, s)(\nu_i^k - s) = \mathcal{U}_i(s | \nu_i^k) \text{ for all } i, \text{ all } 1 \leq k \leq m_i - 1, \text{ and all } s \in \Omega_i \text{ with } s \geq k.$$

Putting these together, we see that (a) - (c) imply that

$$\mathcal{U}_i(\nu_i^k | \nu_i^k) \geq \mathcal{U}_i(s | \nu_i^k) \text{ for all } i, \text{ all } \nu_i^k \in \Omega_i, \text{ and all } s \in \Omega_i. \text{ But, these are just the I.C. constraints. } \mathbf{QED}$$

Armed with this equivalence result, we can recast the principal's mechanism design problem as maximizing $U_0(p, x)$ subject to the following constraints

- 4) $p_i(\nu) \geq 0$ for all i and ν , and $\sum_{i=1}^n p_i(\nu) \leq 1$ for all ν .
- 5) $U_i(p, x, \nu_i) \geq 0$ for all agents i and values $\nu_i \in \Omega_i$.
- 6) $Q_i(p, \nu_i)$ is nondecreasing in ν_i on Ω_i for all agents i .
- 7) $\mathcal{U}_i(\nu_i^k | \nu_i^k) \geq \mathcal{U}_i(\nu_i^{k-1} | \nu_i^k)$ for all agents i such that $m_i \geq 2$ and all $2 \leq k \leq m_i$.
- 8) $\mathcal{U}_i(\nu_i^k | \nu_i^k) \geq \mathcal{U}_i(\nu_i^{k+1} | \nu_i^k)$ for all agents i such that $m_i \geq 2$ and all $1 \leq k \leq m_i - 1$.

A key result allowing further analysis of this problem in closed form is the following.

Lemma 2: In the principal's optimal mechanism design problem of maximizing $U_0(p, x)$ subject to constraints (4) - (8), for all agents i with $m_i \geq 2$, constraints (7) must be binding at optimality, and when these are binding then constraints (8) are automatically satisfied.

Proof: Suppose (p, x) is an optimal mechanism, and constraints (7) are not binding. Then, we must have $\mathcal{U}_i(\nu_i^k | \nu_i^k) = U_i(p, x, \nu_i^k) > \mathcal{U}_i(\nu_i^{k-1} | \nu_i^k) = U_i(p, x, \nu_i^{k-1}) + Q_i(p, \nu_i^{k-1})(\nu_i^k - \nu_i^{k-1})$. This is equivalent to

$U_i(p, x, \nu_i^k) - U_i(p, x, \nu_i^{k-1}) > Q_i(p, \nu_i^{k-1})(\nu_i^k - \nu_i^{k-1}) \geq 0$. So the left hand side must be strictly greater than zero, or

$$\sum_{\nu_{-i}} \pi_{-i}(\nu_{-i}) [\nu_i^k p(\nu_{-i}, \nu_i^k) - x_i(\nu_{-i}, \nu_i^k) - \nu_i^{k-1} p(\nu_{-i}, \nu_i^{k-1}) + x_i(\nu_{-i}, \nu_i^{k-1})] > 0.$$

Then, the principal can raise $x_i(\nu_{-i}, \nu_i^k)$ slightly, increasing his objective but violating no constraints. This contradicts the assumption of optimality. So, at optimality constraints (7) must be binding.

But, now that we know (7) are binding let's look at

$$\mathcal{U}_i(\nu_i^{k+1} | \nu_i^k) - \mathcal{U}_i(\nu_i^k | \nu_i^k) \text{ for any } 1 \leq k \leq m_i - 1.$$

$$\mathcal{U}_i(\nu_i^{k+1} | \nu_i^k) - \mathcal{U}_i(\nu_i^k | \nu_i^k)$$

$= U_i(p, x, \nu_i^{k+1}) + Q_i(p, \nu_i^{k+1})(\nu_i^k - \nu_i^{k+1}) - U_i(p, x, \nu_i^k)$. Because we know (7) are binding, we know that

$$U_i(p, x, \nu_i^{k+1}) = U_i(p, x, \nu_i^k) + Q_i(p, \nu_i^k)(\nu_i^{k+1} - \nu_i^k) \text{ so that}$$

$$\mathcal{U}_i(\nu_i^{k+1} | \nu_i^k) - \mathcal{U}_i(\nu_i^k | \nu_i^k)$$

$$= U_i(p, x, \nu_i^k) + Q_i(p, \nu_i^k)(\nu_i^{k+1} - \nu_i^k) + Q_i(p, \nu_i^{k+1})(\nu_i^k - \nu_i^{k+1}) - U_i(p, x, \nu_i^k)$$

$= (\nu_i^{k+1} - \nu_i^k)(Q_i(p, \nu_i^k) - Q_i(p, \nu_i^{k+1})) \leq 0$ where the last inequality follows because Q_i is nondecreasing in ν_i . So, when (6) is satisfied and constraints (7) are binding, (8) is satisfied. **QED**

Hence, for any agent with $m_i \geq 2$ we can without loss of optimality replace (7) and (8) with the single constraint that (7) is binding, which is equivalent to

$$U_i(p, x, \nu_i^{k+1}) = U_i(p, x, \nu_i^k) + Q_i(p, \nu_i^k)(\nu_i^{k+1} - \nu_i^k) \text{ for all } i \text{ and all } 1 \leq k \leq m_i - 1.$$

This is an expression that is recursive in k and can be rewritten as

(9) $U_i(p, x, \nu_i^k) = U_i(p, x, a_i) + \sum_{j=1}^{k-1} Q_i(p, \nu_i^j)(\nu_i^{j+1} - \nu_i^j)$ for all i and all $1 \leq k \leq m_i$ where we define a sum from 1 to 0 (that is, $\sum_{j=1}^0$ which is the required sum for $k = 1$) to be zero.

Note that (9) is the discrete version of Myerson's equation (4.3). Also, by defining $\sum_{j=1}^0 = 0$ (9) holds for all $m_i \geq 1$ and we can from this point forward analyze any mixture of agents with singleton types ($m_i = 1$) and more ($m_i \geq 2$) in one unified way.

So facing agents with finite type spaces and $m_i \geq 1$ for all i , the principal's mechanism design problem can be expressed as choosing (p, x) to maximize $U_0(p, x)$ subject to the constraints (4), (5), (6), and (9). It is apparent that (9) and $U_i(p, x, a_i) \geq 0$ for all i will imply (5). So, we can write the principal's mechanism design problem as follows.

$$\text{Maximize}_{(p, x)} \sum_{\nu} \pi(\nu) [\nu_0(1 - \sum_{i=1}^n p_i(\nu)) + \sum_{i=1}^n x_i(\nu)]$$

Subject to

$$10) p_i(\nu) \geq 0 \text{ for all } i \text{ and } \nu, \text{ and } \sum_{i=1}^n p_i(\nu) \leq 1 \text{ for all } \nu.$$

$$11) U_i(p, x, a_i) \geq 0 \text{ for all agents } i.$$

$$12) Q_i(p, \nu_i) \text{ is nondecreasing in } \nu_i \text{ on } \Omega_i, \text{ for all agents } i.$$

$$13) U_i(p, x, \nu_i^k) = U_i(p, x, a_i) + \sum_{j=1}^{k-1} Q_i(p, \nu_i^j)(\nu_i^{j+1} - \nu_i^j) \text{ for all } i \text{ and all } 1 \leq k \leq m_i.$$

Now we can pick up Myerson's reasoning and carry it forward, yielding at each step a finite type counterpart to Myerson's results. The next lemma shows that for any p the principal can define an x to satisfy (11) and (13).

Lemma 3: If for any p satisfying (10) and any i and $\nu = (\nu_{-i}, \nu_i^k)$ the principal defines $x_i(\nu) = \nu_i^k p_i(\nu) - \sum_{j=1}^{k-1} (\nu_i^{j+1} - \nu_i^j) p_i(\nu_{-i}, \nu_i^j)$ then $U_i(p, x, a_i) = 0$ and $U_i(p, x, \nu_i^k) = U_i(p, x, a_i) + \sum_{j=1}^{k-1} Q_i(p, \nu_i^j)(\nu_i^{j+1} - \nu_i^j)$ for all i and all $1 \leq k \leq m_i$

Proof: We have by definition

$$U_i(p, x, \nu_i^k) = \sum_{\nu_{-i}} \pi_{-i}(\nu_{-i}) [\nu_i^k p_i(\nu_{-i}, \nu_i^k) - x_i(\nu_{-i}, \nu_i^k)] \text{ and want to choose } x \text{ so that (13) holds. That is, we want to choose } x \text{ so that}$$

$$\begin{aligned} U_i(p, x, \nu_i^k) &= \sum_{\nu_{-i}} \pi_{-i}(\nu_{-i}) [\nu_i^k p_i(\nu_{-i}, \nu_i^k) - x_i(\nu_{-i}, \nu_i^k)] \\ &= U_i(p, x, a_i) + \sum_{j=1}^{k-1} Q_i(p, \nu_i^j)(\nu_i^{j+1} - \nu_i^j) \\ &= U_i(p, x, a_i) + \sum_{j=1}^{k-1} \sum_{\nu_{-i}} \pi_{-i}(\nu_{-i}) p_i(\nu_{-i}, \nu_i^j)(\nu_i^{j+1} - \nu_i^j) \end{aligned}$$

But these equalities are equivalent to

$$U_i(p, x, a_i) = \sum_{\nu_{-i}} \pi_{-i}(\nu_{-i}) [\nu_i^k p_i(\nu_{-i}, \nu_i^k) - x_i(\nu_{-i}, \nu_i^k)] - \sum_{\nu_{-i}} \pi_{-i}(\nu_{-i}) \sum_{j=1}^{k-1} (\nu_i^{j+1} - \nu_i^j) p_i(\nu_{-i}, \nu_i^j)$$

$$= \sum_{\nu_{-i}} \pi_{-i}(\nu_{-i}) [\nu_i^k p_i(\nu_{-i}, \nu_i^k) - x_i(\nu_{-i}, \nu_i^k) - \sum_{j=1}^{k-1} (\nu_i^{j+1} - \nu_i^j) p_i(\nu_{-i}, \nu_i^j)].$$

So if for any $\nu = (\nu_{-i}, \nu_i^k)$ we set

$$x_i(\nu) = \nu_i^k p_i(\nu) - \sum_{j=1}^{k-1} (\nu_i^{j+1} - \nu_i^j) p_i(\nu_{-i}, \nu_i^j)$$

then we simultaneously get $U_i(p, x, a_i) = 0$ and $U_i(p, x, \nu_i^k) = U_i(p, x, a_i) + \sum_{j=1}^{k-1} Q_i(p, \nu_i^j) (\nu_i^{j+1} - \nu_i^j)$. **QED**

We are now ready to prove the main result, which is the discrete analog to Myerson's Lemma 3. For any $\nu_i = \nu_i^k$ and $k \leq m_i - 1$ define $\Delta \nu_i = \nu_i^{k+1} - \nu_i^k$. Define this to be zero for $k = m_i$. That is, for any $\nu_i \in \Omega_i$, $\Delta \nu_i$ is the upward difference to the next highest type for agent i , and for the highest type $\Delta \nu_i = 0$.

Theorem 1: Suppose p maximizes

$$\sum_{\nu} \pi(\nu) \sum_{i=1}^n p_i(\nu) \left[\nu_i - \nu_0 - \frac{1 - F_i(\nu_i)}{\pi_i(\nu_i)} \Delta \nu_i \right]$$

subject to (10) and (12). Suppose also that for any i and $\nu = (\nu_{-i}, \nu_i^k)$

$$x_i(\nu) = \nu_i^k p_i(\nu) - \sum_{j=1}^{k-1} (\nu_i^{j+1} - \nu_i^j) p_i(\nu_{-i}, \nu_i^j).$$

Then (p, x) is an optimal auction.

Proof: The key is to rewrite the principal's objective function as choosing (p, x) to maximize

$$\nu_0 - \sum_{i=1}^n U_i(p, x, a_i) + \sum_{\nu} \pi(\nu) \sum_{i=1}^n p_i(\nu) \left[\nu_i - \nu_0 - \frac{1 - F_i(\nu_i)}{\pi_i(\nu_i)} \Delta \nu_i \right]$$

subject to the constraints (10) - (13). But, x only appears in the second term in the objective function and in the constraints (11) and (13). So, if we choose x to minimize $\sum_{i=1}^n U_i(p, x, a_i)$ subject to (11) and (13) we are doing the best we can. If for any p we define x as in Lemma 3, we have $\sum_{i=1}^n U_i(p, x, a_i) = 0$ which is the best possible outcome for the principal, and (11) and (13) are automatically satisfied. The result then follows. So, the key is to show that the objective function can be written as claimed. The remainder of the proof justifies this. The principal's objective is

$$\begin{aligned} & \sum_{\nu} \pi(\nu) \left[\nu_0 (1 - \sum_{i=1}^n p_i(\nu)) + \sum_{i=1}^n x_i(\nu) \right] \\ &= \nu_0 + \sum_{i=1}^n \sum_{\nu} \pi(\nu) [x_i(\nu) - \nu_0 p_i(\nu)] \\ &= \nu_0 + \sum_{i=1}^n \sum_{\nu} \pi(\nu) [x_i(\nu) - \nu_i p_i(\nu) + \nu_i p_i(\nu) - \nu_0 p_i(\nu)] \\ &= \nu_0 + \sum_{i=1}^n \sum_{\nu} \pi(\nu) [-(\nu_i p_i(\nu) - x_i(\nu)) + p_i(\nu) (\nu_i - \nu_0)] \end{aligned}$$

$$\begin{aligned}
&= \nu_0 + \sum_{i=1}^n \sum_{\nu} \pi(\nu) [p_i(\nu)(\nu_i - \nu_0)] \\
&\quad - \sum_{i=1}^n \sum_{\nu_{-i}} \pi_{-i}(\nu_{-i}) \sum_{k=1}^{m_i} \pi_i(\nu_i^k) [\nu_i^k p_i(\nu_{-i}, \nu_i^k) - x_i(\nu_{-i}, \nu_i^k)].
\end{aligned}$$

Note that this last term is just

$$- \sum_{i=1}^n \sum_{k=1}^{m_i} \pi_i(\nu_i^k) U_i(p, x, \nu_i^k).$$

We will now expand this using (13) to get

$$\begin{aligned}
&- \sum_{i=1}^n \sum_{k=1}^{m_i} \pi_i(\nu_i^k) U_i(p, x, \nu_i^k) \\
&= - \sum_{i=1}^n \sum_{k=1}^{m_i} \pi_i(\nu_i^k) [U_i(p, x, a_i) + \sum_{j=1}^{k-1} Q_i(p, \nu_i^j)(\nu_i^{j+1} - \nu_i^j)] \\
&= - \sum_{i=1}^n \sum_{k=1}^{m_i} \pi_i(\nu_i^k) U_i(p, x, a_i) - \sum_{i=1}^n \sum_{k=1}^{m_i} \sum_{j=1}^{k-1} \pi_i(\nu_i^k) Q_i(p, \nu_i^j)(\nu_i^{j+1} - \nu_i^j).
\end{aligned}$$

It can be verified that we can resequence the sums $\sum_{k=1}^{m_i} \sum_{j=1}^{k-1} = \sum_{j=1}^{m_i-1} \sum_{k=j+1}^{m_i}$ so that the term is

$$- \sum_{i=1}^n \sum_{k=1}^{m_i} \pi_i(\nu_i^k) U_i(p, x, a_i) - \sum_{i=1}^n \sum_{j=1}^{m_i-1} \sum_{k=j+1}^{m_i} \pi_i(\nu_i^k) Q_i(p, \nu_i^j)(\nu_i^{j+1} - \nu_i^j).$$

Because $\sum_{k=1}^{m_i} \pi_i(\nu_i^k) = 1$ and $\sum_{k=j+1}^{m_i} \pi_i(\nu_i^k) = (1 - F_i(\nu_i^j))$ this term is

$$= - \sum_{i=1}^n U_i(p, x, a_i) - \sum_{i=1}^n \sum_{j=1}^{m_i-1} Q_i(p, \nu_i^j)(\nu_i^{j+1} - \nu_i^j)(1 - F_i(\nu_i^j)).$$

By the definition of Q_i this is

$$= - \sum_{i=1}^n U_i(p, x, a_i) - \sum_{i=1}^n \sum_{j=1}^{m_i-1} \sum_{\nu_{-i}} \pi_{-i}(\nu_{-i}) p_i(\nu_{-i}, \nu_i^j)(\nu_i^{j+1} - \nu_i^j)(1 - F_i(\nu_i^j))$$

Noting that for any $\nu = (\nu_{-i}, \nu_i^j)$, $\pi_{-i}(\nu_{-i}) = \frac{\pi(\nu_{-i}, \nu_i^j)}{\pi_i(\nu_i^j)}$ by the independent values assumption, we have that this term is

$$\begin{aligned}
&= - \sum_{i=1}^n U_i(p, x, a_i) - \sum_{i=1}^n \sum_{j=1}^{m_i-1} \sum_{\nu_{-i}} \frac{\pi(\nu_{-i}, \nu_i^j)}{\pi_i(\nu_i^j)} p_i(\nu_{-i}, \nu_i^j)(\nu_i^{j+1} - \nu_i^j)(1 - F_i(\nu_i^j)) \\
&= - \sum_{i=1}^n U_i(p, x, a_i) - \sum_{i=1}^n \sum_{\nu_{-i}} \sum_{j=1}^{m_i-1} \frac{\pi(\nu_{-i}, \nu_i^j)}{\pi_i(\nu_i^j)} p_i(\nu_{-i}, \nu_i^j)(\nu_i^{j+1} - \nu_i^j)(1 - F_i(\nu_i^j))
\end{aligned}$$

Recalling that $\Delta\nu_i = 0$ when $\nu_i = \nu_i^{m_i}$ (and, anyway, $1 - F_i(\nu_i^j) = 0$ under those conditions), we can equivalently take the sum from $j = 1$ to m_i , and the term is

$$= - \sum_{i=1}^n U_i(p, x, a_i) - \sum_{i=1}^n \sum_{\nu} \frac{\pi(\nu)}{\pi_i(\nu_i)} p_i(\nu_{-i}, \nu_i) \Delta\nu_i (1 - F_i(\nu_i))$$

Now, adding this term back we get the principal's objective function as

$$\begin{aligned}
&\nu_0 + \sum_{i=1}^n \sum_{\nu} \pi(\nu) p_i(\nu)(\nu_i - \nu_0) - \sum_{i=1}^n U_i(p, x, a_i) - \sum_{i=1}^n \sum_{\nu} \frac{\pi(\nu)}{\pi_i(\nu_i)} p_i(\nu_{-i}, \nu_i) \Delta\nu_i (1 - F_i(\nu_i)) \\
&= \nu_0 - \sum_{i=1}^n U_i(p, x, a_i) - \sum_{\nu} \pi(\nu) \sum_{i=1}^n p_i(\nu) \left[\nu_i - \nu_0 - \frac{1 - F_i(\nu_i)}{\pi_i(\nu_i)} \Delta\nu_i \right]
\end{aligned}$$

which is the desired expression, completing the proof. **QED**

Note that for any agent with a singleton type, $m_i = 1$, principal will ask for the transfer $x_i(\nu) = \nu_i p_i(\nu)$ and $U_i(p, x, \nu_i) = U_i(p, x, a_i) = 0$. That is, with complete information about an agent's type, the principal extracts all potential rents from that agent, as expected.

Optimal mechanisms in the regular case

The beauty of this result is the simplicity with which optimal mechanisms can be identified, almost by inspection. Myerson, again, provides the conceptual groundwork. However, direct translation of Myerson's results is complicated by the fact that Myerson's paper contains a minor error in his section on regular auctions, and in his continuous type case ties need not be considered because they are probability zero events. This is not the case with finite type spaces.

Refining Myerson's results

We begin by refining Myerson's results for continuous type spaces. Recall that we are ignoring Myerson's linear revision functions e_i . In this case, Myerson's "virtual value" for agent i is defined as

$$\nu_i - \frac{1 - F_i(\nu_i)}{f_i(\nu_i)}.$$

An auction with continuous type spaces is called "regular" if this virtual value is strictly increasing in type ν_i . It is apparent from the objective function in Myerson's lemma 3 that if all agents have virtual values less than ν_0 , the principal should not award the contract ($p_i = 0$ for all agents i). This defines the reservation price for the auction. Otherwise, the principal will want to award the contract to the agent with the highest virtual value ($p_i = 1$ for that agent). Equivalently, one can consider the principal as another bidder, who submits a bid with virtual value ν_0 , and the contract goes to the bidder with the highest virtual value. As noted, in the continuous type case ties are probability zero events and can be ignored in expectation.

The problem with this easy prescription is that in the general case we might violate the constraint that the probability of winning be nondecreasing in type (Myerson's equation 4.2). However, this is not a problem for regular auctions. If the agent with the highest virtual value gets the contract in a regular auction, then an agent's probability of winning is automatically nondecreasing in type (because the virtual value is). Hence, the optimal allocation p is easy in the regular case, and then the optimal transfers x can be computed directly from p as in Myerson's lemma 3.

To make this explicit, Myerson defines the function $z_i(\nu_{-i})$ to be the infimal type for agent i such that agent i 's virtual value meets or exceeds both the principal's value ν_0

and the virtual value of all other agents. This infimum is attained if the virtual value is a continuous function, which is commonly assumed. In that case, $z_i(\nu_{-i})$ is the minimal winning type for agent i given the values of all other agents. Myerson's equation (5.4) is appropriate in his setting, in which ties can be ignored. However, with either continuous or finite type spaces, the influence of a_i (the lower support of the beliefs regarding agent i 's type) on the optimal transfers cannot be ignored, as Myerson apparently does in his equations (5.5) and (5.6). Myerson derives equations (5.5) and (5.6) as if $z_i(\nu_{-i}) \geq a_i$ always. It can be shown that this is guaranteed in the symmetric case, but not in general. With continuous types, the correct expression for Myerson's (5.5) should be

$$\int_{a_i}^{\nu_i} p_i(\nu_{-i}, s) ds = [\nu_i - \max\{a_i, z_i(\nu_{-i})\}]^+$$

where $[x]^+$ denotes the greater of x and zero. It follows that the $z_i(\nu_{-i})$ term should be replaced by $\max\{a_i, z_i(\nu_{-i})\}$ in Myerson's optimal transfer equation (5.6). It is apparent that if agent i cannot value the contract below a_i dollars, and the principal knows this, then if the principal awards the contract to agent i he will ask for a_i dollars even if all competing bids are strictly below a_i . This revision is not required for symmetric auctions, which dominate current applications. In the asymmetrical case, however, the difference can be significant.

Optimal mechanisms in the regular case with finite agent types

To construct optimal auctions in the finite type case we follow Myerson's logic refined as suggested above, using our theorem 1 rather than Myerson's lemma 3 as a point of departure. Define agent i 's virtual value by

$$V_i(\nu_i) = \nu_i - \frac{1 - F_i(\nu_i)}{\pi_i(\nu_i)} \Delta \nu_i.$$

We call the auction "regular" if $V_i(\nu)$ is nondecreasing in ν for all agents i . It is easily shown that if all agents have only two possible types the auction is always regular, whether or not the agent types are symmetrical.

Again, we can consider the principal to submit a "bid" with virtual value ν_0 and we will want to award the contract to the bidder with the highest virtual value if that bidder is unique. We discuss ties below. As before, by awarding the contract to agents with the highest virtual value, constraint (12) is automatically satisfied in the regular case.

The simplest case is where a single agent, i say, has the unique maximal virtual value. Then $p_i(\nu) = 1$ and $p_j(\nu) = 0$ for all $j \neq i$. If the lowest possible value in agent i 's feasible set, $a_i = \nu_i^1$, is higher than any competing bid, it can be verified that the optimal transfer

in Theorem 1 reduces to $x_i(\nu) = a_i$ and $x_j(\nu) = 0$ for all $j \neq i$. If, on the other hand, there is an integer r such that $\nu_i^r > a_i$ is the smallest feasible winning bid for agent i , yet still a unique maximum among the competing bids, then the optimal transfer from the winning agent is i is ν_i^r instead of a_i . This is consistent with Myerson's results, refined as above.

With finite agent types ties may occur with positive probability. This will always be the case with symmetric auctions. The principal can break ties arbitrarily, but the choice can affect transfers and hence the potential information rents to agents. We illustrate this with an example motivated by Harris and Raviv (1981). Those authors prove the form of an optimal mechanism by constructing an upper bound and then proving their mechanism attains it. Here we will derive the result directly from Theorem 1.

There are two symmetrical agents each drawing her type uniformly from a finite set $[\nu^1, \nu^2, \dots, \nu^r]$. Hence, the probability that any one type from this set is drawn is $1/r$. Further, the type levels are equally spaced, so that $\nu^k - \nu^{k-1} = \delta$ a constant, for $k \geq 2$. The principal's value ν_0 is fixed and known to all. In this case, $1 - F_i(\nu^k) = (r - k)/r$ and the virtual value for agent i with $\nu_i = \nu_i^k$ is $\nu_i^k - (r - k)\delta$ for $1 \leq k \leq (r - 1)$. For the highest type the virtual value is just ν_i^r . Hence, the virtual value is strictly increasing in type, and this is a regular auction. The principal should award the contract to the highest bidder, if that bidder is unique and has a virtual value exceeding ν_0 (Harris and Raviv assume that $\nu_0 > 2\nu^1 - \nu^r$ which is sufficient to imply that the contract is awarded with probability one). It is implicit in Harris and Raviv that each agent has an equal probability of winning in the case of a tie, but here we allow potentially weighted schemes. Assume that in the case of a tie, the principal will give the contract to agent 1 with probability q , and to agent 2 with probability $(1 - q)$. We now look at the potential resolutions for this auction.

Suppose that agent 1 wins the auction outright because $\nu_1 = \nu^k > \nu_2 = \nu^{\hat{k}}$ where $\hat{k} < k$. Hence, k denotes the winning bid and \hat{k} the second price. In this case agent 1 is asked to pay $x_1(\nu) = q\nu^{\hat{k}} + (1 - q)\nu^{\hat{k}+1}$. So, when $q < 1$ she is asked to pay more than the second price, the extra payment decreasing in her probability of winning ties. This is not the case with a continuous type space. If agent 2 wins ($\nu_2 = \nu^k > \nu_1 = \nu^{\hat{k}}$, so \hat{k} still denotes the second price) she is asked to pay $x_2(\nu) = (1 - q)\nu^{\hat{k}} + q\nu^{\hat{k}+1}$, with the same interpretation except the payment increases in q . If they tie ($\nu^1 = \nu^2$) then the contract is awarded according to q and the agents are asked to pay (in expectation) $x_1(\nu) = q\nu_1$ and $x_2(\nu) = (1 - q)\nu^2$. One obvious implementation in the case of a tie is to award the contract according to q and ask for full value from the winning agent. So, agents earn no rent in the case of a tie. Harris and Raviv show that an optimal mechanism will ask for payment equal (in expectation) to the second price plus half of δ from a unique winner,

and half the value from each agent in a tie. This is consistent with the above optimal mechanism with $q = 1/2$. The discrete nature of the type space decreases the information rents to agents relative to a second price auction.

With the more general tie-breaking rule $q \neq .5$ and a discrete type space, more information rents accrue to the favored agent unless the agents tie.

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